

Signal and Information Processing

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Presentations

Data, Signals, and Information



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- Webpage: https://alelab.seas.upenn.edu/
- My research is on signal and information processing. Best possible class for me to teach
- We meet online \Rightarrow Mondays, Wednesdays, Fridays 10 am to 11 am
- ▶ I would like to see you in class. A minimum level of social connection is important!
- There is ample empirical evidence on the large correlation of grades and class attendance



- ► Course's website ⇒ https://ese224.seas.upenn.edu/
- ► Discussion forum in Piazza ⇒ https://piazza.com/class/kjuecpey9hb4hr



- Luana Ruiz
- Vinicius Lima
- Zhan Gao
- Zhiyang Wang
- Juan Cerviño



- First midterm will be in class, during the lecture on Friday, March 6
 - \Rightarrow This is the Friday right before the Spring break. Plan accordingly!
- ► Second midterm on the Wednesday, April 29 lecture ⇒ Worth 26 points. Not cumulative.
 ⇒ This is the last day of classes. Plan accordingly!
- ➤ You will be handing in 13 lab reports which we will grade from 0 to 4 ⇒ No show (0). Poor (1). Fair (2). Good (3). Excellent (4)
- Maximum possible points: $104 = 26 + 26 + 13 \times 4$
- > You pass with 60 points and 13 points in midterm and final (each)
- C requires at least 70 points. B at least 80. A at least 90
- Don't worry. We will work hard. We will learn. We will get A's.



Presentations

Data, Signals, and Information



- Data and signals are collections of values (numbers) we have acquired
 - \Rightarrow A conversation recording, an image, car fuel efficiency
- ▶ People tend to say "signal" if they are electrical engineers. and to say "data" if they are not





▶ In any event, this disagreement is irrelevant \Rightarrow Signals and data are functions



- Information is the answer to a question of interest
 - \Rightarrow For example, these two are pictures of the Eiffel tower
 - \Rightarrow But also, "the tower is illuminated" (left) or "the tower is under construction (right)





The goal of signal processing is to extract information from data



- ▶ The pragmatist: Data is different from information
- ► The purist: Information is not created by processing ⇒ Data IS information
- I am a pragmatist, but the purist is right
- Dodo principle: Everybody has won and all must have prizes
 - \Rightarrow Process signals and information to uncover patterns of interest



 \Rightarrow Dimensionality \Rightarrow The number of possible signals is too large



► This is one signal with 16 components



 \Rightarrow Dimensionality \Rightarrow The number of possible signals is too large



► This is a second signal with 16 components



 \Rightarrow Dimensionality \Rightarrow The number of possible signals is too large



► This is a third signal with 16 components



 \Rightarrow Dimensionality \Rightarrow The number of possible signals is too large



► This is a fourth signal with 16 components



 \Rightarrow Dimensionality \Rightarrow The number of possible signals is too large



• There are 10^{16} signals that we can generate with 16 components



 \Rightarrow Dimensionality \Rightarrow The number of possible signals is too large



• There are 10^{32} signals that we can generate with 32 components



 \Rightarrow Dimensionality \Rightarrow The number of possible signals is too large



• There are 10^{64} signals that we can generate with 64 components



 \Rightarrow Dimensionality \Rightarrow The number of possible signals is too large



• Let's stop. There are 10^{78} to 10^{82} atoms in the universe



- ▶ Impossible to explore any large dimensional space with any meaningful degree of accuracy.
- **>** Data points are like the stars in the sky \Rightarrow A lot. But far between.
- And yet, our eyes can identify patterns with so much ease...



When you look at this signal ALL of you can tell me what it is



Original signal x(n). It moves randomly, but not that much



When you look at this signal ALL of you can tell me what it is



Signal y(n) reconstructed from cleaned spectrum

The slow oscillation is a pattern. The fast oscillation is noise

The Fourier transform



▶ We represent the signal in the frequency domain.

 \Rightarrow We decompose into different modes of variability (slow and fast)



▶ The pattern (spikes) is clearly separated from the noise (the floor)

The Fourier transform



▶ We represent the signal in the frequency domain.

 \Rightarrow We decompose into different modes of variability (slow and fast)



DFT Y(k) of signal with reshaped spectrum

► Keep the pattern. Remove the noise



- ▶ The Fourier transform is an alternative representation of the signal
- \blacktriangleright It is equivalent \Rightarrow Can go back and forth. Information is preserved
- ▶ But patterns are easier to explain ⇒ Information is uncovered
- ► Fourier representation is called frequency (variability) representation
- Literally a new sense to view things that are otherwise invisible

"On ne voit bien qu'avec le coeur. L'essentiel est invisible pour les yeux."

The Little Prince

• Coeur being, of course, the french word for frequency



▶ This does work. This is how I generated the red (clean) signal



 \blacktriangleright Don't have to undo transformation $\ \Rightarrow$ Except for the benefit of our eyes



- Indeed. If you are not in awe, you don't have a heart.
 - \Rightarrow Change majors. We don't want you here.
 - \Rightarrow Just kidding! Stay here, and let me convince you!
- But magic is not real magic \Rightarrow It's a trick the magician can explain
- > The Fourier transform reduces the dimensionality of the problem
 - \Rightarrow There are only a few Fourier coefficients that are meaningful
 - \Rightarrow These are not stars in the sky anymore
 - \Rightarrow The Fourier transform concentrates information



- Frequency representations is how we uncover patterns in time signals
 Noise removal. Compression. Speech recognition.
- Images \Rightarrow Two (multi) dimensional Fourier transforms
- ► Random Signals ⇒ Principal Component Analysis
- ► Signals with arbitrary strucutres ⇒ Graph Signal Processing



- ► We talk representations and their role in concentrating information
 - \Rightarrow But we don't dwell too much on how to exploit representations
- ESE 305 Foundations of Data Science
- ESE 545 Data Mining: Learning from Massive Datasets
- CIS 520 Machine Learning
- ► Deep Learning ⇒ Convolutional Neural Networks (CNNs)
 - \Rightarrow Fourier transforms and Convolutional filters are equivalent
 - \Rightarrow CNNs compose convolutional filters with local nonlinearities
- ▶ Graph Neural Networks ⇒ https://gnn.seas.upenn.edu