



Applications

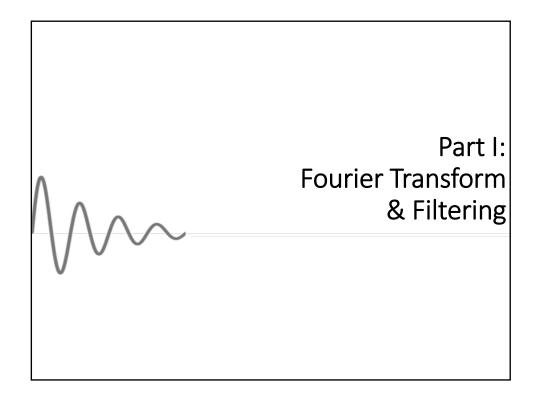
☐ Application fields of signal processing

- > Audio/Speech signal processing
- > Image processing
- ➤ Wireless communication
- ➤ Control systems
- > Array processing
- ➤ Seismology
- > Financial signal processing
- > Feature extraction, Quality improvement, Compression



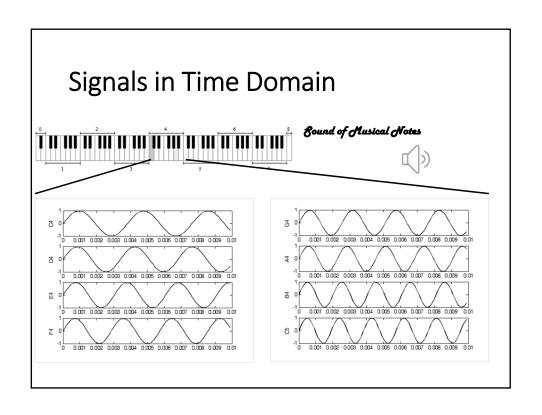
References

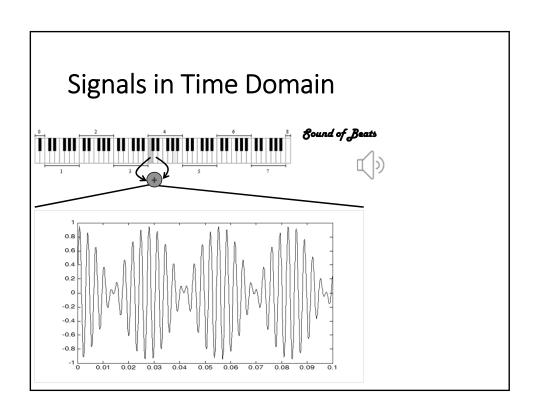
- 1. A. V. Oppenheim, A. S. Willsky and S. Hamid Nawab, Signals and Systems, Pearson Education Ltd., 2nd Edition, 2014.
- 2. EE16A in Berkeley: http://inst.eecs.berkeley.edu/~ee16a/sp15/
- 3. Single-Pixel Camera: http://dsp.rice.edu/cscamera
- 4. Emmanuel Candes, Compressive Sensing A 25 Minute Tour, First EU-US Frontiers of Engineering Symposium, Cambridge, September, 2010
- 5. Xuedong Huang and Li Deng, An Overview of Modern Speech Recognition
- 6. Wikipedia on various topics: compressive sensing, speech recognition, noise cancellation
- 7. E. Perahia and R. Stacey, Next Generation Wireless LANs, 2nd Ed., Cambridge



What is frequency?

How to characterize frequency?





Question:

How to define and visualize the frequency in a signal?

Answer: Fourier Transform



/ˈfʊəriˌeı, -iər/ 1768-1830 French Mathematician, Physicist, Historian

Frequency Domain

Fourier Transform of a signal is defined as:

Signal analysis:
$$X(f) = \int_{-\infty}^{+\infty} x(t)e^{-j2\pi ft}dt$$

f: is the Frequency in Hertz = 1/second

Correspondingly, X(f) can be used to recover the signal as:

Signal synthesis:
$$x(t) = \int_{-\infty}^{+\infty} X(f)e^{+j2\pi ft}d\omega$$

Some notations we OFTEN use

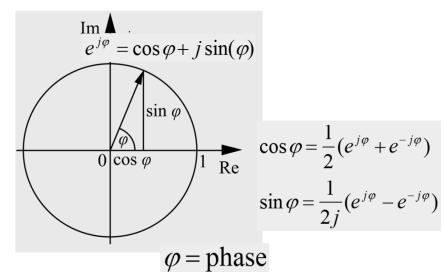
u(t) = step function = 1 only for positive t.

 $\delta[n]$ = delta function (known as Kronecker Delta) = impulse function = 1 only for n=0.

f= frequency = tells you how fast signals change

= not just how fast

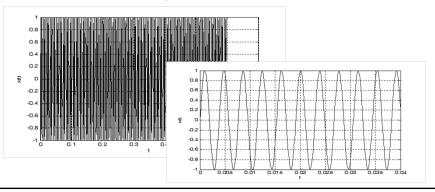
Euler's Formula (frequency)



Frequency Domain

Example 1: Fourier Transform of the C4 tone

$$x(t) = \begin{cases} \sin(2\pi \cdot f_0 \ t), & t \in [-1,1] \\ 0, & o. \ w. \end{cases} f_0 = 261.626 \text{Hz}$$



Frequency Domain

Example 1: Fourier Transform of the C4 tone

$$X(f) = \int_{-1}^{1} \sin(2\pi f_0 t) \cdot \exp[-j2\pi f t] dt$$

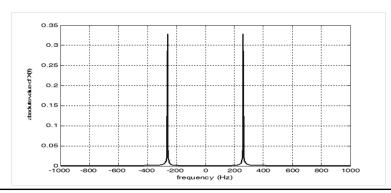
$$= \frac{1}{2j} \left[\int_{-1}^{+1} \exp[j2\pi (f_0 - f)t] dt - \int_{-1}^{+1} \exp[j2\pi (-f_0 - f)t] dt \right]$$

$$= \frac{\sin(2\pi (f - f_0))}{j2\pi (f - f_0)} - \frac{\sin(2\pi (f + f_0))}{j2\pi (f + f_0)}$$

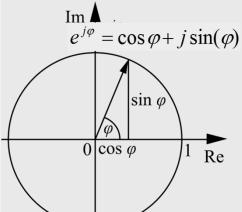
Frequency Domain

Example 1: Fourier Transform of the C4 tone

$$x(t) = \begin{cases} \sin(2\pi \cdot f_0 \ t), & t \in [0,1] \\ 0, & o.w. \end{cases}$$
 $f_0 = 261.626$ Hz



Euler's Formula (frequency)



$$\cos\varphi = \frac{1}{2}(e^{j\varphi} + e^{-j\varphi})$$

$$\cos \varphi = \frac{1}{2} (e^{j\varphi} + e^{-j\varphi})$$

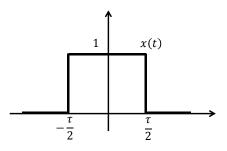
$$\sin \varphi = \frac{1}{2j} (e^{j\varphi} - e^{-j\varphi})$$

 $\varphi = 2\pi f_0 \bullet t = \text{linear phase change in } t$

Frequency Domain

Example 2: Fourier Transform of the rectangular pulse

$$x(t) = \begin{cases} 1, & t \in \left[-\frac{\tau}{2}, \frac{\tau}{2} \right] \\ 0, & o.w. \end{cases}$$



Frequency Domain

Example 2: Fourier Transform of the rectangular pulse

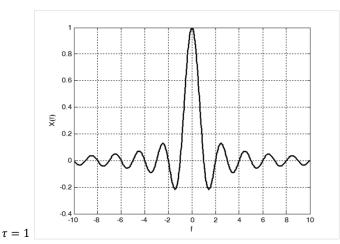
$$X(f) = \int_{-\frac{\tau}{2}}^{\frac{\tau}{2}} 1 \cdot \exp[-j2\pi f t] dt$$

$$= \frac{1}{-j2\pi f} \left[\exp\left(\frac{-j2\pi f \tau}{2}\right) - \exp\left(\frac{+j2\pi f \tau}{2}\right) \right]$$

$$= \frac{\sin(\pi f \tau)}{\pi f}$$

Frequency Domain

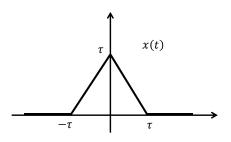
Example 2: Fourier Transform of the rectangular pulse



Frequency Domain

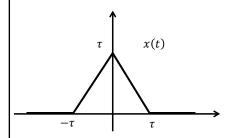
Example 3: Fourier Transform of the triangular pulse

$$x(t) = \begin{cases} \tau - |t|, & t \in [-\tau, \tau] \\ 0, & o.w. \end{cases}$$

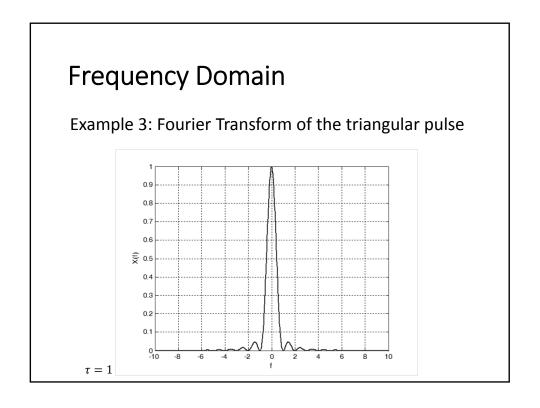


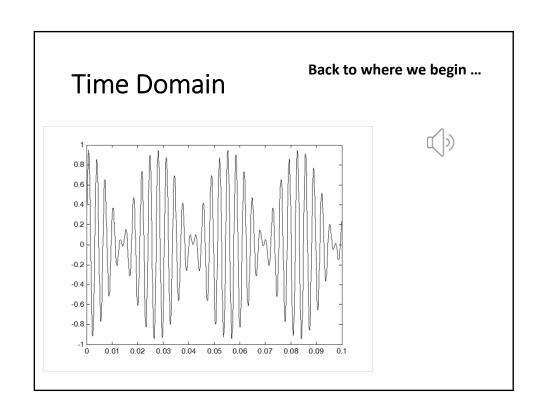
Frequency Domain

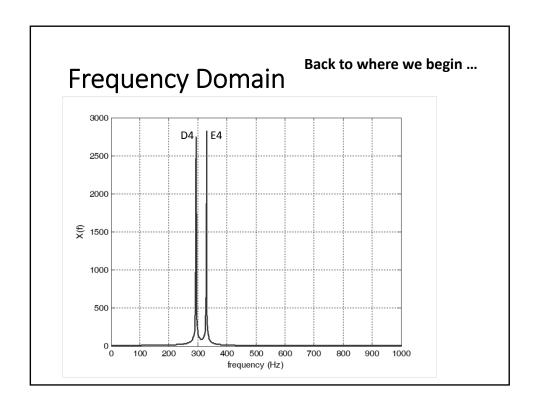
Example 3: Fourier Transform of the triangular pulse



$$X(\omega) = \int_{-\infty}^{+\infty} x(t) \cdot \exp[-j2\pi f t] dt$$
$$= 2 \int_{0}^{\tau} (\tau - t) \cos(2\pi f t) dt$$
$$= \left(\frac{\sin(\pi f \tau)}{\pi f}\right)^{2}$$







Define discrete time signals

x(t) is continuous time signal;

x[n] = x(nT) discrete signal sampled periodically

If your T is small enough, no information loss: Nyquist sampling theorem says T < 1/(2W).

W=Bandwidth (in frequency) of a signal

Digital signal processing only deal with discrete signals

Some notations we OFTEN use

u(t) = step function = 1 only for positive t.

 $\delta[n]$ = delta function (known as Kronecker Delta) = impulse function = 1 only for n=0.

W= $Bandwidth\ of\ x(t)$

= maximum positive frequency for which X(f) is not zero.

Digital signal processing only deal with discrete signals

Discrete Fourier Transform

In computer, we have to discretize the signal and the following continuous Fourier Transform:

$$X(f) = \int_{-\infty}^{+\infty} x(t)e^{-j2\pi ft}dt$$

Sampling every T_s ($F_s=1/T_s$) seconds:

$$X(k) = \sum_{\substack{n = -\infty \\ +\infty}}^{+\infty} x(nT_s)e^{-j2\pi\left(\frac{kF_s}{N}\right)(nT_s)}$$
$$= \sum_{n = -\infty}^{+\infty} x(nT_s)e^{-j2\pi\left(\frac{kn}{N}\right)}$$

Fast Fourier Transform (FFT)

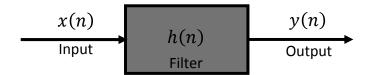
Discrete Fourier Transform:

$$X(k) = \sum_{n=-\infty}^{+\infty} x(nT_s)e^{-j2\pi\left(\frac{kn}{N}\right)}$$

- N is the power of 2 \rightarrow very efficient algorithms
 - DFT is also called FFT!

Note: You will learn more during the practice session!

Filtering

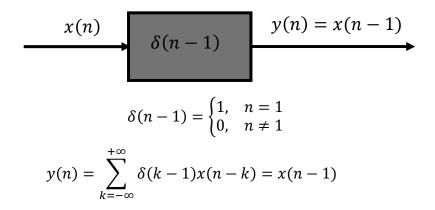


The role of a filter is to perform the following operation called "Convolution":

$$y(n) = \sum_{k=-\infty}^{+\infty} h(k)x(n-k)$$

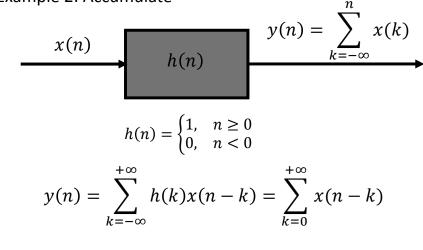
Filtering

Example 1: Unit Delay



Filtering

Example 2: Accumulate



Filtering

Example 3: Difference

$$x(n)$$

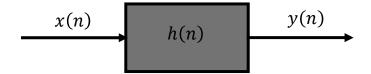
$$h(n)$$

$$y(n) = x(n) - x(n-1)$$

$$h(n) = \begin{cases} 1, & n = 0 \\ -1, & n = 1 \\ 0, & o.w. \end{cases}$$

$$y(n) = \sum_{k=-\infty}^{+\infty} h(k)x(n-k) = x(n) - x(n-1)$$

Filtering Changes Fourier Transform



$$y(n) = \sum_{k=-\infty}^{+\infty} h(k)x(n-k)$$

$$Y(k) = H(k)X(k)$$

Time Domain

Frequency Domain

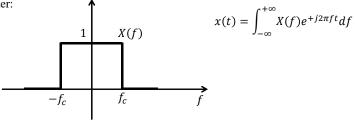
Note: You will learn more during the practice session!

Part II: Homework

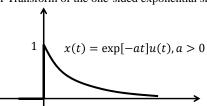


HW Problems

Find the time domain signal that has the following Fourier Transform, which is called Low-Pass Filter:



2. Find the Fourier Transform of the one-sided exponential signal.





HW Problems

3. Find the Discrete Fourier Transform of the following signals:

3.a
$$x(n) = \delta(n-1), n = 0,1,..., N-1$$

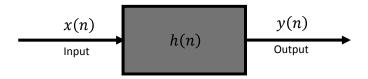
3.b
$$x(n) = a^n, n = 0, 1, ..., N - 1$$

3.c
$$x(n) = \sin(2\pi f_0 n)$$
, $n = 0,1,...,N-1$



HW Problems

4. Let y(n) be the filtering output when inputting x(n), determine the output with the following inputs:



- 4.a $x(n-n_0)$, n_0 is a fixed natural number
- 4.b ax(n), a is a fixed real number
- 4.c $x(n)e^{j2\pi f_0 n}$, f_0 is a fixed real number

Part III: Overview of Signal Processing Applications



Example System: IPAD Air 2

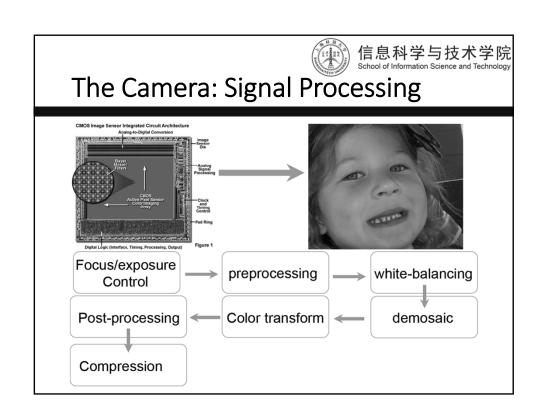


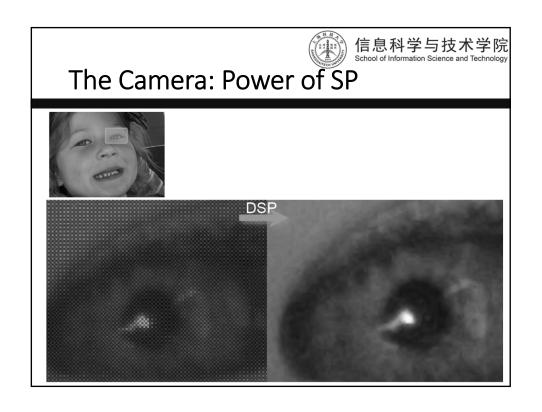
☐ Great stuff running apps, but:

- o What makes the display tick?
- o How does the Wi-Fi work?
- o How does it sense touch on the screen?
- o How does it sense the motion?
- o How does Siri work?
- o ..

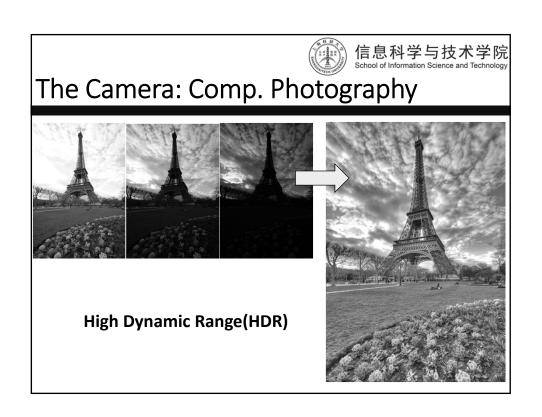














The Microphone: Siri



Siri. Your wish is its command.

Siri lets you use your voice to send messages, schedule meetings, place phone calls, and more. Ask Siri to do things just by talking the way you talk. Siri understands what you say, knows what you mean, and even talks back. Siri is so easy to use and does so much, you'll keep finding more and more ways to use it.





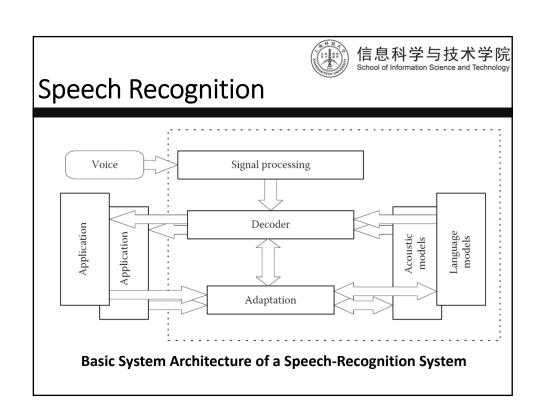
Speech Recognition

- ☐ From the technology perspective, speech recognition has been going through several waves of major innovations since over some 50 years ago.
- ☐ The most recent wave of innovations since 2009 is based on **deep learning** concepts, architectures, methodologies, algorithms, and practical system implementations enabled by big training data and by GPU-based big compute
 - defines the current state of the art in speech recognition accuracy and has been in dominant use (e.g. Siri) since 2013 throughout the speech industry worldwide



Speech Recognition

- Applications
 - ➤ Dictation
 - ➤ Telephone-based Information (directions, air travel, banking, etc)
 - ➤ Hands-free (in car)
 - Second language ('L2') (accent reduction)
 - > Audio archive searching
 - ➤ Linguistic research
 - Automatically computing word durations, etc





Speech Recognition

☐ Fundamental of statistical speech recognition:

$$\hat{\mathbf{W}} = \underset{\mathbf{w}}{\operatorname{arg\,max}} P(\mathbf{W}|\mathbf{A}) = \underset{\mathbf{w}}{\operatorname{arg\,max}} \frac{P(\mathbf{W})P(\mathbf{A}|\mathbf{W})}{P(\mathbf{A})}$$

A: acoustic observation

W: word sequence

The goal of speech recognition is to find out the word sequence that has the max posterior probability.



Speech Recognition

The New York TimesScientists See Promise in Deep-Learning Programs



Richard F. Rashid

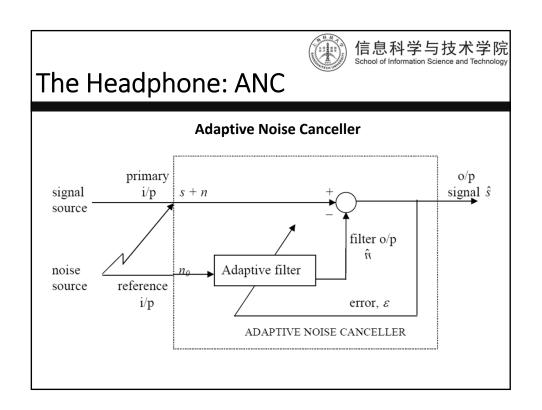
SVP, Microsoft A voice recognition program translated the speech into Mandarin Chinese Tianjin, Oct. 25, 2012

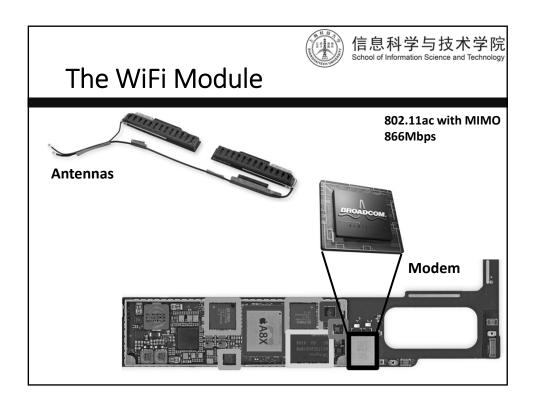
<u>Deep learning</u> technology enabled speech-to-speech translation













The WiFi Module

- ☐ A brief WiFi History
 - IR

clause 15

☐ 1997: 802.11 standard

· FHSS,

clause 14

DSSS in 2.4GHz

2Mbps

- clause 16
- □ 1999: 802.11b

 DSSS in 2.4GHz 11Mbps, clase 17

□ 1999: 802.11a

 OFDM in 5GHz 54Mbps, clase 18

- □ 2003: 802.11g
 - OFDM in 2.4GHz 54Mbps

clase 19

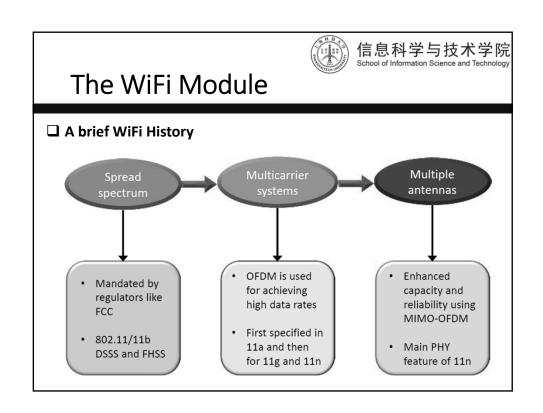
- □ 2009: 802.11n
 - OFDM+MIMO SM

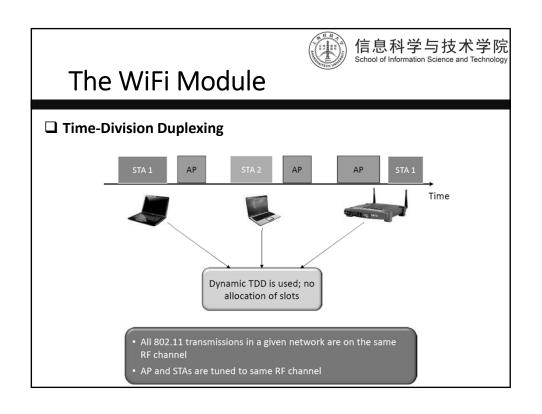
20MHz BW → 300Mbps 40MHz BW → 600Mbps

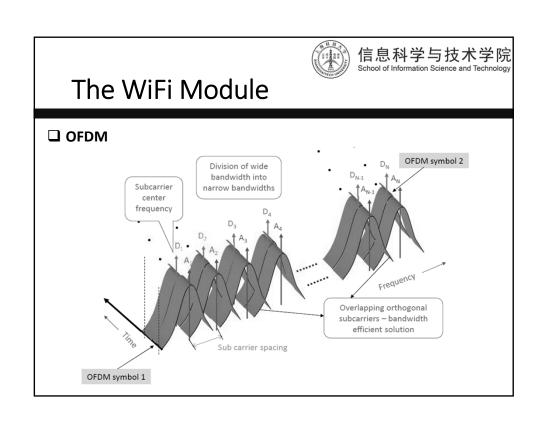
clause 20 (HT PHY) in the 2012 revision of IEEE 802.11

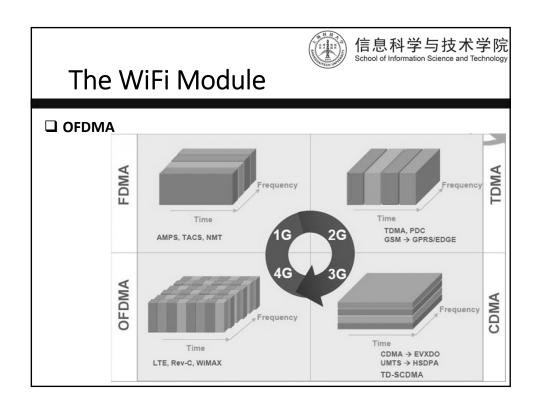
standard

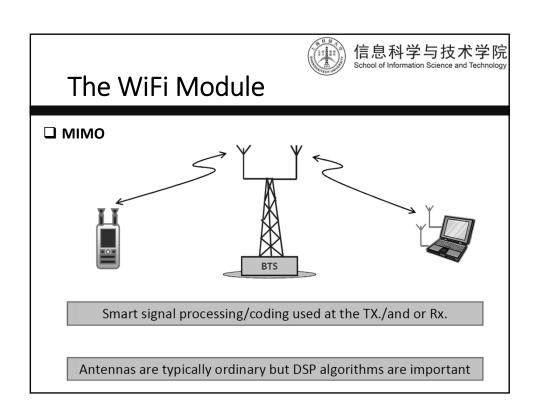
- ☐ 2007~now: 802.11ac
 - VHT
 - enhancing 802.11n in 5GHz

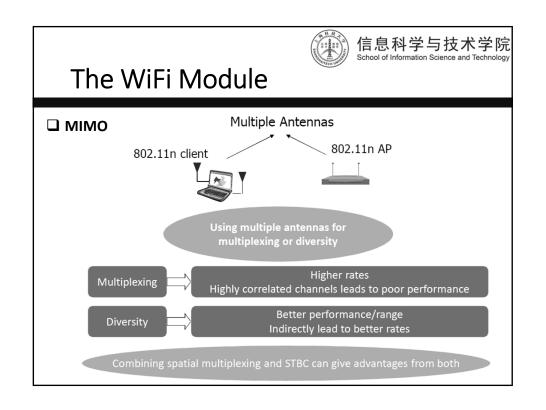


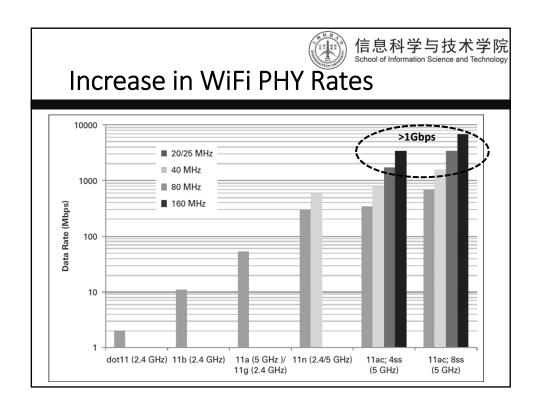












Quiz

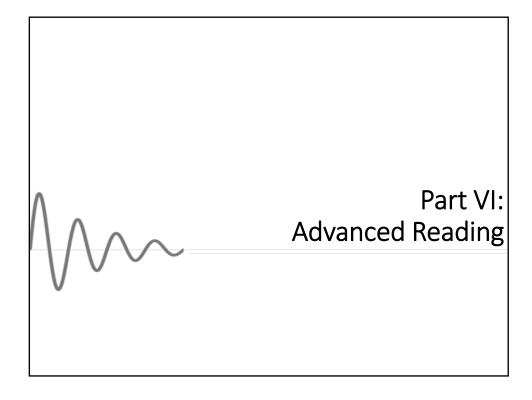
• Find the Fourier transform of two signals

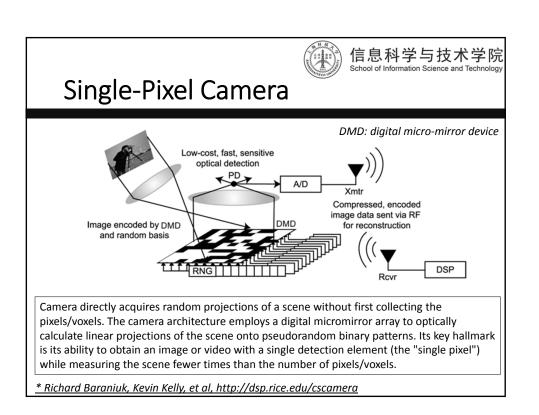
g(t)=A[u(t-1)-u(t-5)]

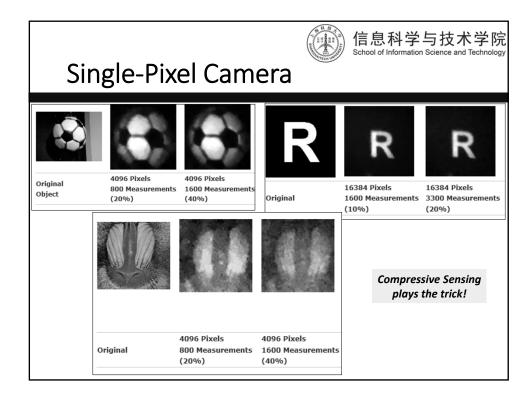
G(f) = integration g(t) exp(-j....) dt continue

 $p(n)=b^n u(n-1), n=0, 1,2,3,... N-1, |b| < 1.$

P(k)= sum p(n) exp(-j 2 pi nk/N)? Continue.... 几何级数









Compressive Sensing

Compressed sensing (also known as compressive sensing, compressive sampling, or sparse sampling) is a signal processing technique for efficiently acquiring and reconstructing a signal, by finding solutions to underdetermined linear systems

In a nutshell ...

- ☐ Can obtain super-resolved signals from just a few sensors
- ☐ Sensing is nonadaptive: no effort to understand the signal
- ☐ Simple acquisition process followed by numerical optimization

First papers:

- ☐ Candes, Romberg and Tao, 2006
- ☐ Candels and Tao, 2006
- ☐ Donoho, 2006

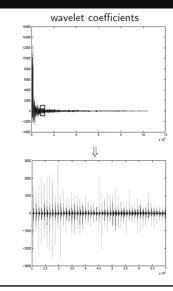
* Emmanuel Candes, Compressive Sensing – A 25 Minute Tour



Compressive Sensing



1 megapixel image



信息科学与技术学队 School of Information Science and Technology

Compressive Sensing

- Compute 1,000,000 wavelet coefficients of mega-pixel image
- Set to zero all but the 25,000 largest coefficients
- Invert the wavelet transform



original image



after zeroing out smallest coefficients



Compressive Sensing

- ullet Take $96\mbox{K}$ incoherent measurements of "compressed" image
- Compressed image is perfectly sparse (25K nonzero wavelet coeffs)
- Solve ℓ_1



original (25k wavelets)



perfect recovery



Introduction to Communications

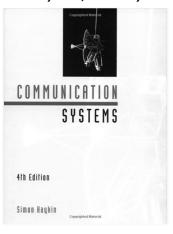
by Yanlin Geng & Xiaojun Yuan





References

1. Simon Haykin, Communication Systems, John Wiley & Sons, 4th Edition, 2001.



Word Counting



A sequence: "ACBACABA"

Q: How many A, B, C? A: A = 4, B = 2, C = 2

Q: How frequently?

A: pA = 0.5, pB = 0.25, pC = 0.25



Word Counting

Use {0, 1}'s to represent A, B, C:

? minimum length can recover "ACBACABA"

Observation:

more frequently \rightarrow shorter $\{0, 1\}$ sequence

Say: A \rightarrow 0, B \rightarrow 1, C \rightarrow 01

Then: "ACBACABA" → "0011001010"

Q: can we recover "ACBACABA" uniquely?

Word Counting



A: no, since $001 \rightarrow AC$ or AAB

Q: how to solve?



Huffman coding

A: Huffman coding

 $A \rightarrow 1$

B → 00

C → 01

Q: any pattern?

A: prefix-free

no codeword is prefix of others

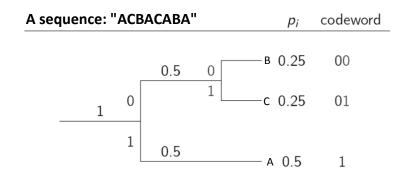
prefix code

信息科学与技术学 School of Information Science and Techn

Huffman coding

Q: how Huffman coding works?

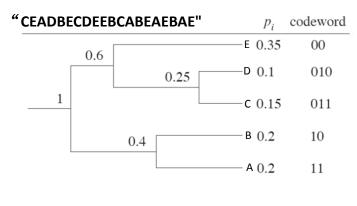
A: Keep merging two smallest frequencies:





Huffman coding

A slightly more complicated example.

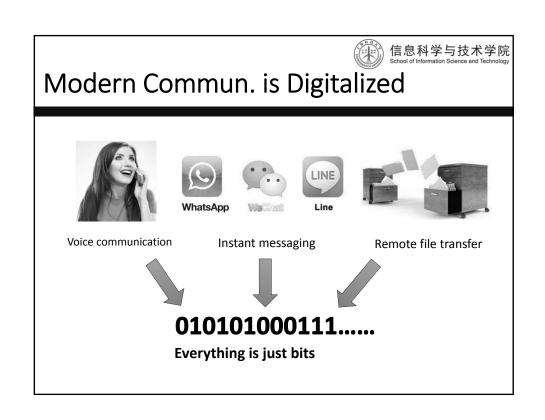


Huffman coding



Q: codewords using {0, 1, 2}? find out

The End-to-End Commun. System Original source Digitize (if necessary) Source coding Communication network The following lecture is about the oval. The simplest network is a single physical communication link.





Physical Links Inherently Analog



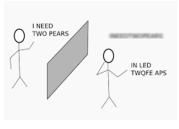
Analog = continuous-valued and continuous-time

- Voltage waveform on a cable
- Light on a fiber, or in free space
- Radio waves through the air

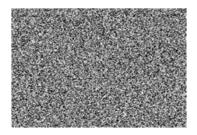
Solution → Modulation



...and Physical Links Inherently Noisy



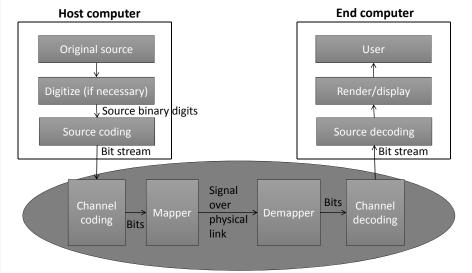




White noise

- Noise usually comes from the ambient environment and communication media.
- Communication errors occur due to channel noise.
- Solution → Channel coding







Modulation: Map Bits to Signals

Key Idea: Map or modulate the desired bit sequence onto a (continuous-time) analog signal

For ease of extracting the intended bits from the noisy received signals, we map bits to signals using a fixed set of discrete values.

For example, a two-level signaling scheme uses two "voltages":

- V₀ is the binary value for "0"
- V₁ is the binary value for "1"

If $V_0 = -V_1$, we refer to this as bipolar signaling.



Digital Signaling: Receiving

At the receiver, process and sample to get a "voltage"

- Voltages near V₀ would be interpreted as "0"
- Voltages near V₁ would be interpreted as "1"

If V_0 and V_1 are spaced far enough apart, we can tolerate some degree of noise – but there will be occasional errors!



Digital Signaling: Receiving

We can specify the behavior of the receiver with the following decision rule (that shows how incoming voltages are mapped to "0" and "1").

Decision Rule:

If the received voltage is below d, then "0" is transmitted; otherwise, "1" is transmitted.

In the above, d is called the threshold voltage.

The threshold voltage is usually chosen as the middle of the two voltage levels, i.e., $d = (V_0 + V_1)/2$. Why?



How to Reduce Decoding Error?

One simple way to reduce decoding error is by repetition.

Code: Bit b coded as bb...b (n times)

Exponential fall-off (log scale)

But huge overhead (low code rate)

We can do much better!



Hamming Distance

The Hamming distance (HD) between two strings of equal length is the number of positions at which the corresponding symbols are different.

Examples:

- HD between "karolin" and "kerstin" is 3.
- HD between 1011101 and 1001001 is 2.

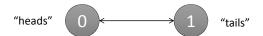
The HD between a valid binary codeword and the same codeword with $\it e$ errors is $\it e$.



Hamming Distance

The Hamming Distance (HD) between a valid binary codeword and the same codeword with e errors is e.

The problem with no coding is that the two valid codewords ("0" and "1") also have HD = 1. So a single-bit error changes a valid codeword into another valid codeword.



Q: What is the Hamming distance of a repetition code?

信息科学与技术学院 School of Information Science and Technology

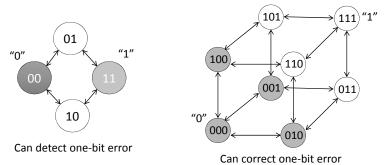
Hamming Distance and Coding

Encode so that the codewords are "far enough" from each other.

Likely error patterns shouldn't transform one codeword to another.

Code: nodes chosen in hypercube + mapping of message bits to nodes

We choose 2k out of 2n nodes, meaning that we can map all k-bit message strings in a space of n-bit codewords. The code rate is k/n.





Minimum HD vs. Detection & Correction Capabilities

If d is the minimum Hamming distance between codewords, we can detect all patterns of <= (d-1) bit errors

If d is the minimum Hamming distance between codewords, we can correct all patterns of (d-1)/2 or fewer bit errors

But how to construct codes with the above properties?



A Simple Code: Parity Check

Add a parity check to message of length k to make the total number of "1" bits even.

If the number of "1"s in the received word is odd, then there has been an error.

- 010111100010 → original word with parity bit
- 010011100010 → single-bit error (detected)
- 010001100010 → two-bit error (not detected)

Minimum hamming distance of parity check code is 2.

- · Can detect all single-bit errors
- · Can detect all odd numbers of errors
- · Cannot detect even numbers of errors
- · Cannot correct any errors



Modulo-2 Algebra

Computations with binary numbers in code construction will involve Boolean algebra, or algebra in "GF(2)" (Galois field of order 2) or modulo-2 algebra:



Linear Block Codes

Block Code: k message bits encoded to n code bits, i.e., each of 2k messages encoded into a unique n-bit combination via a linear transformation with GF(2) operations:



C is an n-element row vector containing the codeword

D is a k-element row vector containing the message

G is a k-by-n generation matrix

Each codeword bit is a specified linear combination of message bits.



Minimum HD of Linear Code

- Key property: Sum of any two codewords is still a codeword
 → necessary and sufficient for code to be linear
 - So the all-zero codeword must be in any linear code --- why?
- (n, k) code has rate k/n
- Sometimes written as (n, k, d), where d is the minimum Hamming distance of the code.
- The weight of a codeword is the number of "1"s in it.
- The minimum HD of a linear code is the minimum weight found in its nonzero codewords. Why?



Examples: What are n, k, d?

```
{000, 111} (3, 1, 3), rate = 1/3

{0000, 1100, 0011, 1111} (4, 2, 2), rate = 1/2

{1111, 0000, 1100} non-linear code

{0000, 1000, 0011, 1111} non-linear code

0000000 1100001 1100110 0000111 (7, 4, 3) code, rate = 4/7

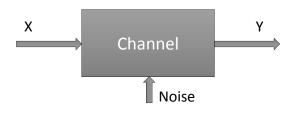
0101010 0110011 0110100 1010101

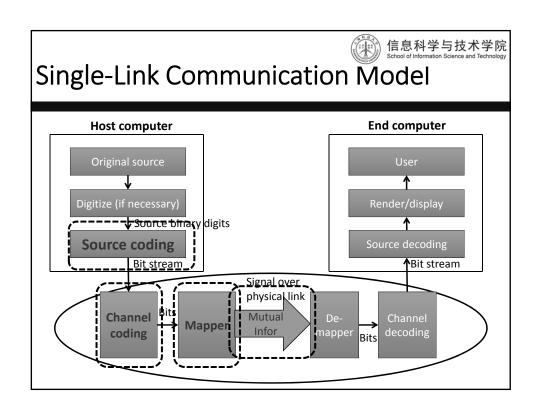
1111000 0011001 0011110 1111111
```

The HD of a linear code is the minimum number of 1's in all codewords.

信息科学与技术学院 School of Information Science and Technology How Fast We can Deliver Bits • H(X):

- Entropy = the uncertainty in X
- H(X|Y):
 - · Uncertainty about X reduced by knowing Y
- I(X;Y) = H(X) H(X|Y):
 - Mutual information = RATE supported by the channel







Homework

- 1. For the following text file, find out the number of occurrences of A, B, ..., Z and the space ('a' taken as 'A', 'b' as 'B', etc.)
 - a). do Huffman coding and calculate the number of bits needed to store this text file
 - b). how many bits needed is we simply apply the ASCII coding scheme?

Samsung still appears to be exclusively using its eight core Exynos five four two zero as the S six app processor Though the WSJ does not state whether Qualcomm will do so there is a chance the company will also supply complementary RF transceiver power management and envelope tracking ICs in units containing its modems as is the case with the iPhone six which relies on a Qualcomm modem and several complementary chips to go with Apple A eight app processor



Homework

- 2. Calculate the HD between the following two bit sequences:
 - a) 1001100111000
 - b) 0101110001001
- 3. Generate a code, i.e., a set of codewords, with the following property:
 - 4-bit codewords with three information bits and with single error correction capability

i.e. min HD = 3

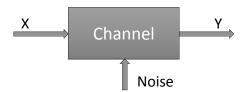




Advanced Reading

Channel Capacity





To characterize the channel, define

 $C = \max \{I(X; Y)\} = \max \{H(X) - H(X|Y)\}$

where the maximization is over all possible distributions of X.

This is the most we can expect to reduce our uncertainty about X through the knowledge of Y, and so must be the most information we can expect to send through the channel.

An Example: Binary Symmetric Channel



Suppose that during transmission a "0" is turned into a "1" or a "1" is turned into a "0" with probability p, independently of transmissions at other times.

This is a binary symmetric channel (BSC) – a useful and widely used abstraction.



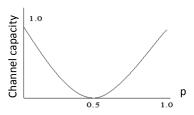
Capacity of the Binary Symmetric Channel





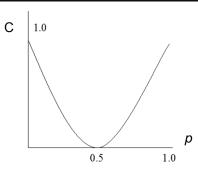
 $C = max \{H(Y) - H(Y|X)\}$, where the maximization is over all possible distributions of X.

The second term doesn't depend on this distribution, and the first term is maximized when 0 and 1 are equally probable at the input.



Capacity for Binary Symmetric Channel





For low noise channel, significant reduction in uncertainty about the input after observing the output. Explain why C = 1 for p = 0.

For high noise channel, little reduction. Explain why C = 0 for p = 0.5.

What happens to p = 1?



Channel capacity tells us how fast and how accurately we can communicate ...

Magic of Asymptotically Error-Free if R < C

信息科学与技术学院 School of Information Science and Technology

Shannon showed that one can theoretically transmit information (i.e., message bits) at any rate R < C per use of the channel, with arbitrarily low error.

He also showed the converse, that transmission at any average rate $R \ge C$ incurs an error probability that is low-bounded by some positive number.

The secret: Encode blocks of k message bits into n-bit codewords, so R = k/n, with k and n very large.

Encoding blocks of k message bits into n-bit codewords to protect against channel errors is an example of channel coding.



Introduction to Networks

Professor: Ziyu Shao

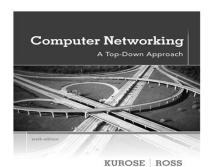




Reference

Acknowledgement:

Main contents of Parts 2,3,4 adopt materials from the official slides to accompany the following book, which is also the reference book



Computer Networking: A Top Down Approach

6th edition Jim Kurose, Keith Ross Addison-Wesley March 2012



Outline

- I Complex Networks
- 2 Foundation of Internet
- 3 Performance Metrics of Networks
- 4 Protocols & Layers
- 5 History

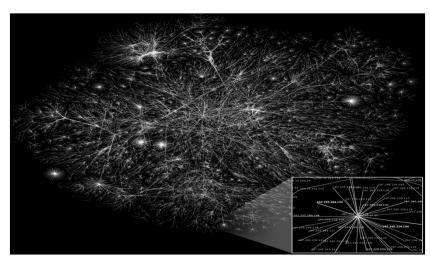


Outline

- I Complex Networks
- 2 Foundation of Internet
- 3 Performance Metrics of Networks
- 4 Protocols & Layers
- 5 History

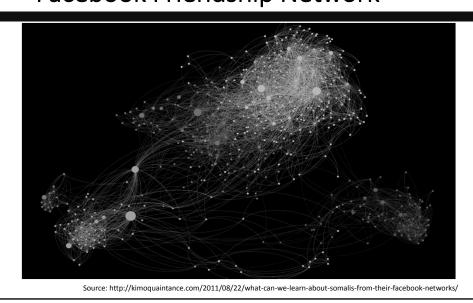


World Wide Web(WWW)

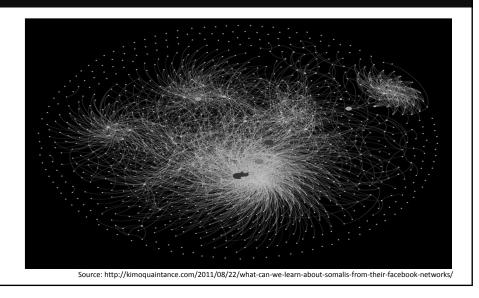


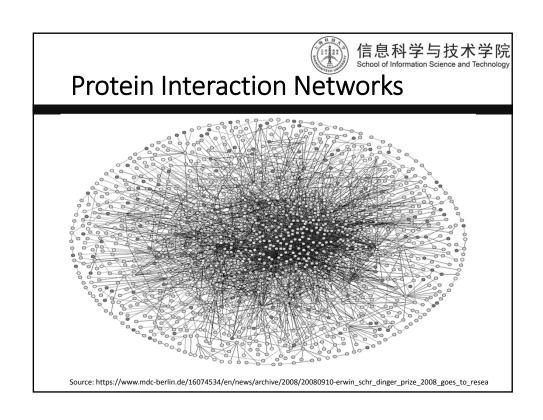
Source: http://en.wikipedia.org/wiki/File:Internet_map_1024.jpg

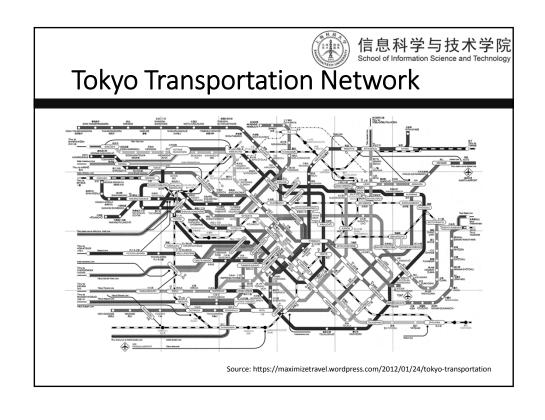




Facebook Friendship Network



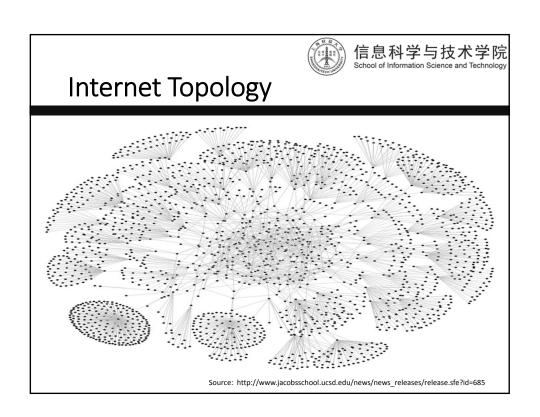






Outline

- I Complex Networks
- 2 Foundation of Internet
- 3 Performance Metrics of Networks
- 4 Protocols & Layers
- 5 History





Elements of Internet



Millions of connected computing devices (hosts)



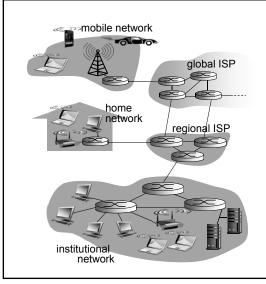
Communication Links (optical fiber, copper, radio, satellite)



Packet Switches (routers & switches)



Structure of Internet



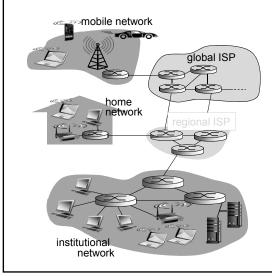
Network Edge: end systems with hosts & access networks

Access Network: connect end systems to edge routers

Network Core: interconnected routers network of networks



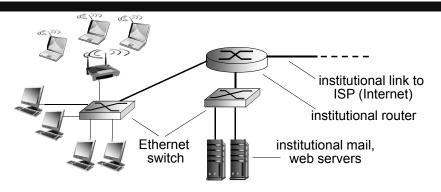
Access Network



The Last Mile Problem: bandwidth (bits per second) of access network



Enterprise Access Network (Ethernet)



Typically used in companies, universities, etc 10 Mbps, 100Mbps, 1Gbps, 10Gbps transmission rates End systems typically connect into Ethernet switch



Wireless Access Network

- · Shared wireless access network connects end system to router
 - via base station aka "access point"

wireless LANs:

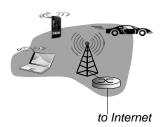
- within building (100 ft)
- 802.11b/g (WiFi): 11,54 Mbps transmission rate

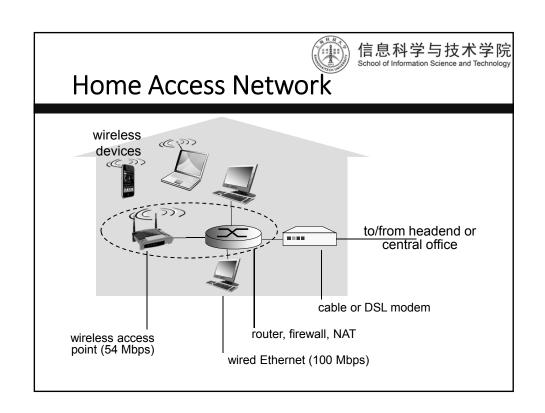


to Internet

wide-area wireless access

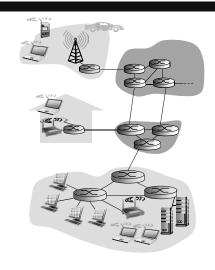
- provided by telco (cellular) operator, 10's km
- between I and I0 Mbps
- 3G,4G: LTE







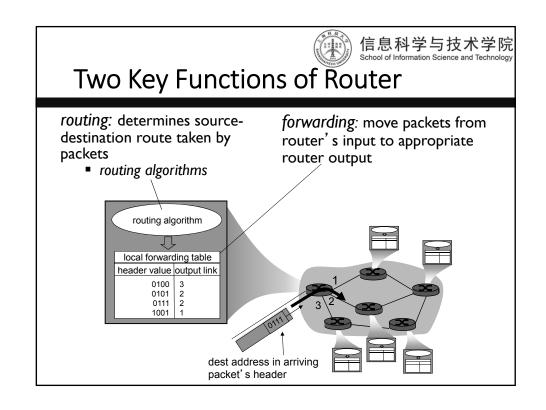
Network Core



Mesh of interconnected routers

Packet Switching:

- Hosts break messages into packets
- Forward packets from one router to the next, across links on path from source to destination
- Each packet transmitted at full link capacity



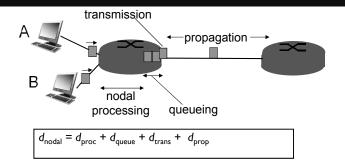


Outline

- I Complex Networks
- 2 Foundation of Internet
- 3 Performance Metrics of Networks
- 4 Protocols & Layers
- 5 History



Metric 1: Packet Delay

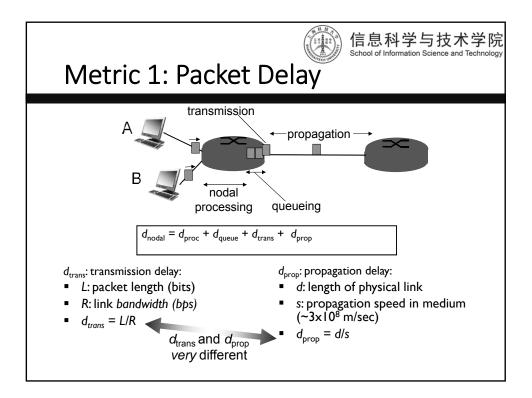


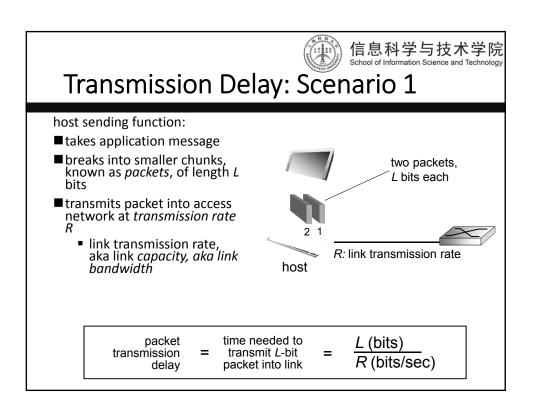
 d_{proc} : nodal processing

- check bit errors
- determine output link
- typically < msec</p>

 $d_{\rm queue}$: queueing delay

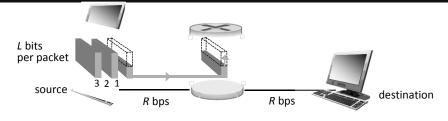
- time waiting at output link for transmission
- depends on congestion level of router







Transmission Delay: Scenario 2



- takes L/R seconds to transmit (push out) L-bit packet into link at R bps
- store and forward: entire packet must arrive at router before it can be transmitted on next link

one-hop numerical example:

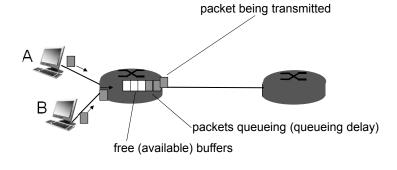
- L = 7.5 Mbits
- R = 1.5 Mbps
- one-hop transmission delay = 5 sec



Queueing Delay

packets queue in router buffers

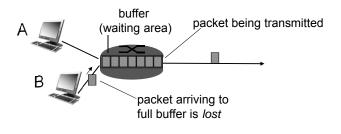
- · packet arrival rate to link exceeds output link capacity
- · packets queue, wait for turn

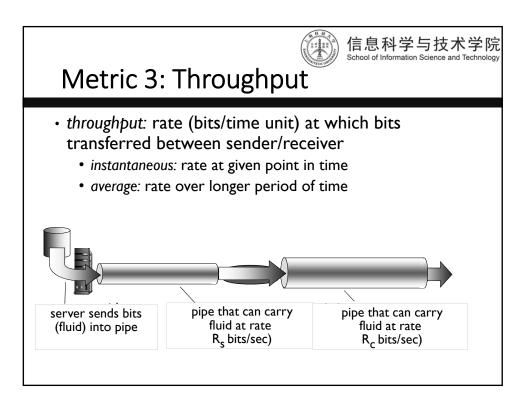


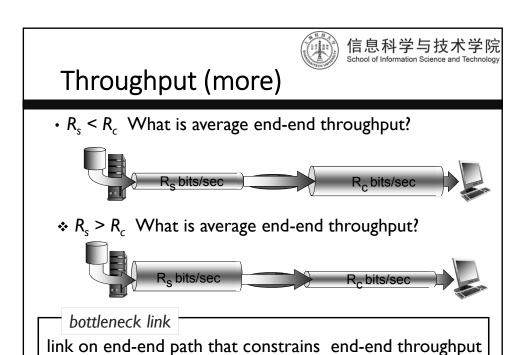


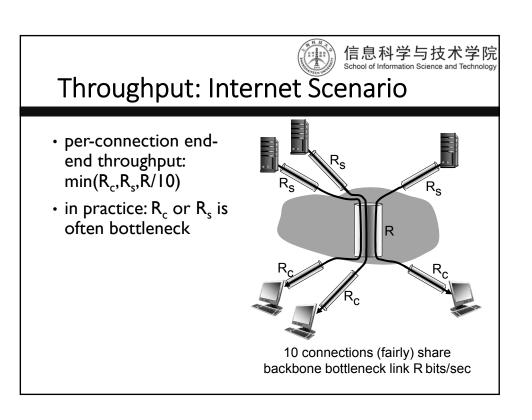
Metric 2: Packet Loss

- queue (aka buffer) preceding link in buffer has finite capacity
- packet arriving to full queue dropped (aka lost)
- lost packet may be retransmitted by previous node, by source of end system, or not at all





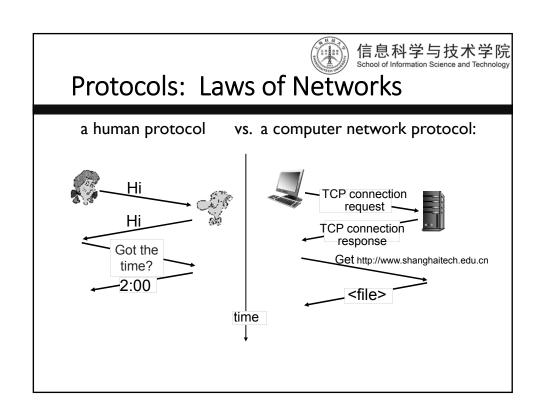






Outline

- I Complex Networks
- 2 Foundation of Internet
- 3 Performance Metrics of Networks
- 4 Protocols & Layers
- 5 History





Protocols: Laws of Networks

Protocols define format, order of messages sent and received among network entities, and actions taken on message transmission & receipt

All communication activity in Internet governed by protocols

Examples: TCP, UDP, IP, BGP, HTTP, 802. I I



The Big Question

Networks are complex, with many "pieces":

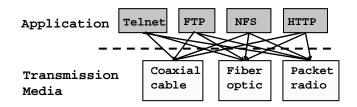
- hosts
- routers
- links of various media
- applications
- protocols
- hardware, software

Question:

is there any hope of organizing structure of network?



The Problem



- Do we re-implement every application for every technology?
- Obviously not, but how does the Internet architecture avoid this?



Architecture

- Architecture is not the implementation itself
- Architecture is how to "organize" implementations
 - what interfaces are supported
 - where functionality is implemented
- Architecture is the modular design of the network



Layering

- Layering is a particular form of modularization
- The system is broken into a vertical hierarchy of logically distinct entities (layers)
- The service provided by one layer is based solely on the service provided by layer below
- Rigid structure: easy reuse, performance may suffers



Layering Model for Internet

- application: supporting network applications
 - FTP, SMTP, HTTP
- transport: process-process data transfer
 - TCP, UDP
- network: routing of datagrams from source to destination
 - IP, routing protocols
- link: data transfer between neighboring network elements
 - Ethernet, 802. I I (WiFi)
- physical: bits "on the wire"

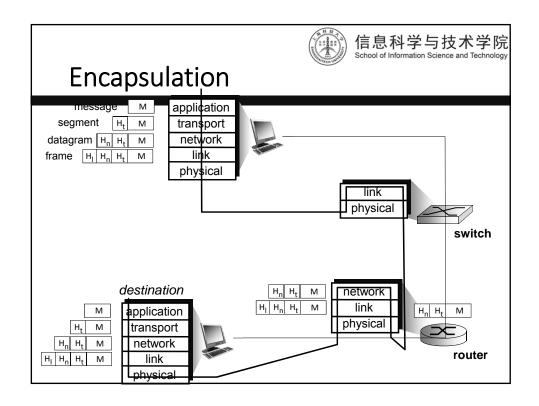
transport
network
link
physical



OSI Reference Model for Layers

- presentation: allow applications to interpret meaning of data, e.g., encryption, compression, machine-specific conventions
- session: synchronization, checkpointing, recovery of data exchange
- Internet stack "missing" these layers!
 - these services, if needed, must be implemented in application
 - · needed?

application
presentation
session
transport
network
link
physical





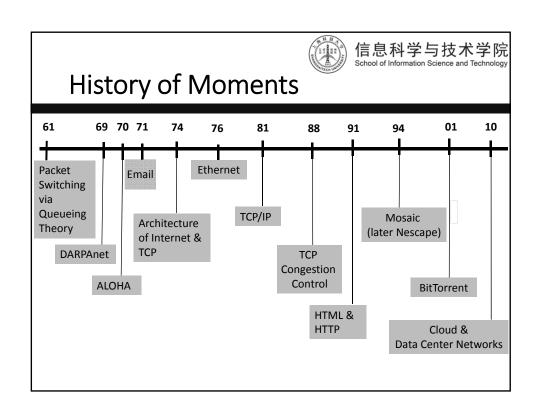
Layering Solves Problem

- Application layer doesn't know about anything below the presentation (or transport) layer, etc.
- Information about network is hidden from higher layers
- This ensures that we only need to implement an application once!



Outline

- I Complex Networks
- 2 Foundation of Internet
- 3 Performance Metrics of Networks
- 4 Protocols & Layers
- 5 History





Father of the Internet







Robert Kahn

Interesting video: https://www.youtube.com/watch?v=xA6Ccg4sdXc

2014 public lecture by Vincent Cerf & Robert Kahn for the 40th anniversary of Internet.

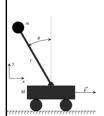


Homework: Simulation

1. Consider a sequence of packets passing through one link.

- Packets arrive one by one, and the time intervals between arriving packets satisfying a probability distribution with a mean equaling to 6 units.
- The bandwidth of the link is random, satisfying a probability distribution with a mean equaling to 0.2 packet per units.
- The size of the link buffer is 3(full buffer with three packets).
- Zero nodal processing delay & propagation delay.

You need to use python to simulate the behavior of such system with various probability distributions. Please record the average queueing delay, average delay, average throughput and loss rate. You are also required to make observations and provide intuitive explanations.



Feedback Control

by Boris Houska



Overview

Why do we need control systems?

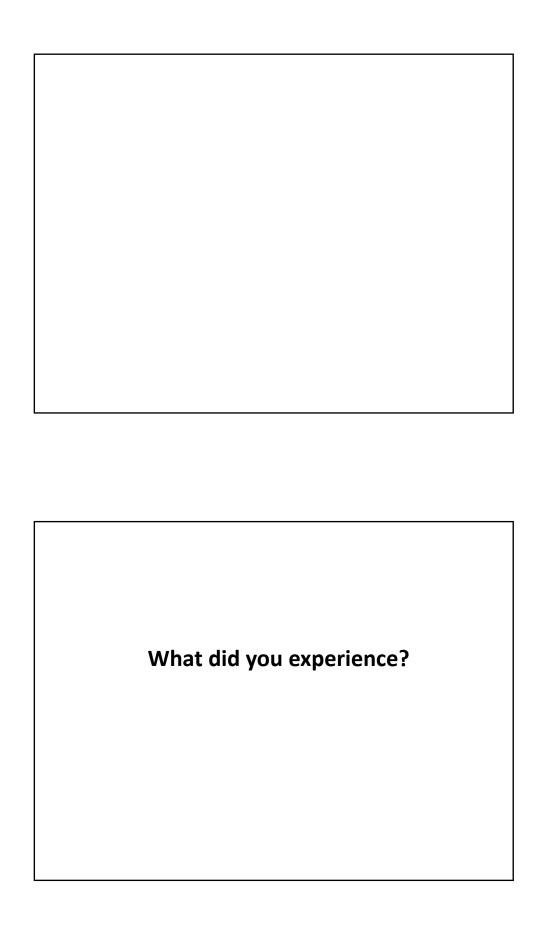
What is a feedback controller?

How do modern feedback controllers work?

Can you stand on one leg?



Now close your eyes...



Why is it difficult to keep your balance with closed eyes?

Basic Question of Control

Sensors ("Eyes")

- Cameras
- Radar
- GPS
- Inertial sensors (IMU)
- Pressures
- Temperatures
- ...

Basic Question of Control

Sensors ("Eyes")

- Cameras
- Radar
- GPS
- Inertial sensors (IMU)
- Pressures
- Temperatures
- ...

Actuators ("Muscles")

- Motors
- Flaps
- Valves
- Propellers
- Heaters
- Pumps
- ...

Basic Question of Control

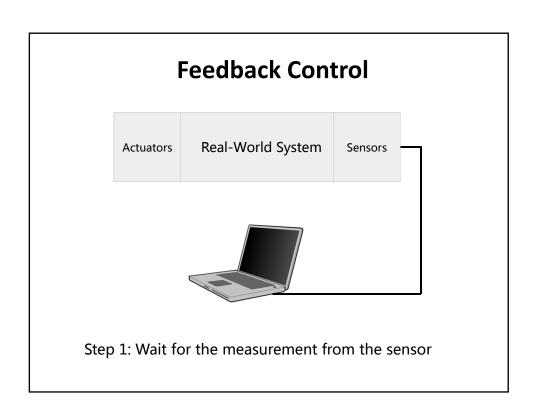
Sensors ("Eyes")

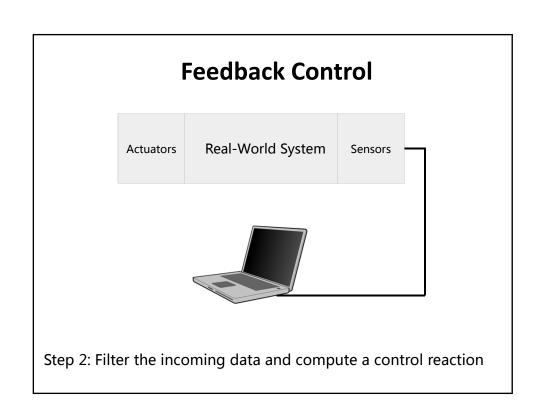
- Cameras
- Radar
- GPS
- Inertial sensors (IMU)
- Pressures
- Temperatures
- ..

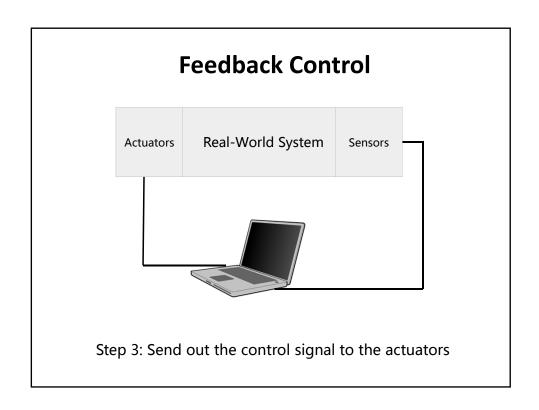
Actuators ("Muscles")

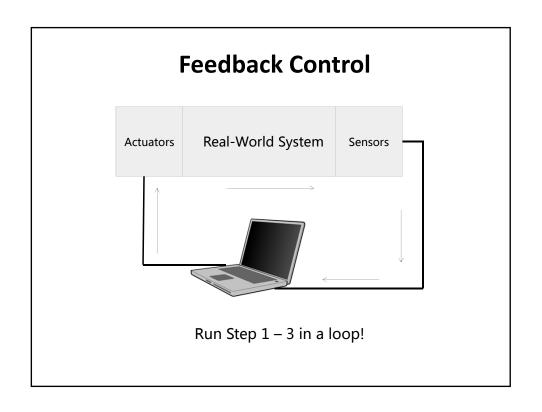
- Motors
- Flaps
- Valves
- Propellers
- Heaters
- Pumps
- ...

How to connect?









Zillions of Applications

Airplanes start and land using auto-pilots

Most chemical production process are optimized and automated

Automatic heating control in building saves huge amounts of energy

... there are many many more !!!











(source: wikipedia

Quiz



What are the actuators?

Quiz



What are the sensors?

Quiz



What are the actuators?

Quiz

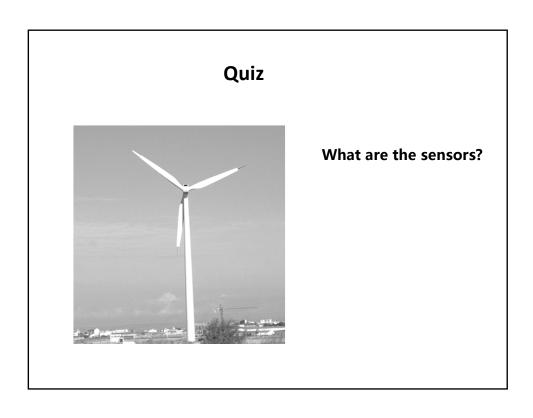


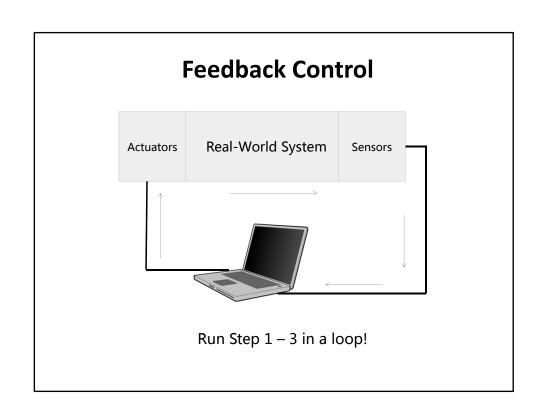
What are the sensors?

Quiz



What are the actuators?





Models of Systems

- Most systems that we study and investigate are linear systems
- Non-linear systems are controlled typically by first linearizing them
- What is a linear system? $x_1(t) \rightarrow y_1(t)$, $x_2(t) \rightarrow y_2(t)$
- $a_1 x_1(t) + a_2 x_2(t)$ -> output $y(t) = a_1 y_1(t) + a_2 y_2(t)$
- What is a linear time-invariant system?

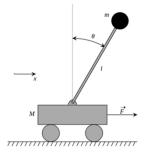
input x(t) -> output y(t) delayed input $x(t-T_d)$ -> delayed output $y(t-T_d)$

How to model a linear system?
 impulse response h(t) or transfer function H(f)

Goal of this lecture

Learn how to design a controller with a few lines of Python code

Motivating Example: Inverted Pendulum



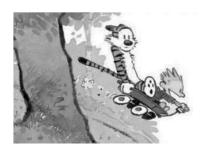
Can we bring the pendulum to its inverted position?

Sensor: Camera measuring positions

and velocities

Actuator: motor adjusting the force F

Potential and Kinetic Energy



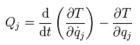
Potential energy:

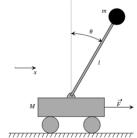
$$V = mgh$$

Kinetic energy:

$$T = \frac{1}{2}mv^2$$

Energy of the Inverted Pendulum





Potential energy:

$$V = mgl\cos(\theta)$$

Kinetic energy:

$$T = \frac{1}{2}M\dot{x}^2 + \frac{m}{2}\left(\dot{x}^2 + 2l\dot{x}\dot{\theta}\cos(\theta) + l^2\dot{\theta}^2\right)$$

Notation: dot above variable means derivative w.r.t. time

Equations of Motion

Lagrange function:

$$L = T - V$$

Lagrange formalism:

$$\frac{\mathrm{d}}{\mathrm{d}t}\frac{\partial L}{\partial \dot{x}} - \frac{\mathrm{d}L}{\mathrm{d}x} = (M+m)\ddot{x} + ml\cos(\theta)\ddot{\theta} - ml\sin(\theta)\dot{\theta}^2 = F$$

$$\frac{\mathrm{d}}{\mathrm{d}t} \frac{\partial L}{\partial \dot{\theta}} - \frac{\mathrm{d}L}{\mathrm{d}\theta} = ml^2 \ddot{\theta} + ml \cos(\theta) \ddot{x} + mgl \sin(\theta) = 0$$

Equations of Motion

Lagrange function:

$$L = T - V$$

Lagrange formalism: Inertia

$$\frac{\mathrm{d}}{\mathrm{d}t} \frac{\partial L}{\partial \dot{x}} - \frac{\mathrm{d}L}{\mathrm{d}x} = \underbrace{(M+m)\ddot{x} + ml\cos(\theta)\ddot{\theta}}_{-ml\sin(\theta)\dot{\theta}^2} = F$$

$$\frac{\mathrm{d}}{\mathrm{d}t} \frac{\partial L}{\partial \dot{\theta}} - \frac{\mathrm{d}L}{\mathrm{d}\theta} \underbrace{ml^2\ddot{\theta} + ml\cos(\theta)\ddot{x}}_{-ml\sin(\theta)} + mgl\sin(\theta) = 0$$

Equations of Motion

Lagrange function:

$$L = T - V$$

Lagrange formalism: Inertia centrifugal force $\frac{\mathrm{d}}{\mathrm{d}t}\frac{\partial L}{\partial \dot{x}} - \frac{\mathrm{d}L}{\mathrm{d}x} = \underbrace{(M+m)\ddot{x} + ml\cos(\theta)\ddot{\theta}}_{}\underbrace{ml\sin(\theta)\dot{\theta}^2}_{} = F$

$$\frac{\mathrm{d}}{\mathrm{d}t}\frac{\partial L}{\partial \dot{\theta}} - \frac{\mathrm{d}L}{\mathrm{d}\theta} \underbrace{ \left\{ ml^2 \ddot{\theta} + ml \cos(\theta) \ddot{x} \right\} + mgl \sin(\theta)}_{} = 0$$

Equations of Motion

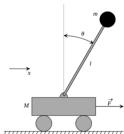
Lagrange function:

$$L = T - V$$

Lagrange formalism: Inertia centrifugal force
$$\frac{\mathrm{d}}{\mathrm{d}t}\frac{\partial L}{\partial \dot{x}} - \frac{\mathrm{d}L}{\mathrm{d}x} = \underbrace{(M+m)\ddot{x} + ml\cos(\theta)\ddot{\theta}}_{}\underbrace{(ml\sin(\theta)\dot{\theta}^2)}_{} = F$$

$$\frac{\mathrm{d}}{\mathrm{d}t}\frac{\partial L}{\partial \dot{\theta}} - \frac{\mathrm{d}L}{\mathrm{d}\theta} \underbrace{(ml^2\ddot{\theta} + ml\cos(\theta)\ddot{x})}_{} \underbrace{(mgl\sin(\theta))}_{} = 0$$
 gravity

Simplified Equations of Motion



For small excitation angles:

$$\sin(\theta) \approx \theta \quad \cos(\theta) \approx 1 \quad \dot{\theta}^2 \approx 0$$

$$\ddot{x} = \frac{F}{M} - \frac{mg}{M}\theta$$

$$\ddot{\theta} = -\frac{F}{Ml} + \frac{M+m}{M}\frac{g}{l}\theta$$

Linear Control System

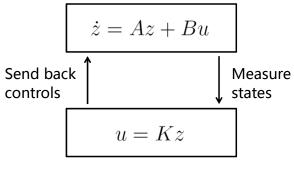
$$\dot{z} = Az + Bu$$
 Measure states

Linear Control System

$$\dot{z} = Az + Bu$$
 Measure states
$$u = Kz$$

Linear feedback law

Linear Control System



Linear feedback law

Inverted Pendulum in Standard Form

$$z = \begin{pmatrix} x \\ \theta \\ \dot{x} \\ \dot{\theta} \end{pmatrix} \qquad \begin{array}{l} \text{Position of trolley} \\ \text{Angle of the pendulum} \\ \text{Velocity of the trolley} \\ \text{Angular velocity of the pendulum} \\ \end{array}$$

System matrices:

$$A = \begin{pmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & -\frac{mg}{M} & 0 & 0 \\ 0 & \frac{M+m}{M} \frac{g}{l} & 0 & 0 \end{pmatrix} \qquad B = \begin{pmatrix} 0 \\ 0 \\ \frac{1}{M} \\ -\frac{1}{Ml} \end{pmatrix}$$

Matrices (Arrays) and Linear Algebra

$$\dot{z} = Az + Bu$$

A, B=? Simple way to represent coefficients

z=? Vector of important information (states)

2 = ? Vector of state changes

Az= how current states impact future state changes?

Bu= how control signals impact future state changes

Matrix Representation of Linear Equations

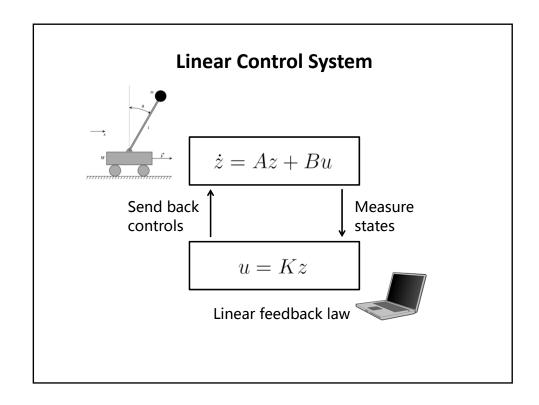
$$y_{1} = -3x_{1} + x_{2} + 1.5x_{3} + u_{1}$$

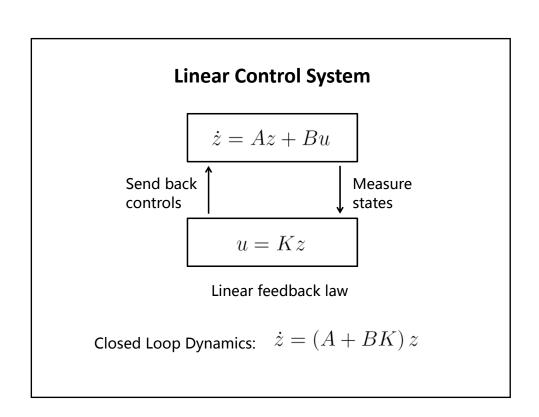
$$y_{2} = 2x_{1} - 4x_{2} + u_{1} - u_{2}$$

$$\mathbf{y} = \begin{bmatrix} y_{1} \\ y_{2} \end{bmatrix}, \quad \mathbf{u} = \begin{bmatrix} u_{1} \\ u_{2} \end{bmatrix} \quad \mathbf{x} = \begin{bmatrix} x_{1} \\ x_{2} \\ x_{3} \end{bmatrix}$$

$$A = \begin{bmatrix} -3 & 1 & 1.5 \\ 2 & -4 & 0 \end{bmatrix} \quad B = \begin{bmatrix} 1 & 0 \\ 1 & -1 \end{bmatrix}$$

$$\mathbf{y} = A\mathbf{x} + B\mathbf{u}$$





Closed Loop Dynamics in the State Space

Closed-loop trajectory satisfies

$$\dot{z} = (A + BK) z$$

Explicit solution

$$z(t) = e^{(A+BK)t}z(0)$$

Simulation of Closed-Loop Dynamics in Python

Math-Syntax:

$$z(t) = e^{(A+BK)t}z(0)$$

Python Syntax:

$$z = la.expm((A+B*K)*t)*z0;$$

Simulation of Closed-Loop Dynamics in Python

Math-Syntax:

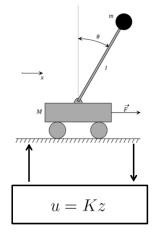
$$z(t) = e^{(A+BK)t}z(0)$$

Python Syntax:

$$z = la.expm((A+B*K)*t)*z0;$$

Simulates linear closed-loop systems with 1 line of code!

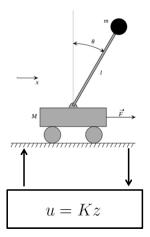
What happens for K = 0?



$$\ddot{x} = \frac{F}{M} - \frac{mg}{M}\theta$$

$$\ddot{\theta} = -\frac{F}{Ml} + \frac{M+m}{M} \frac{g}{l}\theta$$

What happens for K = 0?



$$\ddot{x} = \frac{F}{M} - \frac{mg}{M}\theta$$

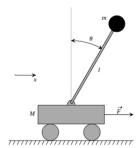
$$\ddot{\theta} = -\frac{F}{Ml} + \frac{M+m}{M} \frac{g}{l}\theta$$

For K = 0 we apply no force u = F = 0

unstable!

Linear feedback law

How to choose K?



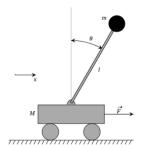
$$\ddot{x} = \frac{F}{M} - \frac{mg}{M}\theta$$

$$\ddot{\theta} = -\frac{F}{Ml} + \frac{M+m}{M} \frac{g}{l}\theta$$

Intuition:

If angle positive, choose F positive
If angle negative, choose F negative

How to choose K?



$$\ddot{x} = \frac{F}{M} - \frac{mg}{M}\theta$$

$$\ddot{\theta} = -\frac{F}{Ml} + \frac{M+m}{M} \frac{g}{l}\theta$$

Intuition:

Let's try the choice $\ K=\left(\,0,\,b,\,0,\,0\,\right)\$ with $\ b>(m+M)g$

Associated control law: $F = Kz = b\theta$

Let's implement this!

```
def simulate( x0, K ):  # inputs: initial values, control gain
  import numpy    as np
  import scipy.linalg as la

M = 0.4;  # mass of the trolley (in kg)
  m = 0.1;  # mass of the pendulum (in kg)
  l = 0.2;  # length of the pendulum (in m)
  g = 9.81;  # gravitational constant (in m/s^2)
```

Let's implement this!

```
def simulate( x0, K ):  # inputs: initial values, control gain
  import numpy    as np
  import scipy.linalg as la

M = 0.4;  # mass of the trolley (in kg)
  m = 0.1;  # mass of the pendulum (in kg)
  l = 0.2;  # length of the pendulum (in m)
  g = 9.81;  # gravitational constant (in m/s^2)
```

Guidelines:

- Give intuitive names to variables and functions
- Don't forget to comment your code

Let's implement this!

Let's implement this!

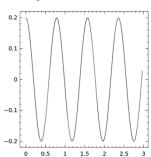
20 lines of self contained code to simulate a closed-loop system for an inverted pendulum

Closed-Loop Simulation Example

$$K = (0, b, 0, 0)$$
 $b > (m + M)g$

Trolley position [m] versus time [s]

0.14 0.12 0.08 0.06 0.04 0.02 0 Angle [rad] versus time [s]



Pendulum does not fall down, but oscillates

How to choose K?

$$\ddot{x} = \frac{F}{M} - \frac{mg}{M}\theta$$

$$\ddot{\theta} = -\frac{F}{Ml} + \frac{M+m}{M} \frac{g}{l}\theta$$

Intuition:

We want to "reduce" the oscillation

Strategy:

If angular velocity positive, increase F

If angular velocity negative, decrease F

Closed-Loop Simulation Example

$$K = (0, b, 0, d)$$

$$K = (0, b, 0, d)$$
 $b > (m + M)g$ $d > 0$

Trolley position [m] versus time [s]

Angle [rad] versus time [s]

Pendulum stabilized at inverted position, but trolley drifts away

How to choose K?

$$\ddot{x} = \frac{F}{M} - \frac{mg}{M}\theta$$

$$\ddot{x} = \frac{F}{M} - \frac{mg}{M}\theta$$

$$\ddot{\theta} = -\frac{F}{Ml} + \frac{M+m}{M}\frac{g}{l}\theta$$

Goal of our lab exercise:

We do not only want to stabilize the pendulum, but also the trolley.

How to choose K?

$$\ddot{x} = \frac{F}{M} - \frac{mg}{M}\theta$$

$$\ddot{\theta} = -\frac{F}{Ml} + \frac{M+m}{M} \frac{g}{l}\theta$$

Goal of our lab exercise:

We do not only want to stabilize the pendulum, but also the trolley.

Strategy:

$$K = (a, b, c, d)$$

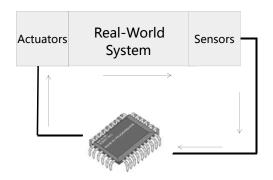
Tune all four coefficients!

Modern Feedback Control

- Nowadays, we do not tune controllers "by hand"
- Instead: use state-of-the-art optimization algorithms

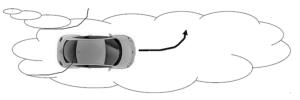
Modern Feedback Control

- Nowadays, we do not tune controllers "by hand"
- Instead: use state-of-the-art optimization algorithms
- Small optimization problems can be solved within microseconds and on embedded hardware



Embedded Optimization

Idea: always look a bit into the future.





Brain predicts and optimizes: e.g. slow down **before** curve

Use repeated computation of optimal controls for feedback control!

Why Optimization + Control ?



Control objective:
Minimize time

Why Optimization + Control ?



Control objective:

Minimize fuel consumption/ safe energy

(subject to other tasks)

Why Optimization + Control ?



Control objective:
Maximize energy

Take Home Points

• Feedback control connects sensors and actuators



We tuned and simulated a closed loop system with 20 lines of code

Take Home Points

• With minor modifications we could use these 20 lines of code to tune and simulate an airplane autopilot



Take Home Points

• With minor modifications we could use these 20 lines of code to tune and simulate an airplane autopilot



Even simple controllers operate faster and more accurately than any human pilot

Take Home Points

• With minor modifications we could use these 20 lines of code to tune and simulate an airplane autopilot



Even simple controllers operate faster and more accurately than any human pilot

Just a few lines of code, no joke!

Take Home Points

Modern control methods use optimization algorithms