# 50.039 Theory and Practice of Deep Learning W12S3 – End and Review

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#### Some admin stuff

#### Final exam

- Content will be everything from W1 to W12 included.
- Similar format as MidTerm.
- Details for exam (location, assignments, etc.) to be sent via email.

#### **Big project**

No extension will be given (have to give grades to OSA!)

#### Some admin stuff

Conferences (to replace classes this week)

Paypal: Friday 22<sup>nd</sup>, 11.30am

Gojek: Thursday 21<sup>st</sup>, 10am

Over Zoom, links to be sent later.

Attendance is mandatory (unless you have clashes with other courses, in which case you will be excused).

## So this is the end

What's next?

#### Advanced attacks and Defense mechanisms (W6++)

Many more mechanisms when it comes to attacking and defending a Neural Network, e.g. new types of attacks, such as:

- Poisoning attacks (attempt to poison the dataset so the NN cannot retrain properly),
- Weights changes (attempt to change a small subset of the weights of the NN to prevent it from working in certain ways),
- Etc.

https://www.comp.nus.edu.sg/~reza/courses/cs6231/

#### Advanced word embedding and NLP problems (W5++ and W8++)

Many more mechanisms when it comes to embedding and language related problems.

- E.g. more advanced embeddings
- Typical tasks in NLP (chatbots, context propagation, sentiment analysis, translation, etc.)
- Go for the Term 8 NLP course!

https://istd.sutd.edu.sg/undergraduate/courses/50040-natural-language-processing

#### **Advanced Graph Neural Networks (W9++)**

We barely scratched the surface of Graph Theory. If you need to study a new math theory, let it be graph theory!

Good graph theory course here:

https://ocw.mit.edu/courses/mathematics/18-217-graph-theory-and-additive-combinatorics-fall-2019/

 More advanced problems and concepts on Graph Neural Networks in lectures 1-9 of course here:

https://www.cs.ox.ac.uk/teaching/courses/2020-2021/advml/

#### **Advanced Graph Neural Networks (W9++)**

Also, keep in mind that Neural Networks are graphs...

- So technically, we could build a Neural Network, which receives another Neural Network as its input...!
- What could be the uses for such a technique?
- Meta-learning? (i.e. training an AI to train another AI?!)
   <a href="https://machinelearningmastery.com/meta-learning-in-machine-learning/">https://machinelearningmastery.com/meta-learning-in-machine-learning/</a>

#### **Advanced Generative Models (W10++)**

- Advanced GANs, operating on other types of data than just images (sound, text, etc.)
- Very good online course here:

https://cs236g.stanford.edu/

#### **Advanced Reinforcement Learning (W11++)**

Barely scratched the surface about Reinforcement Learning.

- Currently considering to create a RL course at SUTD for Term 8.
   Thoughts?
- Otherwise, the reference course on RL is the one from David Sliver (the man behind AlphaGo!)

https://deepmind.com/learning-resources/-introduction-reinforcement-learning-david-silver

And <a href="https://www.davidsilver.uk/teaching/">https://www.davidsilver.uk/teaching/</a>

#### Advanced Interpretability (W12++)

- Rather an ongoing field in research at the moment.
- Not that many course out there, but worth keeping an eye out...

#### More concepts, problems and architectures on Computer Vision

Ask for the course materials of the Computer Vision Term 7 course for more advanced concepts on CV, such as:

- More advanced loss functions like triplet loss,
- Advanced architectures like siamese networks,
- Video data models,
- Etc.

https://istd.sutd.edu.sg/undergraduate/courses/50035-computer-vision

#### More concepts, problems and architectures on Computer Vision

 Also, always good to go for an image processing course to understand typical image transformation and problems out there.

```
https://www.coursera.org/learn/image-
processing?ranMID=40328&ranEAID=*GqSdLGGurk&ranSiteID=.GqSdL
GGurk-GV4LxEnPMuMd1.8y4AurRA&siteID=.GqSdLGGurk-
GV4LxEnPMuMd1.8y4AurRA&utm_content=10&utm_medium=partner
s&utm_source=linkshare&utm_campaign=*GqSdLGGurk
```

Bayesian and Statistical Learning (Variational AutoEncoders were 101, more on diffusion models).

A good entry point for Bayesian Deep Learning

https://medium.com/@ODSC/introduction-to-bayesian-deep-learning-f7568f524c90

Lectures 10-End

https://www.cs.ox.ac.uk/teaching/courses/2020-2021/advml/

#### A bit of advanced optimization and game theory never hurts...

- Especially when trying to optimize two cooperating or competing neural networks! (GANs, actor-critic, etc.)
- Great courses here:

https://oyc.yale.edu/economics/econ-159

And

https://online.stanford.edu/courses/soe-ycs0002-game-theory

## CUDA masters are the king of the world these days...

- BigTech companies are looking for experts that can help with machine learning and custom GPU implementations
- The most obvious way to learn is from Nvidia courses themselves, some give certifications, but it is an investment...

https://developer.nvidia.com/cud a-education-training



#### **CUDA Education & Training**

#### Accelerate Your Applications

Learn using step-by-step instructions, video tutorials and code samples.

- Accelerated Computing with C/C++
- Accelerate Applications on GPUs with OpenACC Directives
- Accelerated Numerical Analysis Tools with GPUs
- Drop-in Acceleration on GPUs with Libraries
- GPU Accelerated Computing with Python

## Could computing is also very valuable...

 Similarly, a certification in AWS or Microsoft Azure of Google Cloud for cloud computing machine/deep learning is of high value these days!



#### Quantum is the next best thing?

Quantum computers are expected to be the next big thing in Computer Science in general.

- This will also apply to AI/ML/DL...
- This means we will get to train larger networks, faster. (This is currently a limit for many applications these days).
- Picking up on quantum computing is never a bad idea (but careful, possibly the most difficult topic out there!)

https://towardsdatascience.com/dont-ask-what-quantum-computing-can-do-for-machine-learning-cc44feeb51e8

https://pennylane.ai/qml/whatisqml.html

#### More stuff

 Advanced Probability and Statistics (a.k.a. Statistical Learning) is always a great plus...

https://www.statlearning.com/

- Neuroscience should probably be part of any serious AI curriculum... [NeuroAI] Barron et al., "What insects can tell us about the origins of consciousness", 2015.
- Etc.

#### More stuff

- Using DL to solve complex differential equations.
   https://medium.com/swlh/artificial-intelligence-can-now-solve-a-mathematical-problem-that-can-make-researchers-life-easier-9602c869128
- General AI, i.e. designing an AI will full human cognitive capabilities (vision, hearing, speech, movement, etc.).
   <a href="https://www.forbes.com/sites/forbestechcouncil/2021/07/16/the-future-of-artificial-general-intelligence/?sh=c9223323ba99">https://www.forbes.com/sites/forbestechcouncil/2021/07/16/the-future-of-artificial-general-intelligence/?sh=c9223323ba99</a>
- Etc.

## The important message is...

Your learning should not stop after SUTD...

Keep learning to stay up to date, this is a very fast evolving field...

So, good luck on your continuing studies!

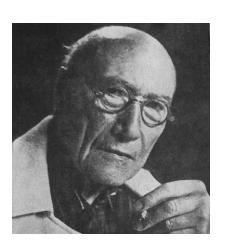
### More importantly

#### DL/AI is a very active and fastpaced field.

- Keep your watchlist of papers and authors up to date.
- I have mentioned researchers, which I believe are among the most notable influencers of the Deep Learning community.

 Will be adding some more on the next slides. "A good professor should have this constant concern: teaching his students how to continue without him."

- André Gide, Nobel Prize of Literature in 1947



- Demis Hassabis: Co-founder of DeepMind, AlphaGo. Several contributions in Reinforcement Learning. <a href="https://scholar.google.com/citations?hl=en&user=dYpPMQEAAAAJ">https://scholar.google.com/citations?hl=en&user=dYpPMQEAAAAJ</a>
- Alex Graves: Professor at University of Toronto. Several contributions in Reinforcement Learning.
   https://scholar.google.co.uk/citations?user=DaFHynwAAAAJ&hl=en
- Michael I. Jordan: Professor at UC Berkeley, co-inventor of LDA. https://scholar.google.com/citations?user=yxUduqMAAAAJ&hl=fr
- Terrence Sejnowski: Professor at UC San Diego, Boltzmann machines.
  - https://scholar.google.ca/citations?user=m1qAiOUAAAAJ&hl=en

- Peter Norvig: Director of Research at Google, co-author of the other Bible of Deep Learning <a href="https://scholar.google.com/citations?user=Ol0vcWgAAAAJ&hl=en">https://scholar.google.com/citations?user=Ol0vcWgAAAAJ&hl=en</a> <a href="http://aima.cs.berkeley.edu/">http://aima.cs.berkeley.edu/</a>
- Stuart Russell: Professor at UC Berkely, co-author of the other Bible of Deep Learning <a href="https://scholar.google.com/citations?user=20y30XYAAAAJ&hl=en">https://scholar.google.com/citations?user=20y30XYAAAAJ&hl=en</a>
- Francois Chollet: Researcher at Google. The man behind the Keras framework and Xception.
  - https://scholar.google.com/citations?user=VfYhf2wAAAAJ&hl=en

- Trevor Hastie: Professor at Stanford, co-autor of the Bible of Statistical Learning.
  - https://scholar.google.ca/citations?user=tQVe-fAAAAAJ&hl=enhttps://hastie.su.domains/ElemStatLearn/download.html
- Robert Tibshirani: Professor at Stanford, co-autor of the Bible of Statistical Learning. Inventor of the LASSO algorithm. https://scholar.google.ca/citations?user=ZpG cJwAAAAJ&hl=en
- Vladimir Vapnik: Retired Professor, inventor of SVMs and many other concepts. Worked with Yann LeCun at Facebook AI. https://scholar.google.com/citations?user=vtegaJgAAAAJ&hl=fr

- Fred Cummins: Professor at University College Dublin, contributions to LSTMs and NLP.
  - https://scholar.google.com/citations?user=E-vg2zQAAAAJ&hl=fr
- Andrej Karpathy: Former Director of AI at Tesla. Many contributions to Computer Vision (Imagenet) and NLP (RNNs).
   (Probably better to follow him than Elon Musk.)

   <a href="https://scholar.google.com/citations?user=l8WuQJgAAAAJ&hl=fr">https://scholar.google.com/citations?user=l8WuQJgAAAAJ&hl=fr</a>
- Li Fei-Fei: Professor at Stanford. Many contributions to Computer Vision (Imagenet). https://scholar.google.com/citations?user=rDfyQnIAAAAJ&hl=fr
- Pieter Abbeel: Professor at UC Berkeley, and a leading researcher in reinforcement learning and robotics.

https://scholar.google.com/citations?user=vtwH6GkAAAAJ&hl=en

- Anil K. Jain: Professor at Michigan State University. Many contributions to Computer Vision and Statistical Learning. <a href="https://scholar.google.com/citations?user=g-ZXGsAAAAJ&hl=fr">https://scholar.google.com/citations?user=g-ZXGsAAAAJ&hl=fr</a>
- Jitendra Malik: Professor at UC Berkeley. Many contributions to Computer Vision and Statistical Learning. <a href="https://scholar.google.com/citations?user=oY9R5YQAAAAJ&hl=fr">https://scholar.google.com/citations?user=oY9R5YQAAAAJ&hl=fr</a>
- **Sebastian Thrun**: **Stanford**, cool stuff on **robotics**. https://scholar.google.com/citations?user=7K34d7cAAAAJ&hl=fr
- Daphne Koller: CEO at InSitro, some cool courses on Coursera, she might be the co-founder of Coursera (?). https://scholar.google.com/citations?user=5lqe53lAAAAJ&hl=en

- Andrew Ng: Professor at Stanford, co-creator of Coursera. Has one of the best online courses on Deep Learning. https://scholar.google.com/citations?user=mG4imMEAAAAJ&hl=en
- Jeremy Howard: Research Scientist at University of San Francisco, a good scout for notable research papers on Twitter and TED talks. <a href="https://scholar.google.com/citations?user=ZWdEJ54AAAAJ&hl=en">https://scholar.google.com/citations?user=ZWdEJ54AAAAJ&hl=en</a>
- Yaser S. Abu-Mostafa: Professor at CalTech, one of the best professors for Deep Learning out there. <a href="https://dblp.org/pid/69/3008.html">https://dblp.org/pid/69/3008.html</a>
- Rachel L. Thomas: University of San Francisco, FastAI, some great TED conferences on AI and Deep Learning. https://scholar.google.com/citations?user=BDsAYUsAAAAJ&hl=en