



Dead-Time Effects on Geiger-Mode APD Performance

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Simulated Linear and Geiger Mode Signals

- ◆ Dead Time: 10 μs
- ◆ Pulse Length: 5 μs
- ◆ BG Flux: 200 kHz
- ◆ #signal photons/pulse: 0.25
- ◆ 2/dead time

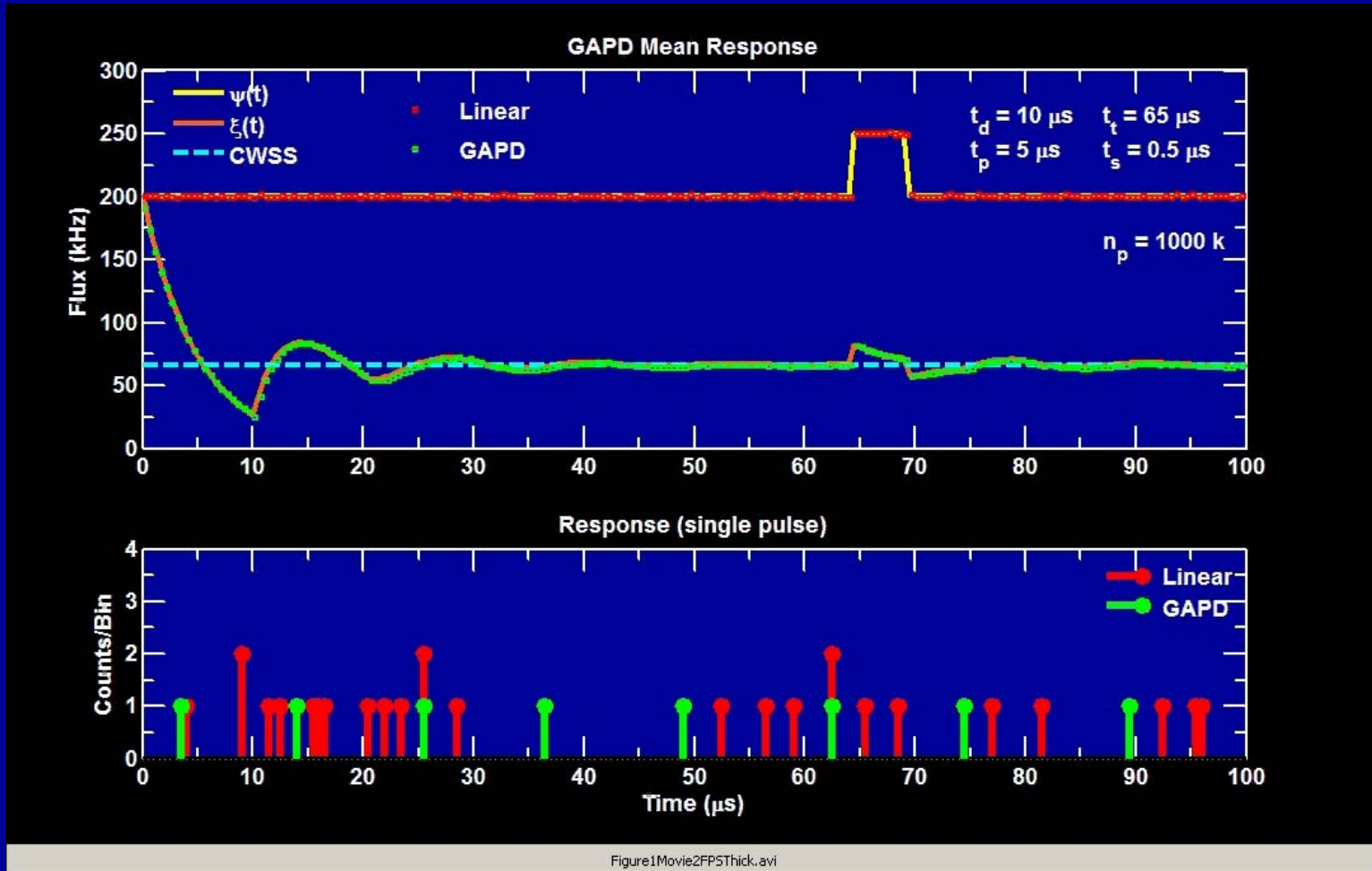


Figure1Movie2FP5Thick.avi

Outline

- ◆ **Problem Overview and Assumptions**
- ◆ **Response Theory**
- ◆ **Efficiency Theory**
- ◆ **SNR Theory**
- ◆ **Detection Statistics**
- ◆ **Advantages of Micro-detectors**
- ◆ **Summary/Conclusions**



Problem Overview

◆ GAPD Model

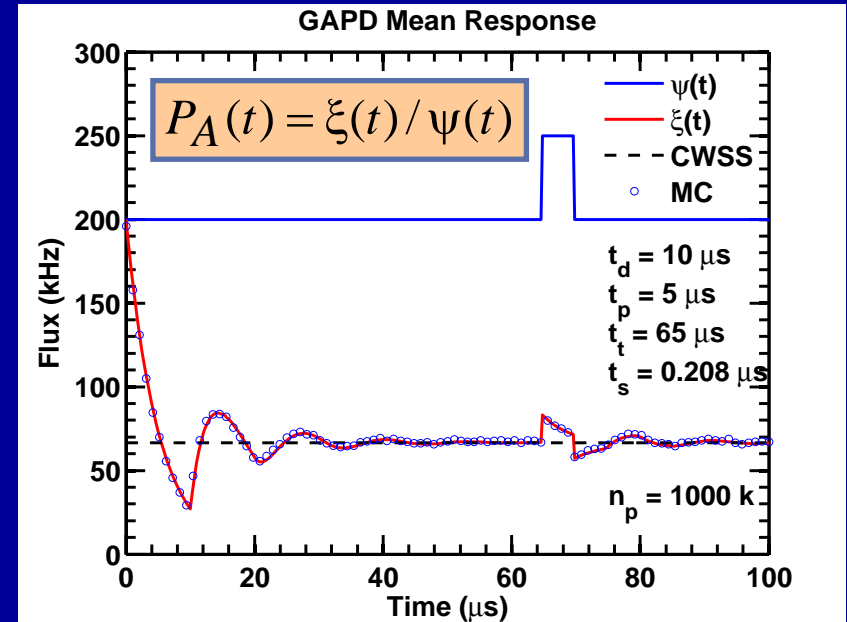
- ❖ Linear front end, PDE or QE
- ❖ High gain region => saturation
- ❖ Detects only 1 event per dead time
 - ⇒ Quench (reset) and trap (after pulses) time drive dead time

◆ Signal Model PE Model

$$\begin{aligned}\psi(t) &= \psi_s(t) + \psi_c(t) + \psi_{bg} + \psi_d \\ &= \eta_q (P_s(t) + P_c(t) + P_{bg} + P_d) / h\nu\end{aligned}$$

◆ Assumptions

- ❖ Poisson noise statistics
- ❖ Negative Binomial signal statistics
- ❖ Short pulse
 - ⇒ $T_p < T_d$, so at most one detection per matched filter width
- ❖ Multi-pulse Accumulation
 - ⇒ Matched Filter Receiver
 - * # noise PEs = $m_n < 1$
 - * # signal PEs = m_s



Probability that the GAPD is Active, $P_A(t)$, is the ratio of the mean-output flux to the input flux



Poisson Properties

◆ Memoryless

- ❖ probability of an event in a given time interval is independent of previous events

◆ Distribution

$$\Pr(k; t_1, t_2) = \frac{e^{-m(t_1, t_2)} m(t_1, t_2)^k}{k!}$$

$$m(t_1, t_2) = \int_{t_1}^{t_2} \psi(t) dt$$

◆ Probability of no event

$$\Pr(k = 0; t_1, t_2) = e^{-m(t_1, t_2)}$$

◆ Probability of one or more events

$$\Pr(k > 0; t_1, t_2) = 1 - e^{-m(t_1, t_2)}$$

◆ Event time relative to a fixed time, t_0 , for constant flux, ψ_0

- ❖ 1st event time is described by an exponential PDF

$$f_1(t - t_0) = \psi_0 \exp[-\psi_0(t - t_0)]$$

- ❖ n th event time is described by a Gamma PDF

$$f_n(t - t_0) = \psi_0^n (t - t_0)^{n-1} \exp[-\psi_0(t - t_0)] / (n-1)!$$

Mean Response Theory: General Flux

- ◆ Total mean response is sum of n^{th} event mean responses

$$\xi(t) = \sum_{n=1}^{N_d(t)} \xi_n(t)$$

- ◆ First event mean response

- ❖ Exponential PDF
- ❖ Exponential Rate

$$\Pr(k > 0; 0, t) = 1 - e^{-m(0,t)}$$

$$\xi_1(t) = \psi(t)e^{-m(0,t)}$$

- ◆ 2nd event mean response, given first event

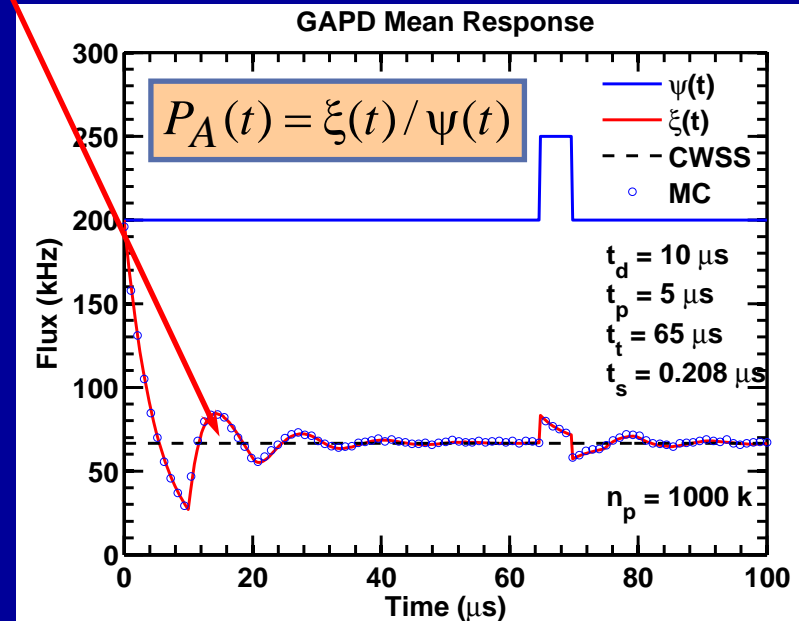
$$\xi_2(t|t_1) = \psi(t)e^{-m(t_1+t_d,t)}$$

- ◆ 2nd event mean response

$$\xi_2(t) = \int_0^{t-t_d} \xi_1(t_1)\xi_2(t|t_1)dt_1 = \psi(t) \int_0^{t-t_d} \xi_1(\tau)e^{-m(\tau+t_d,t)}d\tau$$

- ◆ n^{th} event mean response

$$\xi_n(t) = \psi(t) \int_{t_d(n-2)}^{t-t_d} \xi_{n-1}(\tau)e^{-m(\tau+t_d,t)}d\tau$$



Mean Response Theory: Step Function Input Flux

- ◆ The general theory reduces to a sum of delayed Gamma PDFs

$$\xi_1(t) = \psi_o e^{-\psi_o t}$$

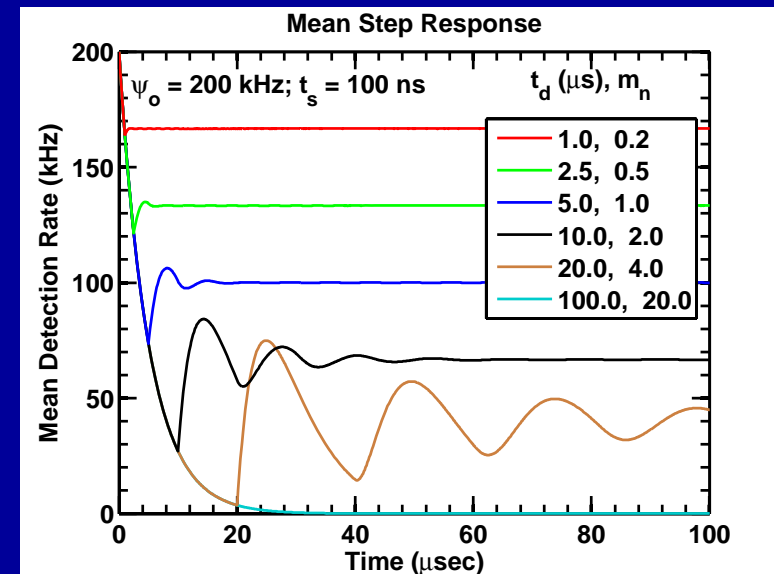
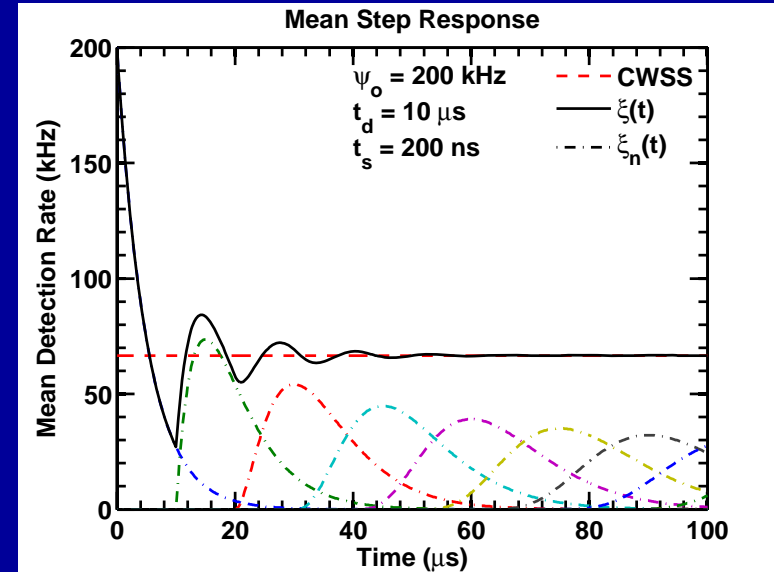
$$\xi_n(t) = \psi_o^n [t - (n-1)t_d]^{n-1} e^{-\psi_o [t - (n-1)t_d]} / (n-1)!$$

- ❖ Mean = $(n-1)t_d + n/\psi_o$ and diversity parameter, n
- ◆ CWSS ($m_d = \psi_o t_d$)
 - ❖ Mean time between detections:

$$E[\Delta T] = (1/\psi_o + t_d) = (1 + m_d) / \psi_o$$
 - ❖ Mean detection rate:

$$\xi_{CWSS} = 1 / (1/\psi_o + t_d) = \psi_o / (1 + m_d)$$
 - ❖ Probability Active:

$$P_{ACWSS} = 1 / (1 + m_d)$$
- ◆ Infinite sum (over diversity) of Gamma PDFs is constant
 - ❖ ψ_o for zero dead time
 - ❖ $\psi_o / (1 + m_d)$ for non-zero dead time
 - ❖ CWSS time scales as $\psi_o t_d^2$



Signal Photon Detection Efficiency (SPDE)

Signal + noise PE Probability of detection

$$PD_m = P_A(t_t)(1 - e^{-m_s - m_n})$$

Signal PE Probability of detection

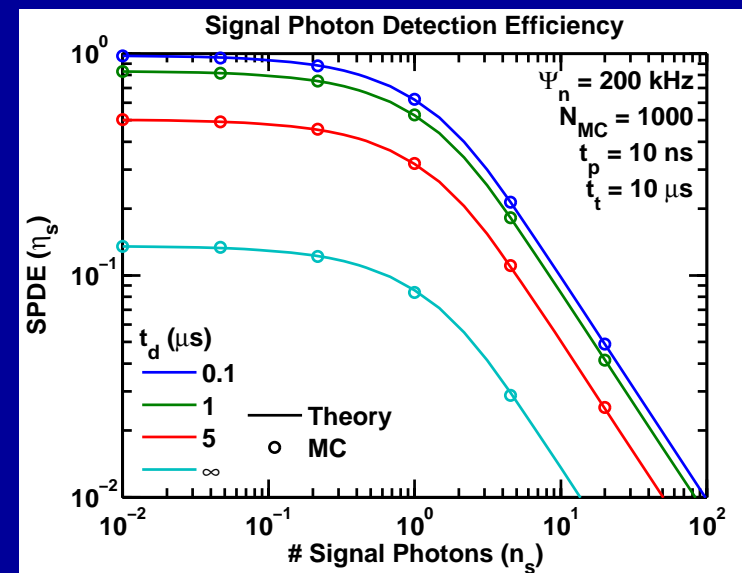
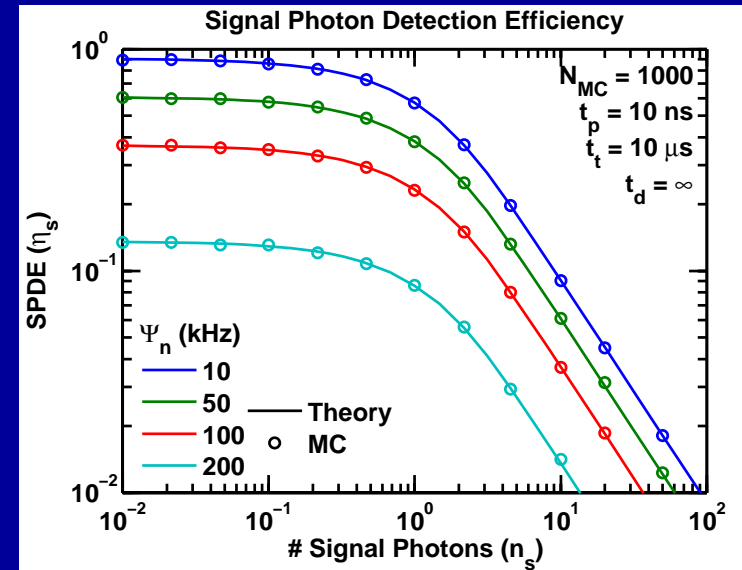
- ❖ If $m_n \ll 1$ then signal and noise detection probabilities are equal and governed by the signal strength

$$PD_s \approx P_A(t_t)(1 - e^{-m_s})$$

Efficiency

$$\eta_s = PD_s / n_s$$

$$\eta_s = P_A(t_t)(1 - e^{-m_s}) / n_s$$



Signal-to-Noise Ratio

◆ SNR Definition

$$SNR = \frac{E[s]^2}{\text{var}[s+n]}$$

◆ Linear Device (Np pulses)

$$SNR = \eta_s N_p \frac{n_s^2}{n_s + n_n}$$

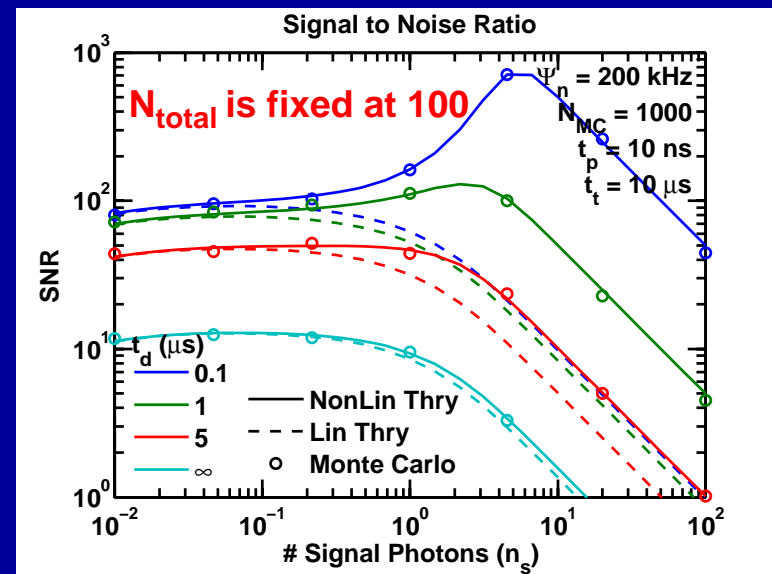
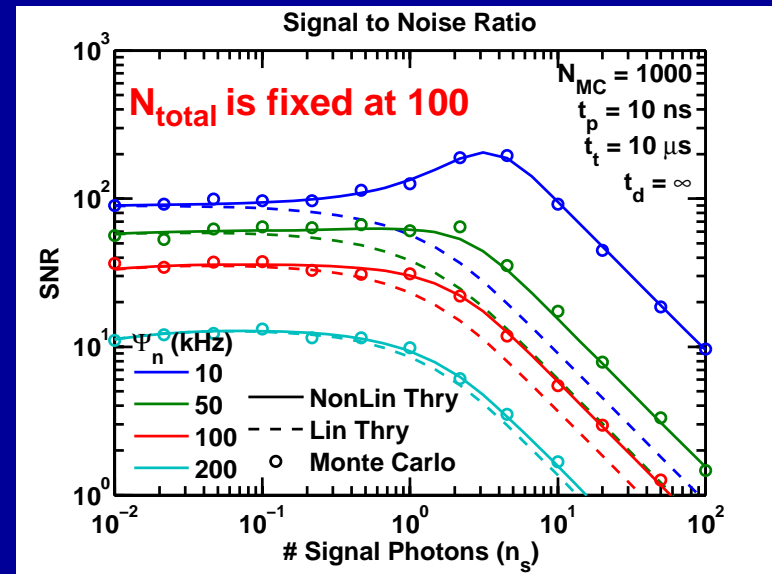
◆ Non-linear model

- ❖ Assume Binomial Distribution
- ❖ Np trials
- ❖ Mean = $N_p * p_s$; $p_s = P_A(1 - \exp[-m_s])$
- ❖ Var = $N_p * p_{sn}(1 - p_{sn})$; $p_{sn} = P_A(1 - \exp[-m_s - m_n])$

$$SNR \approx \frac{N_p P_A (1 - e^{-m_s})^2}{(1 - e^{-m_s - m_n}) (1 - P_A (1 - e^{-m_s - m_n}))}$$

◆ Deviations from linear model

- ❖ Device is behaving non-linearly
- ❖ Detections are not Poissonian
- ❖ Little reflectance information is available



GAPD Detection Statistics: Overview



◆ Assumption

❖ Newman-Pearson detection problem

⇒ Threshold crossing problem with Pd and Pfa per matched-filter bin

❖ Multi-pulse accumulation

⇒ Constitutes a Bernoulli trial (1 of 2 outcomes per pulse, per range-gate)

⇒ Binomial distribution analysis applies for noise and signal+noise statistics

◆ Binomial distribution describes the PDF of the number of successes that occur in n Bernoulli trials, where the probability of success in a trial is, p .

$$p(k; n) = \binom{n}{k} p^k (1-p)^{n-k}$$

$$\binom{n}{k} = \frac{n!}{k!(n-k)!}$$

◆ Detection statistics (Pd or Pfa, denoted Pr below) are equal to the tail of a binomial CDF, which has support up to N_p pulses accumulated

$$\Pr = \sum_{k=k_{th}}^{N_p} p(k; N_p) = 1 - \sum_{k=0}^{k_{th}-1} p(k; N_p)$$

❖ Where k^{th} is a threshold needed to achieve a Pfa specification

GAPD Detection Statistics: Pd and PFA

- ◆ For GAPDs, an appropriate expression for the single trial (1 pulse) probability “p” for noise (Pfa) or signal + noise (Pd) is needed to substitute into the binomial distribution
- ◆ For the noise, Poisson statistics apply. The prob of at least 1 detected PE is

$$p_n = \Pr[k_n > 0] = P_A \left(1 - e^{-m_n} \right)$$

- ❖ m_n is the number of noise PEs in a matched filter integration time
- ❖ P_A is the probability that the GAPD is armed
- ◆ For negative binomial speckled-signal plus Poisson noise, Goodman’s expression with arbitrary diversity, M , (spatial, spectral and polarization) applies: Evaluation of this expression for zero PE’s leads to the following simple expression

$$p_{sn} = \Pr[k_{sn} > 0] = P_A \left(1 - e^{-m_n} \left(1 + m_s / M \right)^{-M} \right)$$

$$\lim_{M \rightarrow \infty} \left(1 + m_s / M \right)^{-M} = e^{-m_s}$$

- ❖ m_s is the number of signal PEs in a matched filter integration time
- ❖ Glint targets are handled by taking the infinite M limit

GAPD Detection Statistics: Detail Math BACKUP

Noise and Signal + Noise Statistics



◆ Noise : Poisson PDF and Prob $m > 0$

❖ m_n is mean number of noise PEs in a matched filter bin

$$p_n(k) = m_n^k \exp(-m_n) / k!$$

$$\Pr[k_n > 0] = 1 - p_n(0) = 1 - \exp(-m_n)$$

◆ Signal + Noise PDF (NegBinomial + Poisson), See Goodman

❖ M is the diversity, m_s is mean number of signal PEs

$$p_{sn}(k) = \left(\frac{M}{M + m_s} \right)^M \frac{e^{-m_n}}{(M-1)!} \sum_{j=0}^k \frac{(k+M-j-1)!}{j!(k-j)!} (m_n)^j \left(\frac{m_s}{M+m_s} \right)^{k-j}$$

$$\Pr[k_{sn} > 0] = 1 - p_{sn}(k=0) = 1 - e^{-m_n} (1 + m_s / M)^{-M}$$

❖ High diversity limit (Boise Einstein)

$$\lim_{M \rightarrow \infty} (1 + m_s / M)^{-M} = e^{-m_s}$$



$$p_{sn}(k > 0) = 1 - e^{-m_s - m_n}$$

◆ These probabilities then need to be multiplied by P_A , the probability that the GAPD is active.

Performance Comparison: Normalized to Matched Filter Bin Width

◆ Example Case

❖ 10ns/bin

❖ $\psi_0 = 100$ kHz $\rightarrow m_n = 0.001$

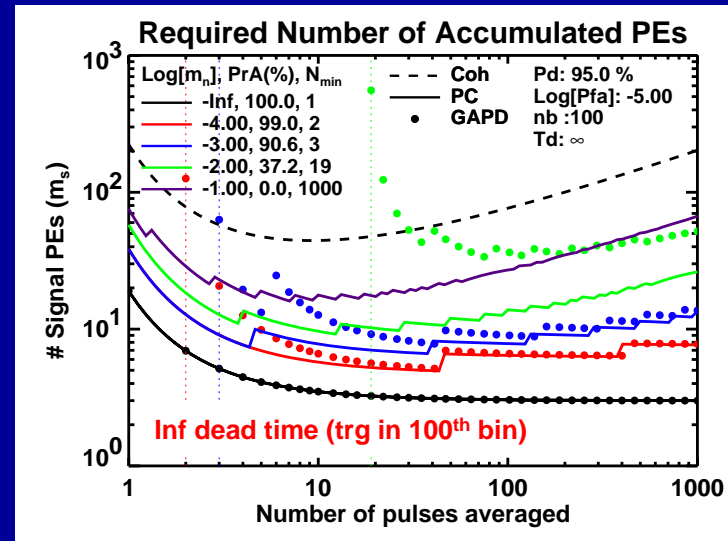
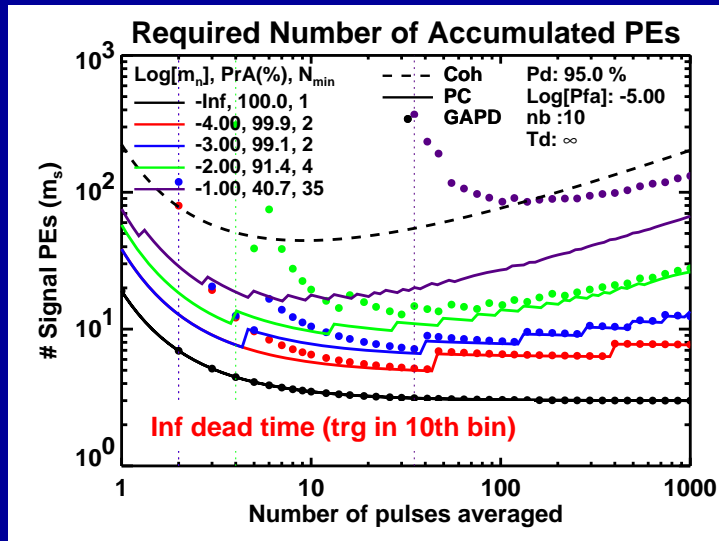
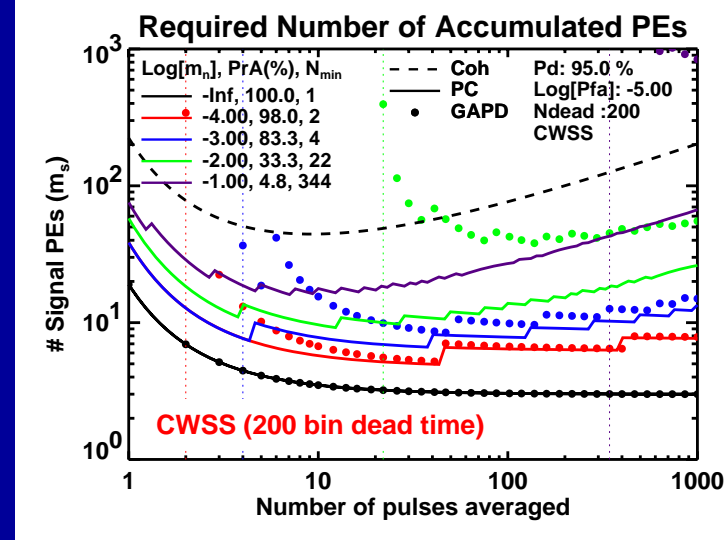
◆ CWSS example

❖ 2 us dead time \rightarrow 200 bins

◆ Long dead time example

❖ 10th bin \rightarrow 15 m gate-to-target range

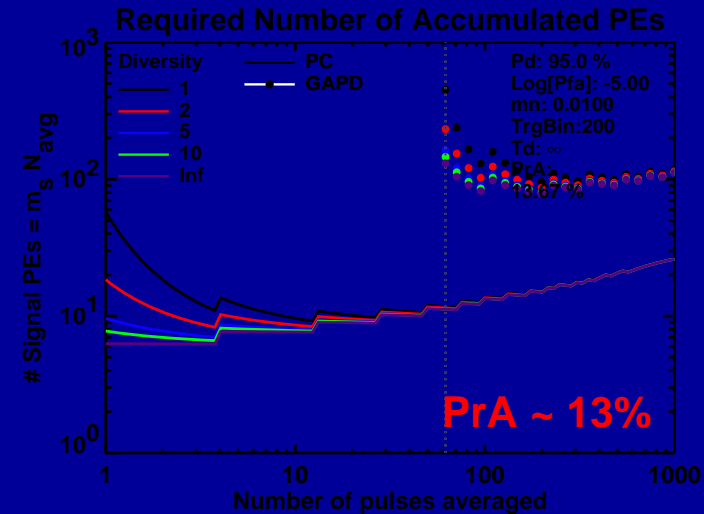
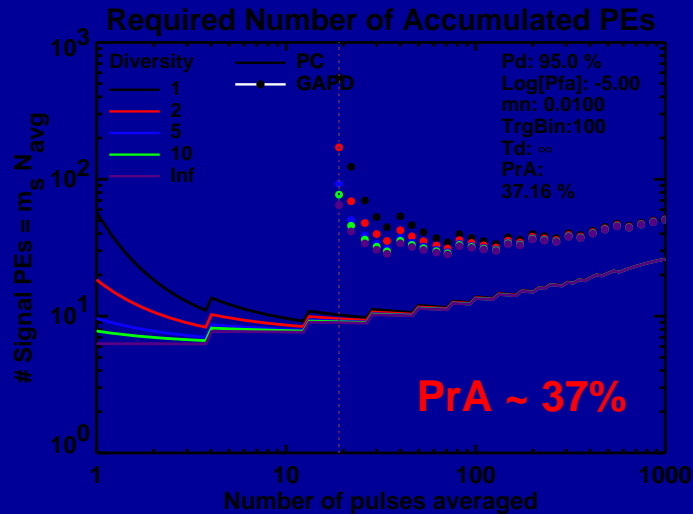
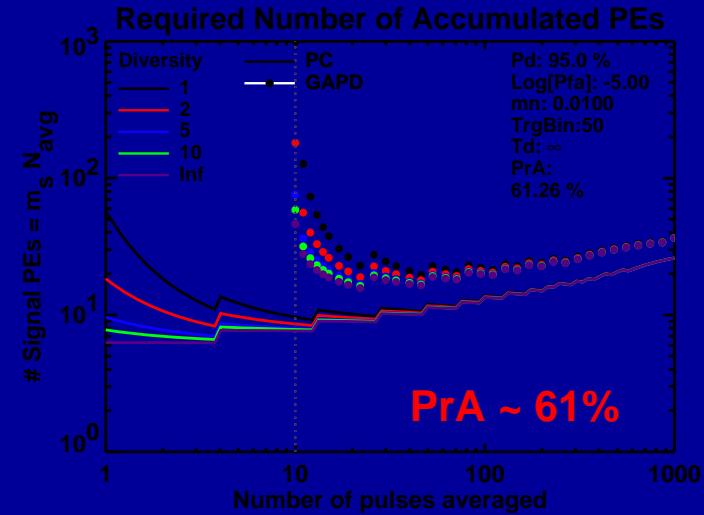
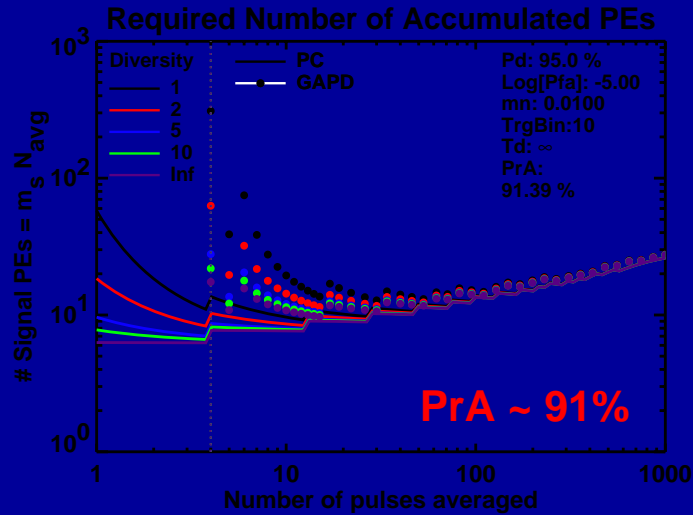
❖ 100th bin \rightarrow 150 m gate-to-target range



Diversity Trades:

◆ Key parameters

- ◆ Pd = 95%
- ◆ Pfa = 10⁻⁵
- ◆ Varying PrA



Advantage of Micro-pixels

- ◆ One approach to increasing SPDE is to combine signals from multiple detectors to form a single pixel

- ◆ **Net SPDE (General illumination)**

- ❖ Weighted sum of micro-pixel efficiencies
- ❖ Weights are fraction of total # signal photons

$$\eta_s = \sum_{i=1}^{N_d} w_i \eta_{si} \quad \text{where } w_i = n_{si} / n_s$$

- ◆ **Net SPDE (Uniform illumination)**

- ❖ Weights are equal (1/N_d)
- ❖ Net SPDE is average of micro-pixel SPDEs, which can be much higher than would be achieved with one large pixel

$$\eta_s = (1/N_d) \sum_{i=1}^{N_d} \eta_{si}$$

Summary/Conclusions



- ◆ **Derived Expression for the Mean GAPD Output Flux**
 - ❖ Response (flux) is sum of individual event response functions or PDFs
 - ⇒ Recursive theory for the n th event PDFs was developed
 - ⇒ For the special case of a constant input flux these PDFs are delayed Gamma PDFs with mean $(n-1)t_d + n/\psi_0$ and diversity, n
 - ❖ CWSS response was derived

- ◆ **P_A = Probability that GAPD is active**
 - ❖ Shown to be ratio of output to input flux
 - ❖ This probability depends on number of noise PEs prior to target within a dead time
 - ❖ Shorter dead time devices have increased P_A

- ◆ **Signal Photon Detection Efficiency derived and shown to agree with MC measurements**
 - ❖ Good efficiency requires
 - ⇒ High P_A
 - ⇒ $m_s < 1/(1+\alpha)$; α is the clutter to signal ratio

- ◆ **SNR theory derived and shown to agree with MC measurements**
 - ❖ Best SNR not necessarily at best efficiency
 - ❖ Linear SNR theory valid when $m_s < 1/(1+\alpha)$
 - ❖ Non-linear device SNR theory based on binomial statistics agrees with measurements
 - ❖ When SNR > Linear Theory limit the device is saturated and target reflectance information is lost!

- ◆ **Detection Statistics Derived and compared to Coherent Detection and Linear Photon Counting DD**
 - ❖ Linear PC is better than GAPD due to PrA