

IMPLEMENTATION OF AN UNDERWATER
DIGITAL ACOUSTIC TELEMETRY
RECEIVER

By

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IMPLEMENTATION OF AN UNDERWATER DIGITAL ACOUSTIC TELEMETRY RECEIVER

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An Abstract of the Thesis Presented
in Partial Fulfillment of the Requirements for the
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This thesis presents the design and software implementation of an underwater acoustic modem receiver. Communication links in underwater environments face several undesired effects. These include multipath signal reflections, intersymbol interference, and channel fading. This receiver design uses a combination of time and spatial diversity inputs combined with an adaptive feedback equalizer to counteract those effects.

The design is based on three modules. A front-end module demodulates and Doppler-compensates the incoming data. A channel combiner module receives data from one or more front ends for spatial diversity and combines repeated transmissions for time diversity. The data from each input channel is time aligned and stored in a 'job' structure. The channel combiner also calculates tap sizes and locations for the feedback equalizer. Completed 'job' structures from the channel combiner are then sent to an equalizer module.

The modules are implemented in C language code written and compiled for Analog Devices SHARC digital signal processors. The hardware consists of several

processors that are interconnected via link ports. This allows each module to run on a separate processor. It also allows for multiple instances of certain modules to be run simultaneously to provide real-time operation.

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CHAPTER 1

Introduction

1.1 Background

Submarine warfare simulation exercises conducted by the Naval Undersea Warfare Center (NUWC) have stimulated an interest in reliable underwater communication links. NUWC requires the ability to exchange data between submerged submarines, surface ships, and seafloor hydrophone arrays in shallow water environments [1].

While acoustical signals propagate reasonably well in water, several issues must be addressed when designing an underwater communications link. Tests conducted for the Seaweb '98 program [2] and NUWC [3] have revealed that multipath signal spread is a major problem associated with underwater communications in shallow water environments. Inter-symbol interference and signal fading are other issues that must be dealt with as well.

A study conducted by NUWC [4] has shown that a decision feedback equalizer (DFE) coupled with a digital phase locked loop (DPLL) can compensate for these undesired effects in underwater environments. The DPLL helps track phase shifts that are too rapid for the DFE. The equalizer design proposed in [4] also makes use of multiple diversity inputs to help combat problems due to channel fading.

1.2 Purpose of the Research

The Test and Evaluation Department at the NUWC Newport Division originally developed and tested a prototype Underwater Digital Acoustic Telemetry (UDAT) system using Texas Instruments TMS320C40 digital signal processors

[5, 1]. This thesis presents the design and implementation of an improved UDAT system using Analog Devices SHARC processors.

NUWC's prototype UDAT system, or modem, utilized two TMS3240C40 DSPs for each of its four functional blocks: (1) packet detection and synchronization, (2) Doppler estimation and compensation, (3) complex demodulation, and (4) equalization. The modem presented in this thesis uses a redesigned Doppler compensation algorithm that has been combined with a redesigned complex demodulation algorithm, thereby allowing the two to run on a single SHARC processor. The packet detection and synchronization algorithms have also been redesigned, allowing them to run real-time on a single SHARC DSP. The redesign of the packet detection and synchronization includes two additional capabilities. First is the ability to automatically choose equalizer parameters based on channel information. Second is the capability to handle multiple acoustic sensors.

A special initialization sequence and manual fine tuning were required to select equalizer parameters in NUWC's prototype modem. That design results in fixed equalizer parameters that may not remain optimal as the channel changes with time. The modem described in this thesis automatically gathers channel characteristics from the synchronization pings and uses them to periodically update the equalizer parameters.

The modem presented in this thesis is capable of handling signals from multiple physically separated acoustic sensors. This spatial diversity technique can help lower error rates. Another technique known as time diversity also lowers error rates. Time diversity requires the transmitter to repeat the message several times (usually twice). The modem described in this thesis is capable of using both spatial and time diversity. In contrast, the prototype modem developed by NUWC utilized time diversity operation only.

Tests of the UDAT system conducted by NUWC [5] revealed that strong multipaths arriving several milliseconds after the main signal created problems with equalization. Increasing the equalizer's memory to include the late arriving paths usually is not practical in real-time systems [5]. The equalizer used in the modem presented in this thesis utilizes a sparse feedback section to include late arriving signal paths without adding excessive computational complexity.

The modem implementation discussed in this thesis is expected to allow a maximum data rate of 1800 bits/sec at a range of 2 nautical miles in shallow water. It also has the capability to track Doppler shifts up to $\pm 2\%$ in order to compensate for transmitters and receivers aboard moving submarines and ships.

1.3 Thesis Organization

Chapter 2 provides an introduction to the modules that make up the UDAT system and shows how they are interconnected. The system's data frame layout is also presented in Chapter 2. Descriptions of each module appear in greater detail in later chapters.

In chapter 3, the front-end module is described. This module demodulates and Doppler-compensates the incoming data. The front-end module also provides channel information in the form of Target ID (TID) correlations.

Chapter 4 presents the implementation and benchmarking of the decision feedback equalizer-phase locked loop (DFE-DPLL). The DFE-DPLL, or equalizer, is used to correct for multipath channel and fading effects.

Chapter 5 discloses the channel combiner module that collects data and channel information from one or more front-end modules. This module time aligns the incoming data frames and submits them to an equalizer module. The channel combiner also gathers information about the channel characteristics that are used to configure the equalizer.

Testing procedures and results used to verify proper operation of the system are discussed in Chapter 6. This chapter also presents recommendations for future changes and improvements.

CHAPTER 2

UDAT System Overview

This chapter provides an overview of the three modules that comprise the Underwater Digital Acoustic Telemetry (UDAT) receiver. It also introduces some notation that will be used throughout the later chapters. A description of the data frame format used by the UDAT receiver is also described in this chapter.

The software implementation of this receiver is implemented on Analog Devices Super Harvard Architecture Computer (SHARC) digital signal processors. The modular design of the software makes use of link port interconnections to exchange data between the various modules. That design allows for flexibility in the system configuration. The original design was constructed and tested on a Morocco II carrier board with 8 SHARC processors [6]. Work is currently under way to move the system to a Hammerhead platform that supports four faster SHARC processors [7].

2.1 UDAT Receiver Modules

The UDAT receiver is comprised of one or more front-end modules, a channel combiner module, and one or more equalizer modules. These modules are connected as shown in Figure 2.1.

One front-end module is required for every acoustic input. Exploitation of spatial diversity requires the use of multiple, physically separated, acoustic sensors - each with a dedicated front-end module. Exploitation of time diversity can be achieved with a single front-end module where repeated transmissions of the same data helps combat time varying channel fading effects.

The equalizer modules implement the mathematical operations required to determine and track dynamic channel characteristics. The equalizers make use of

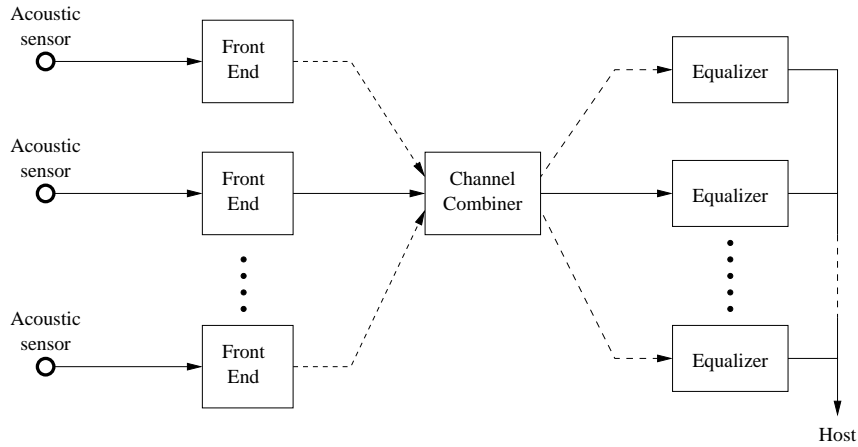


Figure 2.1: Interconnection of UDAT receiver modules.

known training data to learn the channel characteristics. Due to the numerically intense nature of these calculations, multiple equalizers on separate processors are supported. The clock speed of the SHARC processors along with the channel parameters are the key factors that determine how many equalizer modules must be used to provide real-time operation. Chapter 4 contains equalizer benchmarks that give examples for some typical equalizer parameters. Each equalizer module submits its demodulated data output to a host system. The host system is typically a Unix system that is responsible for controlling and receiving data from a group of SHARC processors.

The channel combiner module ties the multiple front-end and equalizer modules together. It is responsible for performing time alignment of the incoming data. The time-aligned receptions from various acoustic sensors are then submitted as “equalizer jobs” to one of the equalizer modules. The time alignment is based on a synchronization ping that appears at the beginning of each data packet. That same ping and its echoes are also used to gather channel information used to configure the equalizers.

2.2 Sensor, Channel, and Job Notation

As was presented in Section 1.2, the UDAT system uses both time and/or spatial diversity to reduce reception errors. Time diversity is implemented by combining two transmissions of the same data into two equalizer input “channels.” Spatial diversity involves combining receptions of the same data frame from two physically separated “sensors” into two equalizer input “channels.” It is necessary at this time to distinguish between “sensors” and “channels”. Sensors are physical input devices (such as hydrophones) which provide an acoustic input to a front-end module. The sensor outputs are demodulated and Doppler compensated by the front end. Multiple copies of these outputs for the same transmitted data (either from spatial or time diversity) are time aligned by the channel combiner and submitted to an equalizer for demodulation. For a given equalizer job, each copy of the time aligned receptions is known as a “channel”.

The equalizer’s input channels can come from any combination of time and/or spatial diversity inputs. For example, two sensors could provide four input channels to the equalizer when the system is operating in both time and spatial diversity modes. Or, two sensors could provide two input channels when operating in spatial diversity mode. Alternatively, one sensor could provide two input channels when operating in time diversity mode.

In addition to time aligning the sensor waveforms, the channel combiner is responsible for determining the dominant channel characteristics and passing this information along to the equalizer. This information is stored in a structure called an “equalizer job”. Equalizer jobs are written by the channel combiner and sent to equalizers as described in Chapter 5.

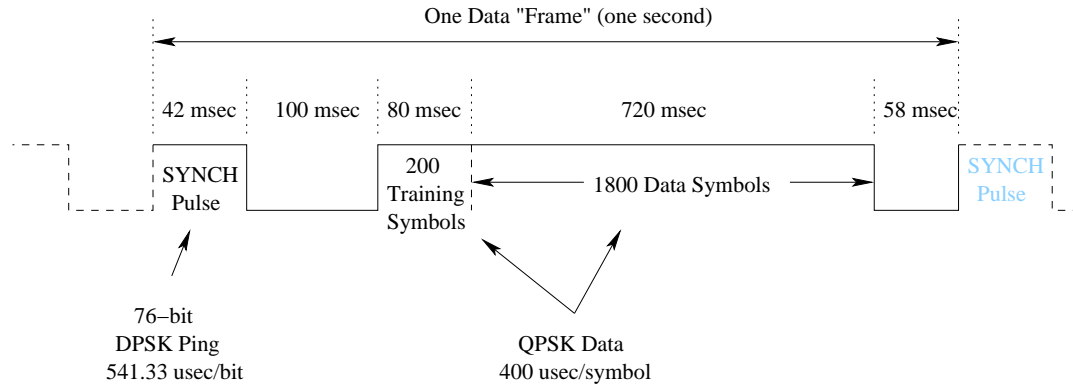


Figure 2.2: Format of a single telemetry data frame.

2.3 Data Frame/Modulation Format

The UDAT system transmits data in a data frame format illustrated in Figure 2.2. Each data frame is one second long and contains the following sections.

- 42 msec target identification (TID) ping
- 100 msec quiet time
- 800 msec modulated data
- 58 msec quiet time

2.3.1 TID Ping

The data frame is initiated using a synchronization pulse consisting of a 76-bit Differential Phase Shift Keyed (DPSK) ping. The receiver, described in Chapter 3, supports the detection of two possible “target ID” bit sequences within this synch pulse. To be compatible with tracking ping formats currently used by NUWC, the DPSK modulation format for the synchronization ping is different from that of the information portion of the frame. The bit duration within the synchronization pulse is 541.33 μ sec, giving a ping duration of 41.682 msec.

These TID pings are used by the channel combiner (described in Chapter 5) to time align the received data frames. These pings are also used to identify data frame numbers when the system is operating in time diversity mode.

2.3.2 Quiet Times

The synchronization ping is followed by a 100 msec quiet time in which no data is transmitted. The receiver utilizes this period to obtain possible multipath delays for the channel. The channel combiner uses synchronization ping detections within this time period to indicate the channel delays that are likely to include significant energy for the following data. Further details of this operation appear in Section 5.4. The channel combiner then passes this delay information on to the equalizer.

The 58 msec quiet time at the end of the frame helps prevent echoes of one frame from interfering with the next frame.

2.3.3 Modulated Data

The information period of the data frame consists of 2000 Quadrature Phase Shift Keyed (QPSK) symbols, 200 “training symbols” followed by 1800 data symbols. The training symbols are known by both the transmitter and receiver. They are used to “train” the adaptive equalizer to “learn” the channel characteristics. Training data is included with each data frame to allow the equalizer to “re-learn” the channel characteristics once every second, thereby compensating for rapidly changing channels.

The symbol duration for this portion of the frame is 400 μsec , so that the information portion of the frame lasts a total of 800 msec. As was stated above, this information period is broken down into training and real data sections. The

200 symbols of training data occupy the first 80 msec, while the real data occupies the last 720 msec of this interval.

Chapter 3 presents the UDAT receiver front-end module used to receive this specialized signal format.

CHAPTER 3

Receiver Front End

This chapter describes the theory behind the receiver front-end module of the UDAT system. It also presents details of the software implementation on the SHARC processors.

3.1 Signal Format

In addition to the modulated data frame described in Section 2.3, the transmitter sends a fixed CW “pilot tone”, which is exploited by the receiver to aid in synchronization. The pilot tone frequency is selected as one of the nulls in the spectrum of the QPSK information portion of the frame. For the 400 μ sec symbol duration, the null-to-null bandwidth of the QPSK signal is 5 kHz, so that the pilot frequency is selected as 2.5 kHz above or below the QPSK carrier frequency. The primary role of the pilot tone is to allow the receiver to cope with unknown Doppler shifts in the modulated signal. The receiver has been designed to allow Doppler shifts of up to $\pm 2\%$, for carrier frequencies ranging from 10 kHz to 40 kHz.

3.2 Front End Tasks

The front-end module is responsible for processing samples associated with a particular hydrophone, and delivering complex matched filter outputs to the channel combiner, which is running on a separate processor. In particular, the front end must perform the following tasks:

- Convert samples of the bandpass transmitted signal to a complex representation.
- Compensate for the Doppler shift of the received signal.

- Detect the presence of the pilot tone, which is an indication that telemetry data is being transmitted.
- Normalize the received waveform by the magnitude of the pilot tone, so that a fixed amplitude received signal is delivered to the channel combiner.
- Calculate the QPSK matched filter outputs, and deliver these values to the channel combiner.
- Find the correlation of the received waveform with two different 76-bit target ID's. The maximum of the two correlations and the associated target ID are delivered to the channel combiner.

Note that although the synchronization ping correlation is performed in the front end, the actual detection and frame alignment is performed by the channel combiner. The receiver front end produces a continuous output stream of correlation values and matched filter outputs, without any regard for the particular data frame structure shown in Figure 2.2.

The front end output data streams are delivered to the channel combiner at a rate of 5000 samples/sec. Each complex matched filter output symbol is accompanied by a correlation value (indicating the larger of the two calculated 76-bit correlations), the ID associated with this maximum, and an indication of whether the pilot tone has been detected.

3.3 Front End Overview

Figure 3.1 shows a block diagram of the receiver front-end module. The receiver input consists of a stream of samples from an acoustic sensor using a sample frequency of $F_s = 104.4375$ ksps.

The signal of interest is assumed modulated at a carrier frequency $\Omega_c = 2\pi F_c$ rad/sec, and has a null-to-null bandwidth of 5 kHz. For efficiency, bandpass

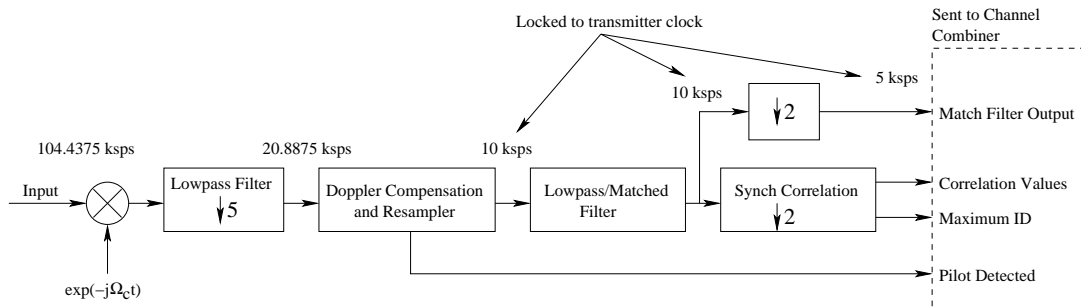


Figure 3.1: Functional block diagram of the telemetry receiver front-end.

signals are manipulated within the receiver front end using a complex representation. For a real bandpass signal $x(t)$, the complex representation is given by

$$\tilde{x}(t) = \text{L.P.P.} \{x(t)e^{-j\Omega_c t}\} \quad (3.1)$$

where the notation $\text{L.P.P.}\{\}$ denotes the “low pass portion” of the signal in the argument. The original signal $x(t)$ can be reconstructed from $\tilde{x}(t)$ using

$$x(t) = 2\text{Re} \{ \tilde{x}(t)e^{+j\Omega_c t} \} \quad (3.2)$$

$$= 2|\tilde{x}(t)| \cos(\Omega_c t + \angle\tilde{x}(t)) \quad (3.3)$$

The magnitude and phase of the complex representation $\tilde{x}(t)$ provide the envelope and phase of the corresponding bandpass signal. No information about $x(t)$ is lost in manipulating $\tilde{x}(t)$, and reduced sample rates can be used to describe the complex representation (saving computations). The mixer and lowpass filter portions of the front end serve to extract the complex representation of the input sample sequence, and reduce the sample frequency by a factor of five to 20.8875 kbps (complex).

Doppler shifts in the received signal cause the apparent carrier frequency and bandwidth of the received signal to differ from the values generated by the transmitter. The “Doppler Compensation and Resampler” block of the receiver

compensates for this effect by modifying the sample frequency implemented at the receiver. In short, the receiver sampling frequency is adjusted until the transmitted pilot tone is phase-locked to a locally generated pilot tone at the appropriate frequency. This procedure effectively locks the receiver sample rate to that of the transmitter. Since Doppler compensation requires resampling of the input signal, it is straightforward to also implement an additional decimation by a factor of (approximately) 2 within this stage. This relaxes the requirements of the first decimation filter, saving computations. Doppler compensation implementation using the complex representations are presented in Section 3.5.

The output of the Doppler Compensation block is a stream of complex samples at sample rate of 10 ksps. A lowpass filter is used to limit the bandwidth of this signal to approximately 2.5 kHz to avoid aliasing in the final decimation stage. The QPSK matched filter output for a given time index involves summing the filter outputs over the most recent four samples (400 μ sec). This operation is combined with the calculation of the lowpass filter output, so that a single filter is used to implement both the lowpass and matched filter operations. This sequence is then decimated (by 2) to form the “Matched Filter Output” that is provided to the channel combiner.

“Synch Correlation” outputs of the receiver front end provide data frame synchronization. The matched filter outputs are correlated with the known bit sequences for the possible synchronization pings. Since phase ambiguity exists between the transmit and receive clocks, the magnitude of the (complex) correlation is used to detect the presence of the synchronization ping. Correlation values are calculated for every other matched filter output, giving an output sample rate of 5 ksps. For each output sample, the maximum correlation value is reported, along with the ping ID associated with the maximum.

The following sections provide additional details regarding the design and implementation of the individual components of the receiver front-end module.

3.4 Calculation of the Complex Representation

The complex representation of the input signal is given in equation (3.1). Samples of the complex representation are calculated from the sequence of input samples by filtering the signal $x(kT_s)e^{-j\Omega_c kT_s}$, where $T_s = 1/F_s$. Values of the exponential can be calculated recursively by setting

$$e_0 = 1 \tag{3.4}$$

$$e_k = e_{k-1} \exp(-j\Omega_c T_s) \quad k = 1, 2, \dots \tag{3.5}$$

$$= e^{-j\Omega_c kT_s} \tag{3.6}$$

The transcendental function $\exp(-j\Omega_c T_s)$ can be evaluated once, so that calculation of the exponential portion of the sequence involves a single complex multiplication. The sequence $e_k x(kT_s)$ must then be filtered to produce samples of $\tilde{x}(t)$.

The null-to-null spectrum of $x(t)$ occupies the band $F_c - 2.5$ kHz to $F_c + 2.5$ kHz. Doppler shifts may shift the apparent frequency at the receiver by up to $\pm 2\%$, so that for a carrier frequency of $F_c = 40$ kHz, the band of interest occupies roughly $F_c \pm 3.3$ kHz. The lowpass filter used to create $\tilde{x}(t)$ must pass frequencies below 3.3 kHz. To reduce the sample rate by a factor of five to 20.08875 kHz, the filter must eliminate frequencies within 3.3 kHz of 20.08875 kHz to avoid the undesired signal from aliasing into the band of interest. A 25 coefficient FIR lowpass filter was designed to provide 80 dB of stopband rejection. The magnitude response of the filter is shown in Figure 3.2.

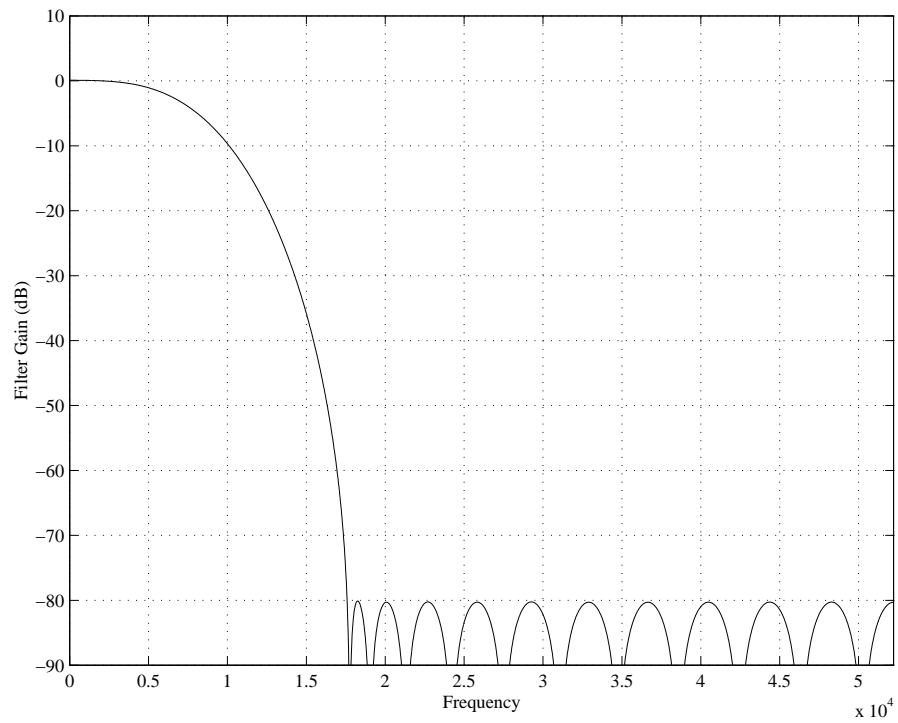


Figure 3.2: Magnitude response of the first FIR decimation filter, used to reduce the sample rate by a factor of five.

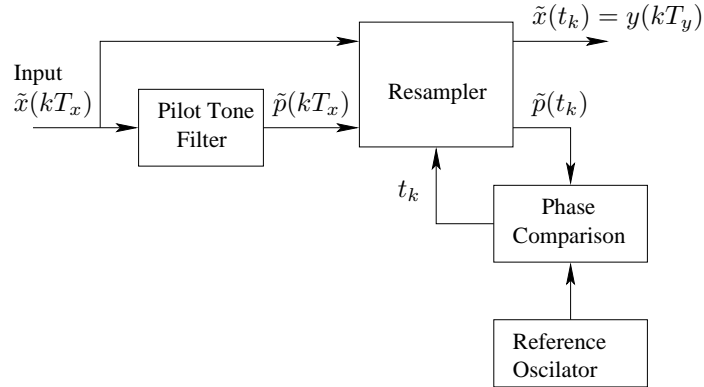


Figure 3.3: Block diagram of the Doppler compensator and resampler.

3.5 Doppler Tracking

Figure 3.3 provides a conceptual block diagram of the Doppler compensation and resampler portion of the receiver front-end module. The system exploits the presence of a pilot-tone at one of the QPSK spectral nulls to lock the sampling frequency to that used in the transmitter. The complex input for the compensator consists of samples of $\tilde{x}(t)$ at sample frequency $F_x = F_s/5 = 20.08875$ kHz. This signal is filtered to extract the pilot-tone portion of the signal. (The filter is not strictly required, and the development described in this section does not assume that the pilot tone has been separated from the modulation. However, marginal performance improvements were realized when the filter was included, particularly for low pilot tone amplitudes. As a result, the pilot tone filter was included in the final implementation.)

The resampler allows calculation of the values of $\tilde{x}(t)$ or $\tilde{p}(t)$ at times that are not multiples of the input sample period $T_x = 1/F_x$. The resampler implemented in this receiver is capable of calculating values of these signals with approximately $1 \mu\text{sec}$ granularity in the sample time. (Note that changing the sample rate for the complex representation is slightly more complicated than interpolating the input signals, since the value of T_s used in (3.5) must also be modified—see

Section 3.5.3.) The goal of the Doppler compensation section is to generate a sequence of input sample times t_k so that the output signal sample rate is $F_y = 10$ ksp/s, *locked to the transmitter clock*. To accomplish this, the pilot-tone portion of the transmitted signal is compared to a locally generated pilot tone. Phase errors between the two signals are used to adjust the sampling clock.

The remainder of this section gives the mathematical development needed to implement the resampler and Doppler compensation.

3.5.1 Complex Representations of Bandpass Signals

To develop the Doppler tracking algorithm, this subsection reviews the complex representation notation used in this thesis, and gives a few examples of signals that will prove useful. Let $x(t)$ denote a bandpass signal of interest with center frequency Ω_c rad/s and bandwidth B Hz. The complex representation of $x(t)$ given by (3.1) is restated here

$$\tilde{x}(t) = \text{L.P.P.} \{x(t)e^{-j\Omega_c t}\}. \quad (3.7)$$

The original signal $x(t)$ can be reconstructed from $\tilde{x}(t)$ using

$$x(t) = 2\text{Re} \{\tilde{x}(t)e^{+j\Omega_c t}\} \quad (3.8)$$

$$= 2|\tilde{x}(t)| \cos(\Omega_c t + \angle\tilde{x}(t)) \quad (3.9)$$

.....

EXAMPLE 3.5.1 (PURE COSINE)

For $x(t) = A \cos(\Omega_1 t)$, we have

$$\tilde{x}(t) = \frac{A}{2} \exp(j(\Omega_1 - \Omega_c)t). \quad (3.10)$$

.....

EXAMPLE 3.5.2 (DOPPLER SHIFTED COSINE)

Let $p(t)$ denote a transmitted “pilot tone” $A \cos(\Omega_p t)$. The received signal $x(t)$ contains a Doppler shifted tone at frequency $D\Omega_p$, where D is a constant close to 1. Then

$$x(t) = p(Dt) = A \cos(\Omega_p Dt) \quad (3.11)$$

$$\tilde{x}(t) = \frac{A}{2} \exp(j(\Omega_p D - \Omega_c)t) \quad (3.12)$$

.....

EXAMPLE 3.5.3 (TIME VARYING DOPPLER SHIFTED COSINE)

Example 3.5.2 may be extended by assuming that the Doppler shift is time-varying. This is represented mathematically by

$$x(t) = p\left(\int_0^t D(\lambda) d\lambda\right) = A \cos\left(\Omega_p \int_0^t D(\lambda) d\lambda\right) \quad (3.13)$$

This gives

$$\tilde{x}(t) = \frac{A}{2} \exp\left(j\Omega_p \int_0^t D(\lambda) d\lambda - j\Omega_c t\right). \quad (3.14)$$

3.5.2 Time Scaling of Signals

In our implementation, timescaling the signal $x(t)$ compensates for the Doppler shift. For a constant Doppler shift factor D , it is desired to create the signal

$$y(t) = x(t/\delta) \quad (3.15)$$

where δ is close to D (δ will represent our estimate of the unknown factor D at a given time). To implement the time scaling, we develop the relationships between the complex representations of $x(t)$ and $y(t)$. The complex representation of $y(t)$

is

$$\tilde{y}(t) = \text{L.P.P.} \{x(t/\delta)e^{-j\Omega_c t}\} \quad (3.16)$$

$$= \text{L.P.P.} \{x(t/\delta)e^{-j\Omega_c(t/\delta)}e^{-j\Omega_c(t-t/\delta)}\}. \quad (3.17)$$

For δ close to 1 (the usual case), the second exponential term of (3.17) is a low frequency term and does not alter the portion of the spectrum selected by the L.P.P. operator. We then obtain

$$\tilde{y}(t) = \text{L.P.P.} \{x(t/\delta)e^{-j\Omega_c(t/\delta)}\} e^{-j\Omega_c(t-t/\delta)} \quad (3.18)$$

$$= \tilde{x}(t/\delta)e^{-j\Omega_c(t-t/\delta)}. \quad (3.19)$$

One can readily observe that the complex representation of $y(t)$ cannot be obtained by simply time scaling the signal $\tilde{x}(t)$. The exponential factor in (3.19) is required to obtain $\tilde{y}(t)$ for the desired center frequency Ω_c .

3.5.3 Nonuniform Sampling of $x(t)$

Note that the development in Section 3.5.2 assumed a constant time-scale factor. In our implementation the scale factor must be adjusted to track a changing Doppler shift factor. To do so, define the samples of $y(t)$ in terms of the (unequally spaced) samples of $x(t)$:

$$y(kT_y) = x(t_k). \quad (3.20)$$

Repeating the development of Section 3.5.2 gives

$$\tilde{y}(kT_y) = \tilde{x}(t_k)e^{-j\Omega_c(kT_y-t_k)}. \quad (3.21)$$

Equation (3.21) will be used extensively. The goal is to determine time sample values t_k so that the signal $\tilde{y}(kT_y)$ contains the pilot tone at the expected frequency. Under this condition, the nonuniform sampler has successfully compensated for the time-varying Doppler shift on the input signal.

In (3.21), t_k denotes the time values used to sample $x(t)$. For Doppler shift factors close to 1, the time step values will be close to T_y , giving $t_{k+1} - t_k \approx T_y$. We define Δ_k as being the correction factor for the k th sample giving the actual sample separation:

$$t_{k+1} = t_k + (1 + \Delta_k)T_y. \tag{3.22}$$

Δ_k is the (small) deviation in the sample step size from the k th sample to the $k + 1$ st sample. For Doppler shifts of up to $\pm 2\%$, the magnitude of Δ_k should remain below 0.02. Given the values of Δ_k , the time sample values are given by

$$t_k = kT_y + \sum_{\ell=0}^{k-1} \Delta_\ell T_y \tag{3.23}$$

.....

EXAMPLE 3.5.4 (SAMPLING OF THE PILOT TONE)

Sampling the time-varying Doppler shifted signal from Example 3.5.3 as indicated above gives

$$\tilde{y}(kT_y) = \frac{A}{2} \exp \left(j\Omega_p \int_0^{t_k} D(\lambda) d\lambda - j\Omega_c t_k \right) e^{-j\Omega_c(kT_y - t_k)} \tag{3.24}$$

$$= \frac{A}{2} \exp \left(j\Omega_p \int_0^{t_k} D(\lambda) d\lambda - j\Omega_c kT_y \right) \tag{3.25}$$

Let $D_k = 1 - d_k$ denote the average (unknown) Doppler shift factor for $t_k < t < t_{k+1}$.

$$\tilde{y}(kT_y) = \frac{A}{2} \exp \left(j\Omega_p \sum_{\ell=0}^{k-1} (T_y(1 + \Delta_\ell)D_\ell) - j\Omega_c kT_y \right) \quad (3.26)$$

$$= \frac{A}{2} \exp \left(jT_y\Omega_p \sum_{\ell=0}^{k-1} (1 + \Delta_\ell)(1 - d_\ell) - j\Omega_c kT_y \right) \quad (3.27)$$

Note that Δ_ℓ and d_ℓ are generally small (< 0.02). In our case, d_ℓ represents a system input—it is the actual Doppler shift at the k th sample. We wish to determine Δ_ℓ such that $(1 + \Delta_\ell)(1 - d_\ell) \approx 1$, so that $\tilde{y}(kT_y)$ will represent the pilot tone at the (known) pilot frequency.

3.5.4 A Phase Locked Loop to Adjust the Sample Rate

To determine Δ_ℓ so that $y(kT_y)$ contains the expected pilot tone at the appropriate frequency, the following strategy is adopted:

1. measure the phase error between $\tilde{y}(kT_y)$ and the desired $\exp(j(\Omega_p - \Omega_c)kT_y)$ terms.
2. Filter this result and adjust Δ_ℓ to keep this phase error small.

Figure 3.4 shows a phase-locked loop structure designed to accomplish these tasks. To analyze the loop, each block is considered separately. The nonuniform sampler described by (3.27) shows how changes in the Doppler shift d_k or the loop output Δ_k effect $\tilde{y}(kT_y)$. The loop filter will be denoted

$$H_L(z) = \frac{B(z)}{A(z)}. \quad (3.28)$$

where $B(z)$ and $A(z)$ are polynomials. These polynomials must be properly selected to ensure the stability of the PLL.

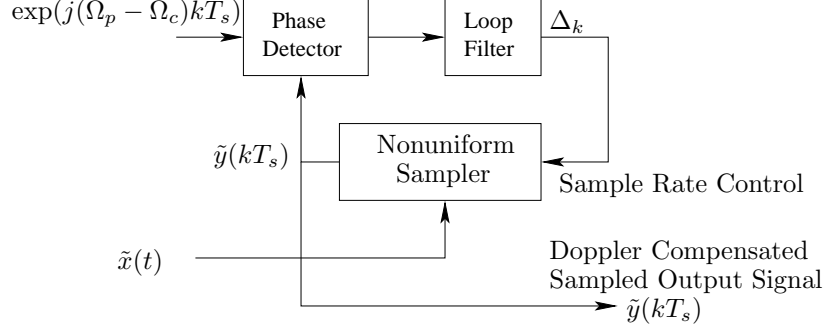


Figure 3.4: Complex representation of the phase-locked loop used for Doppler tracking.

The phase detector measures the phase difference between $\tilde{y}(kT_y)$ and the desired $\exp(j(\Omega_p - \Omega_c)kT_y)$ terms. In implementation, we calculate

$$\tilde{\gamma}(kT_y) = \tilde{y}(kT_y) \exp(-j(\Omega_p - \Omega_c)kT_y) \quad (3.29)$$

The phase difference is then approximated (for small phase errors, while the loop is locked) by the imaginary part of the normalized $\tilde{\gamma}(kT_y)$:

$$\phi(kT_y) = \text{Im} \left(\frac{\tilde{\gamma}(kT_y)}{|\tilde{\gamma}(kT_y)|} \right) \quad (3.30)$$

We can now obtain the closed-loop behavior of the PLL. Substituting (3.27) into (3.29) gives

$$\tilde{\gamma}(kT_y) = \frac{A}{2} \exp \left(jT_y \Omega_p \sum_{\ell=0}^{k-1} [(1 + \Delta_\ell)(1 - d_\ell) - 1] \right) \quad (3.31)$$

The phase detector output is then

$$\phi(kT_y) \approx T_y \Omega_p \sum_{\ell=0}^{k-1} [(1 + \Delta_\ell)(1 - d_\ell) - 1] \quad (3.32)$$

$$= T_y \Omega_p \sum_{\ell=0}^{k-1} (\Delta_\ell - d_\ell - d_\ell \Delta_\ell) \quad (3.33)$$

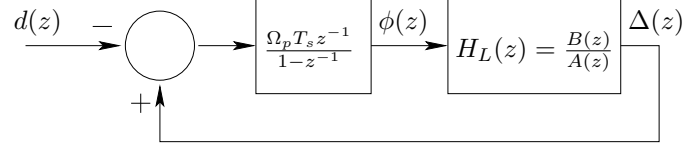


Figure 3.5: Frequency-domain representation of the (linearized) phase-locked loop.

The cross-product term $d_\ell \Delta_\ell$ does not significantly influence the behavior of the loop since Δ_ℓ and d_ℓ are both small. This term is dropped from the analysis, giving

$$\phi(kT_y) \approx \Omega_p T_y \sum_{\ell=0}^{k-1} (\Delta_\ell - d_\ell) \quad (3.34)$$

In difference equation form, we have

$$\phi(kT_y) = \phi((k-1)T_y) + \Omega_p T_y (\Delta_{k-1} - d_{k-1}). \quad (3.35)$$

Taking the z-transform of this result gives a frequency domain description of the phase detector combined with the nonuniform sampler:

$$\phi(z) = (\Omega_p T_y) \frac{z^{-1}}{1 - z^{-1}} (\Delta(z) - d(z)). \quad (3.36)$$

Figure 3.5 gives a frequency domain depiction of the linearized phase-locked loop just developed. The linearized model shows how external changes in the Doppler shift of the sampled signal (d_k) are tracked by the sampler. Successful loop operation is indicated by small values of $\phi(kT_y)$ (small phase errors). The closed-loop transfer function from the Doppler disturbance to the phase error is given by

$$\frac{\phi(z)}{d(z)} = \frac{-\Omega_p T_y A(z)}{(z-1)A(z) - \Omega_p T_y B(z)} \quad (3.37)$$

The loop filter polynomials $A(z)$ and $B(z)$ should be selected to make this small at low input frequencies (near $z = 1$). A second transfer function of interest is the

relationship between d_k and Δ_k . The closed-loop relationship is given by

$$\frac{\Delta(z)}{d(z)} = \frac{-\Omega_p T_y B(z)}{(z-1)A(z) - \Omega_p T_y B(z)} \quad (3.38)$$

This filter should be lowpass in nature, so that the (relatively low frequency) changes in d_k are accurately tracked by the loop, while portions of the re-sampled signal not near the pilot tone frequency are rejected.

3.5.4.1 Selection of Loop Filter Coefficients

Selection of the loop filter coefficients was largely a trial and error process. The following constraints are observed:

1. The loop must be stable. Given the poles and zeros of $H_L(z)$, the gain of the filter may be modified while monitoring the magnitudes of the roots of $(z-1)A(z) - \Omega_p T_y B(z)$. Gain values are identified so that these roots lie within the unit circle under all expected values of $\Omega_p T_y$.
2. The steady-state phase error must be reasonable. The phase detector given in (3.30) is based on an assumption that the phase error will be small. Phase errors above $\pm\pi/2$ violate the linearity assumptions used in the approximation given in (3.32). The nonlinearity will cause the loop to lose lock. The steady-state phase error for a constant Doppler shift factor $D_{ss} = 1 - d_{ss}$ is obtained from the DC gain of the closed-loop response

$$\phi_{ss} = \left(\frac{\phi(z)}{d(z)} \Big|_{z=e^{j0}=1} \right) d_{ss} \quad (3.39)$$

$$= \frac{1}{H_L(1)} d_{ss} \quad (3.40)$$

For example, to obtain a maximum phase error of $\pm\pi/3$ for a maximum Doppler shift of 2% ($d_{ss} = 0.02$) gives the requirement

$$\frac{\pi}{3} \geq \frac{0.02}{|H_L(1)|} \quad (3.41)$$

To meet this requirement, the DC gain of the loop filter must exceed -34 dB.

3. The roots of $A(z)$ are selected close to $z = 1$, making $H_L(z)$ lowpass and making $\frac{\phi(z)}{d(z)}$ small near $z = 1$.
4. The roots of $B(z)$ are selected near the unit circle to control the bandwidth of $\frac{\Delta(z)}{d(z)}$. This transfer function should be lowpass, with bandwidth sufficient to pass the Doppler drift frequencies, but narrow enough to reject any other modulation present in the signal.

3.5.4.2 Doppler Tracking PLL Summary

For further reference, the iterative Doppler tracking PLL is summarized by the equations (3.42) to (3.46), which are restated from the above development.

- Nonuniform Sampler

$$t_k = t_{k-1} + (1 + \Delta_{k-1})T_y \quad (3.42)$$

$$\tilde{y}(kT_y) = \tilde{x}(t_k) \exp\{-j\Omega_c(kT_y - t_k)\}. \quad (3.43)$$

- Phase Detector

$$\tilde{\gamma}(kT_y) = \tilde{y}(kT_y) \exp(-j(\Omega_p - \Omega_c)kT_y) \quad (3.44)$$

$$\phi(kT_y) = \text{Im} \left(\frac{\tilde{\gamma}(kT_y)}{|\tilde{\gamma}(kT_y)|} \right) \quad (3.45)$$

- Loop Filter

$$H_L(z) = \frac{B(z)}{A(z)} = \frac{b_0 + b_1 z^{-1} + \dots + b_n z^{-n}}{1 + a_1 z^{-1} + \dots + a_n z^{-n}}$$

$$\Delta_k = \sum_{i=0}^n b_i \phi((k-i)T_y) - \sum_{i=1}^n a_i \Delta_{k-i} \quad (3.46)$$

3.5.5 Nonuniform Sampler Implementation

Assume that samples of $\tilde{x}(t)$ are available at sample frequency F_x samples per second, and sample period $T_x = 1/F_x$. That is, $\tilde{x}(t)$ has been uniformly sampled, and $\tilde{x}(\ell T_x)$ is known. To implement the nonuniform sampler required by (3.43), samples between times ℓT_x can be calculated using polyphase interpolators. For an interpolation rate of I , samples at time $\ell T_x + m(T_x/I)$ can be obtained as follows:

1. Design a lowpass FIR filter at sample rate IF_x that passes the desired signal band and rejects (at least) frequencies above $F_y - B$, where F_y is the desired output sample rate for $\tilde{y}(t)$ (possibly different from F_x) and B is the bandwidth of $\tilde{x}(t)$. Select the number of filter coefficients as a multiple of I . Let $h(k)$ denote the impulse response of this filter ($k = 0, 1, \dots, MI - 1$).
2. The set of I polyphase filters are designed by decimating $h(k)$. The impulse response of the m th filter is

$$p_m(j) = h(m + jI) \quad m = 0, 1, \dots, I - 1 \quad j = 0, 1, \dots, M - 1 \quad (3.47)$$

3. To evaluate $\tilde{x}(\ell T_x + m T_x / I)$, use the m th polyphase filter at time origin ℓT_x :

$$\tilde{x}(\ell T_x + m T_x / I) = \sum_{j=0}^{M-1} p_m(j) \tilde{x}((\ell - j) T_x) \quad (3.48)$$

$$= \sum_{j=0}^{M-1} h(m + jI) \tilde{x}((\ell - j) T_x) \quad (3.49)$$

To evaluate $\tilde{x}(t_k)$ as required in (3.43), t_k is rounded to a close time value of the form $\ell_k T_x + m_k T_x / I$. The required values of ℓ_k and m_k are found recursively. Assume that the t_{k-1} value is known by

$$t_{k-1} = \ell_{k-1} T_x + r_{k-1} T_x \quad 0 \leq r_{k-1} < 1. \quad (3.50)$$

From (3.42) the next time sample is

$$t_k = t_{k-1} + (1 + \Delta_{k-1}) T_y \quad (3.51)$$

$$= t_{k-1} + \left((1 + \Delta_{k-1}) \frac{T_y}{T_x} \right) T_x \quad (3.52)$$

$$= \ell_{k-1} T_x + \left(r_{k-1} + (1 + \Delta_{k-1}) \frac{T_y}{T_x} \right) T_x \quad (3.53)$$

The term in parenthesis in (3.53) can be broken into its integer and fractional part, giving the desired step sizes:

$$\left(r_{k-1} + (1 + \Delta_{k-1}) \frac{T_y}{T_x} \right) = \Delta \ell_k + r_k, \quad 0 \leq r_k < 1. \quad (3.54)$$

$$t_k = (\ell_{k-1} + \Delta \ell_k) T_x + (r_k I) T_x / I \quad (3.55)$$

The required values of ℓ_{k-1} and m_{k-1} to evaluate (3.49) are given by setting

$$\ell_k = \ell_{k-1} + \Delta\ell_k \quad (3.56)$$

$$m_k = \lfloor r_k I \rfloor \quad (3.57)$$

where

$$\Delta\ell_k = \left\lfloor \left(r_{k-1} + (1 + \Delta_{k-1}) \frac{T_y}{T_x} \right) \right\rfloor \quad (3.58)$$

$$r_k = \left(r_{k-1} + (1 + \Delta_{k-1}) \frac{T_y}{T_x} \right) - \Delta\ell_k \quad (3.59)$$

and $\lfloor \cdot \rfloor$ denotes rounding down to the nearest integer.

Evaluation of $\tilde{y}(kT_y)$ in (3.43) also requires an exponential phase correction term. The exponential evaluation may be avoided in the polyphase filter case as follows. Substituting $t_k = \ell_k T_x + m_k (T_x/I)$ into (3.43) gives

$$\tilde{y}(kT_y) = \tilde{x}(t_k) \exp \{-j\Omega_c(kT_y - \ell_k T_x)\} \exp \{+j\Omega_c T_x m_k / I\} \quad (3.60)$$

The value of the second exponential in (3.60) can be obtained from a lookup table, since there are only I possible values of m_k . The value of the middle exponential term may be calculated recursively. Let c_k denote the value of this correction term for the k th sample:

$$c_k = \exp \{-j\Omega_c(kT_y - \ell_k T_x)\} \quad (3.61)$$

Using the update equations for ℓ_k gives

$$c_k = \exp \{-j\Omega_c((k-1)T_y - \ell_{k-1} T_x)\} \exp \{-j\Omega_c(T_y - \Delta\ell_k T_x)\} \quad (3.62)$$

$$= c_{k-1} \exp \{-j\Omega_c(T_y - \Delta\ell_k T_x)\} \quad (3.63)$$

The exponential term of (3.63) can also be tabulated, since (for small Doppler shifts) only a few values of $\Delta\ell_k$ are possible.

The nonuniform sampler given in (3.43) is implemented by using (3.56) through (3.59) to find ℓ_k , m_k , and $\Delta\ell_k$. These values are used to approximate $\tilde{x}(t_k)$ using (3.49). The integer $\Delta\ell_k$ is used to obtain c_k as in (3.63) (where the exponential factor is obtained from a table). The Doppler corrected output sample is then obtained from

$$\tilde{y}(kT_y) = \tilde{x}(t_k)c_k \exp \{j\Omega_c T_x m_k / I\} \quad (3.64)$$

where again the exponential term is found from an I -element table indexed by the value of m_k .

3.5.6 Filter Designs

A 6th order elliptical highpass filter was used to implement the pilot tone filter shown in Figure 3.3. The filter passband includes frequencies above 1.7 kHz (the lower limit for the Doppler shifted pilot tone location). This filter, combined with the lowpass filter implemented within the resampler, serves to isolate the pilot tone from the bulk of the modulated signal. The filter was implemented as a cascade of three second-order sections. The filter allows 0.3 dB of passband ripple, and attenuates the stopband by 40 dB. A plot of the filter characteristic is shown in Figure 3.6.

The polyphase interpolator used to implement the resampler is based on an interpolation rate of $I = 50$, giving an oversampled sampling frequency of $50F_x = 1044.375$ ksps. The filter design described in Step 1 of Section 3.5.5 was accomplished using a 800-coefficient FIR filter with 80 dB of stopband rejection. The magnitude response of this interpolation filter is illustrated in Figure 3.7.

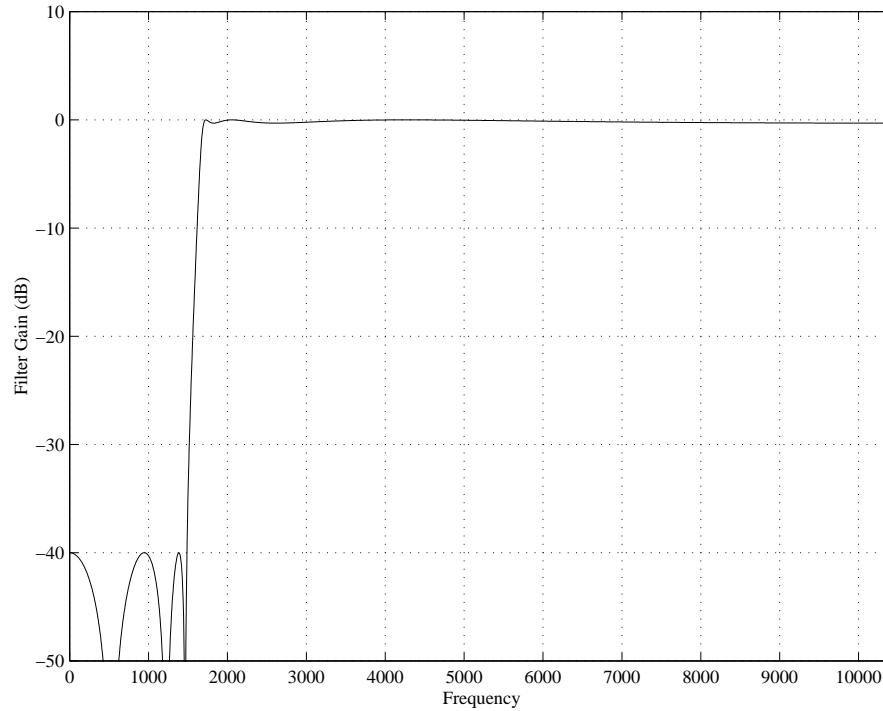


Figure 3.6: Magnitude response of the 6th order elliptical pilot tone filter.

The filter impulse response was decimated as described in (3.47) to give a set of 50 separate 15-coefficient polyphase filters. Calculation of each resampled output value involves finding the output of one of these 15-coefficient FIR filters.

3.6 Lowpass and Matched Filters

The output lowpass filter illustrated in Figure 3.3 is required to avoid aliasing when the sample rate is reduced to 5 ksp/s. This filter was designed to pass frequencies below 2.2 kHz, and reject frequencies above 2.8 kHz. Figure 3.8 shows a plot of the magnitude response of this filter, obtained using a 40-coefficient FIR filter. In implementation, this filter is combined with the QPSK matched filter by convolving the designed impulse response with the impulse response of the matched filter (a sequence of four 1's). The resulting combined lowpass and matched filter is a 43-coefficient FIR filter. Note that although the magnitude response shown in

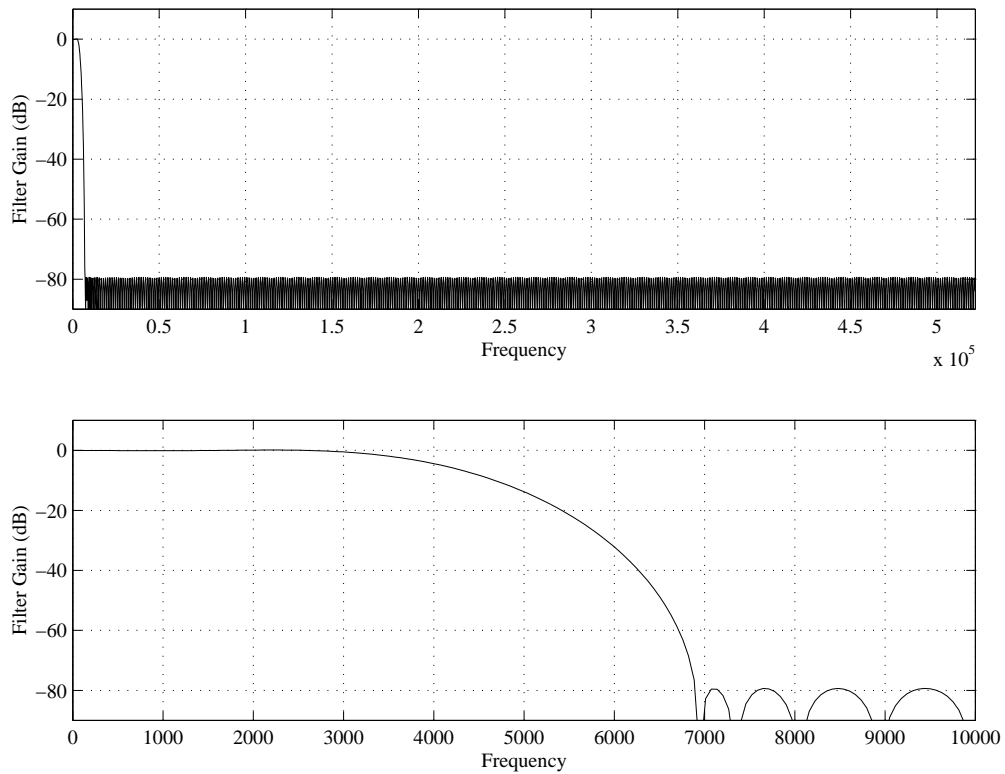


Figure 3.7: Magnitude response of the 800-coefficient FIR interpolation filter used to implement the resampler.

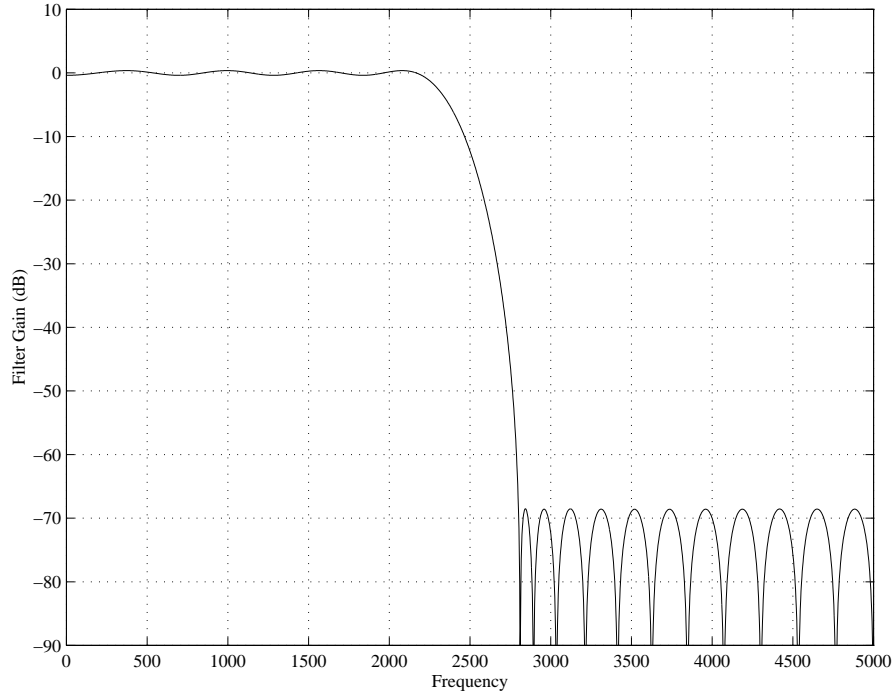


Figure 3.8: Magnitude response of the output FIR anti-aliasing filter.

Figure 3.8 shows less than 70 dB of stopband rejection, the combined magnitude response of the lowpass and matched filter does have more than 80 dB rejection throughout the stopband.

The Lowpass/matched filter outputs are then decimated by a factor of two, giving an output data stream of 5 ksps that is delivered to the channel combiner. The decimation is *not* implemented as a part of the lowpass/matched filter, since the 10 ksps output is required for accurate implementation of the ping synchronization correlator.

3.7 Frame Synch Ping Correlation

The frame-synchronization ping consists of a $N_b = 76$ -bit Differential Phase Shift Keyed (DPSK) modulated ping (or equivalently, a 77-bit BPSK ping). The

ping may be described by

$$p(t) = \sum_{k=0}^{N_b} b_k p_0(t - kT_b) \quad (3.65)$$

where $b_k = \pm 1$ indicates the transmitted bit, $T_b = 541.33 \mu\text{sec}$ is the bit period, and $p_0(t)$ describes the transmitted waveform for each bit.

$$p_0(t) = \begin{cases} \sin(\Omega_c t) & 0 \leq t \leq T_b \\ 0 & \text{elsewhere} \end{cases} \quad (3.66)$$

Note that (3.66) describes a “phase jammed” signal, in which the phase of the transmitted signal is reset for each transmitted bit. The complex representation of $p_0(t)$ is given by

$$\tilde{p}_0(t) = -(j/2)w(t) \quad (3.67)$$

$$w(t) = \begin{cases} 1 & 0 \leq t \leq T_b \\ 0 & \text{elsewhere} \end{cases} \quad (3.68)$$

The complex representation for a frame synchronization ping that occurs at $t = 0$ is

$$\tilde{p}(t) = \sum_{k=0}^{N_b} b_k (-j/2)w(t - kT_b)e^{-j\Omega_c kT_b}. \quad (3.69)$$

The desired correlator output for a received signal $s(t)$ is given by

$$c(t) = \int_t^{t+(N_b+1)T_b} s(\lambda)p(\lambda - t)d\lambda. \quad (3.70)$$

Substituting (3.3) into this expression gives the result in terms of the magnitude and phase of the complex representations

$$c(t) = \int_t^{t+(N_b+1)T_b} 4|\tilde{s}(\lambda)||\tilde{p}(\lambda-t)| \left(\frac{1}{2} \cos(\angle\tilde{s}(\lambda) - \angle\tilde{p}(\lambda-t) + \Omega_c t) + \frac{1}{2} \cos(\angle\tilde{s}(\lambda) + \angle\tilde{p}(\lambda-t) + \Omega_c t + \Omega_c \lambda) \right) d\lambda. \quad (3.71)$$

The second cosine is a high frequency term with an integral that is approximately zero (and equals zero for the ideal case in which there are an integer number of cycles in each bit period). This term is dropped from the integration giving

$$c(t) = \int_t^{t+(N_b+1)T_b} 2\text{Re} \{ \tilde{s}(\lambda)\tilde{p}^*(\lambda-t)e^{j\Omega_c t} \} d\lambda. \quad (3.72)$$

$$= 2\text{Re} \left\{ \left(\int_t^{t+(N_b+1)T_b} \tilde{s}(\lambda)\tilde{p}^*(\lambda-t)d\lambda \right) e^{j\Omega_c t} \right\}. \quad (3.73)$$

The $e^{j\Omega_c t}$ term reflects the phase alignment of the correlation period. Comparing this result to (3.2) gives the desired complex representation of the correlator output

$$\tilde{c}(t) = \int_t^{t+(N_b+1)T_b} \tilde{s}(\lambda)\tilde{p}^*(\lambda-t)d\lambda \quad (3.74)$$

Frame synchronization is obtained by comparing the *envelope* of $c(t)$ to a threshold, or equivalently the value of $2|\tilde{c}(t)|$. The correlator output for the receiver front end are the actual calculated values of $(2|\tilde{c}(t)|)^2$.

The matched filter calculation that has already been completed can be exploited in the evaluation of the above integral. Breaking the integration into

intervals of length T_b seconds gives

$$\tilde{c}(t) = \sum_{k=0}^{N_b} \int_{t+kT_b}^{t+(k+1)T_b} \tilde{s}(\lambda) b_k^*(j/2) e^{j\Omega_c k T_b} d\lambda \quad (3.75)$$

$$= \sum_{k=0}^{N_b} b_k^*(j/2) e^{j\Omega_c k T_b} \int_{t+kT_b}^{t+(k+1)T_b} \tilde{s}(\lambda) d\lambda \quad (3.76)$$

The $(j/2)$ term may be dropped from this expression, since the correlator output used for frame synchronization is $(2|\tilde{c}(t)|)^2$.

$$\text{correlator output} = \left| \sum_{k=0}^{N_b} b_k^* e^{j\Omega_c k T_b} \int_{t+kT_b}^{t+(k+1)T_b} \tilde{s}(\lambda) d\lambda \right|^2 \quad (3.77)$$

The matched filter outputs provide signal integration over a 400 μsec period, which is slightly less than the 541.33 μsec bit period used in (3.77). Ideally, synchronization ping correlation would be performed by integrating the input signal over the entire 541.33 μsec period, multiplying these results by $\pm e^{j\Omega_c k T_b}$ depending upon the bit sequence (b_k) , and summing these results over the entire 42 msec ping duration. In this implementation, the 10 ksps input sample rate implies that each transmitted bit is represented by a non-integer number of samples, and it is difficult to perform the integration over an exact bit period. Instead, the implementation uses the matched filter outputs that have already been calculated. This saves computations, and reduces the sensitivity to the imperfect bit-edge alignment caused by the non-integral relationship between the sample period and the bit duration. The resulting algorithm is expected to perform about 1 dB worse than a more optimal algorithm that integrates over an entire bit period.

CHAPTER 4

Adaptive Feedback Equalizer

This chapter describes the UDAT system's equalizer module. This module consists of a front-end interface to the channel combiner module and a Recursive Least Squares (RLS) equalizer.

Studies conducted by NUWC [4] have shown that an adaptive decision feedback equalizer-digital phase locked loop (DFE-DPLL) works very well in terms of reducing multipath effects, time varying inter-symbol interference, and channel fading inherent in underwater environments. MATLAB prototypes of this equalizer were written by Dr. Nixon A. Pendergrass, Susan M. Jarvis, and Fletcher Blackmon at NUWC using both RLS and Fast Transversal Filters (FTF) weight update algorithms.

Using the RLS algorithm to update the equalizer's filter weights is relatively computationally intensive. It is on the order of N^2 where N is the length of the weight vector (Equation 4.3). The FTF algorithm on the other hand is more computationally efficient (on the order of N). However, since this is a prototype system being developed for the first time on SHARC processors it was decided to start by implementing the equalizer with an RLS algorithm. The straightforward nature of the RLS algorithm simplified the task of translating the MATLAB equalizer code into a C version for the SHARC processors. It was also expected that the increased processing power of the SHARC processors (as compared to the C40 processors used in [1]) might be sufficient to handle the increased computational burden of the RLS algorithm.

Discussion of the equalizer will begin by introducing the notation and presenting the basics of the implemented algorithm. Next, key operations of the C language implementation of the equalizer will be discussed and compared to their

MATLAB equivalents. Finally, a series of benchmarks will be presented to show the equalizer's calculation times for some typical tap weight counts.

4.1 General Equalizer Algorithm Overview

A block diagram of this diversity input adaptive DFE-DPLL is shown in Figure 4.1. Note that this is the same equalizer presented in [4] except for the addition of the sparse feedback sections (detailed in Figure 4.3).

Both the MATLAB and C implementations of this adaptive equalizer use the following notation:

- n is the symbol counter
- $training$ is the training symbol counter
- \mathbf{v} is a matrix with columns of input data for each channel
- L , M , and $M2$ are the number of feedforward, feedback, and sparse feedback tap weights respectively
- $M2OFF$ is a vector of sparse feedback tap locations
- R is the number of diversity inputs (channels)
- \tilde{a} , \tilde{b} , and $\tilde{b}2$ are the feedforward, feedback, and sparse feedback tap weight vectors respectively
- $p(n)$, $q(n)$, and $q2(n)$ are the contributions of the feedforward, feedback, and sparse feedback sections respectively
- $\tilde{W}(n)$ is the weight vector formed from \tilde{a} , \tilde{b} , and $\tilde{b}2$ (Equation 4.8)
- $\tilde{U}(n)$ is the input vector shown in Equation 4.8

- \mathbf{R} is the correlation matrix
- $d(n)$ and $d_{bit}(n)$ are the estimates and bit decisions for symbol n

One important fact to note is the difference in orientation between the C and MATLAB matrices. The reasons for this will be discussed further in Section 4.2.1.

Figures 4.2, 4.3, and 4.4 show the internals of the DFE-DPLL's r^{th} diversity feedforward section as well as the non-sparse and sparse feedback sections respectively.

A digital phase-locked loop is used to provide the phase correction term, $\Theta_r(n)$ as shown in Figure 4.2. The loop for the r^{th} diversity input is described by

$$\Theta_r(n+1) = \Theta_r(n) + K1 \cdot \phi_r(n) + K2 \cdot \sum_{i=0}^n \phi_r(i), \quad (4.1)$$

where $\phi_r(i) = \Im\{p_r(n)[\sum_{r=1}^R p_r(n) + \varepsilon(n)]^*\}$, and $K1$ and $K2$ are constants that govern the loop's tracking characteristics [4].

The received symbol estimate can be written as

$$\begin{aligned} d(n) = & \sum_{r=1}^R \sum_{i=-T}^T a_{r,i}^*(n) \cdot e^{-j\Theta_r(n)} \cdot v_r(2n-i) \\ & - \sum_{\ell=1}^M b_{\ell}^*(n) \cdot d_{bit}(n-\ell) \\ & - \sum_{sp=1}^{SP} \sum_{\ell=1}^{M2(sp)} b_{sp,\ell}^*(n) \cdot d_{bit}(n-M2OFF(sp)-\ell), \end{aligned} \quad (4.2)$$

where R is the number of diversity inputs, $L = 2T + 1$ is the number of tap weights in each feedforward section, M is the number of tap weights in the feedback section, and $M2(sp)$ are the numbers of taps in each of the SP sparse feedback sections.

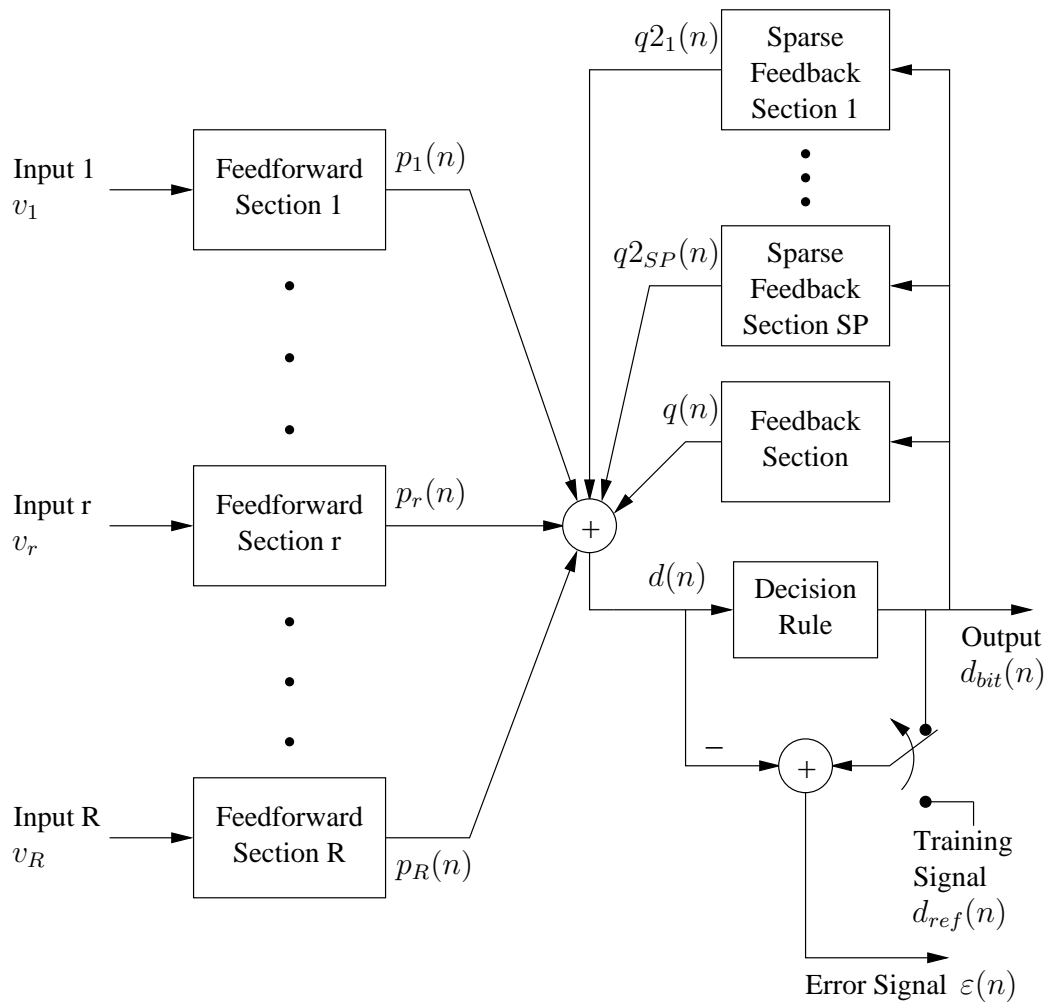


Figure 4.1: Block diagram of the decision feedback equalizer - digital phase locked loop (DFE-DPLL).

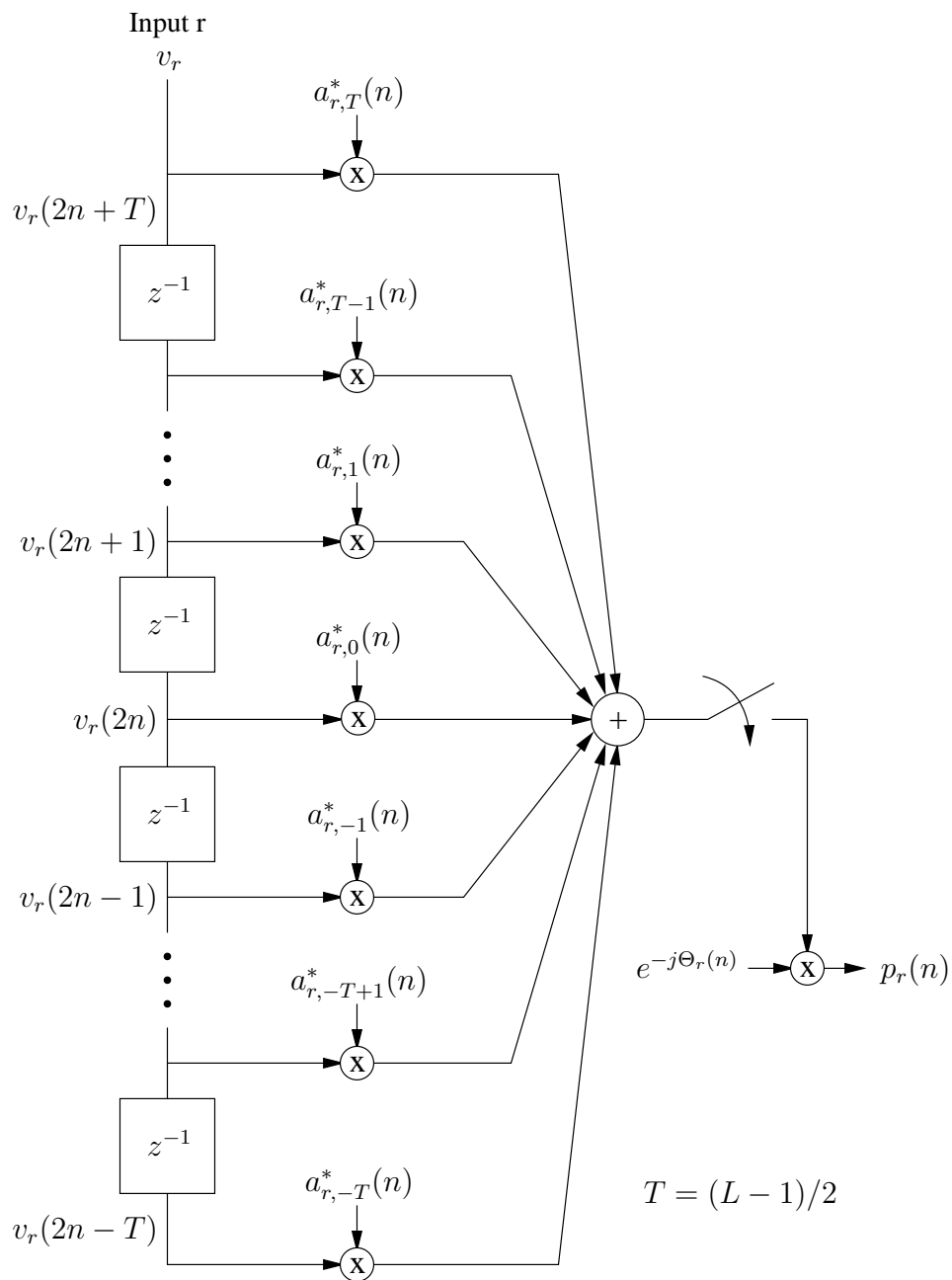


Figure 4.2: Block diagram of the r^{th} diversity feedforward section of the DFE-DPLL.

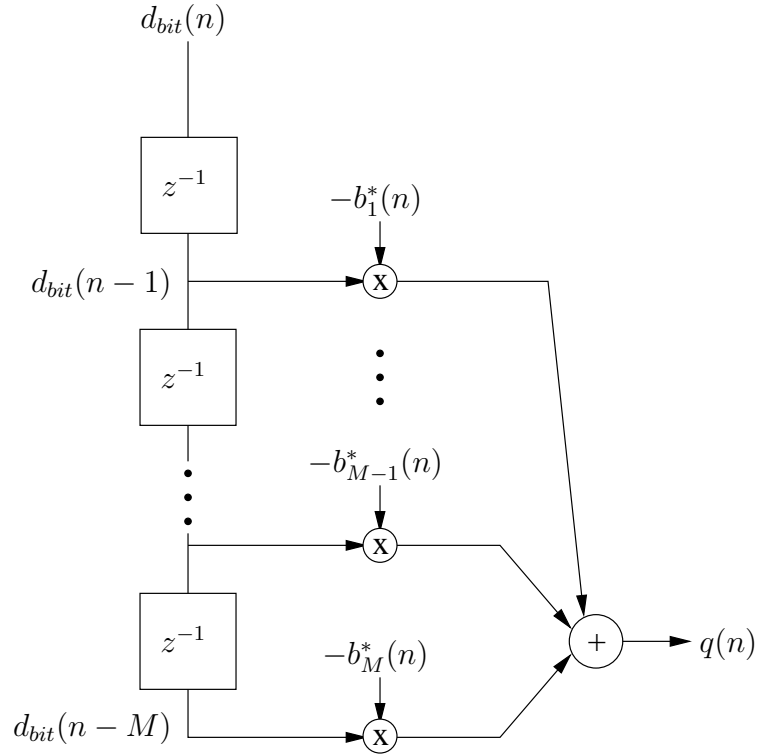


Figure 4.3: Block diagram of the non-sparse feedback section of the DFE-DPLL.

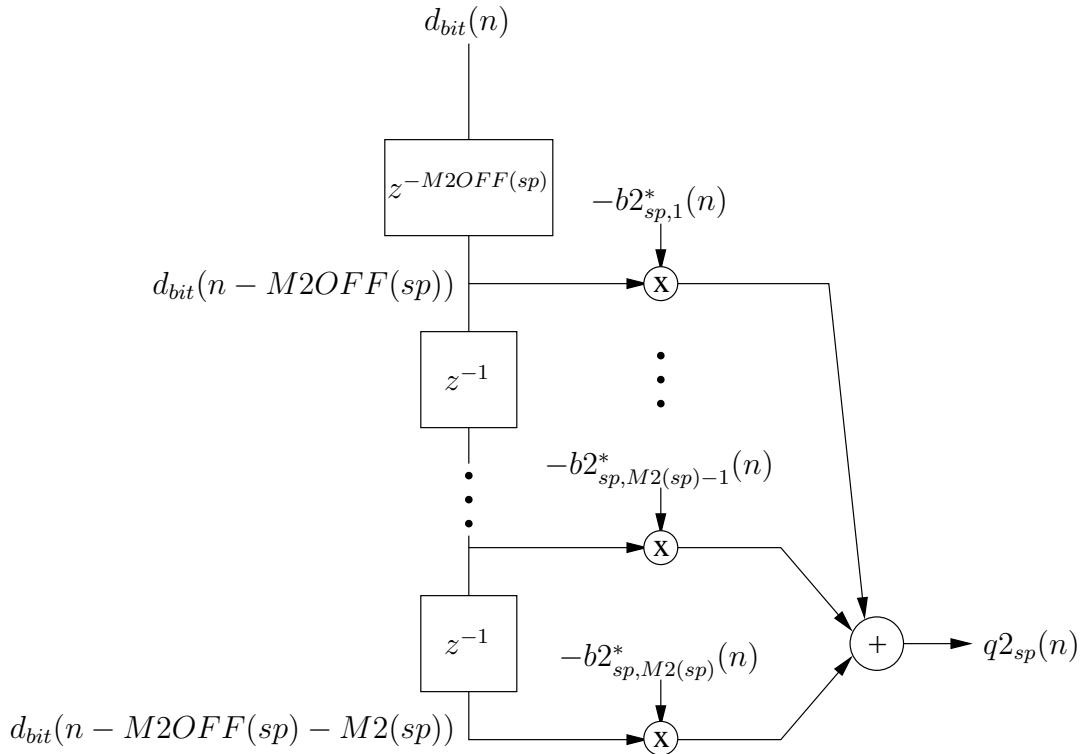


Figure 4.4: Block diagram of the sp^{th} sparse feedback section of the DFE-DPLL.

The total number of tap weights is given by

$$N = L \cdot R + M + \sum_{sp=1}^{SP} M2(sp). \quad (4.3)$$

In order to eliminate recalculating values in the z^{-1} delay lines, column vectors $\tilde{v}_r(n)$ are used to store the $2n + T$ through $2n - T$ values of v_r for each of the R diversity inputs. Values in each $\tilde{v}_r(n)$ vector are shifted down each time a new input sample is added. Similarly, a vector $\tilde{d}(n)$ is used to store the $n - 1$ through $n - M$ samples of d_{bit} for use in the feedback section. And column vectors $\tilde{d}_{2sp}(n)$ are used to store the $n - M2OFF(sp) - 1$ through $n - M2OFF(sp) - M2(sp)$ values of d_{bit} for use in each of the sparse feedback sections. These coefficient vectors are shown in Equation 4.4. The corresponding weight vectors are also stored in column vectors as shown in Equation 4.5.

$$\begin{aligned} \tilde{v}_r(n) &= \begin{bmatrix} v_r(2n + T) \\ \vdots \\ v_r(2n) \\ \vdots \\ v_r(2n - T) \end{bmatrix} & \tilde{d}(n) &= \begin{bmatrix} d_{bit}(n - 1) \\ \vdots \\ d_{bit}(n - M) \end{bmatrix} \\ \tilde{d}_{2sp}(n) &= \begin{bmatrix} d_{bit}(n - M2OFF(sp)) \\ \vdots \\ d_{bit}(n - M2OFF(sp) - M2(sp)) \end{bmatrix} \end{aligned} \quad (4.4)$$

$$\tilde{a}_r(n) = \begin{bmatrix} a_{r,T}(n) \\ \vdots \\ a_{r,0}(n) \\ \vdots \\ a_{r,-T}(n) \end{bmatrix} \quad \tilde{b}(n) = \begin{bmatrix} -b_1(n) \\ \vdots \\ -b_M(n) \end{bmatrix} \quad \tilde{b}2_{sp}(n) = \begin{bmatrix} -b2_{sp,1}(n) \\ \vdots \\ -b2_{sp,M2(sp)}(n) \end{bmatrix} \quad (4.5)$$

Equations 4.4 and 4.5 allow Equation 4.2 to be written as

$$\begin{aligned} d(n) &= \sum_{r=1}^R \tilde{a}_r^H \cdot \tilde{v}_r(n) + \tilde{b}^H(n) \cdot \tilde{d}(n) + \sum_{sp=1}^{SP} \tilde{b}2_{sp}^H(n) \cdot \tilde{d}2_{sp}(n) \\ &= \sum_{r=1}^R p_r(n) + q(n) + \sum_{sp=1}^{SP} q2(n) \end{aligned} \quad (4.6)$$

Equation 4.2 can also be written in terms of the entire $\tilde{W}(n)$ weight vector as

$$d(n) = \tilde{W}^H(n) \cdot \tilde{U}(n) \quad (4.7)$$

where

$$\tilde{W}(n) = \begin{bmatrix} \tilde{a}_1(n) \\ \vdots \\ \tilde{a}_R(n) \\ \tilde{b}(n) \\ \tilde{b}2_1(n) \\ \vdots \\ \tilde{b}2_{SP}(n) \end{bmatrix} \quad \tilde{U}(n) = \begin{bmatrix} \tilde{v}_1(n) \cdot e^{-j\Theta_1(n)} \\ \vdots \\ \tilde{v}_R(n) \cdot e^{-j\Theta_R(n)} \\ \tilde{d}(n) \\ \tilde{d}2_1(n) \\ \vdots \\ \tilde{d}2_{SP}(n) \end{bmatrix} \quad (4.8)$$

In Equation 4.7, $\tilde{W}(n)$ represents the vector of all filter coefficients used in the various systems shown in Figures 4.2, 4.3, and 4.4. To track changing channel conditions, this vector must be modified while the receiver is operating. The

value of the error signal, $\varepsilon(n)$, shown in Figure 4.1 is used to drive this adaptive process. A standard RLS algorithm was implemented to accomplish this task. The algorithm is described by Equations 4.9 through 4.12 where \mathbf{R}^{-1} is the inverse of the correlation matrix and λ is the forgetting factor.

$$\tilde{P}(n) = \mathbf{R}^{-1} \cdot \tilde{U}(n) \quad (4.9)$$

$$\alpha = \Re \left\{ \frac{1}{\lambda + \tilde{U}(n)^H \cdot \tilde{P}(n)} \right\} \quad (4.10)$$

$$\tilde{W}(n) = \tilde{W}(n) + \alpha \cdot \varepsilon(n) \cdot \tilde{P}(n) \quad (4.11)$$

$$\mathbf{R}^{-1} = \frac{\mathbf{R}^{-1} - \alpha \cdot \tilde{P}(n) \cdot \tilde{P}^H(n)}{\lambda} \quad (4.12)$$

Following the weight update, the contributions of the feedforward, feedback, and sparse feedback sections are recalculated using Equation 4.6. The decision rule shown in Figure 4.1 makes a bit decision using

$$d_{bit}(n) = \text{sign}(\Re\{d(n)\}) + j \cdot \text{sign}(\Im\{d(n)\}) \quad (4.13)$$

If the equalizer is still in training mode (first 200 symbols) the error is calculated using $\varepsilon(n) = d_{ref}(n) - d(n)$ where d_{ref} is a known reference symbol. However, if the equalizer is in decision directed mode the error is calculated using $\varepsilon(n) = d_{bit}(n) - d(n)$.

To summarize, the equalizer algorithm operates in the following manner:

- Initialize counters, data, and weight vectors.
- Initialize the inverse of the correlation matrix, \mathbf{R}^{-1} , to an identity matrix multiplied by a small scalar.
- For each QPSK symbol in the data frame, perform the following:
 - Shift and add new data to the data vectors shown in Equation 4.4.
 - Update the $\tilde{U}(n)$ vector as shown in Equation 4.8.
 - Calculate the output, $d(n)$, using the current weights.
 - Form a temporary bit decision (Equation 4.13 and use it to calculate the error, $\varepsilon(n)$).
 - Update the $\tilde{W}(n)$ weight vector using the RLS algorithm as described in Equations 4.9 through 4.12.
 - Recalculate the output using the new weights.
 - Form the bit decision (Equation 4.13 and use it to calculate the error, $\varepsilon(n)$).
 - Update the phase estimate, Θ_r , for the phase locked loop as shown in Equation 4.1.
- When the data frame has been completed, return the resulting vectors of symbol estimates and bit decisions.

Section 4.2 further discusses the MATLAB and C software implementations of this decision feedback equalizer-digital phase locked loop algorithm.

4.2 Equalizer Implementation

The RLS equalizer used in this system is a C language implementation of NUWC's MATLAB version. The code performs an R input channel complex valued exponentially weighted recursive least squares sparse adaptive filter algorithm for a QPSK signal. The software implementations accept fractionally spaced input samples with two samples per QPSK symbol. Therefore, k is used to denote the sample index while n is used for the QPSK symbol index. Wideband libraries [8] were used extensively to duplicate the functionality of the MATLAB code in C. Examples of the Wideband library functions used in the equalizer are shown in the following sections.

4.2.1 Vector and Matrix Storage

As was noted earlier in Section 4.1, both the MATLAB and C equalizer code use the same notation. However, the column vectors in the MATLAB code have been transposed in the C code for memory usage and indexing efficiency. Matrices are stored in C as one long vector containing the matrices' rows. This is not only the optimal storage format but it also allows certain vector operations to be used on matrices.

4.2.2 Initializing Vectors and Matrices

MATLAB uses the functions `eye`, `ones`, and `zeros` to initialize vectors and matrices. The Wideband functions `vfill` and `cvfill`, which stand for vector fill and complex vector fill, can be used as an equivalent replacement. To duplicate the functionality of `eye`, the vector fill command is first used to fill the matrix with zeros by treating the matrix as one long row vector. Then a loop is used to change the diagonal elements to ones.

4.2.3 Copying Sections of Vectors and Matrices

The Wideband functions `vmov` and `cvmov`, which stand for vector move and complex vector move, are used to copy sections of vectors and matrices. As an example, the matrix `v_k` is used to store the \tilde{v}_r vectors as defined in Equation 4.4 is formed in MATLAB using a statement of the form:

```
v_k(1:k,:) = v(k:-1:1,:);
```

to copy the first `k` rows of `v` into `v_k` in the reverse order. This functionality is duplicated in C using the Wideband function `cvmov` within a `for` loop covering the number of diversity inputs.

```
cvmov(&v[index][k-1],-1,&v_k[index][0],1,k);
```

The `for` loop accounts for all diversity input channels just as the colon operator does in MATLAB. Note that the variables `v` and `v_k` are transposed from their MATLAB orientation as was described in Section 4.2.1. The first argument of `cvmov` is the address of the `k`'th column of `v` and the second is a stride of `-1`. This takes the first `k` columns of `v` in the reverse order. The third argument is the destination (columns of `v_k`) and the stride of `1` keeps them in order. The final argument, `k` is an element count.

4.2.4 Dot Products

The Wideband function `ccdopr` to calculate a complex conjugate dot product, is also used in the equalizer code. For example, the received symbol estimate (Equation 4.7) is calculated in MATLAB using:

```
d(n) = W_k' * U_k;
```

Again, these vectors are transposed from their MATLAB orientation requiring the Wideband equivalent to multiply them in the reverse order using:

```
ccdopr(U_k,1,W_k,1,&d_tmp,taps);
```

Here, U_k and W_k are the two vectors to be multiplied and ones are used as the strides. The variable `taps` has been initialized to the length of the vectors being implemented.

4.2.5 Matrix - Vector Multiplication

Another useful Wideband function is `cmvmul` which performs a complex matrix-vector multiplication. It is used to implement Equation 4.9. In MATLAB, this equation takes the form:

```
invR_Uk = inv_R * U_k;
```

Note that `invR_Uk` is the variable used to denote \tilde{P} in the MATLAB and C code. The Wideband equivalent of the above calculation uses:

```
cmvmul(&inv_R[0][0], taps, taps, U_k, invR_Uk);
```

The first argument is the start of the `inv_R` matrix. The next two arguments are the row and columns of `inv_R` which are both `taps` since it is a square matrix. The fourth argument is the source vector, U_k , which is followed by the destination, `invR_Uk`.

4.2.6 RLS Correlation Matrix Update

Updating the correlation matrix (Equation 4.12) is performed using the following MATLAB code:

```
inv_R = (inv_R - alfa * invR_Uk * invR_Uk')/lmda;
```

However, the Wideband library does not provide a function that does both a matrix subtraction and a multiplication. And storing a temporary copy of the `invR_Uk` outer product requires more memory than is available on the SHARC processors. Therefore, the matrix update is performed on a column by column basis.

A `for` loop is used to cycle through the columns. First, a scaling factor of `-alfa * invR_Uk[col_idx]` is calculated. Then the Wideband function `cvsma`

is used to perform a complex vector-scalar multiply and add. Initially, this was implemented using:

```
cvsmma(invR_Uk,1,&scale,&inv_R[0][index],
        &taps,&inv_R[0][index],taps,taps);
```

to calculate the entire column of `inv_R`. However, on the SHARC processors this implementation had some numerical instability problems. So a second method was used to ensure that the matrix would remain symmetric. The new method uses:

```
cvsmma(&invR_Uk[index],1,&scale,&inv_R[index][index],
        &taps,&inv_R[index][index],taps,taps-index);
```

followed by:

```
cvconj(&inv_R[index+1][index],taps,
        &inv_R[index][index+1],1,taps-index-1);
```

inside a loop that covers all but the last two columns. This multiply and add only operates on the upper triangular half of the matrix. The `cvconj` is then used to generate the lower (conjugate) part of the matrix. Since most of the Wideband vector functions have a minimum element count, this method requires the last four elements (last two columns) to be calculated outside the loop.

Finally, the Wideband function `vscmul` is used to multiply the result by $1/\text{lmda}$ to take care of the division by `lmda`. `vscmul` is a vector-scalar multiply, but the entire `inv_R` matrix is treated as one long vector since there is no “matrix-scalar multiply” function in the Wideband library.

4.3 Equalizer Front End

Due to processing and memory requirements, real-time operation requires several equalizers, each running on its own processor. As will be described in Section 5.2.4, jobs are submitted to the equalizers via a link port transfer. This requires a small job receiving front end to configure a link port and DMA in order

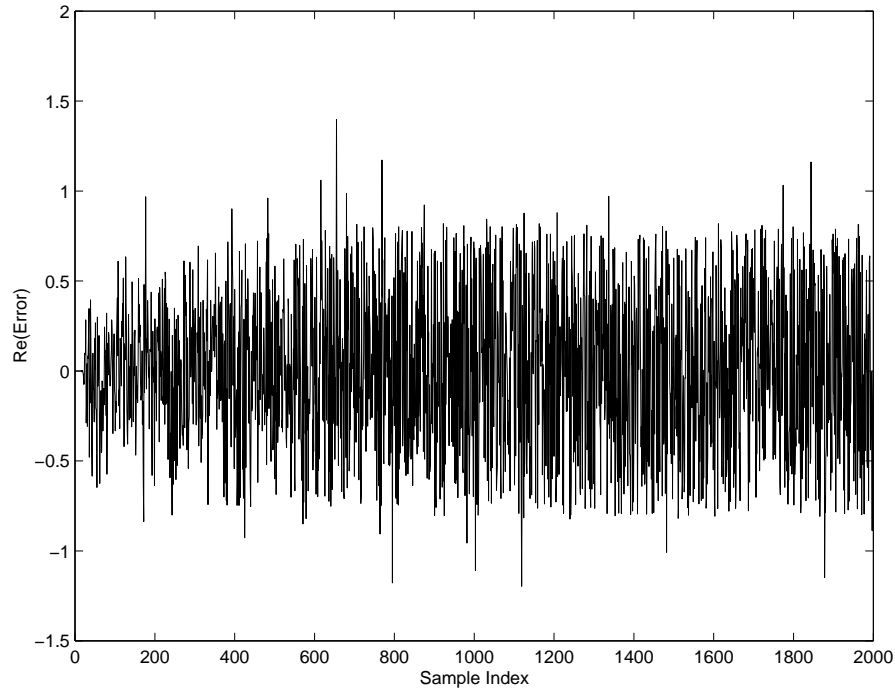


Figure 4.5: RLS equalizer error from C version.

to receive the job. After the DMA reception has been configured, the front end enters a loop in which it polls the link port status. Once a complete job transfer has finished, the equalizer function is called to perform the equalization. Then the loop repeats by reinitializing another link port reception.

An alternative job sending method that uses the processor common memory space to transfer jobs from the channel combiner to the equalizers is described in Section 5.2.3. That job transfer method also requires a job receiving front end to poll the queue in the common memory for a job to equalize.

4.4 Equalizer Testing and Verification

Proper operation of the RLS equalizer code was verified by comparing it to the known MATLAB version provided by NUWC. Figures 4.5 and 4.6 show how the equalizer errors compare between the C and MATLAB versions respectively. Both plots were generated using the same set of testing data.

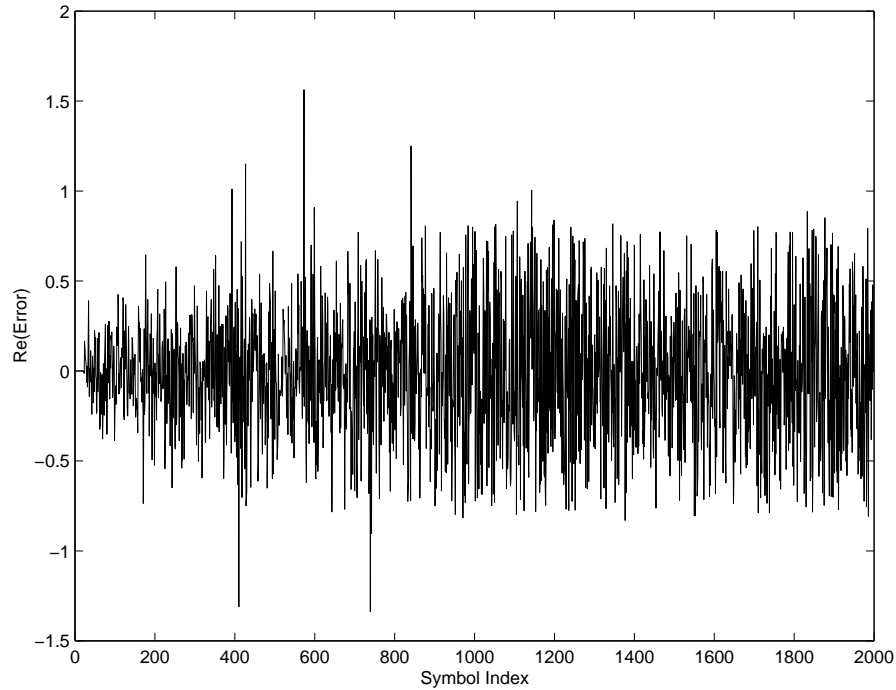


Figure 4.6: RLS equalizer error from MATLAB version.

Due to the fact that MATLAB's floating point precision is greater than that of the SHARC processors, the two error waveforms are not exactly the same. However, they both exhibit the same characteristic shape, indicating that the C version is operating correctly. The actual equalized output vectors were also compared and found to be very close. Again, there was some variation in the exact magnitudes due to precision differences.

4.5 RLS Equalizer Benchmarks

Several benchmarking tests were performed on the RLS equalizer code to determine the time required to equalize jobs with various tap counts. This equalizer design applies the feedback and sparse feedback tap weights equally to all input channels. Therefore, the equalizer's computational complexity is strongly related to the number of feedforward taps and the number of input channels.

The following benchmarks were performed on 40 MHz 21060 SHARC digital signal processors on a Spectrum Morocco II carrier board [6]. The time lines in Figures 4.7 through 4.12 were generated by measuring the time required to execute each of the major steps within the equalizer algorithm. A PMC ADADIO board [9] was connected to the Morocco II carrier board. The DAC on the ADADIO board was monitored with an oscilloscope. The equalizer code was temporarily modified to write a different value to the DAC upon completion of each of the following events:

- entry into the main processing loop
- addition of new samples to the U_k input vector
- calculation of the output and error with current weights
- update of the filter weights, W_k
- update of the inv_R matrix
- calculation of the output with the new weights
- update of the DPLL phase estimate

The computation time required for each event showed up as a plateau on the oscilloscope waveform. Data captured from the oscilloscope trace was used to measure the width of those plateaus and generate the benchmarking timelines.

The above sequence occurs once for each sample. Therefore, a larger spike was also written to the DAC at the completion of each job. Triggering the oscilloscope on the larger peak was used to determine the total equalization time required per job. The total equalization time is also noted in each of the following benchmarks.

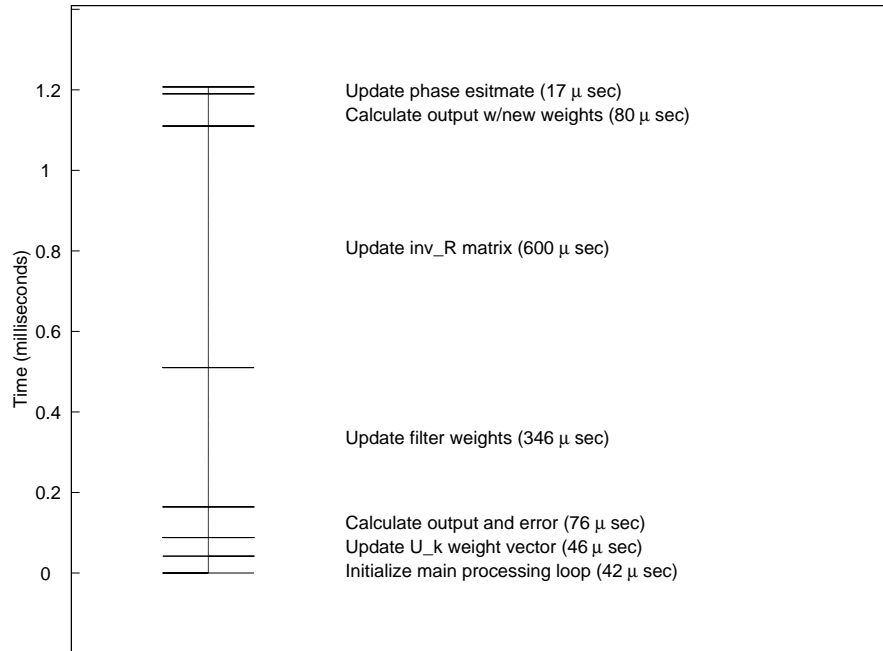


Figure 4.7: RLS benchmark for 41 taps.

4.5.1 41 Taps

According to preliminary tests conducted by NUWC, a typical test case consists of 21 feedforward taps, 10 feedback taps, and two 5 tap sparse feedback sections. For one input channel this results in a total tap count of $21 + 10 + 5 + 5 = 41$ taps. As shown in Figure 4.7, this case requires about 1.5 msec of computation time per sample. The widest ($590 \mu\text{sec}$) section represents the time required to perform the inv_R matrix update. Prior to that is the $350 \mu\text{sec}$ section representing the filter weight update calculations. Since these two operations account for the majority of the equalizer's computation time, they were optimized using Wideband library routines.

Under these conditions, one job requires 2.34 seconds to equalize. Since each job contains one second of data, three equalizers would be required for real-time operation in this case.

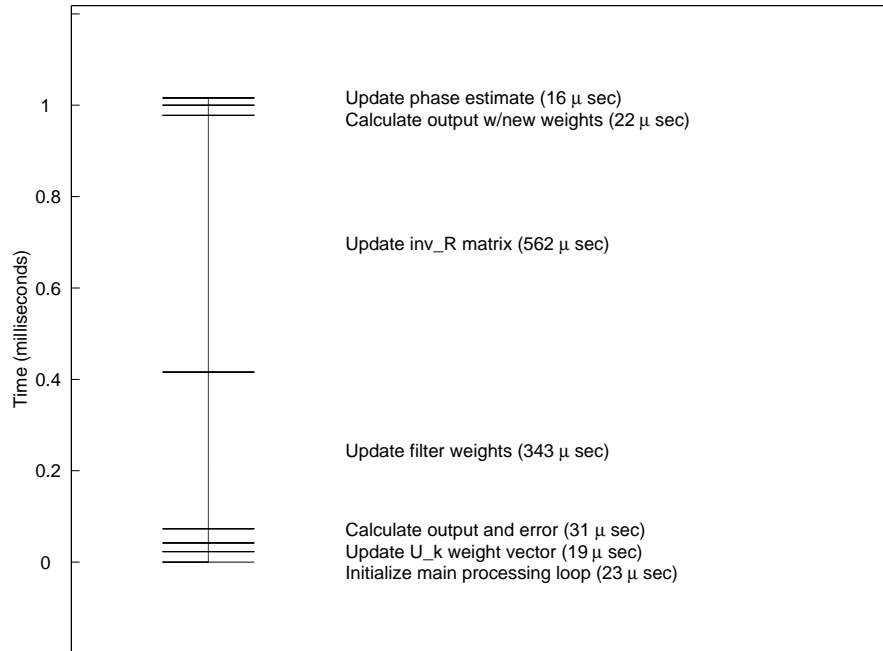


Figure 4.8: RLS benchmark for 41 taps with optimization.

Enabling optimization in the compiler results in the above case requiring 1.95 seconds to equalize one second of input data. As Figure 4.8 shows, the time per sample has been reduced to about 1 msec. The filter weight update still takes about 350 μsec and the `inv_R` matrix update has been reduced by 40 μsec to 550 μsec . This indicates that these operations are already well optimized.

4.5.2 62 Taps

Increasing the number of channels to two doubles the total number of feedforward taps. Feedback taps are shared among channels so their count remains unchanged. This results in a total tap count of $(21 * 2) + 10 + 5 + 5 = 62$ taps. Figure 4.9 shows that the filter weight update now requires just under 800 μsec and the matrix update requires about 1200 μsec . This results in a total equalization time of 4.75 seconds per job.

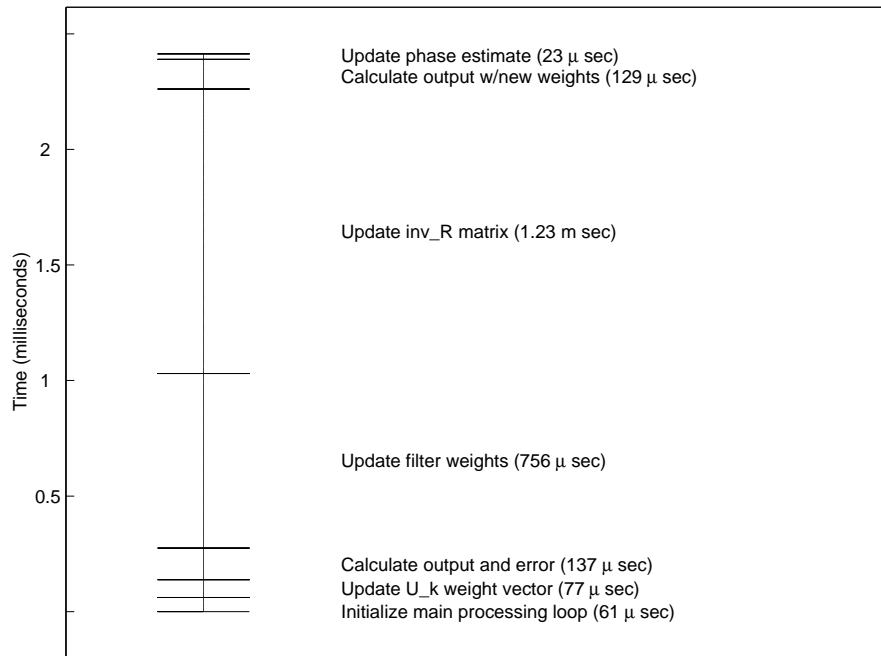


Figure 4.9: RLS benchmark for 62 taps.

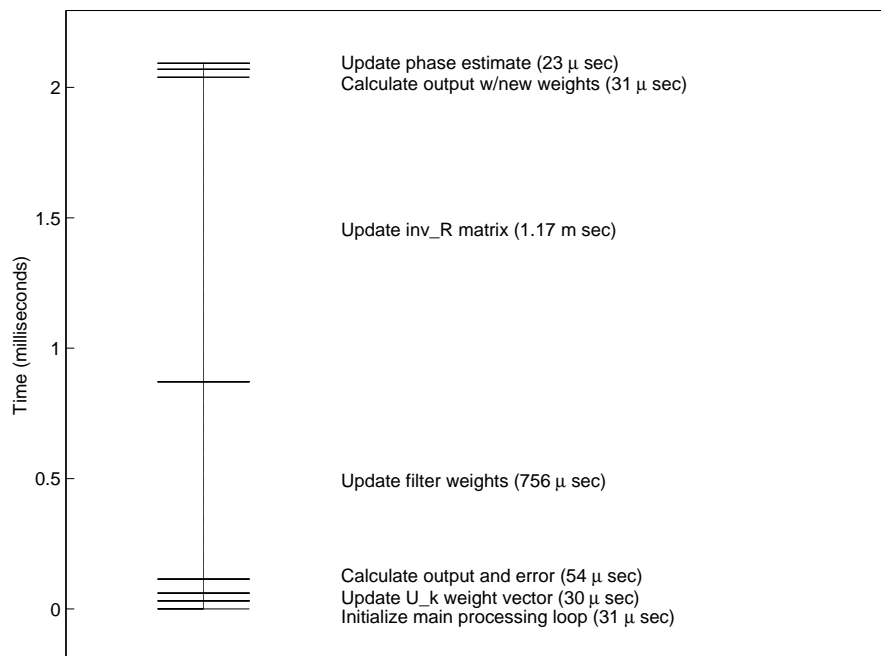


Figure 4.10: RLS benchmark for 62 taps with optimization.

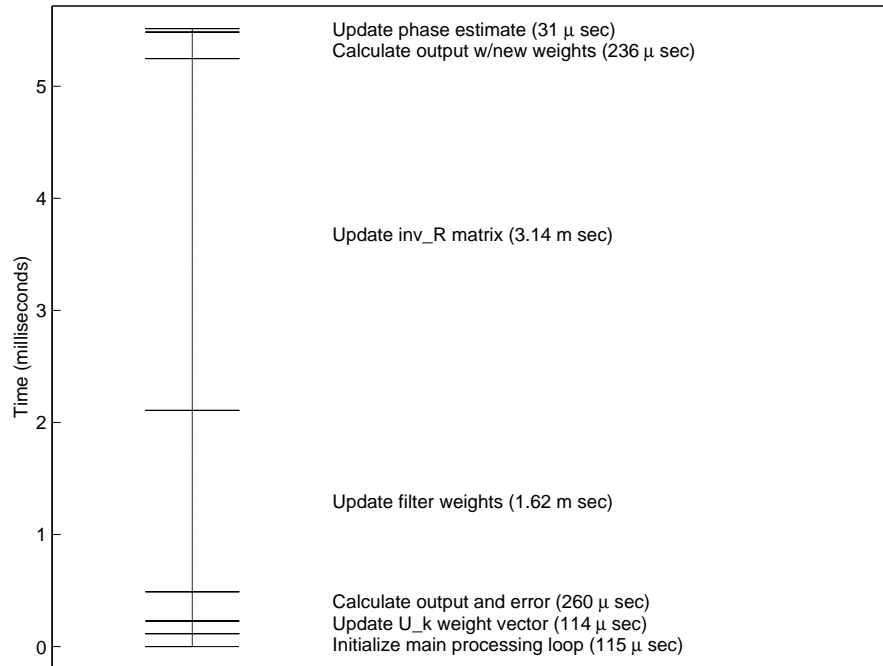


Figure 4.11: RLS benchmark for 104 taps.

Again, enabling optimization in the compiler helps slightly by reducing the overall equalization time to 4.1 seconds per job. The computational details of this case are shown in Figure 4.10.

This two diversity channel example represents operation in either time or spatial diversity mode. Real-time operation with this tap count will require 5 equalizer modules.

4.5.3 104 Taps

On chip memory limits the channel combiner to four channels. This brings the total tap count for this case to $(21 * 4) + 10 + 5 + 5 = 104$ taps. Figures 4.11 and 4.12 show the timing details of this 104 taps case. Clearly, the filter weight and matrix updates account for the majority of computational time per sample.

Considering the fact that equalization times with and without optimization are 11.89 and 10.72 seconds respectively, it is obvious that the RLS equalizer is too

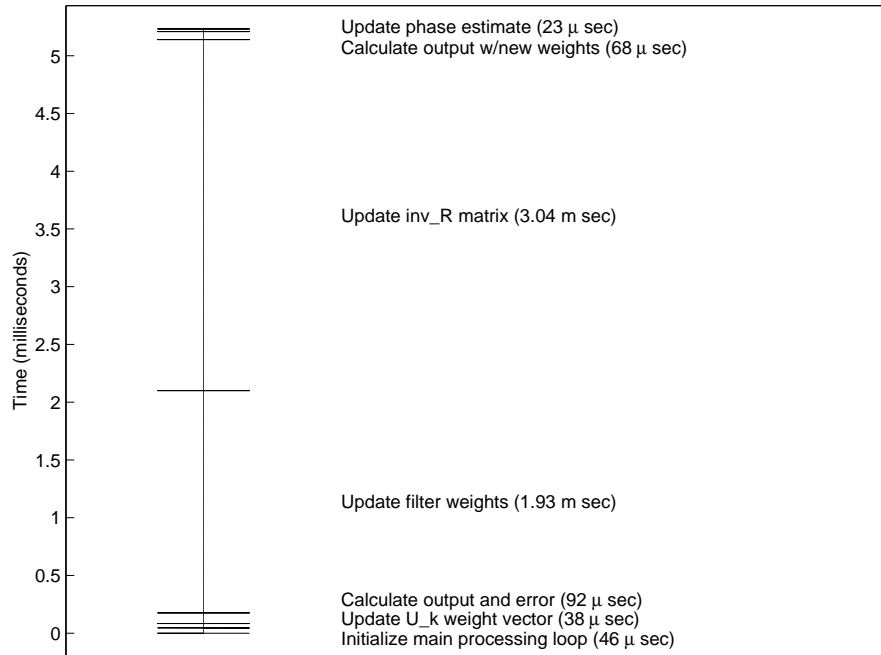


Figure 4.12: RLS benchmark for 104 taps with optimization.

computationally intensive for practical use in this case. Four channel operation is still possible but it requires a tradeoff in the tap count. For example, as shown in Table 4.2, the number of feedforward taps must be reduced to 11 (as opposed to 21) to bring the equalization time down to a practical 4.59 seconds. Equalization times over five seconds are not considered practical since there are only five available SHARC processors on the Morocco II board when the other three are running front-end and channel combiner modules.

4.5.4 Benchmark Summary

Table 4.1 lists the total equalization times required for the above benchmarks without compiler optimization. The times obtained with optimization enabled are summarized in Table 4.2.

Input Channels	Feedforward Taps	Feedback Taps	Sparse Feed-back Taps	Total Taps	Time (sec)
1	21	10	10	41	2.34
2	21	10	10	62	4.75
4	21	10	10	104	11.89

Table 4.1: RLS equalization times without compiler optimization.

Input Channels	Feedforward Taps	Feedback Taps	Sparse Feed-back Taps	Total Taps	Time (sec)
1	21	10	10	41	1.95
2	21	10	10	62	4.1
4	21	10	10	104	10.72
4	19	10	10	96	9.28
4	17	10	10	88	7.97
4	15	10	10	80	6.75
4	13	10	10	72	5.63
4	11	10	10	64	4.59
4	9	10	10	56	3.67

Table 4.2: RLS equalization times with compiler optimization.

CHAPTER 5

Channel Combiner

This chapter presents the channel combiner module, the function of which is tying the front-end and equalizer modules together. As such, it must perform the following tasks:

- Time alignment of data frames
- Channel analysis (based on the target ID correlation waveform) to identify dominant multipath components
- Formation of “equalizer jobs” for distribution to the equalizer modules

As was shown in Chapter 2, several front-end modules send time and/or spatial diversity channels into the channel combiner. Since signals from physically separated sensors (spatial diversity) do not all arrive at the same time, the channel combiner must time align the data frames before presenting them to an equalizer module. Figure 5.1 presents a pictorial view of how the channel combiner time aligns signals from multiple sensors. The channel combiner must also calculate equalizer tap sizes and locations based upon the channel information gathered from the TID ping and its echoes as described in Sections 2.3.1 and 2.3.2.

The inner workings of the channel combiner module are presented by first discussing the interface with front-end modules followed by methods of distributing completed jobs to equalizer modules. Next, the state machines that make up the channel combiner’s inner core are discussed. Finally, the equalizer tap calculations are presented.

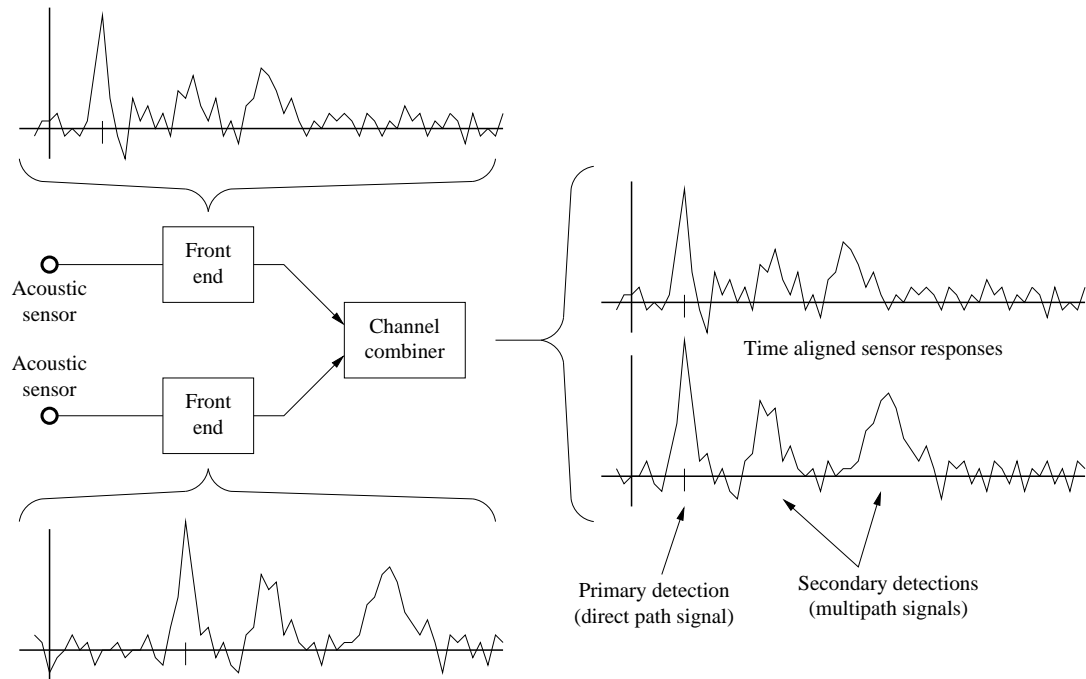


Figure 5.1: Channel combiner functional illustration.

5.1 Interface to Front-End Modules

Each front-end module runs on its own SHARC processor and is connected to one acoustic sensor. The channel combiner also runs on its own processor and receives data from the front-end modules via link ports. A set of ping pong buffers is used in conjunction with a direct memory access (DMA) transfer to bring data into the channel combiner. The ping pong buffers utilize a pointer (`buffer_to_process`) to indicate the working buffer. While the CPU is processing the working buffer, the other buffer is being filled by DMAs transferring data from the front-end modules.

The channel combiner uses data structures to store information from the input sensors (front-end modules). One of these sensor data structures is used for every front-end module that is connected to the channel combiner. Table 5.1 lists the elements contained within each sensor data structure. Each sensor data structure also contains the ping pong buffers mentioned above. Since these

Element Name	Description
front-end data sub-structure	input data (ping pong buffers) associated with this sensor
<code>sensor state</code>	(described in Section 5.3)
<code>current job</code>	pointer to the job associated with this sensor (described in Section 5.2.1)
<code>channel</code>	channel (location) within the job where the sensor data is to be written
<code>transmission number</code>	data frame number for use with time diversity
<code>largest peak and age</code>	used for ping synchronization (described in Section 5.3.2)
<code>TID_array[]</code>	used to track past TIDs for channel alignment

Table 5.1: Elements of the sensor data structure.

“buffers” consist of three arrays plus status information, they are essentially sub-structures of the sensor data structure. Table 5.2 lists the elements contained within each of the front-end sub-structures. Note that these are the same arrays shown as being “Sent to Channel Combiner” in Figure 3.1.

Since the channel combiner can only perform calculations on one input channel at a time it must rapidly alternate among them. A pointer (`curr_sensor`) is maintained in order to keep track of the sensor (input channel) that is currently being processed. Each time the channel combiner moves on to a new buffer of input data, the `curr_sensor` pointer is updated to point to the sensor data structure

Element Name	Description
<code>length</code>	length of the buffer being sent from the front-end module to the channel combiner
<code>td_present</code>	pilot detected flag (switches telemetry on/off)
<code>sensor_id</code>	identifies the sensor associated with a buffer of data
<code>corr[]</code>	buffer of correlation values from the front-end module (to be used for ping synchronization and equalizer tap calculations)
<code>TID[]</code>	buffer of target ID associated with each correlation value
<code>demod[]</code>	buffer of matched filter outputs (demodulated data) (to be time aligned and sent to an equalizer)

Table 5.2: Elements of the front-end data sub-structure.

associated with the sensor (front-end module) supplying the data. So the channel combiner makes use of a C code statement of the form:

```
curr_sensor -> buffer_to_process -> data
```

to access data associated with the working ping pong buffer within the appropriate sensor data structure.

5.2 Equalizer Job Queuing

Real-time operation of the UDAT system requires that the data packets generated by the channel combiner be equalized at a rate of one per second. However, as was shown in Chapter 4, the equalizer algorithm often requires more than one second to equalize a data packet. Therefore, real-time operation requires several copies of the equalizer module to be run on separate processors. Multi-processor equalizer operation requires a queuing scheme to distribute jobs among the equalizers.

5.2.1 Equalizer Jobs

The channel combiner uses structures known as “equalizer jobs” to contain all of the information required by an equalizer to process a packet of data. Elements contained within each of the equalizer job structures are listed in Table 5.3. Two of these job structures must be maintained by the channel combiner since data from the various input sensors does not all arrive at the same time. For example, it is possible for ‘sensor A’ to come to the end of a data frame before ‘sensor B’. The job can’t be submitted to the equalizer until ‘sensor B’ has also reached the end of its data frame. But in the meantime, ‘sensor A’ will be starting on a new data frame which in turn requires a new job. Maintaining two job structures eliminates this problem. Pointers known as `actv_job` and `pend_job` are utilized to keep track of the active and pending jobs respectively. In the above example where ‘sensor

Element Name	Description
pop[]	vector of channel ‘fully populated’ flags
half[]	vector of channel ‘halfway populated’ flags
det[]	vector of channel detected flags
aborted[]	vector of aborted channel flags
L	feedforward symbol count
M	feedback symbol count
M2[]	sparse feedback symbol count vector
sumM2	sum of elements in the M2 vector
lenM2	length of the M2 vector
M2OFF[]	sparse feedback delay/offset vector
v[][]	demodulated signal vectors (data to be equalized)

Table 5.3: Elements of the equalizer job structure.

A’ finishes before ‘sensor B’, the two jobs allow ‘sensor A’ to start writing its new data frame into the pending job while ‘sensor B’ is finishing up the current job. When ‘sensor B’ has finished and the job has been submitted to the equalizer, the `actv_job` and `pend_job` pointers are switched. Once a sensor has been assigned a job as described in Section 5.3.2.4 it continues to use that job until it reaches the end of a data frame. Therefore, the switching of the active and pending job pointers does not cause a sensor to jump between jobs part way through a data frame.

The channel population vector (`pop`) is used to signify that a channel is fully populated with valid data to be used by the equalizer. This vector contains `FULL` flags that correspond to valid demodulated signal vectors. Invalid signal vectors (ones that do not contain data) are marked with an `EMPTY` flag in the corresponding location in this population vector. Similarly, the halfway (`half`) flags are used to indicate when a channel is half full of data. The `det` (detected) flags are used to indicate that a signal has been detected on a particular channel. The `aborted` flags are used to mark channels that were aborted due to the fact that they hadn’t received a signal detection. The use of these flags is described

Element Name	Description
<code>dir_path[]</code> <code>list</code>	storage space for the direct path signal linked list of correlator values for calculating sparse feedback taps (described in Section 5.4.2)
<code>pointers[]</code>	vector of pointers into the linked list (also described in Section 5.4.2)

Table 5.4: Elements of the equalizer scratch space structure.

in further detail in Sections 5.3.2 and 5.3.3. All of the feedforward and feedback symbol counts L through $M2OFF$ are calculated by the channel combiner based on channel information gathered from the target ID correlation waveform. Section 5.4 explains the tap calculation procedures. Finally, the demodulated signal vectors (\mathbf{v}) contain the matched filter outputs (`demod` from Table 5.2) from each channel that have been time aligned by the channel combiner.

Each equalizer job also has an associated scratch space for information that is not passed on to the equalizer. Jobs and scratch space are separate for two reasons. First, it allows them to be stored in different areas of on-chip memory. Second, it prevents the needless transfer of the scratch space information to the equalizer. Each scratch space structure contains the elements listed in Table 5.4.

The direct path storage area is used by the “Record Direct Path” state that will be described in Section 5.3.2. The linked list and its vector of pointers are utilized by the “Build Sparsing List” state also described in Section 5.3.2 as well as the equalizer tap calculations presented in Section 5.4.

5.2.2 Job Queuing Methods

Two methods were developed to transfer jobs from the channel combiner to the equalizer modules. The first method uses the common memory on the Morocco

II board as a queue where jobs are temporarily stored until an equalizer reads them. The second method uses link port transfers to send jobs to the equalizers.

The common memory job queuing method allows equalizers to be run on any processor without the need for any special configurations. However, it does have some limitations. For example, the code used to read data from the ADADIO (input/output) board [9] ties up the processor common bus making it unavailable for use with this job queuing scheme. There are also plans to run this software on platforms other than the Morocco II that do not have the same common memory architecture. Therefore, it was necessary to develop the link port job transfer method that is currently in use.

The link port method works very much like the link port transfer used to bring front end data into the channel combiner. However, instead of using a ping pong buffering scheme, the data is sent in one continuous block. An additional link port status check is also performed prior to sending a job. The status check insures that the equalizer on the receiving link port is ready to accept a new job.

5.2.3 Common Memory Transfer Theory of Operation

With this method, multiple job slots were set aside in the common memory. This allowed new jobs to be continuously written into the queue without overwriting an unequalized job. The channel combiner would write out new jobs at a rate of one per second. Each write consisted of a bus request, the data transfer, and a bus release. Meanwhile each processor running an equalizer module was also running a job scanning front-end routine. The job scanning front end would request the common bus on regular intervals. While it had access to the bus it would check for an available job by looking at a job status flag and copy the first available job it found into the processor's local memory. Once the check and transfer (if applicable) were completed, the job's status flag was changed and the

bus was released. When available jobs were found they were sent to the equalizer and the front end stopped polling for jobs until the equalizer was finished.

5.2.4 Link Port Transfer Theory of Operation

Instead of submitting jobs to a queue in the common memory, the channel combiner writes them directly to an equalizer via a link port transfer. This requires that the channel combiner be provided with a list of link ports that will be connected to processors running equalizers. Under normal operating conditions, the channel combiner cycles through the available link ports in sequence. However, if an equalizer is busy (as determined by the link port status check) an error is flagged and the job is sent to the next free equalizer in the sequence.

The required number of equalizers depends largely on the number of taps selected. As long as a sufficient number of equalizers are in operation, the channel combiner will function under normal operating conditions and no “busy equalizer” error flags will be set. Busy equalizer errors are an indication that more equalizers are required to maintain real-time operation. Since the maximum number of equalizers is limited by the SHARC processor’s 6 link ports, it may be necessary to reduce the maximum tap limit. The benchmarks presented in Chapter 4 give a more detailed account of the number of equalizers required to deal with various tap counts.

5.3 State Machine Implementation

The inner core of the channel combiner consists of two state machines. One state machine handles the various calculations associated with each of the input sensors (front-end modules). The other state machine is used to maintain the equalizer jobs and distribute them to the equalizer modules. The channel

combiner uses an overall status to cycle among gathering new input as described in Section 5.1 as well as updating the appropriate state machines

5.3.1 Overall Status

The channel combiner's overall status is one of three possibilities:

- Monitor Input
- Process Input
- Job Maintenance

Figure 5.2 presents a flowchart showing how the channel combiner cycles through the above modes.

While in the “Monitor Input” mode, the link ports are monitored for incoming data. This is done by looking at a pointer associated with the ping pong buffers described in Section 5.1. New data is ready when the `buffer_to_process` pointer associated with a front-end module changes, indicating that the DMA has completed filling an input buffer. The actual pointer change takes place in an interrupt service routine that handles the DMA transfers to the ping pong buffers.

Once a new input buffer has been collected, the status changes to “Process Input”. The sensor state machine for the appropriate sensor (presented in Section 5.3.2) is executed for the new buffer of input samples. Then, if necessary, the “Job Maintenance” state machine described in Section 5.3.3 is updated.

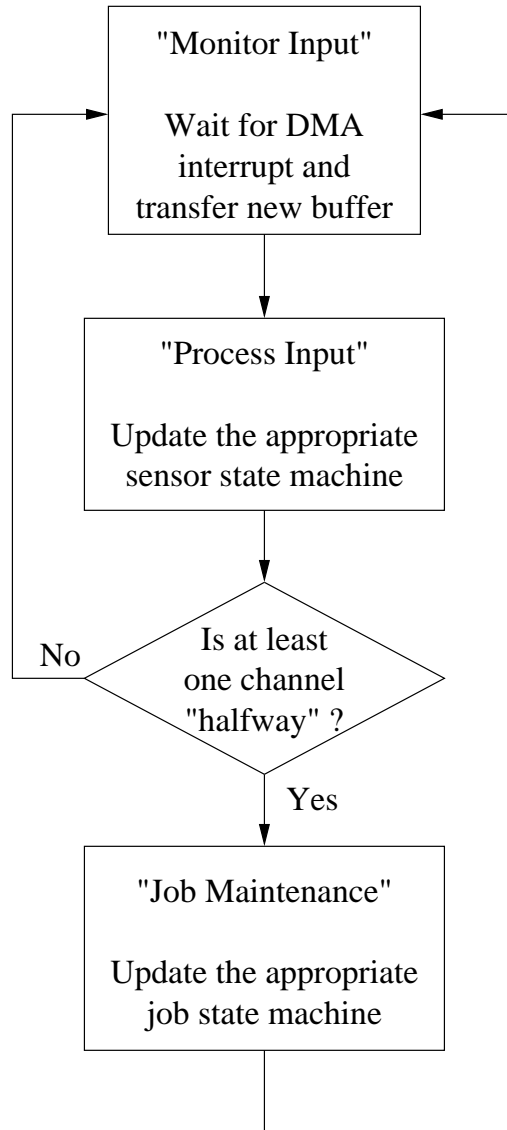


Figure 5.2: Channel combiner status flow chart.

5.3.2 Sensor State Machine

The channel combiner has the capability to maintain multiple sensor state machines. A separate sensor state machine is associated with each front-end module that supplies data to the channel combiner. When the channel combiner status is “Process Input”, the sensor state machine associated with a ping pong buffer of new data is updated. A diagram of the sensor state machine is shown in Figure 5.3. It contains the states discussed in the following subsections, namely, Ping Synchronization, Get Ping TID, Wait for Ping, Watch for Detection, Record Direct Path, Build Sparsing List, and Record Modulated Data.

5.3.2.1 Ping Synchronization

This is the start-up state for the sensor state machine. By looking for pings spaced approximately one second apart, this “Ping Synchronization” state prevents the UDAT receiver from triggering on echoes of pings or false ping detections. This state examines the correlator outputs provided by the front-end modules to locate dominant peaks that are close to one second apart. Since it is impractical to save the correlator outputs for a full second, only the largest correlator peak over the past 1.1 seconds is maintained in memory. A flow chart of the ping synchronization logic is shown in Figure 5.4.

In this state, the sample from the correlator is compared to a detection threshold. Synchronization occurs if the sample is above the threshold and the largest peak is approximately one second old. Upon synchronization, the sensor’s state is changed to “Get Ping TID”. Otherwise, the sample is compared to the largest correlator peak and if appropriate used to replace the largest peak.

Once the sample has been processed, the age of the largest peak is incremented and then checked. A peak age that is significantly older than one second is an indication of a missed ping. So, if the peak’s age is over 1.1 seconds, it is

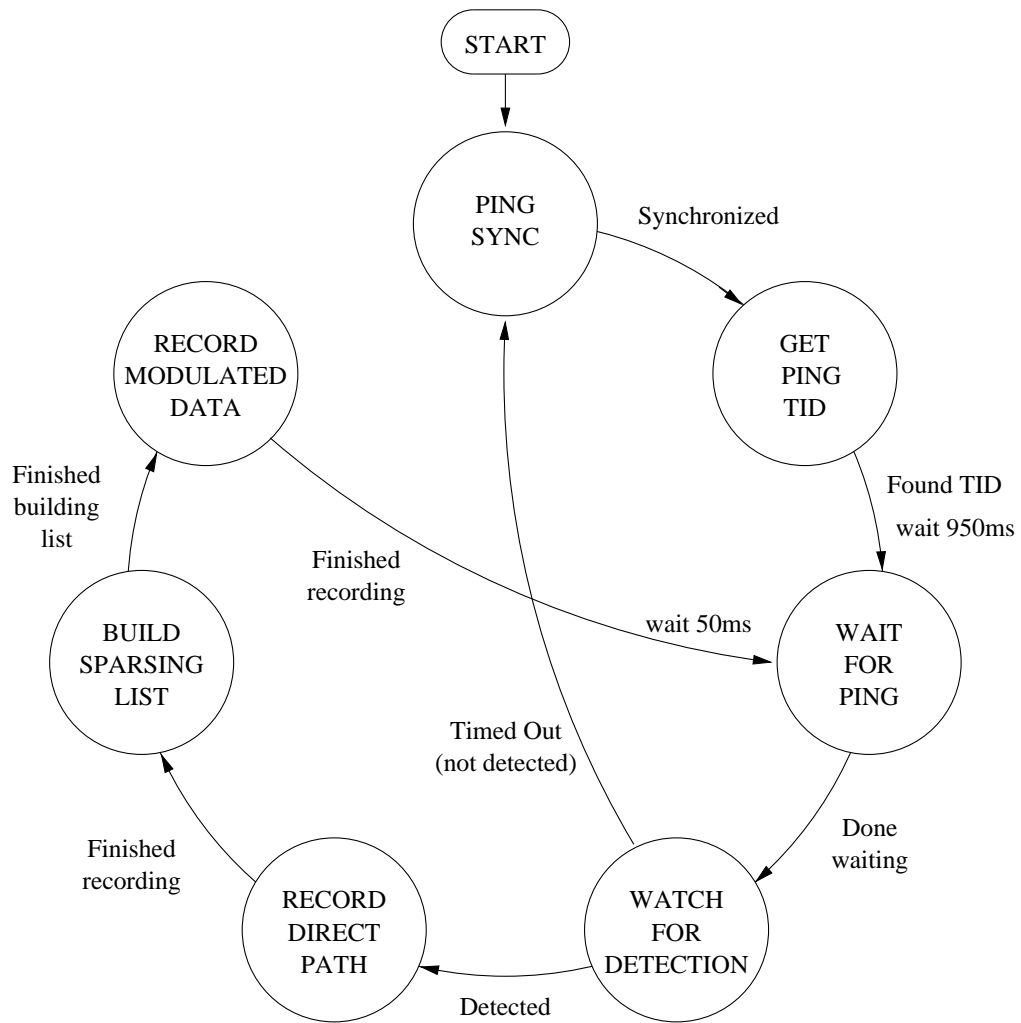


Figure 5.3: State flow diagram for the sensor state machine.

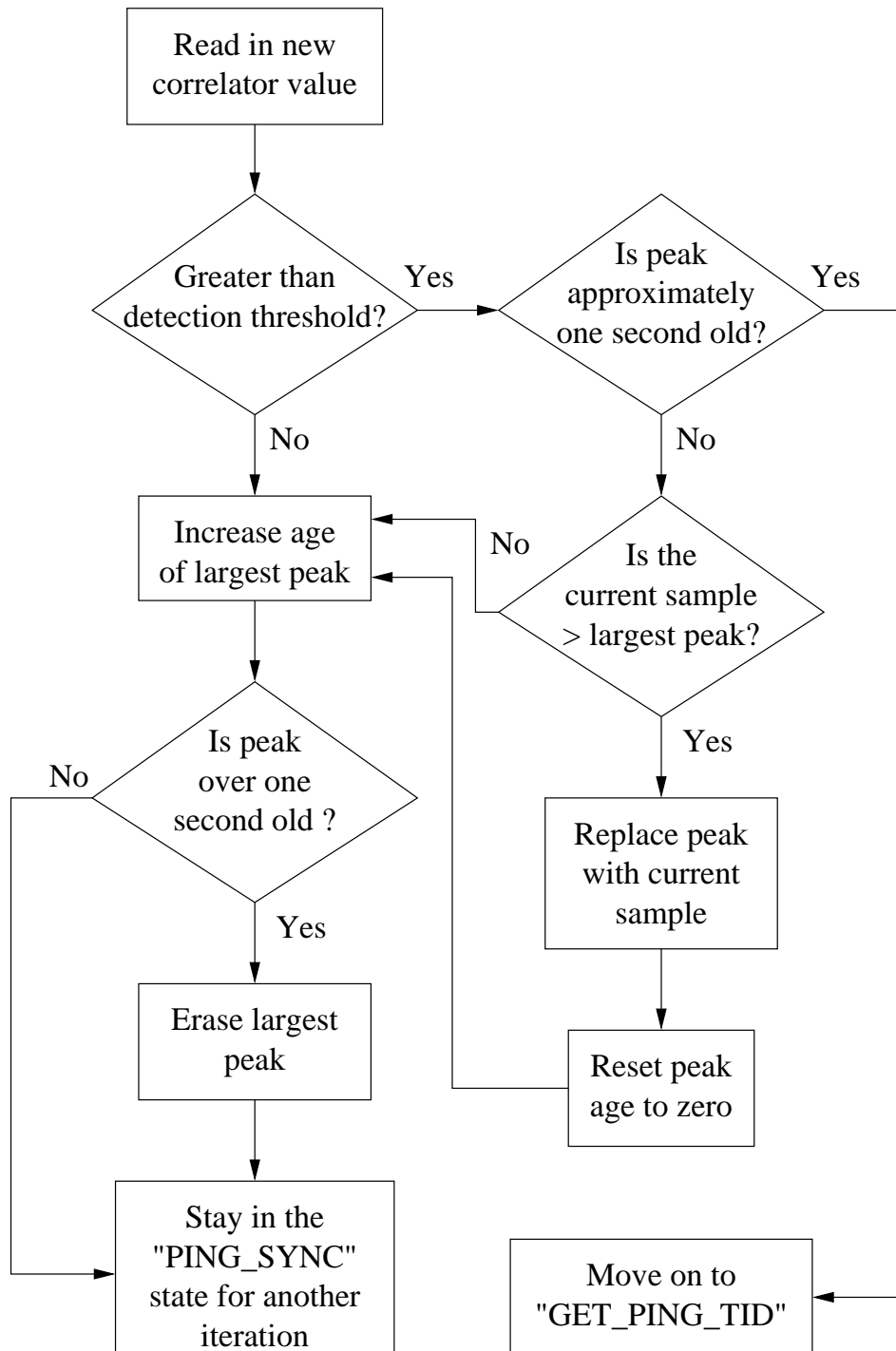


Figure 5.4: Ping synchronization flow chart.

erased. This procedure prevents synchronization to a false correlator peak that is not at the start of a data frame.

5.3.2.2 Get Ping TID

Once ping synchronization has been established, the sensor state machine moves into the target ID detection (“Get Ping TID”) state. This state records the target ID that corresponds to the synchronization ping. When the ping’s TID has been established, it is entered into an array containing the TIDs of the past four pings. The array of past TIDs is used to determine the data frame number of a packet when operating in time diversity mode.

5.3.2.3 Wait for Ping

Following the “Get Ping TID” state, the state machine enters the “Wait for Ping” state. This state simply waits a specified amount of time (determined by a “wait counter”) before moving on to “Watch for Detection”.

“Wait for Ping” can follow either the “Get Ping TID” or “Record Modulated Data” states. Each state sets the “wait counter” differently. “Get Ping TID” sets the “wait counter” to just under one second. That causes “Wait for Ping” to stop just short of the next ping that is due to arrive. “Record Modulated Data” sets the “wait counter” for 50 msec causing “Wait for Ping” to wait out most of the 59 msec quiet time that follows each data frame. Once “Wait for Ping” is finished the state is changed to “Watch for Detection”.

5.3.2.4 Watch for Detection

This state looks for the next detection threshold crossing in the correlator waveform. Once the threshold has been exceeded, the detection is assigned a job and a data channel within that job. This is where data from this detection will

be written. A `detected` flag is also set for this channel at this time. That flag will later be used by the job state machine described in Section 5.3.3 to determine which channels are being filled with valid data. Then the sensor state machine continues on to the state “Record Direct Path”.

However, if a threshold crossing does not occur within a timeout limit the sensor state machine is reset back to the “Ping Synchronization” state. Figure 5.5 contains a flow chart illustrating the “Watch for Detection” state.

5.3.2.5 Record Direct Path

The threshold crossing detected by “Watch for Detection” indicates the start of the direct signal path. This state records the direct signal path section of the correlator waveform for use in determining channel characteristics in the form of feedforward and non-sparse feedback equalizer taps. Section 5.4 discusses the calculations involved in determining the tap sizes based on this direct signal path recording.

In addition, this state also locates the maximum peak of the target ID correlation. The time at which the maximum peak occurs is used as the time origin for this detection.

The target ID corresponding to the maximum peak is also recorded into the array of past TIDs and used to determine the data frame number of this packet. A check of the TID array is performed to ensure that the expected TID sequence is being received. An out of sequence ID will set a `TID_error` flag that is used as a warning that this may be a false detection.

Once the direct path (5 msec time window) has been recorded into the job structure, the sensor’s state is changed to “Build Sparsing List”. In the event that the direct path lasts longer than the 5 msec time window, the rest of the

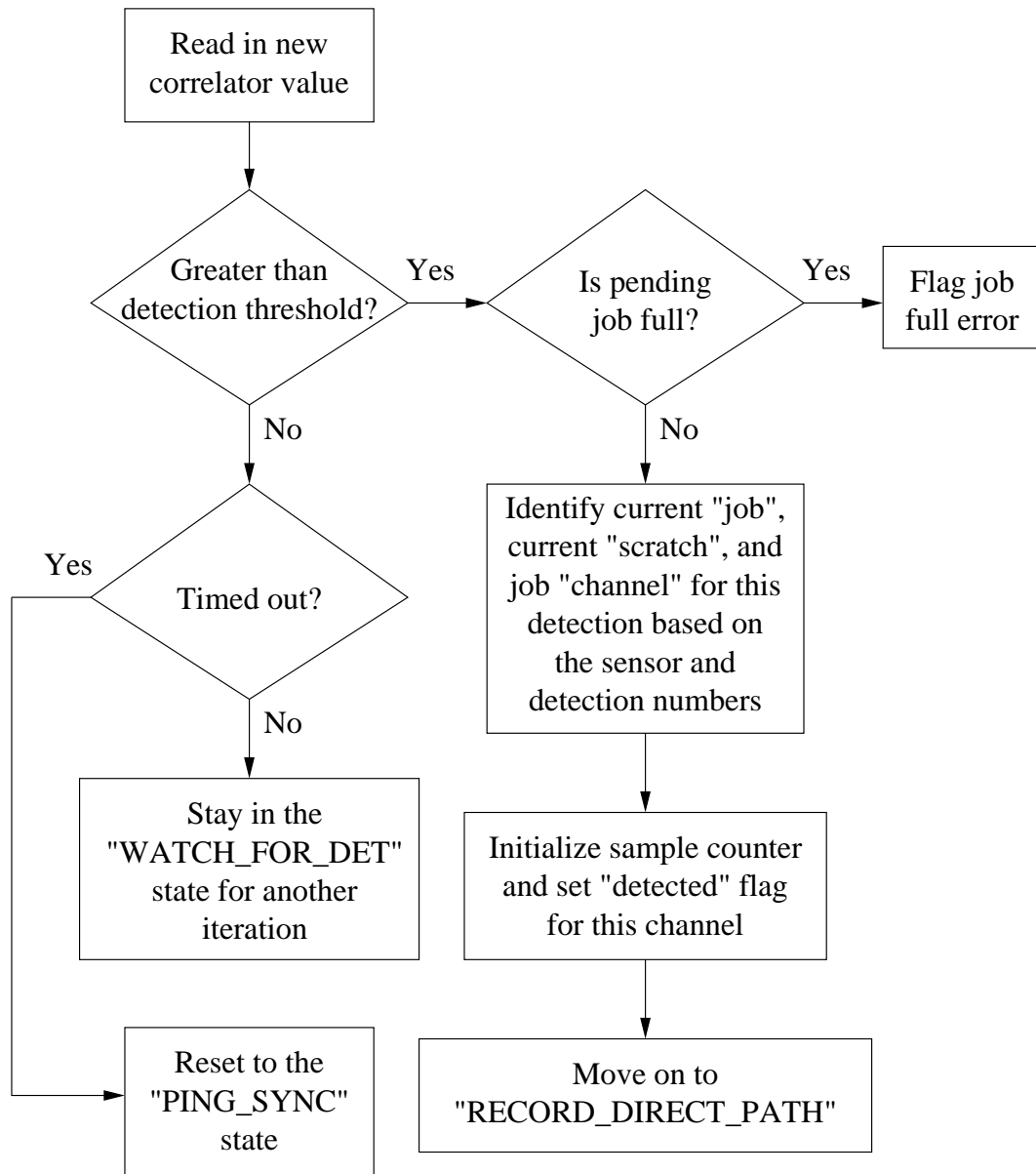


Figure 5.5: Watch for detection flow chart.

direct path will be picked up by the sparsing list and later merged with the direct feedback path during the final tap calculations.

5.3.2.6 Build Sparsing List

Sensors in the “Build Sparsing List” state update the sparsing list (described in Section 5.4.3). This list is later used to extract channel characteristics that are fed into the equalizer in the form of sparse feedback tap sizes and locations. The actual calculation of these sparse feedback taps is discussed in further detail in Section 5.4.

This state lasts until the end of the 100 msec quiet time in the data frame. When completed, the sensor enters the “Record Modulated Data” state.

5.3.2.7 Record Modulated Data

Sensors in this state record the demodulated (and Doppler compensated) message portion of the data frame into an appropriate slot in the current job. The appropriate slot and current job were determined while the sensor was in the “Watch for Detection” state. Recording the demodulated data consists of copying samples of `demod` (Table 5.2) into the appropriate row (slot) of `v` (Table 5.3). Since the “Record Modulated Data” state is always entered exactly 100 msec after the peak correlation with the synchronization ping this effectively time aligns the data.

As was described in Section 2.3, this section of data lasts for 800 msec. A sample counter is used to set a halfway (`half`) flag at the 400 msec point. The `half` flag is later used to see if the job state machine (described in Section 5.3.3) is ready to be updated.

Once the sample counter has reached the 800 msec point, the sensor has completed recording its demodulated data. At this point, the sensor is marked complete by setting its `pop` flag in the job data structure. Then the sensor state

machine is returned to the “Wait for Ping” state to wait out the 59 msec quiet time that follows each data frame.

Since all of the flags such as `half`, `det`, `pop`, and `aborted` reside within the job structure (rather than the sensor structure) there is no need for the sensor state machine to perform any sort of clean up or re-initialization at the end of each data frame. Those tasks are performed at the end of the “Send Job” state described in Section 5.3.3.4.

5.3.3 Job State Machine

Following each update of the sensor state machine, the `half` flags for each channel are examined to see if the job state machine needs to be updated.

In the case of no time diversity, a set `half` flag on at least one of the channels will cause the channel combiner’s status to change to “Job Maintenance” which in turn updates the job state machine. When time diversity is being used, only the channels associated with the second transmitted data frame are examined. A set `half` flag on any one of those channels will move the channel combiner into the “Job Maintenance” status where the job state machine will be updated.

Figure 5.6 presents a diagram of the job state machine. It takes on one of four different states.

5.3.3.1 Abort Channels

While in this state, all the channels associated with the active job are examined for detections. This is done using the `detected` flags that were set in the “Watch for Detection” state of the sensor state machine (Section 5.3.2.4). Any channels that have not been detected by this point are aborted by setting their sensor state back to “Ping Sync” and setting their `aborted` flag.

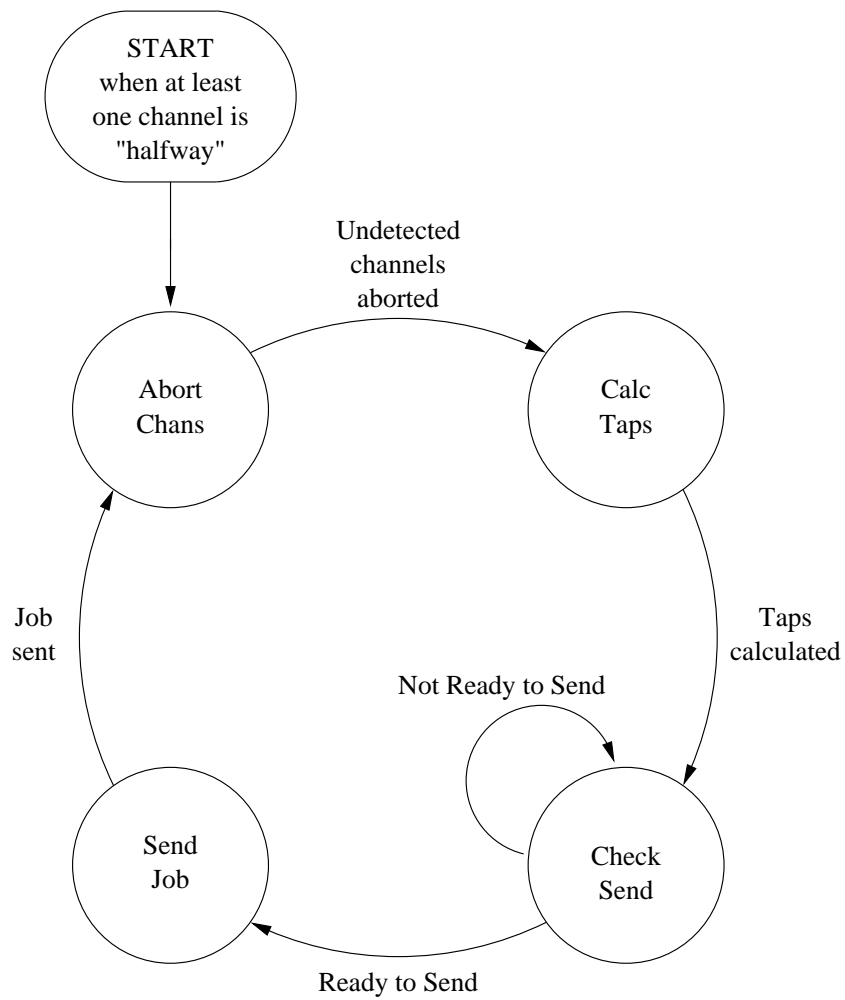


Figure 5.6: State flow diagram for the job state machine.

Since the job state machine is only updated after at least one channel has been detected and reached the halfway point, any undetected channels can be safely aborted at this point. Once this check has been performed and necessary channels aborted, the job state machine moves on to “Calc Taps”.

5.3.3.2 Calculate Taps

The “Calc Taps” state performs the equalizer tap calculations as described in Section 5.4. These calculations are based on the linked list of maximum correlator values described in Section 5.4.2. Following the calculations, the job state is changed to “Check Send”.

5.3.3.3 Check Send

The “Check Send” state determines whether or not a job is ready to be sent to an equalizer. First, the number of full and aborted channels are counted. If the number of full channels plus the number of aborted channels does not add up to the total number of channels, then there are still some channels being filled. Note that for time diversity mode, only the channels associated with the second transmitted data frame are checked in this manner.

In the event that there are still detected channels waiting to finish filling, the job state machine is kept in the “Check Send” state and the channel combiner’s status goes back to “Monitor Input”, which allows another buffer of input data to be processed by the sensor state machine.

Upon returning to “Job Maintenance” status (after processing the new buffer), the job state machine is now in the “Check Send” state where the above check will be performed again.

Once the number of aborted plus full channels matches the total number of channels it is time to send the job. The job is sent by moving into the “Send Job” state.

5.3.3.4 Send Job

Upon entering the “Send Job” state, the job is submitted to the equalizer module via a link port transfer as described in Section 5.2. The old common memory job transfer method (also described in Section 5.2) required the “Send Job” state to search for a free slot in the job queue prior to submitting the job.

After the job has been sent, some cleanup and bookkeeping are done. The local copy of the sent job is re-initialized to prepare it for reuse. Reinitializing the job clears all of the populated (`pop`), detected (`det`), halfway (`half`), and aborted flags as well as clearing out the entries from the sparsing list. Then any uncompleted sensors are returned to the state “Ping Sync”. In addition, the active and pending jobs are switched. Finally, the job state machine returns to the “Abort Chans” state.

Once the job state machine has been updated, the overall status of the channel combiner is set back to “Monitor Input” to read in a new buffer of input data.

5.4 Equalizer Tap Calculations

Section 2.3.2 mentioned that the quiet time between the synchronization ping and the actual data can be used to gather information about the link’s channel. This information must be presented to the equalizer in the form of tap locations and sizes. Information recorded by the “Record Direct Path” and “Build Sparsing List” states is used for just that purpose.

As was mentioned in Section 5.3.2.5 the detection threshold crossing in the “Watch for Detection” state signifies the start of the direct signal path. The feedforward section starts there and lasts until the peak in the correlator waveform. Both of those points were determined in the “Record Direct Path” state so the feedforward tap calculation is simply finding the distance between those two points. A feedforward tap scaling factor is also included so the user can fine tune the number of feedforward taps selected by this calculation. The center-tap equalizer requires an odd number of taps in its feedforward section. Therefore, if the number of taps turns out to be even, one is added to make it odd.

The non-sparse feedback section starts at the correlation peak and lasts until the direct path signal has dropped below a certain threshold. Calculation of the non-sparse feedback tap count involves searching through the recorded direct path for that threshold crossing and determining how far it is from the correlation peak. Again, a scaling factor is used to allow fine tuning of the non-sparse feedback tap count.

The selection of sparse feedback tap locations and sizes is performed using a linked list of the maximum correlator values. Whenever a sensor is in the “Build Sparsing List” state, it updates this list. Once all sensors (for a given job) have finished updating the list, it is then used to calculate the sparse feedback tap information.

5.4.1 Tap Selection Theory of Operation

This method of building and parsing a linked list of correlator values to determine tap locations was developed due to processing time constraints.

Input from the front-end module(s) arrives at a rate of 5000 samples/sec in packets containing roughly 60 samples (12 msec) each. With only 12 msec to process 60 samples (or 120 samples if receiving input from two front-end modules)

the channel combiner does not have much time available for a computationally unbalanced tap calculation algorithm.

An example of a computationally unbalanced algorithm would be storing the raw target ID correlation waveform and then searching through it for peaks. Storing the values would be quick and have time per sample to spare. However, searching the 100 msec (quiet time) vector would likely take longer than 12 msec and result in data loss.

Building and parsing a linked list is a more computationally balanced method. It spreads the processing time over a large number of samples. Updating the list for each new sample makes use of the spare processing time that would be wasted by just recording the sample into an array. The resulting list contains correlator peaks sorted by magnitude eliminating the need to search through a vector. Therefore, the remaining calculations needed to extract the tap information will not last long enough to result in data loss.

A benchmark similar to the ones performed on the equalizer was used to verify that this tap calculation method would fit within the channel combiner's time constraints. The time required for the "Build Sparsing List" function to process one sample was hardly measurable on an oscilloscope (well under the 200 μ sec available per sample). A typical list containing 30 entries required 0.3 msec to process. Again, well within the 12 msec available between input receptions.

The linked list of peak values also provides a memory savings over storing the raw correlator waveforms. By having all sensors associated with a job maintain the same list, it also provides a simple and effective way of combining the information from all channels.

5.4.2 The Linked List Structure

Each equalizer job scratch space (mentioned in Section 5.2.1) contains a linked list of correlator values that were recorded during the “Build Sparsing List” state. Each element of the linked list is a structure containing the following elements.

- pointer to the previous element
- pointer to the next element
- index of the correlator value
- the correlator value

The previous and next pointers are used for list bookkeeping. The index is used to keep track of the value’s original location in the input sequence. The index is taken with reference to the center of the target ID correlation peak found by “Record Direct Path”. This list is used to store a fixed number of the largest peaks (over all sensors associated with a job) in their correlator waveforms.

In addition to the list, a vector of pointers into the list is also maintained for bookkeeping purposes. These pointers are all initialized to NULL. When an element from location N in the original input sequence is added to the list, the N’th pointer is set to point to that list entry. This list of pointers will be used to index into the list during the tap calculation process. Figure 5.7 illustrates the relationship between the linked list and its corresponding vector of pointers.

5.4.3 Maintaining the List

This list of peaks in the correlator waveform is updated by sensors in the “Build Sparsing List” state. As each new correlator sample arrives, it is checked

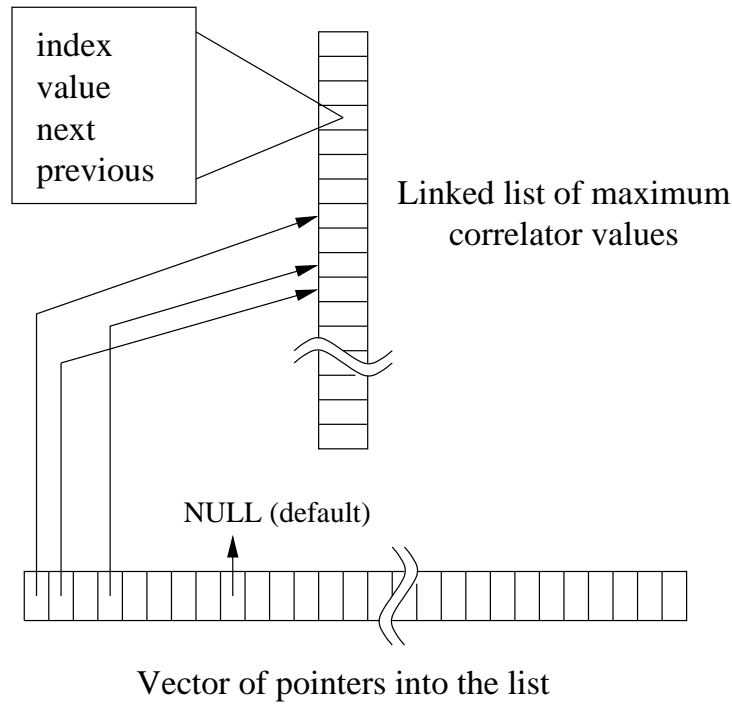


Figure 5.7: Linked list of correlator values and corresponding pointers.

against the last (smallest) element in the list. If the new sample is larger, it is added to the list. If not, the sample is discarded and the next sample is processed.

Since multiple sensors (each associated with the same job) all write to the same list, the possibility of repeated indices arises. This requires an additional check to be performed before writing a new value to the list. If a value is to be written to an already occupied index, the old value is overwritten only if it is smaller than the new one. If the index of the value to be written is not already occupied then the new value (along with its index) is written into the appropriate location based on its magnitude. To maintain a finite list size, smaller elements are deleted from the end of the list whenever a new entry is added. The flow chart in Figure 5.8 illustrates this procedure.

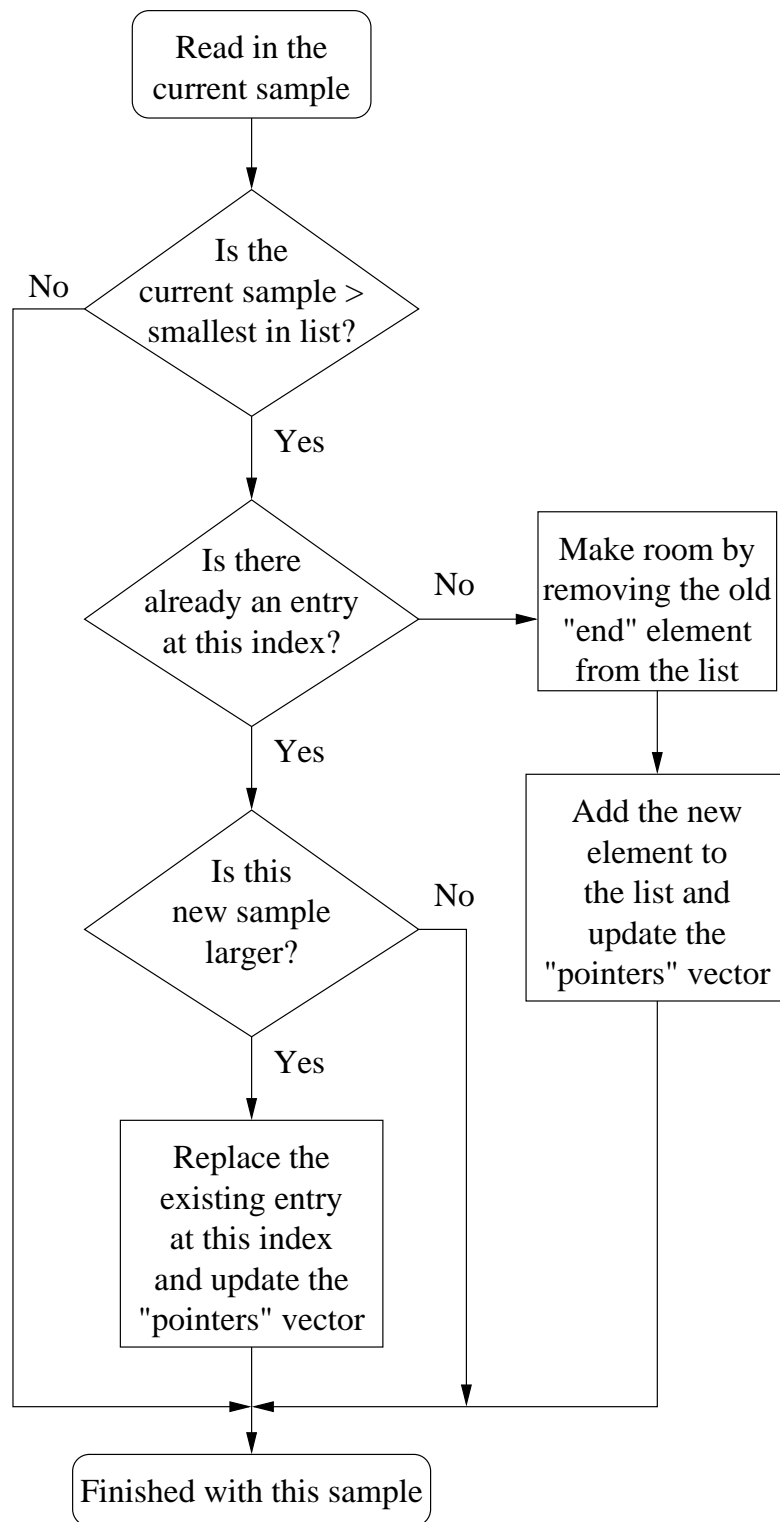


Figure 5.8: List construction and update flow chart.

5.4.4 Selection of Sparse Tap Sizes and Locations

Once all sensors have finished writing to the list, the tap and location calculations can begin. This is done when the job state machine of Section 5.3.3 is in the “Calc Taps” state. The feedforward and non-sparse feedback taps are calculated as described at the beginning of this section. Then the sparse feedback taps are calculated in the following manner.

First, a tap is centered at the index of the maximum value in the list. The width of the tap is initially set to a minimum tap selection width called `MIN_SECT`. These `MIN_SECT` elements are then deleted from the list along with their corresponding pointers in the bookkeeping pointer array. The pointers array is examined to see if the tap’s four neighbors (two on each side) are present in the list. If any of them are present, the tap width is widened by four and those entries are deleted from the list and pointers array as well. The widening continues until there are no more list entries at the neighboring indices. Then, the process repeats with the next tap location being set to the index at the top of the list. New taps are added and widened until either the maximum number of taps are allocated or the maximum number of tap locations are used. This sparse feedback tap calculation procedure is illustrated in Figure 5.9.

5.5 Channel Combiner Testing and Verification

Since the UDAT system makes extensive use of link ports to transfer data into and out of the channel combiner, it was imperative that the ping pong buffer and link port status check routines be thoroughly tested. These tests involve writing two pieces of test code, a sender and a receiver. The sender simply fills a buffer full of easily recognizable data and sends it out a link port by calling the link port sending code to set up a transfer. Meanwhile, the receiver test code

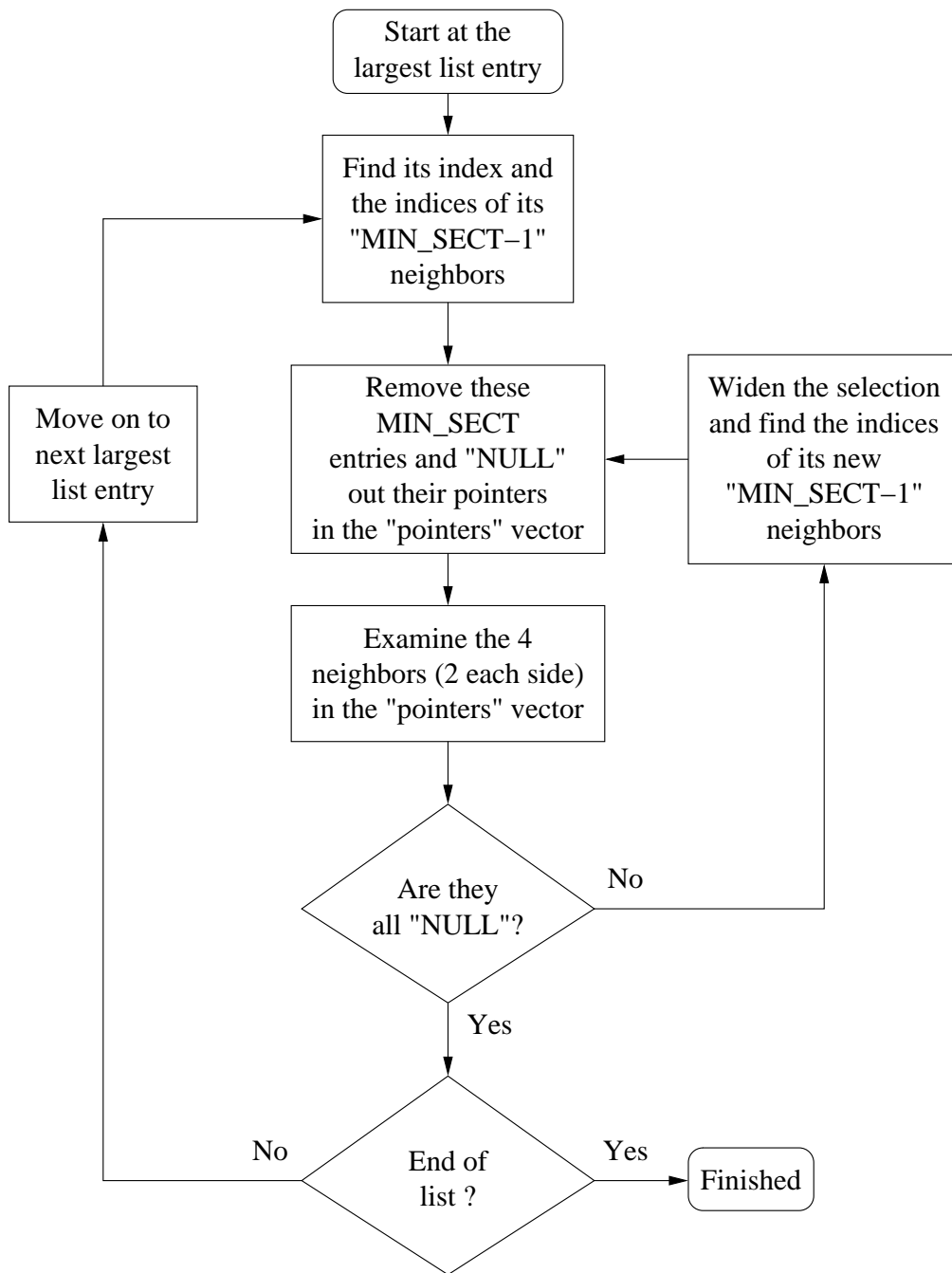


Figure 5.9: Sparse feedback tap calculation flow chart.

calls upon the link port ping pong buffer routines to receive the data. The Analog Devices Visual DSP debugger [10] was used to set a break point in the receiver test code and examine the contents of the ping pong buffers. The presence of the easily recognizable test data in the ping pong buffers confirms that the link port sending and receiving code operates correctly. Similarly, the link port status check (used by the “Send Job” state) was also tested using the Visual DSP debugger. This involves stopping the receiver and leaving the sender running. Examining the status flag proved that the link port status check code functions correctly.

Testing of the channel combiner module was broken down into individual testing of the key states as well as overall testing. The key states of the channel combiner include “Ping Synchronization”, “Watch for Detection”, “Build Sparsing List”, and “Calc Taps”.

5.5.1 Ping Synchronization Testing

The ping synchronization state of the sensor state machine was tested using artificial test data as the input to the channel combiner. The Visual DSP debugger was used to single step through the ping synchronization code and check for proper operation. Due to memory limitations it was not possible to store several seconds of artificial test data so the debugger was used to modify the peak age to simulate peaks that have gotten “too old”. As desired, the sensor state would only advance to “Get Ping TID” when two pings were approximately one second apart.

5.5.2 Watch for Detection Testing

Artificial testing data was also used to make sure the “Watch for Detection” state was working correctly. This testing data consists of a buffer of correlator values containing known peaks exceeding the detection threshold. Single stepping

through the code with the debugger was used to verify that the peaks are being detected and assigned to jobs correctly.

5.5.3 Build Sparsing List Testing

The “Build Sparsing List” code described in Sections 5.3.2 and 5.4 was thoroughly tested using a dedicated test routine. The test procedure involves calling the “Build Sparsing List” function with artificial testing data. This testing data consists of a vector populated with sparsely placed peaks. The resulting list are then examined in the debugger to be sure that the peaks and their locations are being stored correctly.

5.5.4 Calc Taps Testing

Sparse feedback tap calculation for the equalizer (Section 5.4) is based on the list generated by “Build Sparsing List”. Therefore, the same artificial test data and resulting list were also used to test the tap calculation code. Comparing the resulting tap widths and locations to the artificial data shows that the “Build Sparsing List” and “Calc Taps” routines are correctly centering appropriately sized taps on the peak locations.

5.5.5 Overall Channel Combiner Testing

Normal operation of the channel combiner requires a minimum of several seconds worth of input data to become synchronized and generate some equalizer jobs. Therefore memory size constraints limit the degree of testing that can be performed with artificial data.

Overall testing of the channel combiner involved writing an artificial data source module to supply data in the same manner as the actual front-end module. This data source stores one data frame (1 sec) worth of simulated target ID

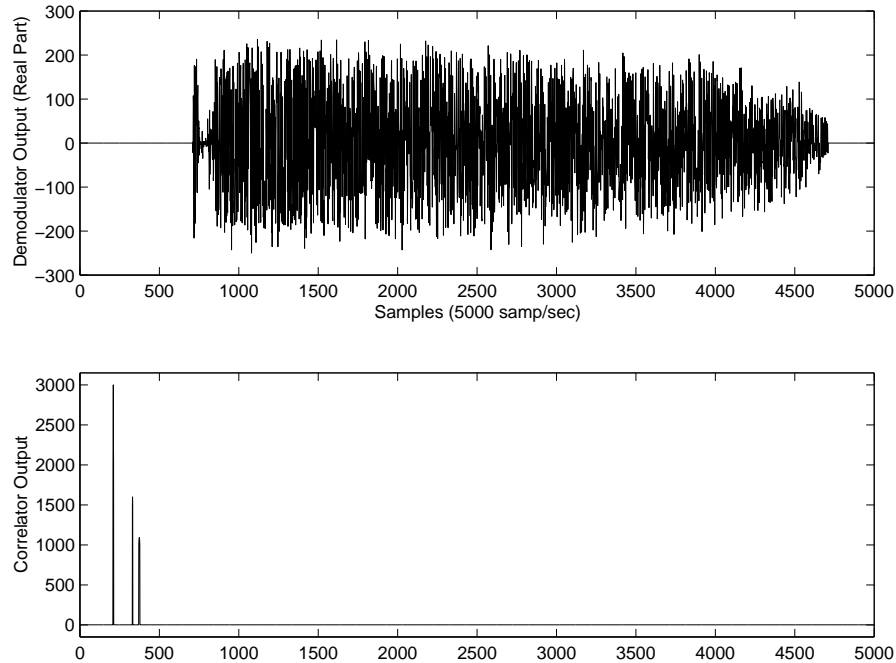


Figure 5.10: Sample of artificial testing demodulator (top) and correlator (bottom) waveforms.

correlator and demodulator output waveforms in local on-chip memory. Those waveforms are continuously transmitted through the link port to the channel combiner. Figure 5.10 shows the simulated waveforms that were repeatedly sent to the channel combiner for this test.

The output of the channel combiner was sent to an equalizer module and a breakpoint was placed in the equalizer module, allowing the resulting equalizer jobs to be analyzed for proper contents. The tap counts and locations in the resulting jobs were compared to the peaks in the simulated correlator waveform to verify that the channel combiner performs the tap calculations correctly. The simulated data source placed zeros in the non-message portions of the demodulator waveform making it easy to check that the channel combiner properly time aligns the data frames.

However, since this test continuously repeated the same exact packet it did not serve as a true test of the channel combiner's synchronization and detection

capabilities. Those features can only be tested by feeding real data into the complete system consisting of a front-end module, the channel combiner module, and one or more equalizer modules. Sections 6.1 and 6.2.2 present the test setup and procedures used for overall system testing.

CHAPTER 6

System Testing and Conclusions

This chapter outlines the hardware and software configuration used during the development and testing stages of this underwater digital acoustic telemetry (UDAT) system.

6.1 Test Setup Configuration

The system was developed and tested on a Spectrum Morocco II carrier board [6] containing eight 21060 SHARC digital signal processors. The system uses a modular design and makes extensive use of link ports to transfer data amongst the processors. This allows the system to be relatively flexible when migrating to other SHARC processor platforms.

These tests conducted on the UDAT system were performed on the setup shown in Figure 6.1. A digital audio tape (DAT) of actual underwater signal receptions (recorded aboard a submarine) is used to provide real data signals to the system. A PMC ADADIO board [9] containing analog to digital converters (ADCs) converts the DAT's audio output and makes it available to the SHARC processors on the Morocco II board. The ADADIO board also contains digital to analog converters (DACs) that were connected to an oscilloscope to conduct the equalizer benchmarks presented in Section 4.5. The front panel LEDs on the Morocco II also proved useful as an additional debugging tool.

Also shown in Figure 6.1 is a Mountain ICE JTAG interface pod [11], which provides the link between the Morocco II board and a PC running Visual DSP development and debugging software. The Visual DSP debugging software was used extensively to test the code for proper operation and store copies of various data buffers.

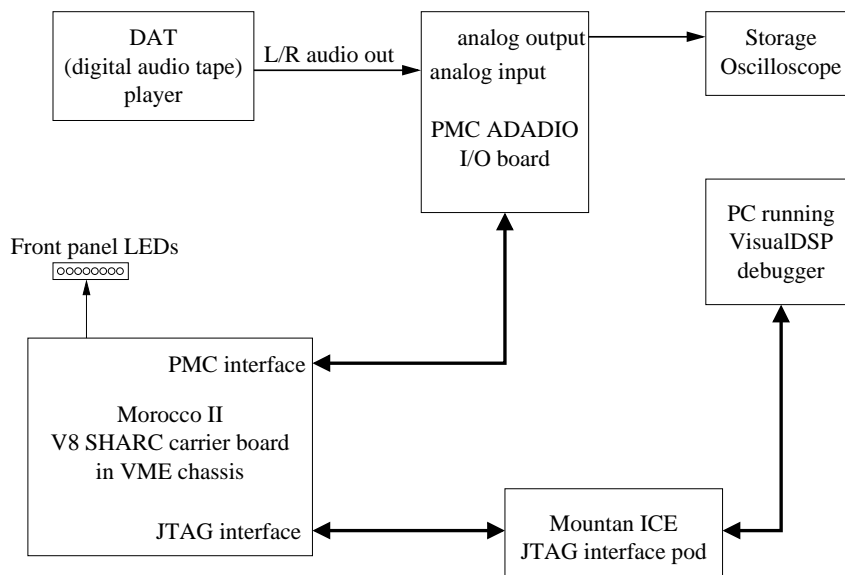


Figure 6.1: Test setup configuration.

Figure 6.2 illustrates the UDAT module layout on the Morocco II board. This configuration makes maximum use of the Morocco II’s link port interconnection structure and allows for five equalizer modules and one front-end module. An alternate configuration (used for two physical input channels) replaces the equalizer on DSP 0, Cluster 3 with a front-end module. That layout is illustrated in Figure 6.3. The “Data Source” shown on DSP 0, Cluster 1 in both configurations is a program (`m2_drv_bb_sim.c`) written by NUWC to read data from the ADADIO board and distribute it (via a link port) to another processor. Since `m2_drv_bb_sim.c` is limited to sending data out one link port, the front-end module running on DSP 0, Cluster 2 in Figure 6.3 must use a `send_to_back` option to pass the second channel of data along to another equalizer module on DSP 0, Cluster 3. Software configuration parameters and their default settings for these tests are listed in Appendix B.

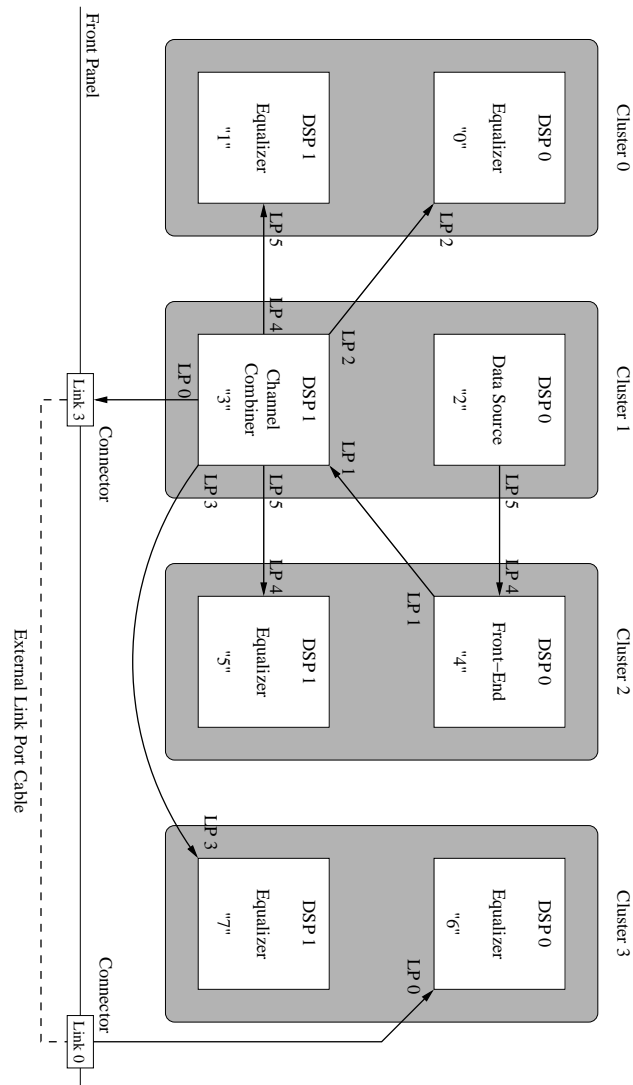


Figure 6.2: Module arrangement on Morocco II for one input channel.

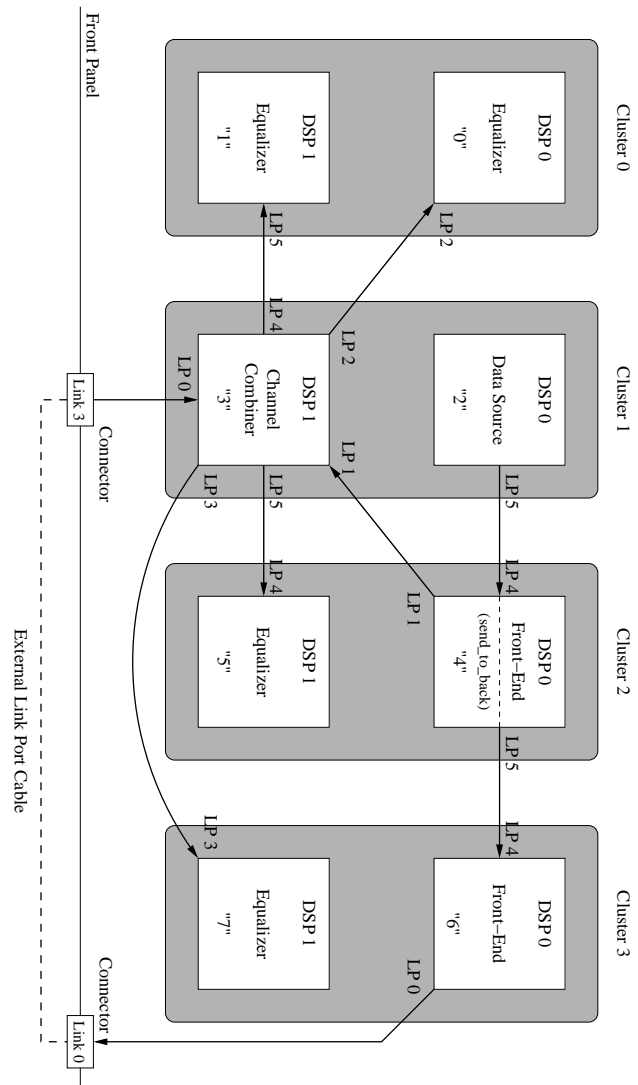


Figure 6.3: Module arrangement on Morocco II for two input channels.

6.2 System Testing Procedures and Results

Several procedures were used to test the UDAT system for proper functionality. These tests will be described in detail in the following sections.

6.2.1 Front-End Module Testing

The front-end module (presented in Chapter 3) was tested using the data on the DAT. The channel combiner module was temporarily replaced by a data gathering test routine that recorded several seconds of the front-end module's output waveforms. A three second section of the demodulator output waveform is shown in Figure 6.4. This waveform matches the structure of the data frame format presented in Section 2.3. Since it was sampled at a random time it starts out near the tail end of a frame. The region starting at around 0.2 sec and ending around 1.0 sec is a QPSK message portion. There is a clear quiet time following that message before the TID ping that occurs shortly after 1.0 sec. There is also a clear quiet time following that ping and the start of another message near 1.2 sec. However, this second message has a significant delayed multipath that obscures the following quiet time and starts to blend with the following TID ping that occurs shortly after 2.0 sec.

In addition to the demodulator output waveform, a corresponding three second section of the correlator output waveform was also recorded. This waveform is shown in Figure 6.5. As expected, the main detections show up as peaks one second apart. Detections from multipath signals are also present and in the case of the second detection (around 1.0 sec) they show up as a weaker peak in the correlation waveform. The third detection in this waveform (around 2.0 sec) shows a weaker main detection followed by a stronger delayed detection. The weakness of the main detection is due to the interference caused by the long delayed multipath signal present in the preceding frame.

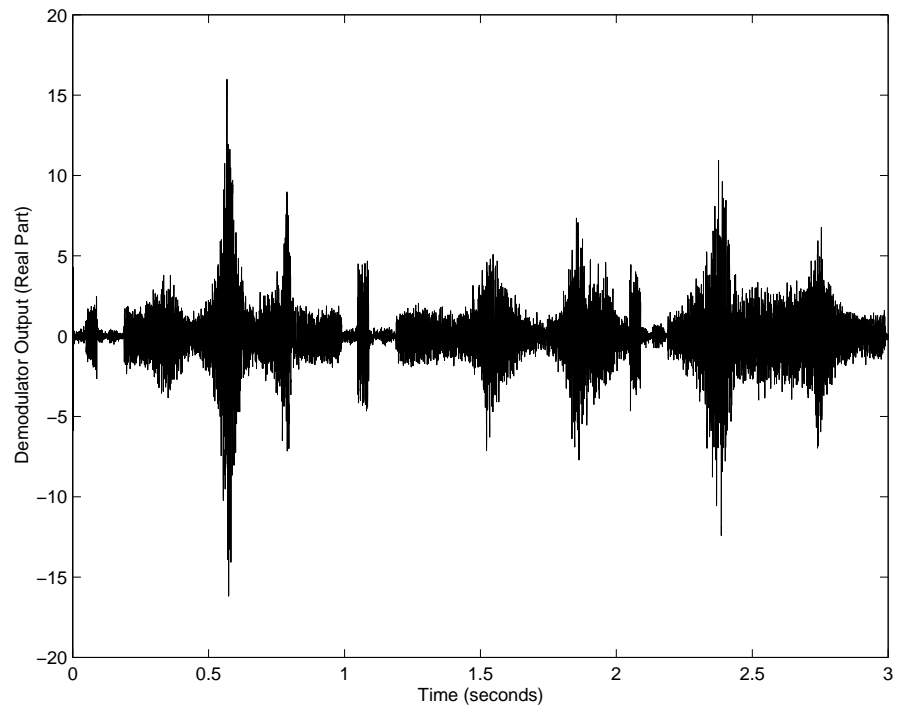


Figure 6.4: Sample of demodulator output waveform.

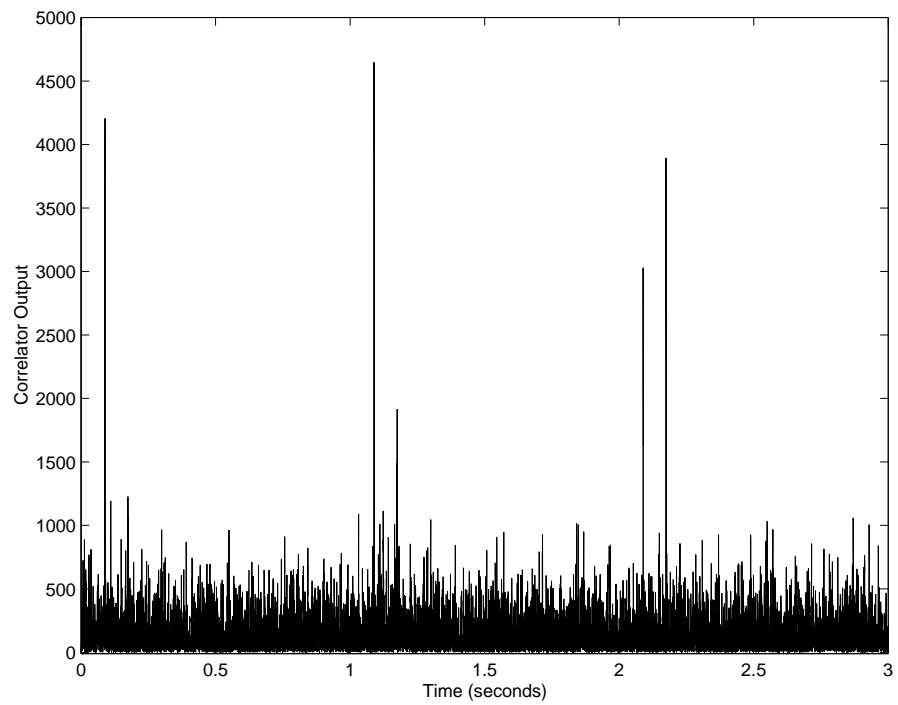


Figure 6.5: Sample of correlator output waveform.

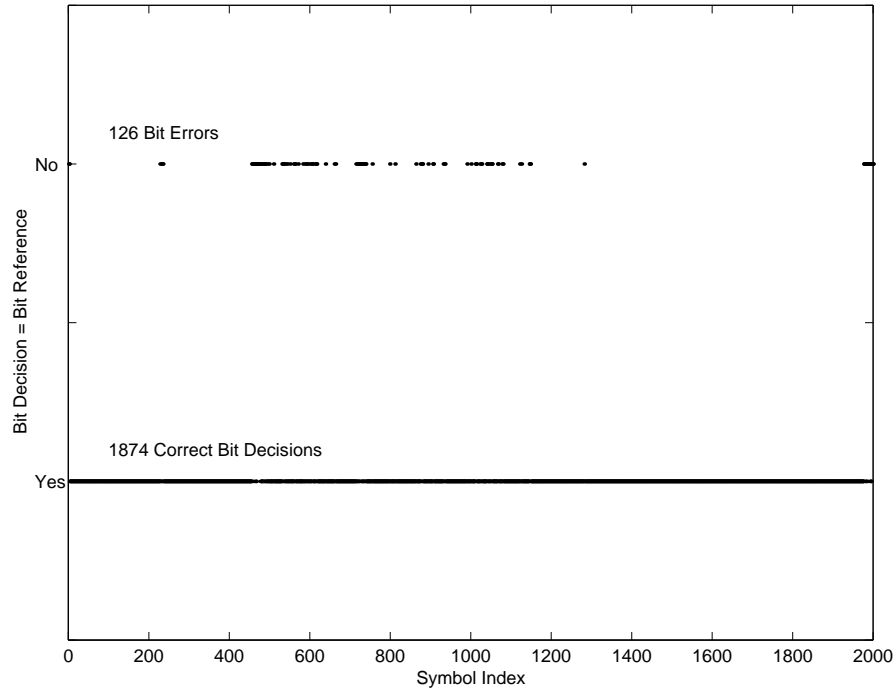


Figure 6.6: Equalization results from one input channel (no diversity).

6.2.2 Overall System Testing

Overall system tests were performed using the complete test setup that was described earlier in this Chapter. The setup is shown in Figures 6.1 and 6.2.

The DAT contains receptions of a known data packet that is repeatedly broadcast for several minutes. The debugger was used to place breakpoints in the equalizer modules to observe and store their outputs. The stored equalizer outputs were then compared against the known message in MATLAB. Figures 6.6 through 6.8 show the results of these tests.

The results of the single channel (no diversity) test shown in Figure 6.6 show that the system is working but the performance is rather poor with 126 bit errors (out of 2000 bits). This poor performance can be attributed to the fact that one input channel (no diversity) was used during that test. Figure 6.7 backs this up by showing how time diversity lowers the number of bit errors to 63. However, 63 bit errors is still a rather high error rate, indicating the system would likely benefit

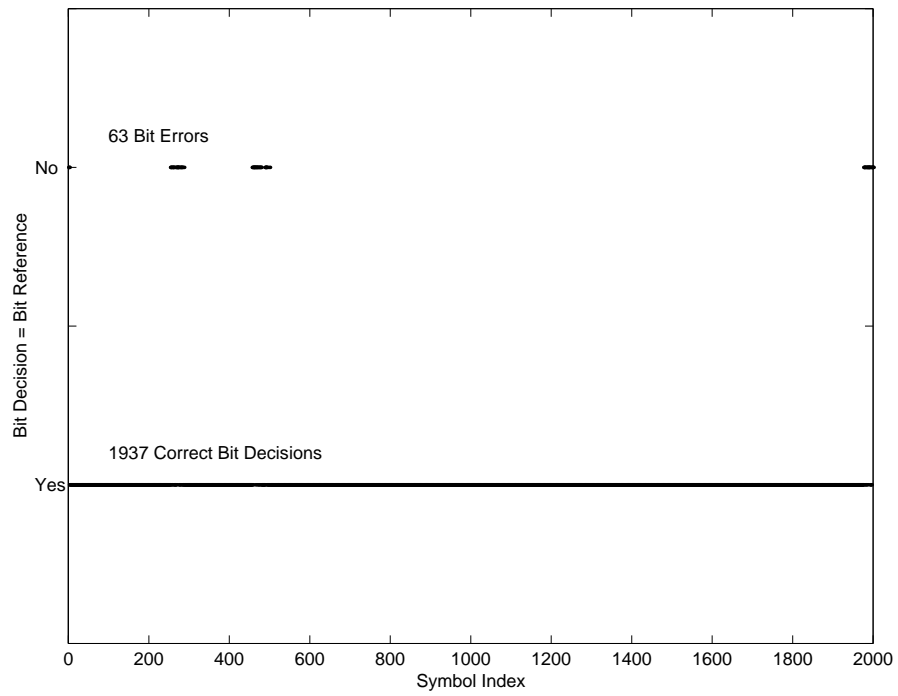


Figure 6.7: Equalization results from one input channel with time diversity.

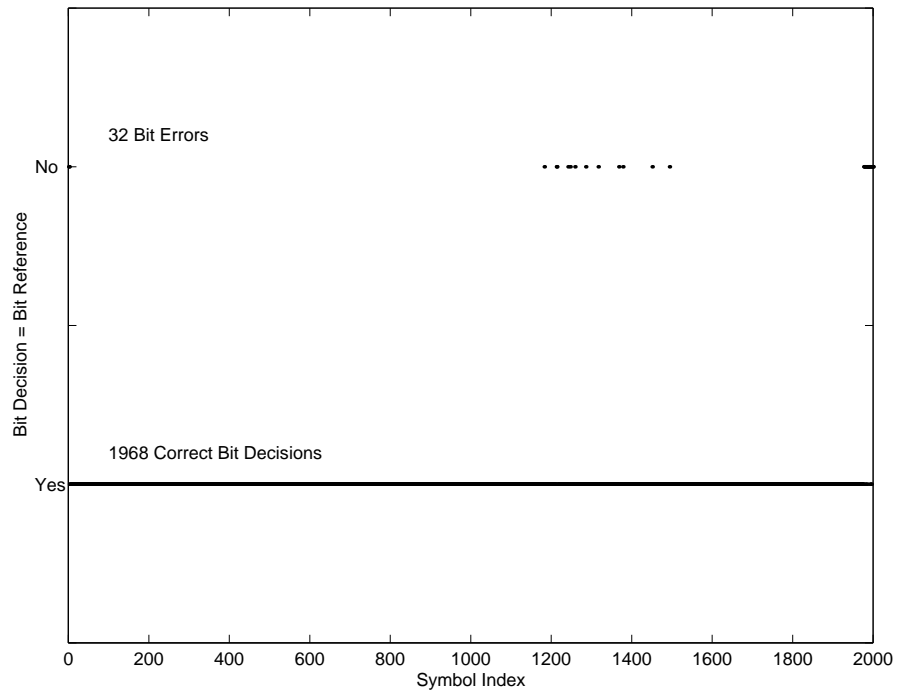


Figure 6.8: Equalization results from two input channels (spatial diversity).

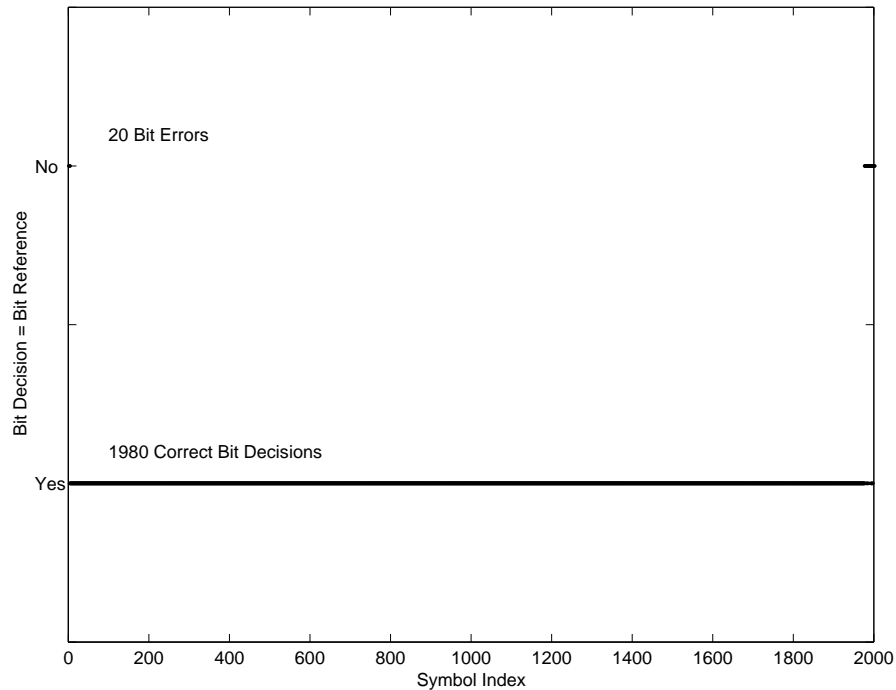


Figure 6.9: Equalization results from four input channels (spatial and time diversity).

from the fine tuning of the various parameters involved with selecting equalizer taps.

Testing the system's functionality for spatial diversity involves utilizing the DAT's left and right channels that contain recordings from two physically separated hydrophones (sensors). The test layout illustrated in Figure 6.3 utilizes a front-end module on DSP 0, Cluster 2 to perform calculations for one sensor and pass the raw data along to DSP 0, Cluster 3 using a `send_to_back` option in the link port setup. The second front-end module on DSP 0, Cluster 3 performs calculations for the second sensor. Both front-end modules send data to the channel combiner module on DSP 1, Cluster 1. The results of a spatial diversity test are presented in Figure 6.8. With only 32 bit errors, this is clearly an improvement over both the single channel and time diversity results. Again, further fine tuning is likely to result in even lower bit error rates.

Combining both spatial and time diversity serves to further lower bit errors. Figure 6.9 shows that two spatial diversity inputs, each using time diversity, brings the number of bit errors down to 20.

Further tests of the entire system were conducted using the Morocco II's front panel LEDs. Unlike the equalizer benchmark testing, these timing tests could not be done using the ADADIO board's DAC since the `m2_drv_bb_sim.c` program being used to read incoming data ties up the processor common bus. That leaves the front panel LEDs (one for each processor) as the only real-time diagnostic output. The first test conducted with the LEDs involved having the channel combiner turn its LED on whenever it entered the "peak too old" portion of the "Ping Synchronization" state. This provided a visual representation of how often the system was not detecting synchronization pings. As was expected, the LED would occasionally flash when running the system with input from the DAT. This indicates that every so often a ping was weak enough to fall below the detection threshold. If the LED had been flashing more frequently there would have been concern for lost detections and the detection threshold could have been lowered to compensate.

A similar test was done using the LED to indicate whenever synchronization had been lost, requiring the sensor to be sent back to the "Ping Synchronization" state. Again, the LED would light occasionally but not often enough to indicate a problem. As desired, playing a portion of the DAT without telemetry data kept the LED constantly illuminated.

Another "LED test" involved turning on the LED whenever the "Watch for Detection" state picked up a detection. The LED was then turned back off at the beginning of a new input data buffer. This test resulted in the LED blinking on once per second indicating that the system was indeed picking up detections at the expected rate for one physical input channel. A very similar test was performed

with the LED being turned on whenever an equalizer job was submitted. Again, the flashes were spaced at one second intervals as expected.

6.3 Conclusions and Future Work

The above tests verify that the UDAT system functions properly on the Morocco II platform. However, there are still several variables such as the detection threshold and tap size scaling factors that require fine tuning to achieve optimal system performance.

One example of parameter fine tuning was discovered during the testing disclosed in Section 6.2.2. It was found that it is desirable to place a limit on the number of sparse feedback taps generated by the channel combiner. Since the sparse feedback tap calculation algorithm (Section 5.4) utilizes peaks in the correlator waveform to place taps, it is possible for it to start including “noise” if the number of sparse feedback sections is too large. Observation of the target ID correlation waveforms (such as those shown in Figure 6.5) indicate that one or two sparse feedback sections should be sufficient for most situations.

As was discussed in the beginning of Chapter 4, the RLS algorithm was chosen to update the equalizer weight vector in this prototype implementation. Benchmarks conducted on this equalizer design (see Section 4.5) revealed that the computational complexity of the RLS algorithm limits real-time operation to about 64 taps. Recall from Table 4.2 that a 64 tap equalizer running on one of the Morocco II board’s eight 40 MHz 21060 SHARC processors requires 4.59 seconds per job. Therefore, real-time operation would require five equalizers (the maximum available since the remaining three processors are being used by other modules).

The 64 tap limit is likely to result in performance limitations. To avoid such limitations, a more computationally efficient algorithm such as the Fast

Transversal Filters (FTF) algorithm proposed in [4] must be implemented on a SHARC processor.

Error-correction encoding and decoding should also help lower error rates. Current receiver implementations all place error-correction decoding on a remote host. A faster equalizer algorithm should also allow time for a convolutional decoder to be integrated with the equalizer.

NUWC already has plans to move this system from the Morocco II 21060 platform to other SHARC platforms with faster processors. Faster processors will help reduce the equalizer computation time but many of the other platforms such as the Hammerhead [7] have fewer processors. So a faster equalizer algorithm will still be required. The faster processors should allow multiple front-end modules to be run on one processor. Such an arrangement would be beneficial in situations utilizing spatial diversity as it would free up a processor that could then be used to run another equalizer module.

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APPENDIX A

Memory Usage and Allocation

A.1 Channel Combiner Memory Usage

The channel combiner must store two equalizer job structures, each of which contains one second's worth of data from up to four input channels. The memory required to store each structure is listed in Table A.1. Storing two such structures (for the current and pending jobs) requires a total of 64,080 words. The 21060 and 21160 SHARC processors have 4 Mbits of internal memory that is divided into two 2 Mbit blocks. When used to store 32 bit words, these blocks have a capacity of 64 K (65,536) words. Therefore, the Linker Description File (LDF) used in conjunction with the channel combiner sets aside one complete block for the Data Memory Data Area (DMDA) used to store the equalizer jobs.

In addition to the equalizer jobs, the channel combiner must also store information required to calculate tap sizes and locations for the equalizer. That data is stored in the scratch space structures described in Chapter 5. Table A.2 lists the memory used to store each scratch space. (*) Note that the list size of 31 entries is only a typical size.

Since each job structure (current and pending) is accompanied with a corresponding scratch space, two scratch spaces must be stored which requires 1,668 words.

The channel combiner is equipped to receive data from two physical sensors. That data is transferred in ping pong buffers. The memory used by each front-end sub-structure (ping pong buffer) is listed in Table A.3. These two ping pong buffers (totaling 468 words) are stored within a sensor structure that also contains about

Variable Name	Data Type	Size	Memory Words
pop[]	int	1 x 4	4
half[]	int	1 x 4	4
det[]	int	1 x 4	4
aborted[]	int	1x4	4
L	int	1	1
M	int	1	1
M2[]	int	1 x 10	10
sumM2	int	1	1
lenM2	int	1	1
M2OFF[]	int	1 x 10	10
v[] []	cplx float	4 x 4000	32000
Total			32040

Table A.1: Memory used by the equalizer job structure.

Variable Name	Data Type	Size	Memory Words
dir_path[]	float	1 x 210	210
list	4-element structure	1 x 31*	124
pointers[]	int	1 x 500	500
Total			834

Table A.2: Memory used by the scratch space structure.

Variable Name	Data Type	Size	Memory Words
length	int	1	1
td_present	int	1	1
sensor_id	int	1	1
corr[]	float	1 x 60	60
TID[]	int	1 x 60	60
demod[]	cplx float	1 x 60	120
Total			243

Table A.3: Memory used by the front-end sub-structures.

20 other scalar elements, some of which are listed in Table 5.1. Therefore, storing two such sensor structures (one for each sensor) requires a total of 1,012 words.

Both the sensor data structure (1,012 words) and the scratch spaces (1,668 words) are stored in the Program Memory Data Area (PMDA). Since the LDF sets aside 32 K words in the PMDA there is plenty of space to store these structures. The LDF used with the channel combiner defines the following sizes:

- 16 K C code space on block 0
- 32 K PMDA on block 0
- 4 K stack on block 0
- 4 K heap on block 0
- 64 K DMDA on block 1

A.2 RLS Equalizer Memory Usage

Similarly, the RLS equalizer must also store large quantities of data. Storage space for the raw input data and the correlation matrix, `inv_R`, account for the majority of the equalizer's memory usage. Unlike the channel combiner, the equalizer receives information that can be used to calculate the memory requirements of a particular job. Therefore, the equalizer is able to use dynamic memory allocation to store most of its data on the heap. Table A.4 lists the items that the equalizer allocates on the heap. Since the sizes of these elements are variable the values in the table are calculated based on a typical maximum of 140 taps ($L = 20$ feedforward, $M = 20$ feedback, and 100 sparse feedback taps in 10 sections). The equalizer's LDF file sets aside a 56 K word heap to store this data.

The equalizer must also store a local copy of the incoming job which contains the raw input data. As was shown in Table A.1, these equalizer jobs require 32,040

Variable Name	Data Type	Size	Memory Words
d_ref2[]	cplx float	lenM2 (10)	20
W_k[]	cplx float	taps (140)	280
U_k[]	cplx float	taps (140)	280
a[]	cplx float	div * L (80)	160
b[]	cplx float	M (20)	40
b2[]	cplx float	sumM2 (100)	200
inv_R[] []	cplx float	taps x taps (19600)	39,200
invR_Uk[]	cplx float	taps (140)	280
v_k[] []	cplx float	div * L (80)	160
d_k[]	cplx float	M (20)	40
d_k2[]	cplx float	lenM2 * maxM2 (400)	800
q2[]	cplx float	lenM2 (10)	20
err[] (optional)	cplx float	lenV/2 (2000)	4000
Total			45,480

Table A.4: Equalizer memory allocated on the heap.

words of storage space. Therefore, the equalizer's LDF file creates a 40 K (40960) word PMDA in which to store this data along with the equalizer's output decision (8000 words). The LDF used with the equalizer defines the following sizes:

- 16 K C code space on block 0
- 40 K PMDA on block 0
- 4 K DMDA on block 1
- 4 K stack on block 1
- 56 K heap on block 1

APPENDIX B

Software Configuration

B.1 Front-End Module Parameters

The front-end modules each have a `Run_Time_Parms` data structure that contains user selectable run-time parameters. The contents of this structure are listed in Table B.1.

Parameter Name	Description	Default Value
<code>carrier_frequency</code>	Carrier frequency (Hz)	12,500
<code>tid_A</code>	First TID for synch ping	1
<code>tid_B</code>	Second TID for synch ping	2
<code>sensor_id</code>	Hydrophone ID number	100
<code>vic_channel</code>	VIC input channel	1
<code>send_to_back</code>	Pass data through to another processor	false
<code>lp_xmt</code>	Link port to use with <code>send_to_back</code>	5
<code>pilot_detection_factor</code>	Pilot detection factor	0.02
<code>pilot_location</code>	Pilot tone above/below the carrier frequency	below (0)
<code>output_destination</code>	DSP ID of the channel combiner processor	3
<code>phase_jam_flag</code>	Phase jamming as described in Section 3.7	off (0)

Table B.1: Front-end module run-time parameters.

B.2 RLS Equalizer Module Parameters

The equalizer modules also contain user selectable run-time parameters shown in Table B.2.

Parameter Name	Description	Default Value
lmda	RLS forgetting factor	0.995
K1	Phase locked loop tracking constant 1	0.01
K1	Phase locked loop tracking constant 2	0.001
lp_rcv	Link port connected to the channel combiner	2

Table B.2: RLS equalizer module run-time parameters.

B.3 Channel Combiner Parameters

The channel combiner software has both compile-time and run-time parameters. The compile-time flag `TIME_DIV` (found in `chan_combiner.c`) is used to enable time diversity operation. When `TIME_DIV` is disabled, the channel combiner operates only in spatial diversity mode and treats consecutive data frames as new messages. When `TIME_DIV` is enabled, the channel combiner operates in time diversity mode and treats pairs of consecutive data frames as time diversity pairs. Note that when the channel combiner is operating in time diversity mode it is still capable of handling spatial diversity inputs as well.

Additional compile-time parameters for the channel combiner are listed in Table B.3.

The run-time parameters for the channel combiner are listed in Table B.4. Note that `num_sensors` has a default value of either 1 or 2. One sensor can allow the system to operate in time diversity mode. At least two sensors are required to operate in spatial diversity (or both time and spatial diversity) mode.

Parameter Name	Description	Default Value	Defined In File
TIMEOUT_LIMIT	Timeout limit used in “Watch for Detection”	50 msec	job.h
LIST_SIZE	Element count of the list used in “Build Sparsing List”	31	job.h
MIN_SECT	Minimum number of taps in one sparse feedback section	5	calc_taps.c

Table B.3: Channel combiner module compile-time parameters.

Parameter Name	Description	Default Value
det_thresh	Detection threshold	2000
tap_thresh	Tap threshold (see Section 5.4)	1000
ff_scale	Feedforward tap scaling factor	5
fb_scale	Feedback tap scaling factor	1
sp_fb_taps	Upper limit on number of sparse feedback sections	1
num_sensors	Number of diversity input sensors supplying data to the channel combiner	1 or 2
num_equalizers	Number of equalizers connected to the channel combiner	4
lp_rcv []	Vector of link ports connecting the channel combiner to front-ends	[1 0]
lp_xmt []	Vector of link ports connecting the channel combiner to equalizers	[2 3 4 5]

Table B.4: Channel combiner module run-time parameters.

BIOGRAPHY OF THE AUTHOR

Raymond McAvoy was born in Houlton, Maine on September 10, 1976. He received his high school education from Katahdin High School in Sherman, Maine in 1995.

He entered the University of Maine in 1995 and obtained his Bachelor of Science degree in Electrical Engineering in 1999.

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