The impact of digitization on the entropy generation rates of physical sources of randomness

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Why Physical RNG?

Any one who considers arithmetical methods of producing random digits is, of course, in a state of sin.

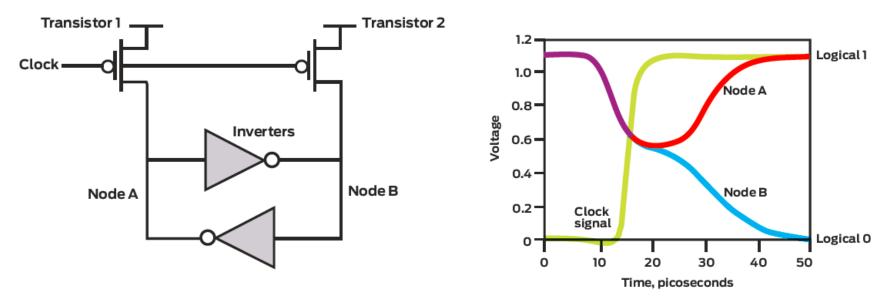
--John von Neumann



Physical RNG

- Algorithms can only produce pseudo-random numbers
- For true random numbers, we turn to physical systems
- Can be FASTER because not limited by CPU clock
- Need to be post-processed to remove bias, etc.
- Important differences between pseudo-RNG and physical RNG should be reflected in evaluation metrics

Electronic Physical RNG Today (Intel Ivy Bridge Processors)

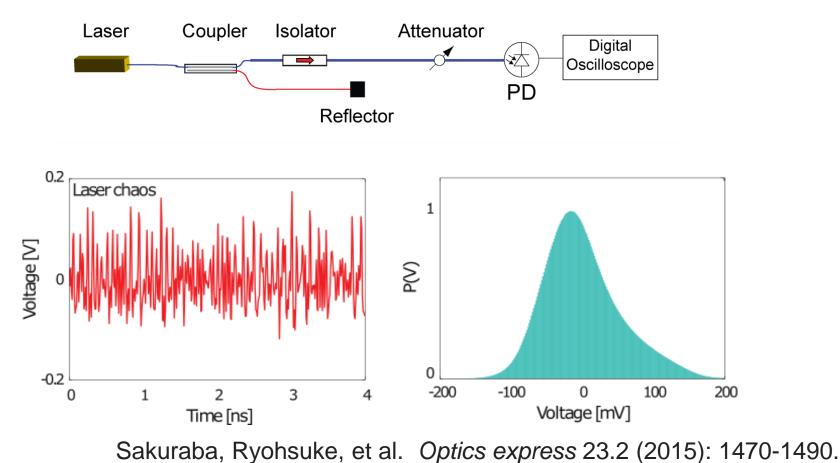


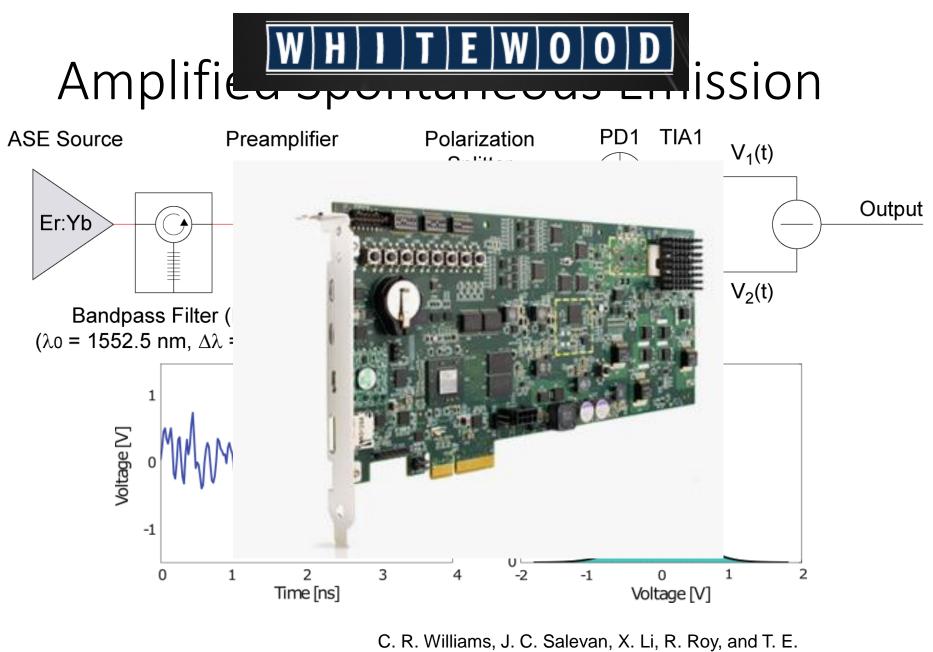
- 3 Gb/s raw RNG rate
- Raw bits are not directly used (nor accessible)
- Continuously re-seeds a pseudo-random generator
- Instruction: RDRND

Chaotic Semiconductor Laser

Fluctuations are fast and chaotic

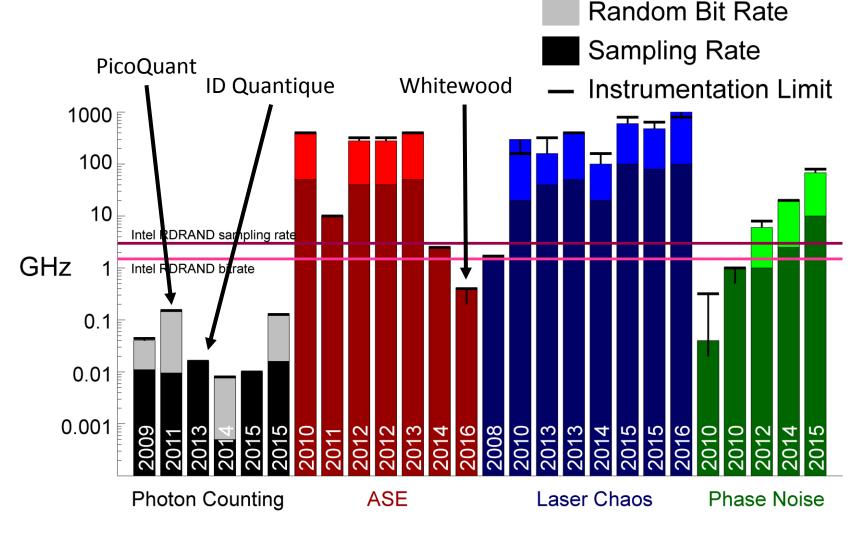
(sensitive dependence on initial conditions)





Murphy, Optics Express 18, 23584–23597 (2010).

Comparison of Optical RNG Methods – Recent Research

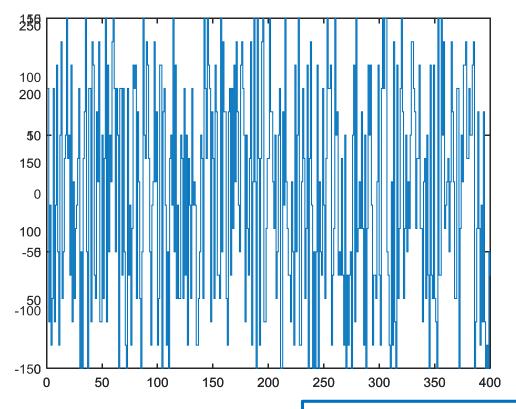


NIST SP 800-22rev1a

- Easy to implement
- Publicly accessible standard
- Even non-cryptographically secure Pseudo-RNG methods (e.g., Mersenne Twister) will pass all tests
- Only works on binary data (1s and 0s), not analog data or waveforms
- Most physical RNG methods require postprocessing to pass tests

Post-Processing of Digitized Waveforms

• Least Significant Bit Extraction:



What is the source of entropy? (waveform, digitizer, thermal noise?)

Entropy estimates

- Try to quantify the number of random bits allowed to be harvested from a physical system
- Works on raw data, not post-processed data
- Can help reveal *where* the entropy is coming from
- Can be slower, require more data than NIST SP 800-22rev1a

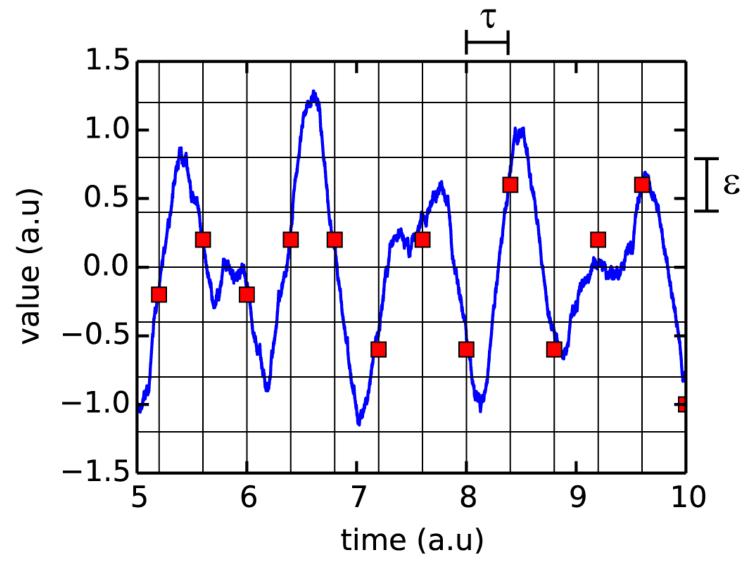
Dynamical systems approach to entropy generation

• Kolmogorov-Sinai (or metric) entropy

$$H = -\frac{1}{d\tau} \sum p(i_1, \dots, i_d) \log_2 p(i_1, \dots, i_d)$$

- Analog of Shannon entropy for dynamical system
- Allows for direct comparison of dynamical processes, stochastic processes, and mixed processes

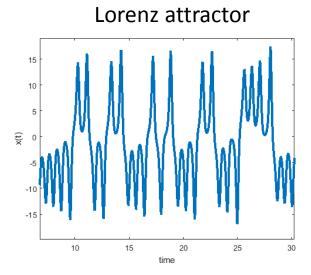
Discretization of analog signals

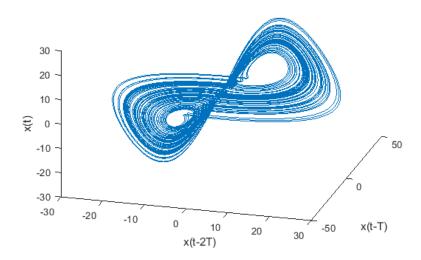


Time-delay embedding

 Reconstruct phase-space of dynamical system from measurement of one variable

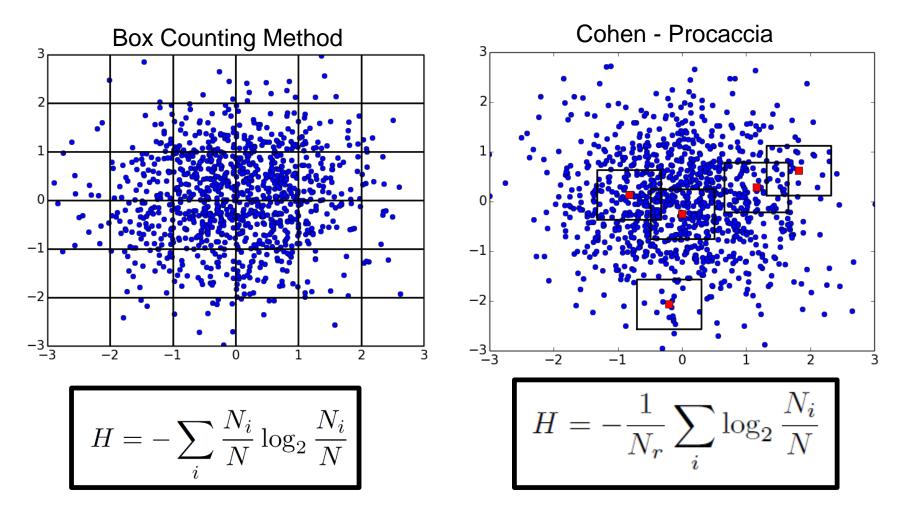
$$\vec{\mathbf{x}}(t) = (x(t), x(t-T), \dots, x(t-(d-1)T))$$





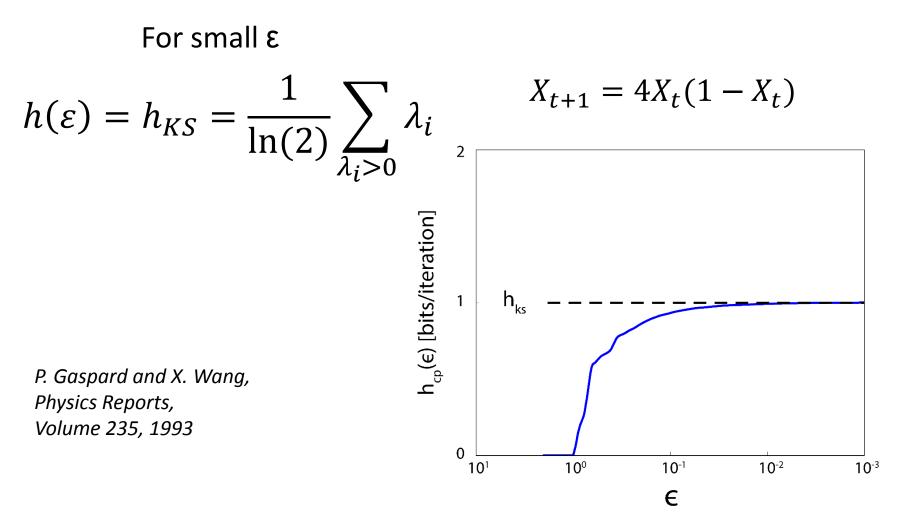
Takens, Floris. *Detecting strange attractors in turbulence*. Springer Berlin Heidelberg, 1981.

Numerically Estimating Entropy



Cohen and Procaccia, "Computing the Kolmogorov entropy from time signals of dissipative and conservative dynamical systems", Phys. Rev. A **31**, 1872 (1985)

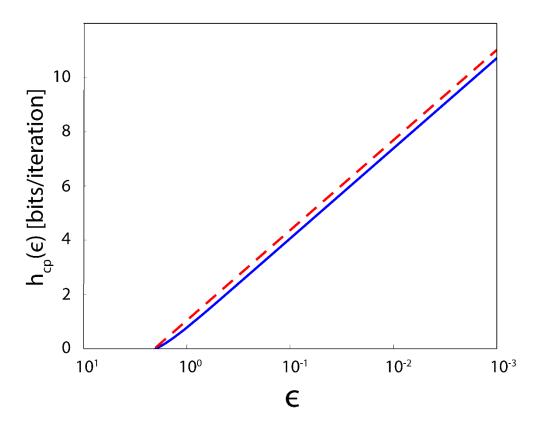
Entropy of chaotic systems



(ε-τ) entropy of noise

 $h(\varepsilon) \sim -\log_2(\varepsilon)$

Gaussian random variable



P. Gaspard and X. Wang, Physics Reports, Volume 235, 1993

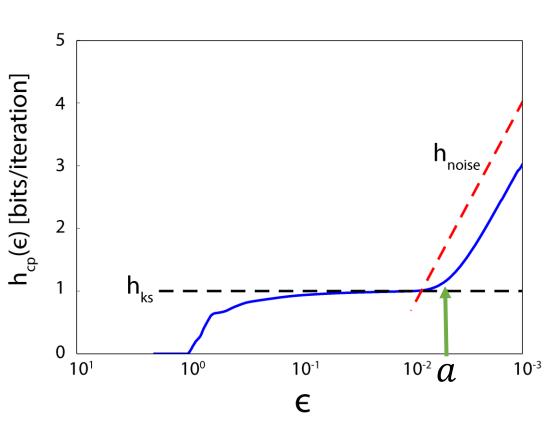
Noisy chaotic systems

$$Z_{t+1} = X_t + aR_t$$

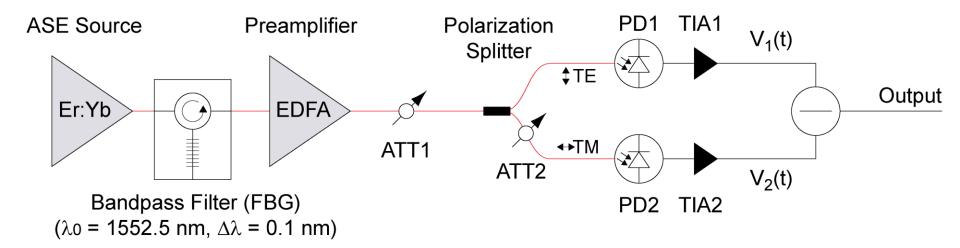
$$X_{t+1} = 4X_t(1 - X_t)$$

R is random Gaussian variable

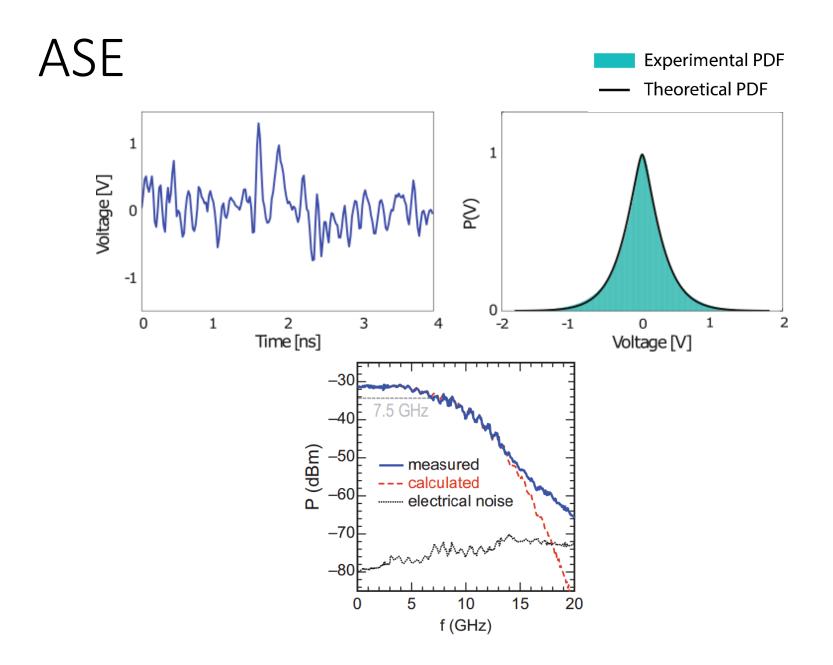
P. Gaspard and X. Wang, Physics Reports, Volume 235, 1993



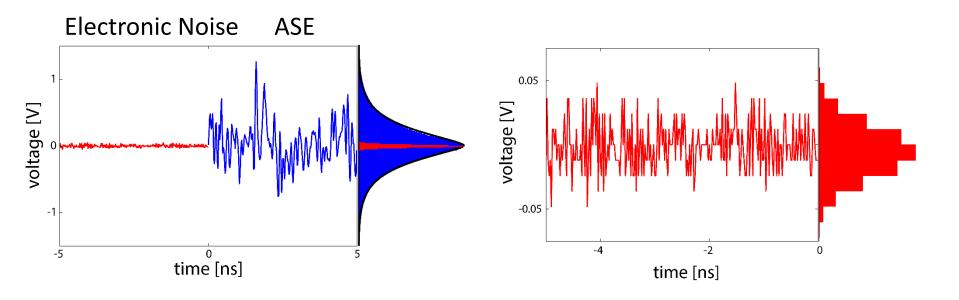
Case study: Amplified Spontaneous Emission (ASE)



C. R. Williams, J. C. Salevan, X. Li, R. Roy, and T. E. Murphy, Optics Express **18**, 23584–23597 (2010).

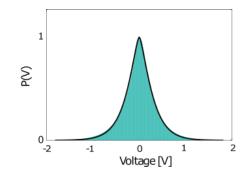


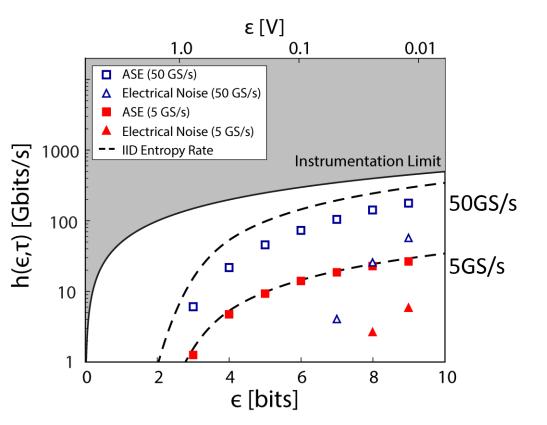
Signal and Noise



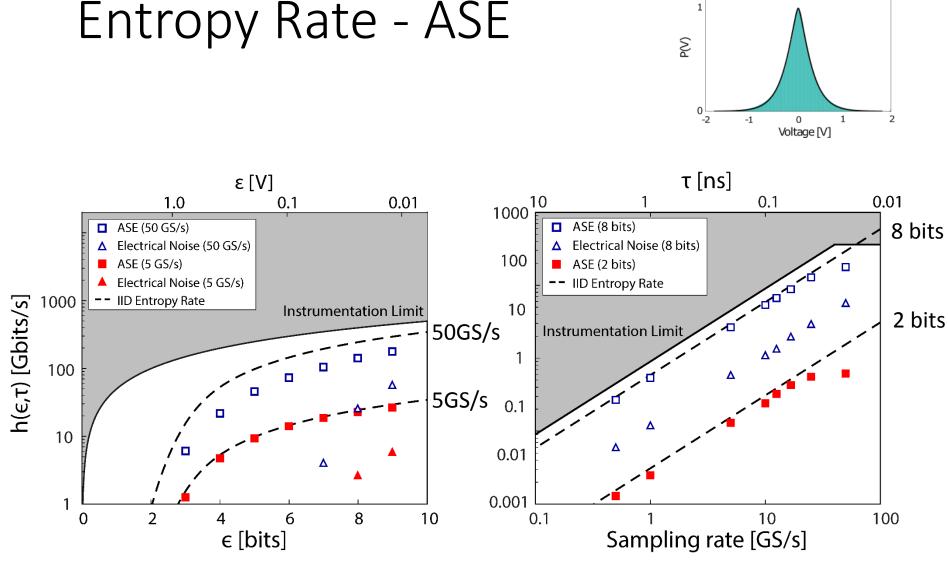
 Least-significant bits contribute considerable entropy (not optical!)

Entropy Rate - ASE





• Entropy rolls off with sample rate

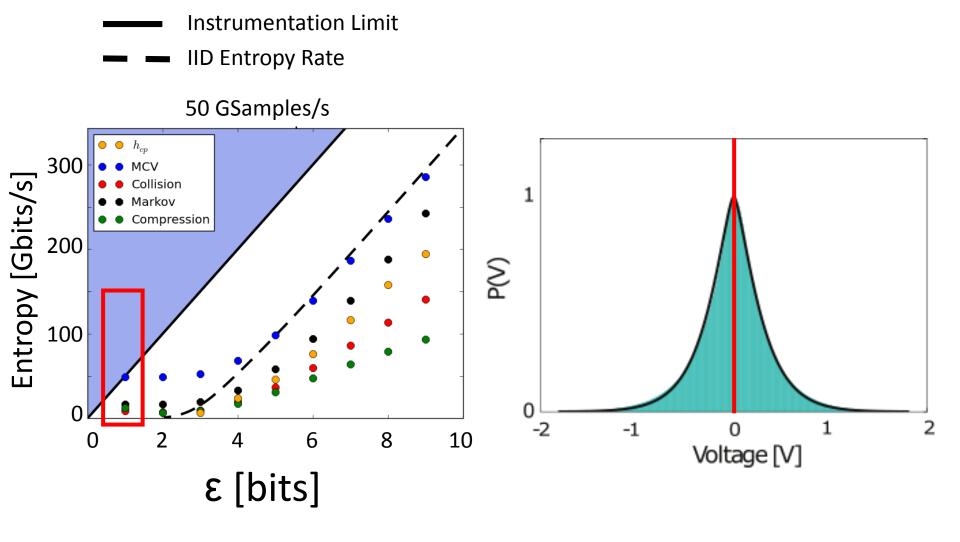


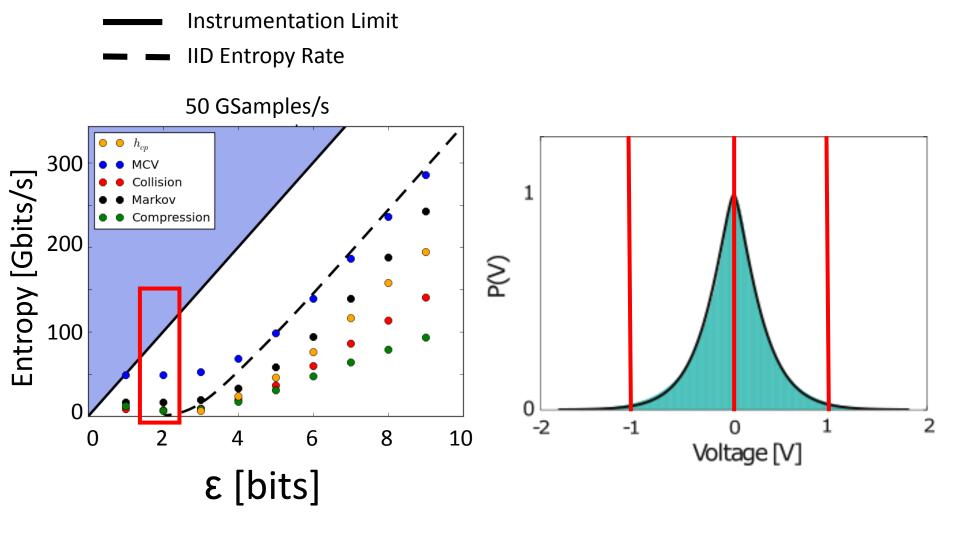
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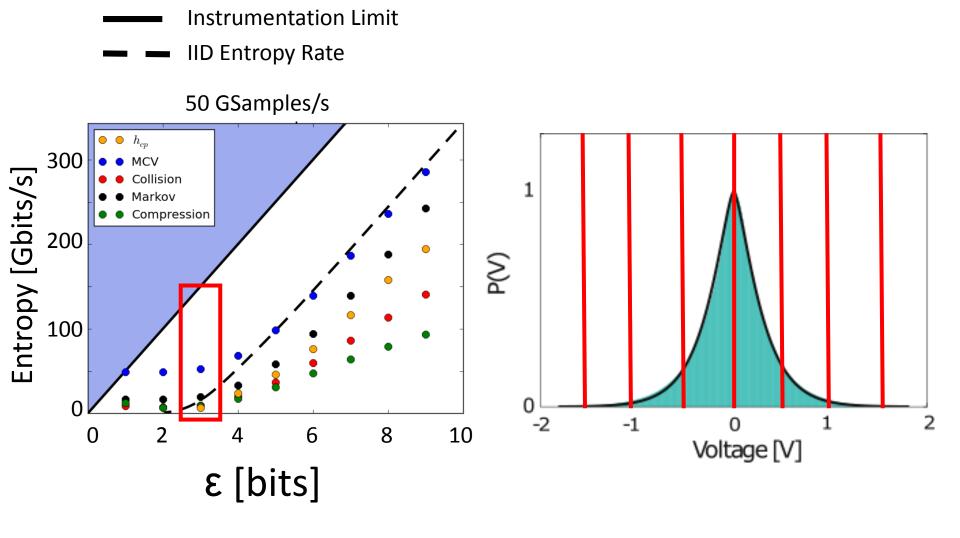
Entropy rolls off with sample rate

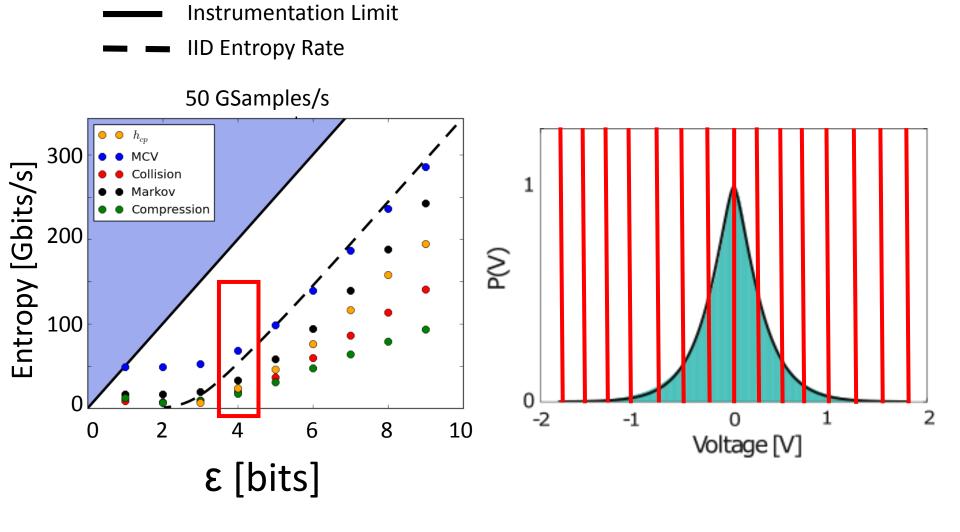
NIST SP 800-90B Entropy Estimates

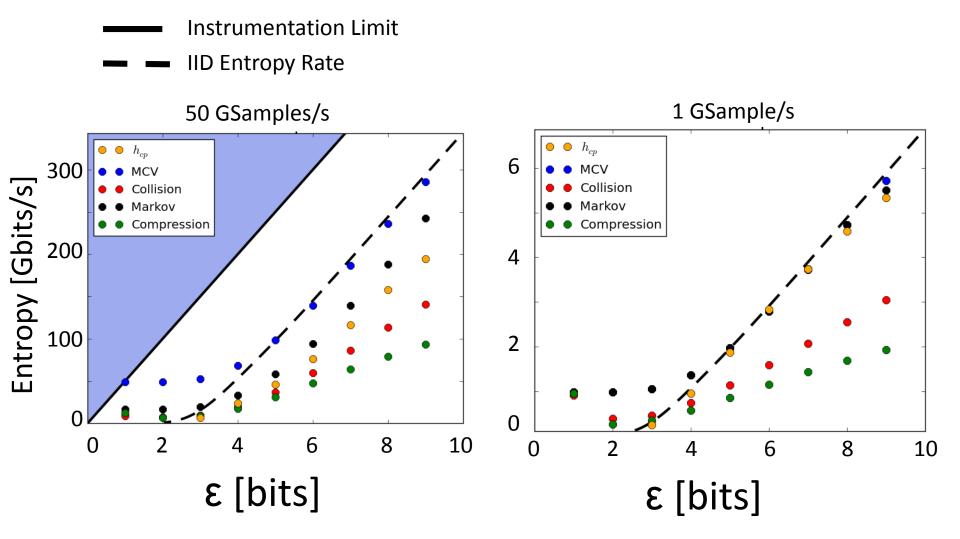
- Most Common Value Estimate
- Collision Estimate—based on mean time until first repeated value
- Markov Estimate—measures dependencies between consecutive values
- Compression Estimate—estimates how much the dataset can be compressed
- Other more complicated tests...



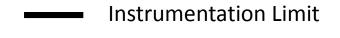




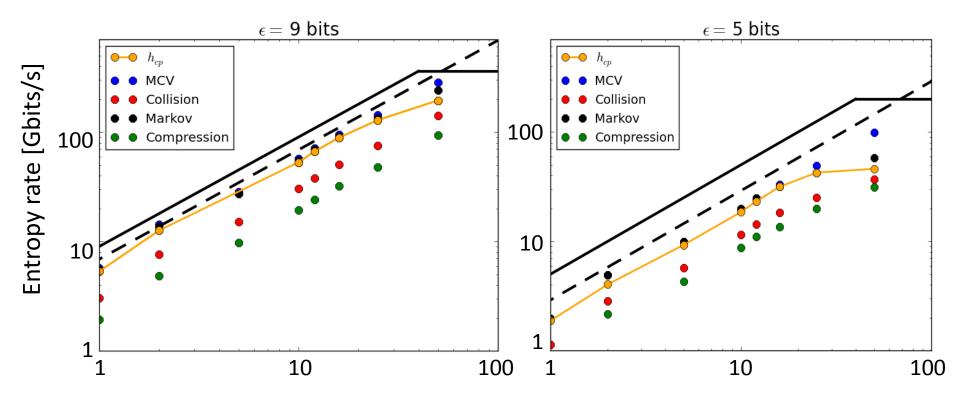




Capturing Temporal Correlations



🗕 🗕 IID Entropy Rate

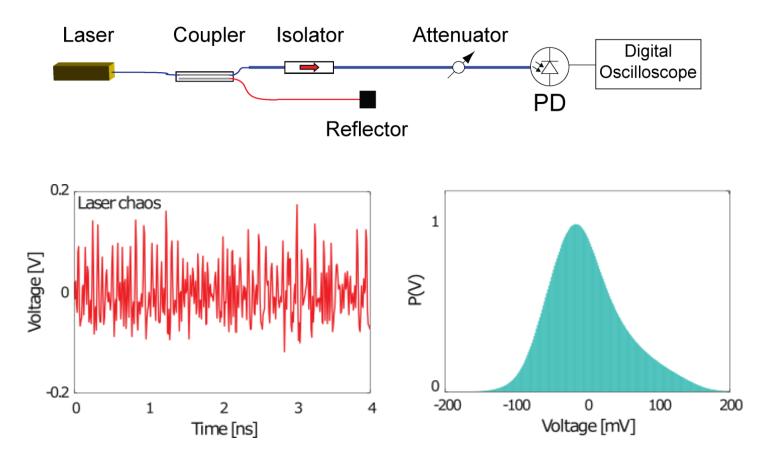


Sampling Frequency [Gsamples/s]

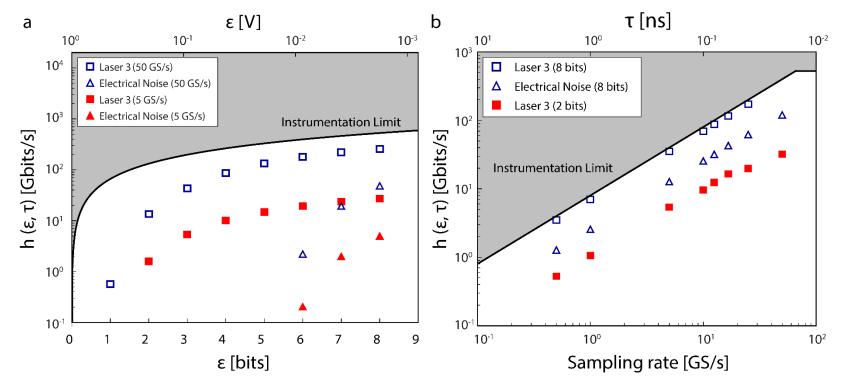
Sampling Frequency [Gsamples/s]

Chaotic Laser

• Fluctuations are fast and chaotic (sensitive dependence on initial conditions)



Entropy rate—laser chaos



 Significant portion of entropy comes from background noise (especially at high resolution)

Conclusions:

- Important to look at entropy as a function of measurement resolution and sampling frequency
- Different physical processes can generate entropy, even within the same experiment
- Measurement determines which physical entropy generation processes you observe
- Entropy estimates should consider analog data, not post-processed bit stream

To learn more about:

Entropy generation in noisy chaotic systems:

Aaron M. Hagerstrom, Thomas E. Murphy, and Rajarshi Roy. "Harvesting entropy and quantifying the transition from noise to chaos in a photon-counting feedback loop." *Proceedings of the National Academy of Sciences* 112.30 (2015): 9258-9263.

Amplified spontaneous emission:

Williams, Caitlin RS, et al. "Fast physical random number generator using amplified spontaneous emission." *Optics express* 18.23 (2010): 23584-23597.

Li, Xiaowen, et al. "Scalable parallel physical random number generator based on a superluminescent LED." *Optics letters* 36.6 (2011): 1020-1022.