

# Equalizer Optimization for TDECQ

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Version 1.1c

IEEE 802.3dj May Meeting

3 May 2026

Comments Addressed: I-92, I-395, I-396

# Supporters

- Chris Cole, Coherent

# Preview

- This presentation suggests changes to the draft specification that enable:
  - Well-defined equalizer convergence
  - Implementation with rapid execution times
- The goal was to remove ambiguity and facilitate repeatability while leaving room for alternative implementations
- This method optimizes mean squared error at a given sampling phase and finds the trial TDECQ at that sampling phase, then optimizes over the sampling phase to achieve the final TDECQ

# Proposed Changes to p.479

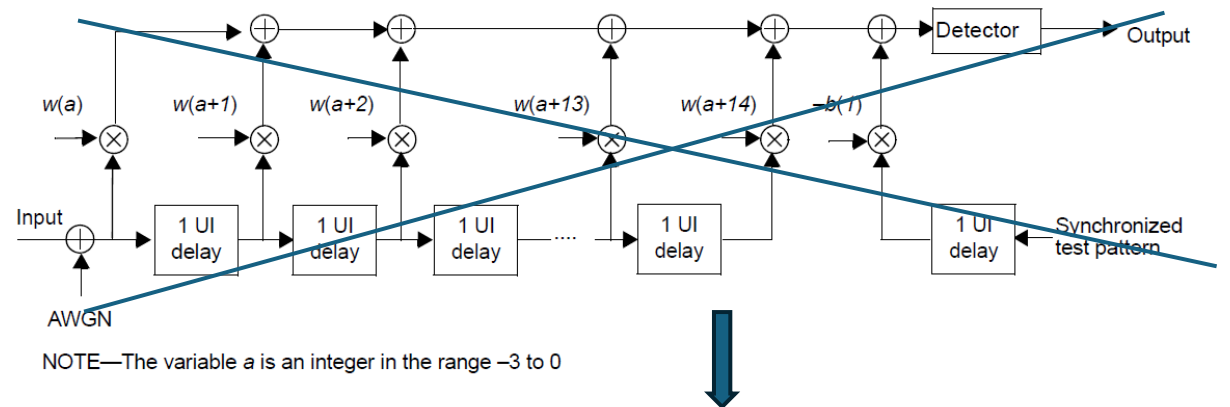
The reference equalizer is a T-spaced discrete-time equalizer with 15 feedforward taps and a single decision feedback tap, where T is the symbol period. Equalizer coefficient constraints are given in Table 180-16. The reference equalizer may be implemented in the oscilloscope. The decisions fed back in the equalizer are modeled as being correct, so the equalizer is modeled as shown in Figure 180-10, where  $x(n)$  is the  $n^{\text{th}}$  symbol in the test pattern sequence,  $r(t)$  is the output of the reference receiver defined in 180.9.2,  $\eta(t)$  is colored Gaussian noise,  $z(t) = r(t) + \eta(t)$ ,  $z_n = z(nT + \phi)$ , where  $0 \leq \phi < T$ , and  $y_n$  is the equalizer output corresponding to  $x(n)$ . The Bessel-Thomson (BT) filter shown is identical to that in 180.9.2. The received signal  $r(t)$  is aligned with the test pattern such that  $r(nT)$  corresponds to  $x(n)$ .

This slide was presented in swenson\_3dj\_01\_2605.pdf

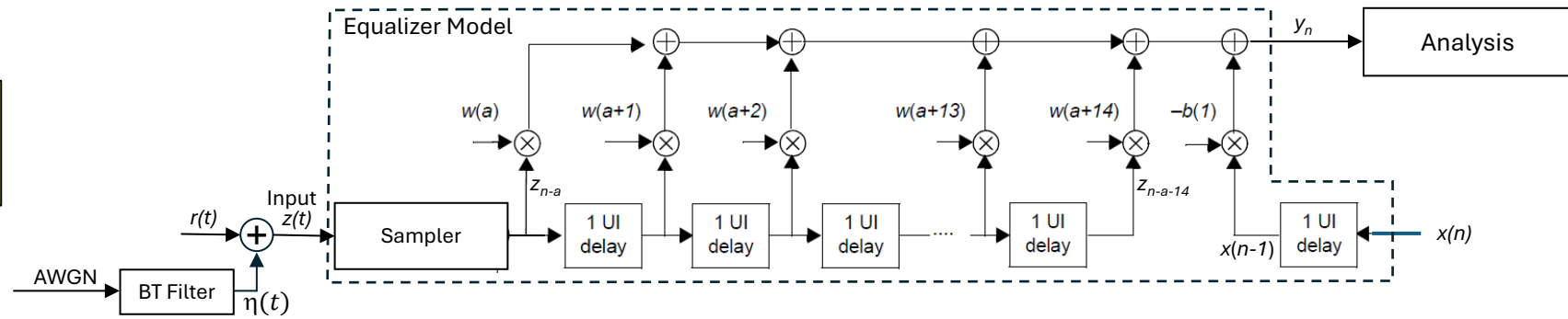
## 180.9.6.3 Reference equalizer

The reference equalizer is a 15-tap, T-spaced, feed-forward equalizer (FFE), followed by a 1-tap decision feedback equalizer (DFE), where T is the symbol period, with equalizer coefficient constraints as shown in Table 180-16. The reference equalizer may be implemented in the oscilloscope.

NOTE—This reference equalizer is part of the test and does not imply any particular receiver equalizer implementation.



NOTE—The variable a is an integer in the range -3 to 0



NOTE—The variable a is an integer in the range -3 to 0

# Proposed changes to p.480

## 180.9.6.4 TDECQ measurement method

The standard deviation of the noise of the reference receiver specified in 180.9.2,  $\sigma_S$ , is determined with no optical input signal and the same settings as used to capture the histograms described below.

The test pattern specified for TDECQ (see Table 180-13) is transmitted repetitively by the optical lane under test and the oscilloscope is set up to capture samples from all symbols in the complete pattern without averaging.

If an equivalent-time sampling oscilloscope is used, the impact of the sampling process and the reference equalizer on transmitter noise has to be compensated for, so that the correct magnitude of noise is present at the output of the equalizer.

The captured waveform is processed to find the largest noise that could be combined with the signal by an a reference receiver when optimally equalized by a reference equalizer. The optimal equalizer tap coefficients are dependent on the amount of noise that can be added to the signal, so finding the noise that can be added and the optimal equalizer setting is an iterative process. One way of doing this, using estimated PAM4 symbol error ratio as the figure of merit for the equalized signal, is described below.

The reference equalizer specified in 180.9.6.3 is applied to the waveform. An eye diagram is formed from the equalized captured waveform.

~~The average optical power ( $P_{ave}$ ) of the equalized eye diagram is determined, and the 0 UI and 1 UI crossing points are determined by the average of the eye diagram crossing times, as measured at  $P_{ave}$ , as illustrated in Figure 180-11.~~

as follows. For a given sampling phase  $\phi_0$  and assumed variance  $\sigma_G^2$  for  $\eta(t)$ , calculate the equalizer coefficients  $\mathbf{w} = [w(a) w(a+1) \dots w(a+14)]$  and b that result in the minimum mean squared error (MMSE) between  $y_n$  and  $x(n)$ , subject to the requirements of Table 180-16. With the coefficients fixed, sweep the sampling phase from  $\phi_0 - \frac{T}{2}$  to  $\phi_0 + \frac{T}{2}$ , and for each phase plot the scatter plot of the  $y_n$  values resulting from the test pattern sequence. An example eye diagram is shown in Figure 180-11. The feedback symbol  $x(n)$  changes at phase  $\phi = \phi_0 - \frac{T}{2}$ . Note: the Gaussian noise  $\eta(t)$  added to  $r(t)$  in Figure 180-10 is not reflected in the eye diagram; it is accounted for separately when estimating the SER in the analysis block.

This slide is a modified version of slide 6 in swenson\_3dj\_01\_2605.pdf. New text is highlighted.

# Proposed changes to p.481

Two vertical histograms are measured through the eye diagram, nominally centered at ~~0.45 UI and 0.55 UI.~~  $.05 \text{ UI}$  before and after sampling phase  $\phi_0$ . Each of the histogram windows spans all of the modulation levels of the eye diagram, as illustrated in Figure 180–11. The precise time position of the pair of histograms is adjusted to minimize TDECQ while keeping the histograms spaced  $0.1 \text{ UI}$  apart.

After the paragraph above, insert the following paragraph:

As described further below,  $\sigma_G$  quantifies the amount of noise that can be added at the receiver while still achieving the target symbol error rate. For a given value of  $\sigma_G$ , the symbol error rate for the left histogram,  $\text{SER}_L$ , and that for the right histogram,  $\text{SER}_R$ , are estimated as described below using the tap coefficients optimized to minimize mean squared error at  $\phi_0$  for that given value of  $\sigma_G$ . The nominal thresholds used in the estimation of the SER are the average power level  $P_{\text{ave}}$  and  $P_{\text{ave}} \pm \text{OMA}_{\text{TDECQ}}/3$ , as shown in Figure 180-11. When the SER is estimated, the thresholds are adjusted within 1% of their nominal values to minimize  $\max(\text{SER}_L, \text{SER}_R)$ . If  $\max(\text{SER}_L, \text{SER}_R)$  is not within 1% of the target PAM4 SER of  $4.56 \times 10^{-4}$ , the process is iterated, changing the value of  $\sigma_G$  and reoptimizing the equalizer taps until  $\max(\text{SER}_L, \text{SER}_R)$  is within 1% of the target. Having found the value of  $\sigma_G$  that causes the SER to fall within the target range,  $\text{TDECQ}(\phi_0)$  is calculated as described below.  $\text{TDECQ}(\phi_0)$  is then minimized over  $\phi_0$ , which finds the optimal location of the histograms. TDECQ is the minimum value thus found.

# Proposed changes to p. 482

Each element of the cumulative probability function,  $CF_{R1}(y_i)$ , is multiplied by a value  $G_{th1}(y_i)$ , and then summed to calculate an approximation for  $SER_{R1}$ , the partial PAM4 SER for threshold 1.  $CF_{R2}(y_i)$  and  $CF_{R3}(y_i)$  are treated similarly to calculate  $SER_{R2}$ , and  $SER_{R3}$ , the partial PAM4 SERs for threshold 2 and threshold 3. The sum of the three partial PAM4 SERs is the PAM4 SER associated with the right histogram,  $SER_R$ .

$G_{th1}(y_i)$  is equivalent to a Gaussian probability density function with an RMS value of  $\sigma_G$ , centered around the sub-eye threshold  $P_{th1}$ .  $G_{th1}(y_i)$  is given by Equation (180-5) and can be estimated by Equation (180-6).

This accounts for the Gaussian noise  $\eta(t)$  added at the input of the reference equalizer in Figure 180-10.

$$G_{th1}(y_i) = \int_{y_i - \frac{\Delta y}{2}}^{y_i + \frac{\Delta y}{2}} \frac{1}{C_{eq} \sigma_G \sqrt{2\pi}} \times e^{-\left(\frac{y - P_{th1}}{C_{eq} \sigma_G \sqrt{2}}\right)^2} dy \quad (180-5)$$

$$G_{th1}(y_i) = \frac{1}{C_{eq} \sigma_G \sqrt{2\pi}} \times e^{-\left(\frac{y_i - P_{th1}}{C_{eq} \sigma_G \sqrt{2}}\right)^2} \times \Delta y \quad (180-6)$$

$G_{th2}(y_i)$  and  $G_{th3}(y_i)$  are similar Gaussian probability density functions with the same RMS value of  $\sigma_G$ , centered around the sub-eye thresholds  $P_{th2}$  and  $P_{th3}$  respectively.  $G_{th2}(y_i)$  and  $G_{th3}(y_i)$  are given by Equation (180-7) and Equation (180-8) respectively.

$$G_{th2}(y_i) = \int_{y_i - \frac{\Delta y}{2}}^{y_i + \frac{\Delta y}{2}} \frac{1}{C_{eq} \sigma_G \sqrt{2\pi}} \times e^{-\left(\frac{y - P_{th2}}{C_{eq} \sigma_G \sqrt{2}}\right)^2} dy \quad (180-7)$$

$$G_{th3}(y_i) = \int_{y_i - \frac{\Delta y}{2}}^{y_i + \frac{\Delta y}{2}} \frac{1}{C_{eq} \sigma_G \sqrt{2\pi}} \times e^{-\left(\frac{y - P_{th3}}{C_{eq} \sigma_G \sqrt{2}}\right)^2} dy \quad (180-8)$$

# Proposed changes to p. 483

where

$C_{\text{eq}}$  is a coefficient which accounts for the reference equalizer noise enhancement

The value of  $C_{\text{eq}}$  can be calculated from the product of the normalized noise power density spectrum  $N(f)$  at the input of the reference equalizer and the normalized frequency response  $H_{\text{eq}}(f)$  of the reference equalizer, as shown in Equation (180–9).  
feedforward section of the

$$C_{\text{eq}} = \sqrt{\int_f N(f) \times |H_{\text{eq}}(f)|^2 df} \quad (180-9)$$

where

$N(f)$  is the normalized noise power density spectrum equivalent to white noise filtered by a fourth-order Bessel-Thomson response filter with a 3 dB bandwidth of 53.125 GHz.

and

$$\int_f N(f) df = H_{\text{eq}}(f=0) = 1 \quad (180-10)$$

# Proposed changes to p. 483 (cont'd)

~~The equalizer tap coefficients are iteratively adjusted and  $SER_L$  and  $SER_R$  calculated until the largest of  $SER_L$  and  $SER_R$  is minimized. Then, if the larger of  $SER_L$  and  $SER_R$  is greater than the target PAM4 SER of  $4.56 \times 10^{-4}$ , the value of  $\sigma_G$  is decreased and the process of equalizer optimization is repeated; If the larger of  $SER_L$  and  $SER_R$  is lower than the target PAM4 SER of  $4.56 \times 10^{-4}$ , then the value of  $\sigma_G$  is increased and the process of equalizer optimization is repeated.~~

~~$P_{th1}$ ,  $P_{th2}$ , and  $P_{th3}$  are varied from their nominal values by up to  $\pm 1\%$  of  $OMA_{TDECQ}$  in order to optimize TDECQ. The same three thresholds are used for both the left and the right histogram.~~

~~When the larger of  $SER_L$  and  $SER_R$  is equal to the target PAM4 SER of  $4.56 \times 10^{-4}$ , and the value of  $\sigma_G$  cannot be increased by further optimization of the equalizer tap coefficients or the sub-eye threshold levels, then TDECQ is calculated.~~

The RMS noise, R, that could be added by a receiver is given by Equation (180-11).

$$R = \sqrt{\sigma_G^2 + \sigma_s^2} \quad (180-11)$$

~~TDECQ is given by Equation (180-12).~~

At a given sampling phase  $\phi_0$ , TDECQ( $\phi_0$ ) is given by Equation (180-12)

$$TDECQ(\phi_0) = 10 \log_{10} \left( \frac{OMA_{outer}}{6} \times \frac{1}{Q_t R} \right) \quad (180-12)$$

where

$OMA_{outer}$  is measured as defined in 180.9.5  
 $Q_t$  is 3.428, consistent with the target symbol error ratio for Gray mapped PAM4, and can be calculated according to Equation (180-27)

Finally, TDECQ is determined by minimizing over  $\phi(0)$ :

$$TDECQ = \min_{\phi_0} TDECQ(\phi_0)$$

~~Alternative optimization methods such as minimum mean squared error (MMSE) may be used to determine equalizer tap weights to reduce test time, and are expected to report equal or higher values of TDECQ. These alternative methods should not be used for receiver sensitivity and stressed receiver sensitivity calibration.~~

The iteration of  $\sigma_G$  and the tap values to hit the target SER is covered in the text on slide 6.

Replace with: “Alternative optimization methods that achieve the same TDECQ are valid.”

# Thank You

# Backup (Presented Earlier)

# Background of Reference Equalizer for TP2 Compliance Testing

- The first reference equalizer for 802.3 compliance testing was TWDP (Transmitter Waveform and Dispersion Penalty) in Clause 68.6.6 for 10GBASE LRM
  - December 2004 contribution by Swenson, Voois, Lindsay, and Zeng
    - <https://grouper.ieee.org/groups/802/3/aq/public/tools/>
    - N. Swenson, et al., "Explanation of IEEE 802.3, Clause 68 TWDP," 2006, available as Clause\_68\_TWDP.pdf at [https://standards.ieee.org/wp-content/uploads/2022/07/802.3-2022\\_downloads.zip](https://standards.ieee.org/wp-content/uploads/2022/07/802.3-2022_downloads.zip)
    - N. L. Swenson, et al., "Standards Compliance Testing of Optical Transmitters Using a Software-Based Equalizing Reference Receiver," in *OFC and NFOEC*, 2007, paper NWC3. <https://opg.optica.org/abstract.cfm?URI=NFOEC-2007-NWC3>
  - The final version of the TWDP reference equalizer was a T/2-spaced DFE with 14 feedforward taps and 5 feedback taps
- TDEC was subsequently introduced for binary NRZ waveforms using a T-spaced FFE reference equalizer
- TDECQ followed for PAM-4, also based on a feed forward equalizer

# The Search

Comments I-395, I-396 (Swenson), Comment I-92 (Adee)

Suggested Resolution: Specify that an MMSE solution is compliant whenever TDECQ or TECQ is evaluated.

# Test Definition Must Be Well-Defined and Implementable for Repeatability

- It is impractical to do a brute force search to find a global optimum with 16+ degrees of freedom<sup>\*</sup>
- MMSE optimization is well established for DFE optimization – why aren't we using it?
  - Algorithm is well-defined, implementable, and readily available

<sup>\*</sup>There are (arguably) 23 degrees of freedom to search over: 16 (taps) + 1 (number of precursors) + 1 (phase) + 1 (noise level) + 3 (threshold levels) + 1 (histogram selection)

# Brute Force Search vs MMSE for Decision Feedback Equalizer (DFE) optimization with 15 feedforward taps and 1 feedback tap (16 taps total)

| Feature                     | Brute Force Search                                     | MMSE (Analytical Solution)                      |
|-----------------------------|--|---|
| <b>Mathematical Goal</b>    | Exhaustive search for the lowest Error/BER.            | Direct solution of the Wiener-Hopf equations.   |
| <b>Search Space</b>         | 16 –dimensional hyper-grid.                            | Single point (intersection of 16 hyper-planes). |
| <b>Complexity Class</b>     | $O(S^N)$ where $S$ is steps and $N = 16$ .             | $O(N^3)$ , where $N = 16$ .                     |
| <b>Estimated Operations</b> | $\approx 1.2 \times 10^{24}$ (assuming 32 values/tap). | $\approx 4,096$ floating-point operations.      |
| <b>Compute Time</b>         | $\approx 38,000$ years (at 1 THz test rate).           | $\approx 100$ microseconds (on a standard PC).  |
| <b>Channel Knowledge</b>    | None required (blind testing).                         | Requires Pulse Response or Training Data.       |
| <b>Global Optimality</b>    | Guaranteed (eventually).                               | Guaranteed (instantly) for Mean Square Error.   |
| <b>Noise Handling</b>       | Evaluates noise impact per iteration.                  | Accounts for noise via diagonal regularization. |