

Information Engineering

Options Talk

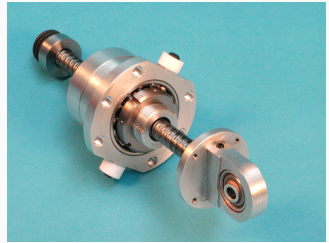
Dr Fumiya Iida (fi224)

21 February 2017

IIA Information Engineering broadly covers:

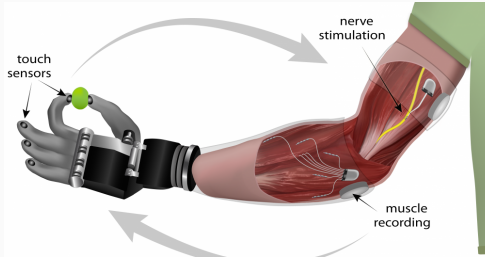
- Control
- Signal Processing
- Information Theory
- Communications
- Inference & Machine Learning

Control



The inerter mechanical device in Formula One racing

Control



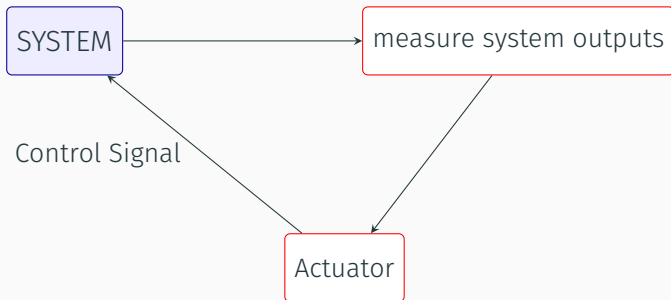
Control in biomedical engineering



In the smart grid

Control Engineering

We get the system to behave in a desired way by:



Part II control modules discuss how to address various challenges within this framework

- Everything is implemented digitally!
Discrete time systems, z-transforms, Fast Fourier Transform and Digital Filters (3F1)

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- Optimal operation is desirable
Feedback system design, Optimal control (3F2, 4F3)

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Feedback system design, Optimal control (3F2, 4F3)
- Fundamental limitations (4F1), nonlinear control (4F2)

It's not all hard Maths!

Modern applications of Control Engineering include:

- Control and dynamics in F1 (McLaren)
- Optimization and control in modern power systems and smart grids
- Motor control of bio-inspired musculoskeletal robots (4M20 Robotics)
- Green aircraft design: stability and control
- Reactivity control of small modular reactors with neutron absorber rods

3F1 Signals and Systems

- Discrete-time signals and systems
- Digital Filters and Fast Fourier Transform
- Continuous-time Random Processes

3F1 is a key module for **control, signal processing, and communications**

3F2 Systems and Control

- State-space models
- Feedback System Design
- State Estimation
- Control in a state-space framework

Signal Processing

Example: Image Denoising

Noisy image



Denoised image

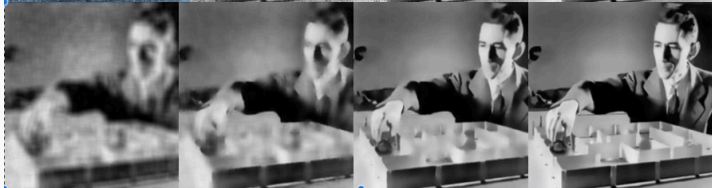
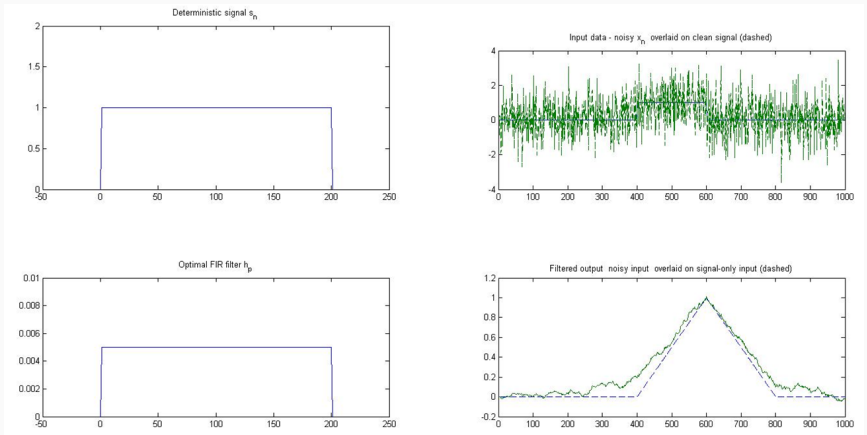


Image credit: A. Montanari (Stanford)

Signal Processing

Optimal detection of signal buried in noise – Matched Filter
3F3 (Applications in Radar, Communications, ...)



3F3 Statistical Signal Processing

- Modelling and Analysis of random, or 'noise-like', signals and systems
- Effects of noise and randomness in digitised signals

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- Modelling and Analysis of random, or 'noise-like', signals and systems
- Effects of noise and randomness in digitised signals
- Module starts with advanced treatment of probability fundamentals
- This leads into the new topic of random processes: *how to characterise random signals through autocorrelation functions, power spectrum ...*
- The second 8 lectures discuss **signal detection, estimation, and inference** using Wiener filters, matched filters, likelihood and Bayesian modelling

This module is fundamental to Communications, Speech & Image Processing, Machine Learning, (Big) Data Analysis, Computer Vision

3F3 Statistical Signal Processing:

- Runs alongside 3F1 (co-requisite) and leads into 3F8 (Inference) in Lent
- Also supports 3F4 (Data Transmission) and 3F7 (Information Theory)

3F1 Signals and Systems:

- Discrete-time signals and systems
- Digital Filters and Fast Fourier Transform
- Continuous-time Random Processes

3F1 + 3F3

Essential signal processing theory & techniques that are used in: speech/audio/video compression, wireless communication, echo cancellation, ...

Gravitational Wave Detection by LIGO



Analysis of LIGO data used classic signal processing techniques (FFTs, Power Spectral Density, Matched Filtering 3F1, 3F3)

https://lsc.ligo.org/s/events/GW150914/GW150914_tutorial.html

Information Theory and Communications

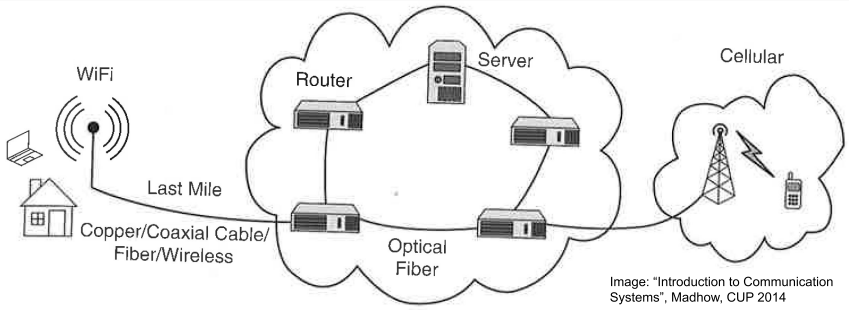
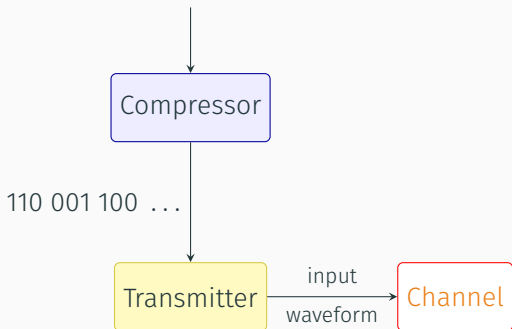


Image: "Introduction to Communication Systems", Madhow, CUP 2014

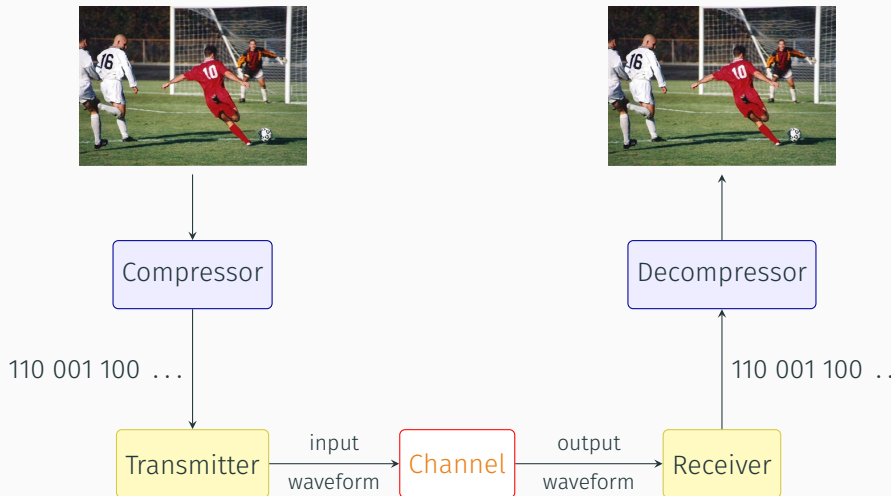
An Point-to-Point Communication System



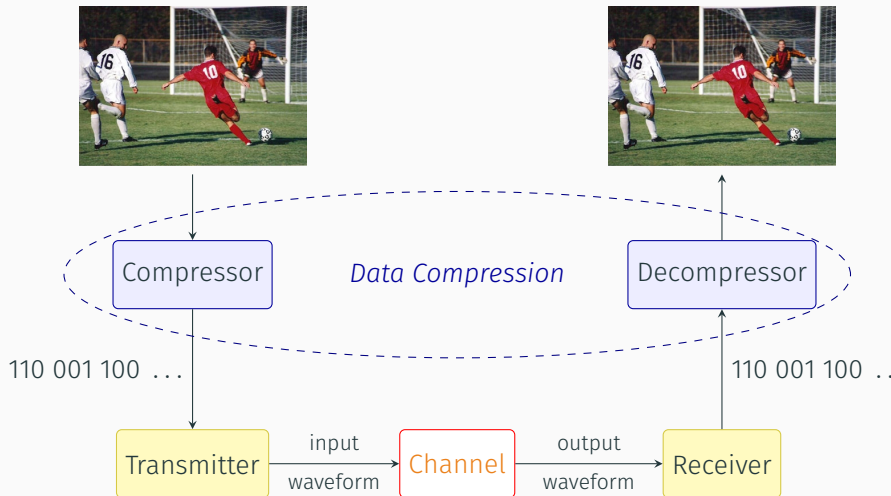
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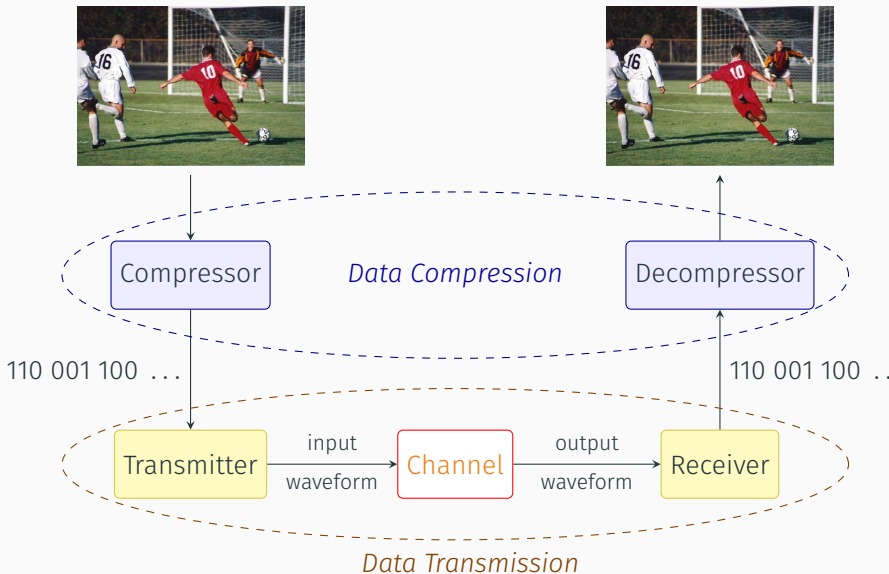
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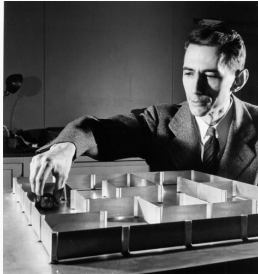


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An Point-to-Point Communication System



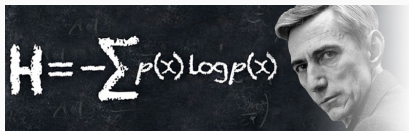


Claude Shannon, in 1948, showed how to quantify information using probability

Answered two fundamental questions:

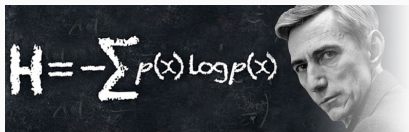
- Given a source of data, how much can you **compress** it?
- Given a noisy communication channel, what is the maximum rate at which you can **reliably transmit** data?

3F7 Information Theory and Coding


$$H = -\sum p(x) \log p(x)$$

- How to quantify information — using *Entropy*, a measure of uncertainty
- What is the fundamental limit of data **compression**?
- The fundamental limit of reliable data **transmission**?

3F7 Information Theory and Coding



- How to quantify information — using *Entropy*, a measure of uncertainty
- What is the fundamental limit of data **compression**?
- The fundamental limit of reliable data **transmission**?

3F7 also covers *practical* techniques to attain optimal limits:

- Practical Data **Compression** algorithms (Huffman coding, Arithmetic Coding)
- State-of-the-art techniques for **Error Correction** (Linear Codes, Sparse Graph codes)

Data Compression vs Error Correction

Compression

Data is compressed by squeezing out redundancy from the data, e.g.

Th_ onl_ wa_ to ge_ ri_ of a tempta_____ is to yie__ to it

PPM compression achieves 1.5 bits per character of English text

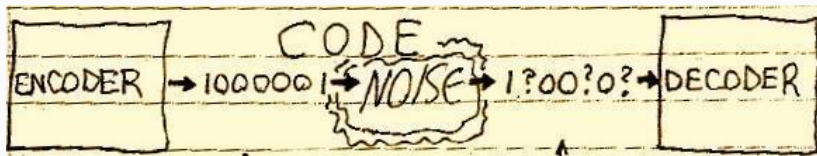
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Error Correction



Encode information by adding redundancy in a controlled way, so that it can be *decoded* from the noisy output at the receiver

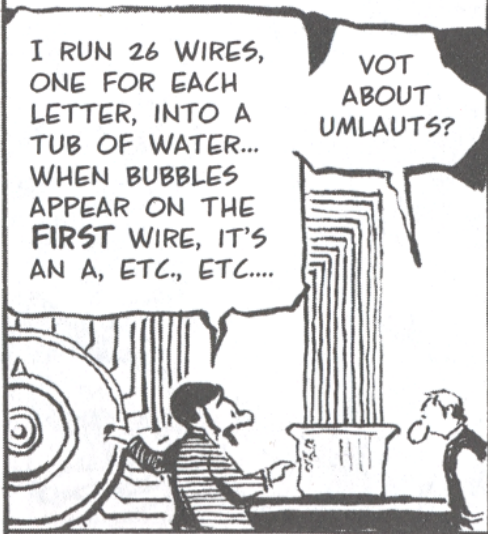


How do we transmit the bits across a real physical channel?

STARTING IN THE EARLY 1800S,
INVENTORS TRIED SEVERAL SYS-
TEMS FOR SENDING ELECTRIC
SIGNALS DOWN WIRES.

I RUN 26 WIRES,
ONE FOR EACH
LETTER, INTO A
TUB OF WATER...
WHEN BUBBLES
APPEAR ON THE
FIRST WIRE, IT'S
AN A, ETC., ETC....

VOT
ABOUT
UMLAUTS?



3F4 Data Transmission



How do we transmit the bits across a real physical channel?

3F4 deals with:

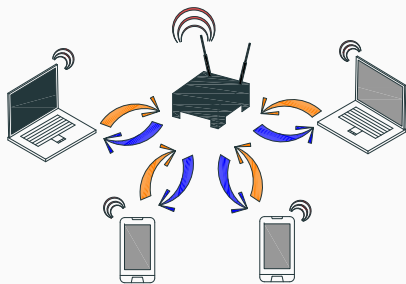
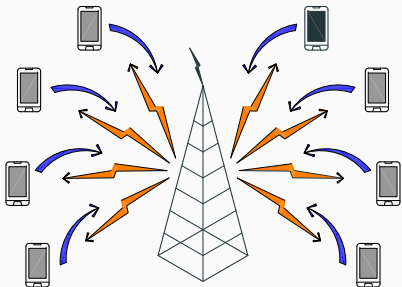
How to design good input waveforms tailored to the channel?

How to detect the bits from the output waveform?



Some applications in your pocket

- Orthogonal Frequency Division Multiplexing: modulation technique used in 4G (3F4)
- Transform coding and MP3 compression (digital filters, FFTs, entropy coding 3F1, 3F7)



Dealing with mobility & channel uncertainty in wireless networks
(frequency diversity & antenna diversity 3F4, 4F5)

Inference & Machine Learning

3F8 Inference

- Non-linear regression
- **Classification**
- **Clustering**
- Dimensionality reduction
- Sequence models

Techniques:

- Optimisation
- Monte Carlo
- Bayesian methods

classification



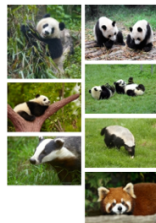
Predicted Tags



cluster 1

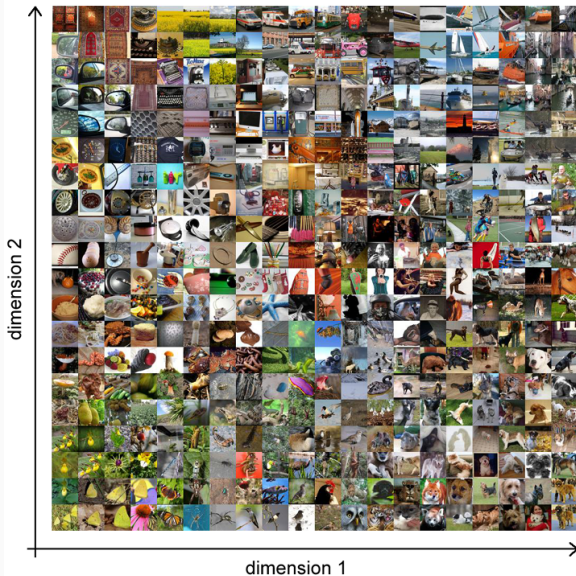


cluster 2



www.clarifai.com

images embedded in 2D



Dimensionality Reduction

3F8 Inference

- Non-linear regression
- Classification
- Clustering
- Dimensionality reduction
- **Sequence models**

Techniques:

- Optimisation
- Monte Carlo
- Bayesian methods

sequences of characters

私はでそれを信じて

predict the next character?

Thomas Bayes



represent uncertainty using
probability distributions

Pre-Requisites and Co-Requisites

Third-year Modules:

3F1 (M) Signals and Systems (no pre-req)

3F3 (M) Statistical Signal Processing (3F1 is a co-requisite)

3F7 (M) Information Theory and Coding (no pre-req)

3F2 (L) Systems and Control (no pre-req)

3F4 (L) Data Transmission (3F1 is a pre-req)

3F8 (L) Inference (3F3 is a pre-requisite)

Fourth Year Modules

Fourth Year Modules

4F1 Control System Design

4F2 Robust and Nonlinear Systems and Control (3F2 pre-req.)

4F3 Optimal and Predictive Control

4F5 Advanced Communications and Coding (3F7 pre-req.)

4F7 Statistical Signal Analysis (3F7 pre-req.)

4F8 Image Processing and Image Coding (3F1 pre-req.)

4F10 Deep Learning and Structured Data

4F12 Computer Vision

4F13 Probabilistic Machine Learning

4M20 Robotics

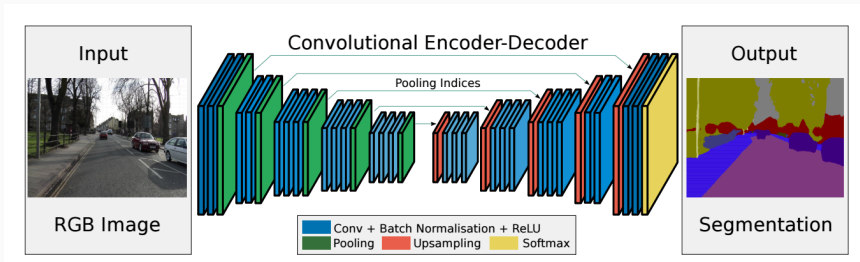


Image Segmentation using convolutional neural nets

You can try a real-time demo at:
<http://mi.eng.cam.ac.uk/projects/segnet/>

Algorithms that use input data to build models and make predictions or decisions

Search engines, spam filtering, computer vision ...

Machine learning algorithm learns to play
Atari Breakout

Part IIA Information Engineering

- Control (3F1, 3F2)
- Signal Processing (3F1, 3F3)
- Information Theory (3F7)
- Communications (3F7, 3F4)
- Inference & Machine Learning (3F3, 3F8)

These slides will be available at:

<http://mi.eng.cam.ac.uk/Main/FI224>

Or email me (fi224)