

# Summary

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# Big Picture

# Why Study This Topic?

- Almost any problem you can think of can be cast as 1 of 2 types of problems
  - Information extraction
  - Action selection
  - Data generation
- These problems can be put into a general framework
  - Pre processing                      - optional domain specific processing
  - Function                              - input output mapping
  - Post processing                      - optional domain specific processing
- xNNs do a great job at input output mapping for many problems of interest
  - xNNs are universal approximators so if a mapping exists they can approximate it (under some relatively mild technical conditions)
  - So potentially they can be used to solve almost any problem you can think of
  - And in practice they're a key component of the top performing methods for many important tasks

# What's The Problem With This?

- We don't know the input output mapping that we need
  - It's not given what the optimal network structure is to approximate the input output mapping (design) and how to find the parameters that control the network structure mapping (training)
  - It's not given how to practically and efficiently realize network structures (implementation)
  - It's not given what different applications require in terms of pre and post processing and along with different design and training strategies (vision, speech, language, games, art, ...)

# The Purpose Of This Course

- So the obvious purpose of this course is
  - To review the basic math needed for pre processing, xNNs and post processing
  - To learn network design strategies to accomplish different goals
  - To learn end to end network training from data
  - To learn efficient network implementations
  - To learn specific strategies for common applications
- However, an equally important purpose of this course is to learn how to learn
  - The last paper isn't written, there are many more each day
  - You have a good base to work from
  - You have the potential to understand new ideas from others and create new ideas yourselves
  - Your ability to do this improves with practice

# Topics We Covered

# We Covered A Lot Of Material

That's by design: if you want to do meaningful work in this area, it helps to know a lot about a lot

- The first 3 topics in the linear algebra class were sets, fields and vectors
- The last 4 flagship topics in the application classes were object based image segmentation, language to language translation, speech to text transduction and playing Atari and Go
- And somewhere in the middle we talked a little bit about semiconductor device physics

# The Course Was Divided Into 3 Sections

- Math
  - Linear algebra
  - Algorithms
  - Probability / info theory
  - Calculus
  - Analysis
- Networks
  - Design
  - Training
  - Implementation
- Applications
  - Vision
  - Language
  - Speech
  - Games
  - Art



# Homework

- 1 purpose was to complement in class material through reading and expand theory through problems
- Another purpose was to provide an opportunity for practical experience
  - PyTorch and the Colab environment by default
  - Templates to provide a starting point and connect to theory
  - A gradual ramping up in terms of complexity to build confidence
  - Some amount of training pain to better appreciate why we discuss implementations

# Topics We Didn't Cover

# There's A Lot Of Material We Didn't Cover

- Not sure if this is good or bad
  - Perhaps it's just the reality of a large field viewed through the lens of a 1 semester class
- Realize that these uncovered topics are everywhere
  - Within the topics areas that we covered
  - Topics that we didn't cover
- Realize that there are new developments every day in this field
- Again, your only hope is to learn how to learn
  - Hopefully this course contributes positively to your ability to do that
  - In general, the more stuff you know the easier it is to learn something new (as it's likely similar to something you already know)

# Ex: Networks

- While it feels like we covered a lot of network topics
  - Layers, design, training, implementation
- There's a lot more that we didn't cover
  - Networks with an external memory developed in the context of language applications, interesting to think about in the context of other applications
  - Networks for processing non Euclidean data
- On the upside, the network information presented during this semester should provide a good basis for understanding new network designs

# Ex: Vision, Language, Speech And Games

- While it feels like we covered a lot of vision topics
  - Classification, pixel segmentation, multiple object detection, object based segmentation, depth estimation and motion estimation
- There's a lot more that we didn't cover
  - Variations within the topics that we covered
  - Other topics that we didn't cover like point clouds, facial recognition, handwriting recognition, optimizing object boundaries, ...
- The same observation is true for language, speech and games
- On the upside, the vision, language, speech and games info presented during this semester should provide a good basis for understanding different topics in these fields

# Ex: Art

- A lot of art related work is leading to interesting network structures for generative systems
  - Style transfer
  - GANs
  - Music, images and video
- Maybe I'll add some additional info on this at some point in time
  - Right now there's no way for me to fit it in the semester so it's at a lower priority

# The Future

# Interpolation

These are pretty easy to guess

- Gradually more and more theoretical questions are answered
  - People will quit complaining about a lack of theory
  - Note: there's already a lot of theory and the complaining seems unreasonable even now
- Everything gets better in proportion to gradual advances in computing
  - Maybe slightly faster than process technology but not radically faster
  - Will get a boost for going to lower precisions and building more efficient architectures
- Basic methods are applied to more and more fields
  - Just like there was computational <science> that became <science>, there will be machine learning for <science> that eventually becomes commonplace enough that it's again called <science>
  - It's likely that some developments in some of these fields will propagate back to the core and become used in many fields
  - In general there will be convergence towards a few core techniques with some specialization (e.g., pre and post processing) around them



# Extrapolation

These are comments and guesses with a much higher variance and by choice are somewhat constrained

- Theory
  - It would radically change what we're capable of if algorithms are developed with processing complexity proportional to the information rate vs the data rate
- Computation
  - Over the course of a longer time scale what happens if compute capability increased by 1000? 1000000? 1000000000?
  - What happens if a different computing paradigm (maybe quantum) becoming commonplace and let's us design algorithms that solve complementary problems and providing another fundamental step up in AI
- Applications
  - We figure out how to handle more and more complex problems by putting together more and more simple problems
  - To an outsider this starts to feel like general intelligence vs specific / isolated problem solving
- Humanity
  - We become much more closely integrated with AI; everyone's hand is forced in this direction or they're left out / disadvantaged
  - We observe a person over the course of their life and get good at simulating them
  - We question our own intelligence

# Your Place In It

- A few suggestions
  - Add a new hobby every few years
  - Figure out what's at the intersection of what you love to do and what you're good at
  - Do that
  - Be great (know you can be)
  - Don't be afraid to change

# Next Time

# Content Refinements

- Continue to improve the code examples and homework
  - Improve code examples for design, training, implementation, vision, language, speech, games and art
  - Figure out ways to improve training efficiency
- Continue to improve the slides
  - It would be nice to do some general notational cleanup in a few places
- Continue to look for ways to use the board vs slides whenever possible

# Consider An Expanded Offering

- I probably have enough material for 2 semesters
- Perhaps the ideal
  - Go a bit slower through the current material for 1.5 semesters
  - Take the last 0.5 semester to go in depth into a different topic each week

# One More Thing

# Thank You!

- You're all special to me and I appreciate you sticking with me though this
- I wish you all the best for happiness in your lives and careers
  - You're capable of great things
  - I look forward to seeing what you accomplish academically and professionally
- Keep in touch!

# The Final Slide

## Number of slides

Introduction	69
Linear algebra	137
Algorithms	33
Probability	97
Calculus	61
Design	205
Training	175
Implementation	298
Vision	137
Language	102
Speech	143
Games	109
Art	20
Summary	24

<b>Total</b>	<b>1610</b>
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