

# What are the Differences between White, Blue and Pink Noise?

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# Outline

1. Motivation
2. Spectral Density
3. White, Blue and Pink Noise
4. Conclusion



## Sound That We Are Familiar With...



What is that?



## Process of Audio Snippet

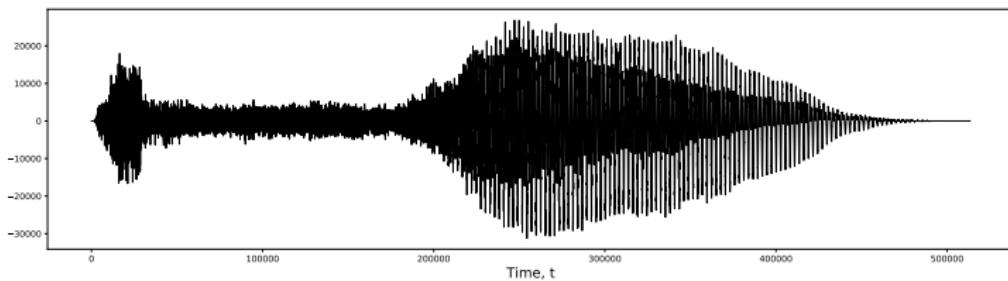


Figure 1: HBO Opening Audio Signal



- Time domain representation of HBO opening audio process  
 $\{X_t\}_{t=0}^{\infty}$
- Periodic, stationarity

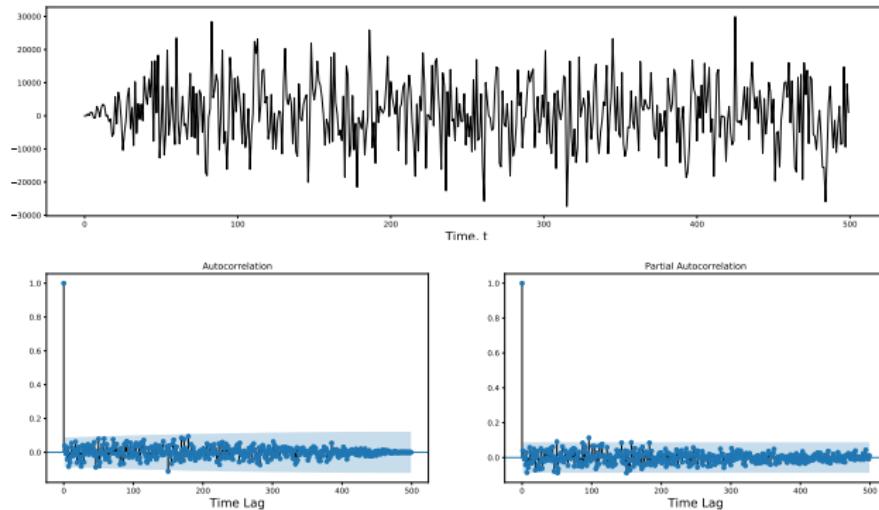


# What do White, Blue and Pink Noise sound like?

*White Noise, Blue Noise And Pink Noise*



## Time Domain of White Noise Sample

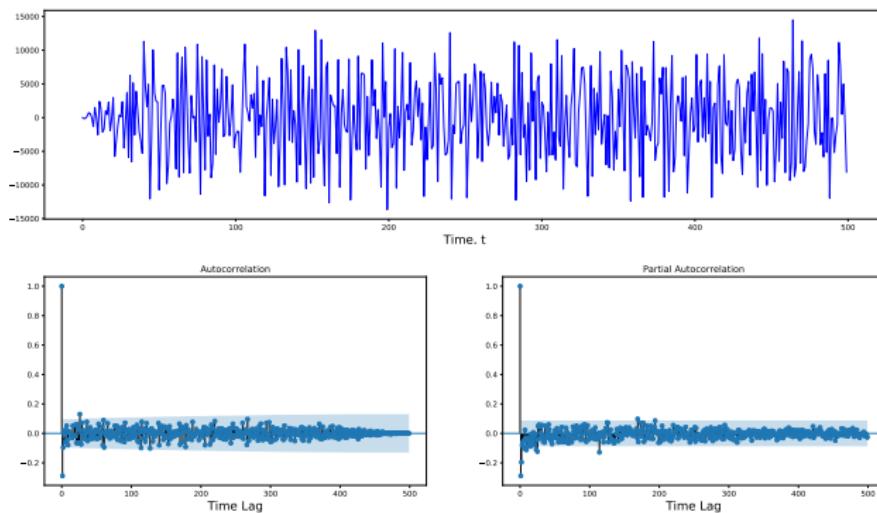


- $\{X_t\}_{t=0}^{500} \sim N(0, \sigma^2)$ , stationary process,  $AR(0)$  process

White, Blue and Pink Noise – Differences



## Time Domain of Blue Noise



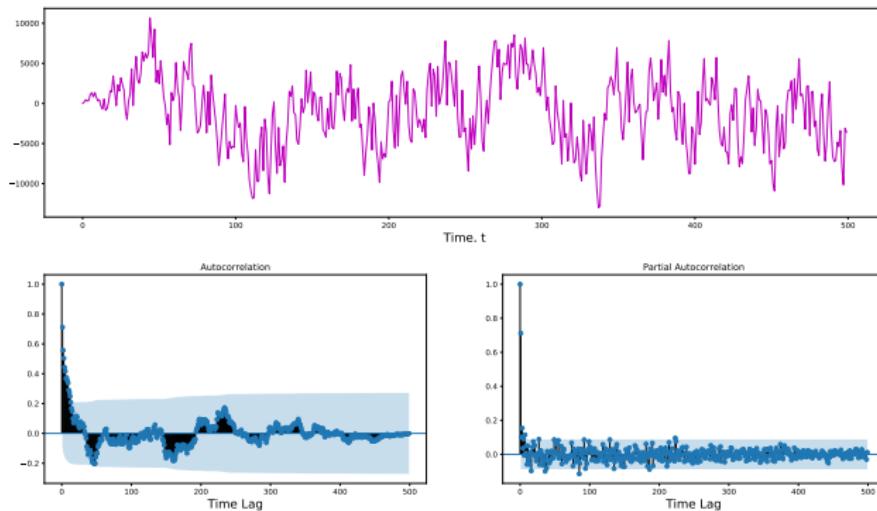
 WBPNoise

- Non-normal distributed, stationary process,  $AR(p)$  process ( $p > 0$ )

White, Blue and Pink Noise – Differences



## Time Domain of Pink Noise



 WBPNoise

- Non-normal distributed, stationary process,  $AR(p)$  process ( $p > 0$ ), long memory

White, Blue and Pink Noise – Differences



## Decomposition of Time Series

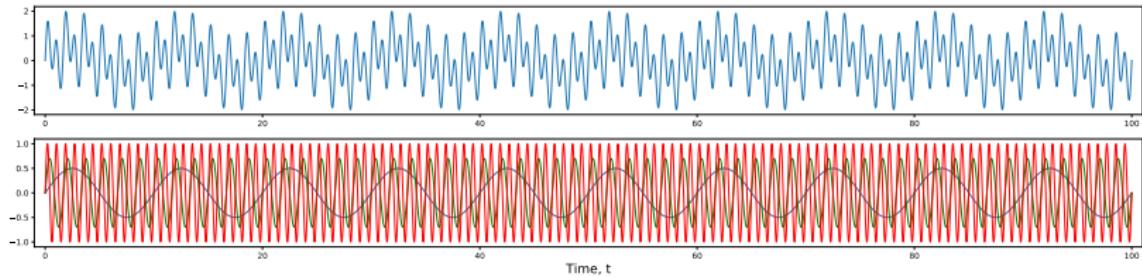


Figure 2: Time series  $\{X_t\}_{t=0}^{100}$  (upper panel) can be decomposed into sinusoids with frequencies  $\omega = 1.2$ ,  $\omega = 0.6$  and  $\omega = 0.1$  (lower panel)



## Fourier Transform

- Fourier analysis: approximate a function by weighted sums of trigonometric functions
- A stationary process  $x(t)$  can be written as
  - ▶ function of time  $x(t)$  (time domain)
  - ▶ function of frequency  $S(\omega)$  (frequency domain)

*Jean Baptiste Joseph Fourier* on BBI:



## Fourier Transform ctd

- Conversion of time function into frequency domain function and vice versa

$$S(\omega) = \int_{-\infty}^{\infty} x(t)e^{-2\pi\omega it} dt$$

$$x(t) = \int_{-\infty}^{\infty} S(\omega)e^{2\pi\omega it} d\omega$$

- $\omega$  = frequency



## Spectral Density

- Stationary process  $X_t$  with autocovariance  $\gamma(\tau)$ , ( $\tau \in \mathbb{Z}$ )

$$\gamma(\tau) = E[(x_{t+\tau} - \mu)(x_t - \mu)], \sum_{\tau=-\infty}^{\infty} |\gamma(\tau)| < \infty \quad (1)$$

- Spectrum  $S(\omega)$  as Fourier transform of  $\gamma(\tau)$ :

$$S(\omega) = \sum_{\tau=-\infty}^{\infty} \gamma(\tau) e^{-2\pi\omega i\tau} \quad (2)$$

- $S(\omega) \in \mathbb{R}$ ,  $S(\omega) \geq 0$



## Properties

- Symmetric of  $S$

$$S(-\omega) = S(\omega) \quad (3)$$

- Periodicity of  $S$

$$S(\omega + 2\pi t) = S(\omega) \quad (4)$$

for every  $t \in \mathbb{Z}$

- Inverse Fourier transform

$$\gamma(\tau) = \int_{-0.5}^{0.5} e^{2\pi\omega i\tau} S(\omega) d(\omega) \quad (5)$$



## Periodogram

- Periodogram  $P(\omega)$  for realizations  $\{x_t\}$ :

$$P(\omega) = \frac{1}{N} \left| \sum_{t=0}^{N-1} x_t e^{-2\pi\omega it} \right|^2, \quad 0 \leq \omega \leq \frac{1}{2} \quad (6)$$

- Unbiased (but inconsistent) estimator of spectral density



**Spectrum:**  $\sum_{i=1}^3 \gamma_i \sin(2\pi\omega_i t)$

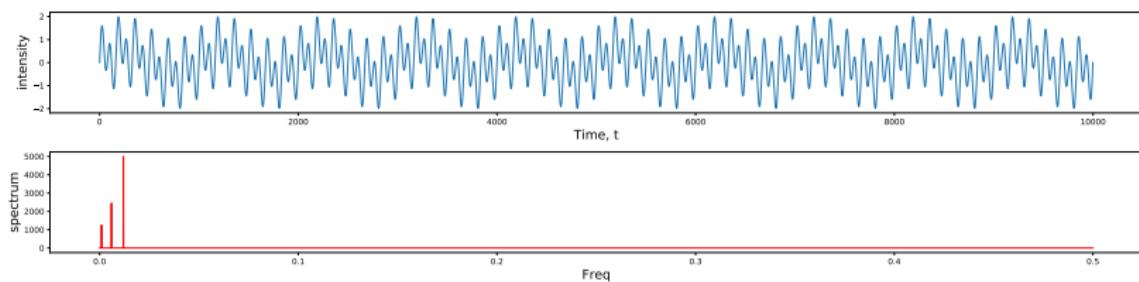


Figure 3: Spectrum of sine waves



- $\gamma_i \in \{1, 0.7, 0.5\}, \omega_i \in \{1.2, 0.6, 0.1\}$  (upper panel)
- Periodogram detects the exact frequency of sine waves (lower panel)



## Spectrum: White Noise

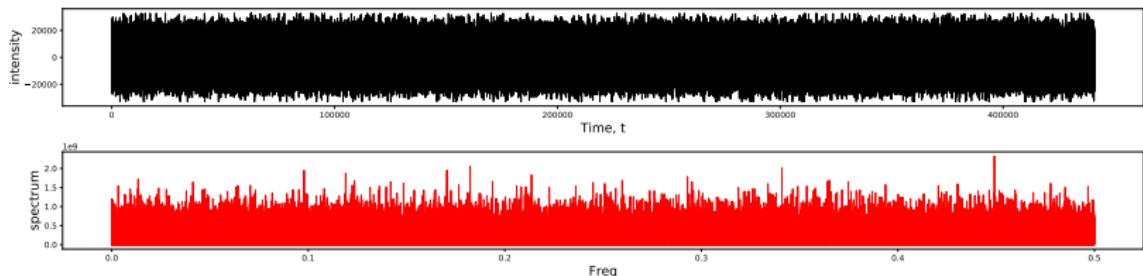


Figure 4: Spectrum of white noise



- Periodogram (lower panel) shows that white noise dominate in whole spectrum
- Theoretically,  $S(\omega) = \sigma^2 / 2\pi$



## Spectrum: Blue Noise

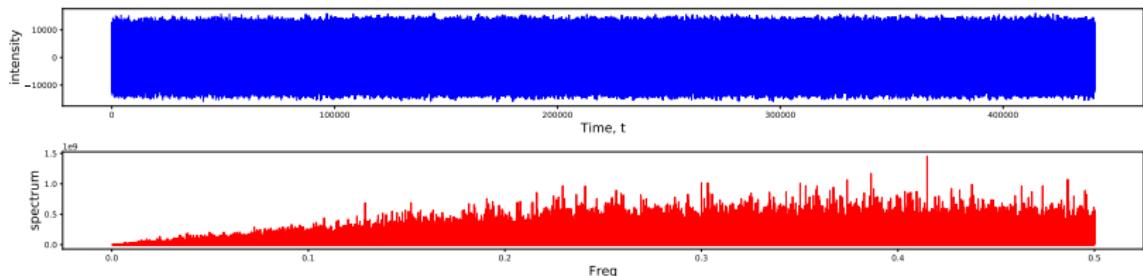


Figure 5: Spectrum of blue noise



- Periodogram (lower panel) shows that blue noise has more energy in higher frequency domain
- Theoretically,  $S(\omega) = \eta\omega$ ,  $\eta \in \mathbb{R}$



## Spectrum: Pink Noise

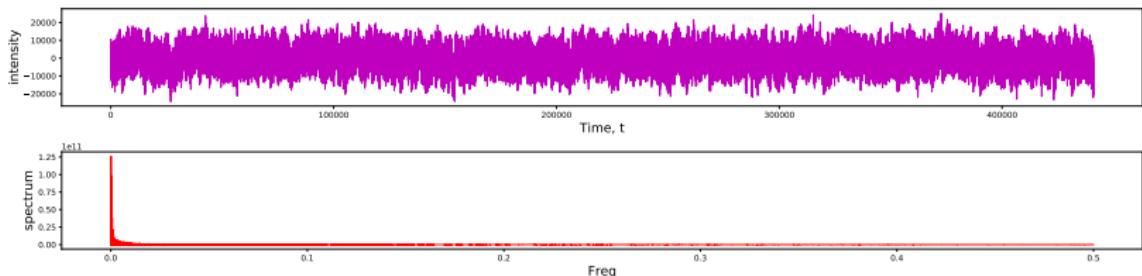


Figure 6: Spectrum density of pink noise

WB  
WBNoise

- Periodogram (lower panel) shows clearly that pink noise is low frequency signal with no high frequency energy
- Theoretically,  $S(\omega) = \eta (1/\omega)$ ,  $\eta \in \mathbb{R}$



# Conclusion

- Different characteristics on time series perspective:
  - ▶ White noise are shown to be  $AR(0)$  process
  - ▶ Pink noise tends to have long memory
- Spectral density perspective
  - ▶ White noise is uniformly distributed in whole spectrum
  - ▶ Blue noise has more energy in higher frequency domain
  - ▶ Pink noise is purely low frequency sound



## Bibliography

- █ Oppenheim, A.V., Willsky, A.S. and Hamid, S., 1996  
*Signals and Systems, 2<sup>nd</sup> edition*  
Pearson Press
- █ Tsay, R. S., 2016  
*Analysis of financial time series, 3<sup>rd</sup> Edition*  
John Wiley & Sons



## Normality Test

Using the omnibus normality test in *DAgostino, R.B.* (1971, 1973), we have following result:

Signal	Statistics	P-Value
HBO audio	$4.48 \times 10^4$	0.00
White noise	$5.60 \times 10^{-2}$	0.97
Blue noise	21.60	0.00
Pink noise	5.22	0.07

