

HOW TO DO DEEP LEARNING WITHOUT A PHD.

MATH FOR DEEP LEARNING:

- 1) Deep Learning, Ian Goodfellow (www.deeplearningbook.org) - Chapter 1 of this book covers Linear Algebra, Probability and Calculus. Short, Simple and Comprehensive.
- 2) Elements of Statistical Learning, Friedman - (web.stanford.edu/~hastie/Papers/ESLII.pdf) Comprehensive Guide to learning the math behind Machine Learning.
- 3) Math and Computation, Avi Wigderson (www.math.ias.edu/files/mathandcomp.pdf)

BASICS OF COMPUTER SCIENCE

- 1) Introduction to Algorithms by Cormen - Universal Guide to Learning the basics of computer science. What came first? Algorithms or Cormen? Nobody knows....

MATLAB

- 1) Just Joking, MATLAB sucks. We only use Python.
- 2) Python Course from Codecademy (<https://www.codecademy.com/learn/learn-python>)
- 3) Python Documentation (<https://docs.python.org/3/>)
- 4) On a side note, when learning any new package or framework, the best place to get started is the documentation. The second best place is an introductory blog (Medium is a good place).

STATISTICAL LEARNING/MACHINE LEARNING INTRO

- 1) Andrew Ng's Machine Learning Course (<https://www.coursera.org/learn/machine-learning>)
- 2) Machine Learning by ColumbiaX on EdX
- 3) Siraj Raval's Live Videos on YouTube - He explains complex problems in a simple way. His live videos are great. His video descriptions also have good learning resources.

*Our take: Watch a week of videos in Andrew Ng's course and then watch Siraj Raval's videos and read blogs to get a deeper understanding of the concepts. Then implement them as a project.

NEURAL NETWORKS/DEEP LEARNING

- 1) Andrew Ng's Deep Learning Specialisation Course (www.coursera.org/specializations/deep-learning) - Yes, his courses are *that* good.
- 2) Neural Networks and Deep Learning by Michael Nielsen (www.neuralnetworksanddeeplearning.com) - Lucid explanation of Neural Networks
- 3) Hinton's Neural Network Course (www.coursera.org/learn/neural-networks) - Godfather of Deep Learning. Need I say more?
- 4) Stanford Computer Vision Course - <http://cs231n.stanford.edu/>

- 5) Stanford NLP Course - <http://web.stanford.edu/class/cs224n/>
- 6) Reinforcement Learning by David Silver

GETTING STARTED WITH PROJECTS IN DEEP LEARNING

- 1) Fast.ai's courses are based around projects (<http://www.fast.ai>)
- 2) Siraj Raval's Youtube Channel has a lot of projects that you can practice
- 3) Sentdex's Youtube Channel also has a lot of projects
- 4) Machine Learning Mastery (www.machinelearningmastery.com)

SOME ESSENTIAL PACKAGES/Frameworks

- 1) Numpy - <http://www.numpy.org/>
- 2) Pandas - <https://pandas.pydata.org/>
- 3) Matplotlib - <https://matplotlib.org/>
- 4) Keras - <https://keras.io/>
- 5) NLTK - <http://www.nltk.org/>
- 6) OpenCV - <https://opencv.org/>
- 7) TensorFlow - <https://www.tensorflow.org/>
- 8) Sci-kit Learn - <http://scikit-learn.org/>

BLOGS

- 1) Math - <http://www.inference.vc>
- 2) NLP and lit surveys: <http://runder.io>
- 3) Almost everything AI&DL: <http://www.wildml.com>
- 4) Best journal hands down: <https://distill.pub>
- 5) Intuition behind DL architectures: <http://colah.github.io>
- 6) Learn ML concepts easily and quickly - <http://fastml.com/>

RESEARCH PAPERS

- 1) <http://deeplearning.net/reading-list/>
- 2) Google Brain - <https://research.google.com/teams/brain/>
- 3) Deep Mind - <https://deepmind.com/research/publications/>
- 4) Arxiv Sanity - <http://www.arxiv-sanity.com/>
- 5) Paper summaries: <http://blog.acoyer.org>

SYSTEMS/DEPLOYMENT/CLOUD

- 1) Architecture of Open Source Applications - <http://aosabook.org/en/index.html>
- 2) FloydHub - <https://www.floydhub.com/>
- 3) Google Cloud Platform - <https://cloud.google.com/>
- 4) Amazon Web Services - <https://aws.amazon.com/>

MACHINE LEARNING NEWS & COMPETITIONS

- 1) KDNuggets - <https://www.kdnuggets.com/>
- 2) Hacker News - <https://news.ycombinator.com/>
- 3) Indian Deep Learning Initiative (IDLI) & AIDL FB Page
- 4) Reddit - r/MachineLearning, r/learnmachinelearning, r/datascience
- 5) Twitter is where researchers share and discuss papers, concepts and ideas
- 6) ML Challenges, Competitions & Datasets - <https://www.kaggle.com/>