

Notes - Machine Learning

Friday, September 15, 2017 10:44 PM

Quick Overview

NVIDIA Blog - [What's the Difference Between Artificial Intelligence, Machine Learning, and Deep Learning?](#)

NVIDIA Webinar - [Deep Learning Demystified](#) (23 min)

Microsoft Azure: Data Science for Beginners Video Series

[The Five Questions Data Science can Answer](#) (5 min)

- 1) Is this A or B? - Classification
- 2) Is this weird? - Anomaly Detection
- 3) How much - or - How many? - Regression - Numerical Predictions
- 4) How is this organized? - Clustering
- 5) What should I do next? - Reinforcement Learning

[Is Your Data Ready for Data Science?](#) (5 min)

- Data Must be:
 - o Relevant
 - o Connected
 - o Accurate
 - o Enough to work with

[Ask a Question You can Answer With Data](#) (4 min)

- Ask a very specific that can be answered with a name or number
- How you ask a question can give you direction on what algorithm to use

[Predict an Answer with a Simple Model](#) (8 min)

- Example of using regression and SSE to fit parameters to a model
- Not that insightful of video

[Copy other People's work to do Data Science](#) (3 min)

- Sales pitch for Microsoft Azure Machine Learning and Cortana Machine Learning Gallery
- Not that insightful of video

Conceptual Resources

Brandon Rohrer's Blog - [Data Science and Robots](#)

- Blog filled with videos and posts about ML and data science
- Very good conceptual explanations without getting very technical

3 Brown 1 Blue - [Neural Network Video Series](#)

- Great animations and visualizations
- A little technical but not overwhelming
- Play at 1.5x speed

- 1) [But what *is* a Neural Network?](#) (19 min)
- 2) [Gradient Descent, How Neural Networks Learn](#) (21 min)
- 3) [What is Backpropagation and What is it Actually Doing?](#) (14 min)

- 4) [Backpropagation Calculus](#) (10 min)

YouTube Channel - [DeepLearning.TV](#)

- Cartoon animations on the concepts behind Artificial Neural Networks

Technical Resources

YouTube Channel - [Mathematicalmonk](#)

- Very advance lectures on ML algorithms, information theory, and statistics
- Play at 2x speed

Course Website - [AP Monitor](#)

- Dr. Hedengren's course website on dynamic optimization and control
- Modeling/optimizing for physical systems

Tutorials

Google Developers - [Machine Learning Recipes Tutorial Series \(python\)](#)

- 1) [Hello World - Machine Learning Recipes #1](#) (7 min)
- 2) [Visualizing a Decision Tree - Machine Learning Recipes #2](#) (7 min)
 - You might not have the library to perform the visualization but that's okay
- 3) [What Makes a Good Feature? - Machine Learning Recipes #3](#) (6 min)
- 4) [Let's Write a Pipeline - Machine Learning Recipes #4](#) (8 min)
- 5) [Writing Our First Classifier - Machine Learning Recipes #5](#) (9 min)
- 6) [Train an Image Classifier with TensorFlow for Poets - Machine Learning Recipes #6](#) (7 min)
 - Requires TensorFlow
- 7) [Classifying Handwritten Digits with TF.Learn - Machine Learning Recipes #7](#) (7 min)
 - Requires TensorFlow
- 8) [Let's Write a Decision Tree Classifier from Scratch: Machine Learning Recipes #8](#) (10 min)
 - A much more technical tutorial involving the inner workings of a decision tree

Buzzwords - Have these on your resume

Artificial Intelligence (AI)

- Computers being programmed to do something "smart"

Machine Learning (ML)

- Subfield of AI
- Statistical analysis of data, lots and lots of data
 - o Can be very simple
 - o Example: Plotting points and drawing a line through it
 - o Can be very complex
 - o **Artificial neural networks (ANN)**
- Essential it's optimization

Deep Learning (DL)

- Subfield of ML
- An ANN that is many layers "deep" in complexity
- Requires sample data and GPUs to "train" the ANN

Big Data

- Three V's: Velocity, Variety, & Volume

- Velocity: Data comes in at a high rate
- Variety: Incoming data might be different structure
- Volume: Data is large in size
- Massive amounts of data
 - 100s of Terabytes or so in size

Cloud Computing

- Outsourcing your computation/hardware to a third party service that you connect to through the internet
 - Amazon Web Services (AWS) - Dominant Cloud Provider
 - Microsoft Azure - Closest competitor to AWS
 - IBM and Google Cloud - Trying to catch up to AWS and Azure
- The Cloud providers offer many different services that range from storage, to webhosting, to computation and such
- Three main "levels" of service
 - Infrastructure as a Service (IaaS)
 - Platform as a Service (PaaS)
 - Software as a Service (SaaS)

Internet of Things (IoT)

- Connecting things to the internet that normally aren't connected
 - Examples: fridge, thermostat, car engine
 - Examples: General Electric is starting to embed its products with sensors and internet connection
 - Examples: "Smart" devices - Smart phone, Smart Watch, Smart Home - Basically it's all IoT
- Arduinos and Raspberry Pis are good at this
- **Industrial Internet of Things (IIoT)**
 - IoT but for industry application

Edge Computing

- Collect small amounts data from sensors and perform some computation onsite before sending it to a larger collection of data for storage or further computation - Usually sent to the Cloud

Hadoop Distributed File System (HDFS)

- Sometimes referred to as "Hadoop"
- An opensource database framework designed for storing data across a cluster of machines
 - YARN and HDFS are the core components in the architecture
 - YARN
 - The Data Operating System that runs on top of HDFS
 - It does Cluster Resource Management
 - Just a technical portion of Hadoop to be aware of
- HDFS is a distributed data base - The data is distributed across a cluster of machines
 - Good with Big Data
- There are many appendage-like services that may or may not be installed on top of HDFS and YARN
 - **Apache Spark**
 - An important in-memory data processing engine that people use
 - It's kind of like a HDFS version of a SQL database
 - Experience with Apache Spark is considered very valuable

Data Science

- Basic Workflow: Data Capture -> Storage -> Processing -> Visualization
 - Data Capture

- Implementing sensors to collect data - usually IoT
 - Storage
 - Basically storing data
 - There might be some quick preprocessing done to format the data
 - Processing
 - Basically it's high performance computing
 - Ideally you would skip the storage step and feed your data into processing as it is captured
 - Visualization
 - Making the processed data in human readable form - usually charts, graphs, and heat maps
 - This can be very hard for high dimensional results
- Any parts of the workflow could be considered Data Science