

Getting started with Machine Learning on AWS

Julien Simon Global Evangelist, AI & Machine Learning, AWS @julsimon



A primer on Machine Learning



Al vs. Machine Learning vs. Deep Learning

Artificial Intelligence: design software applications which exhibit human-like behavior, e.g. speech, natural language processing, reasoning or intuition

Machine Learning: using statistical algorithms, teach machines to learn from featurized data without being explicitly programmed

Deep Learning: using neural networks algorithms, teach machines to learn from complex data where features cannot be easily expressed



Types of Machine Learning

Supervised learning

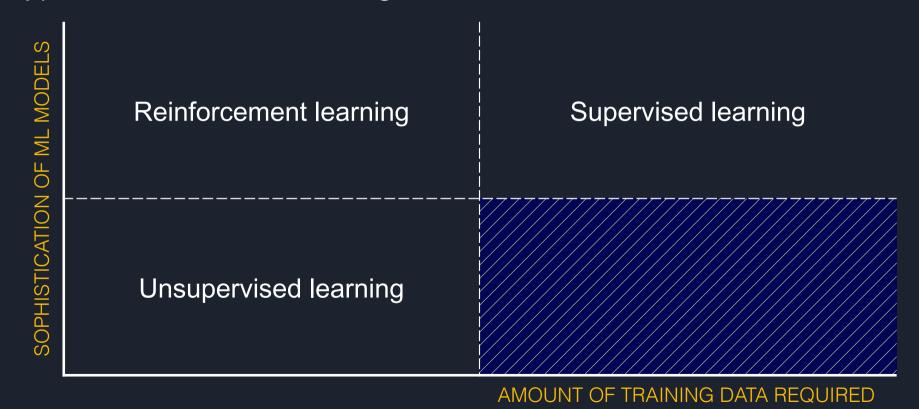
- Run an algorithm on a labeled data set.
- The model learns how to correctly predict the right answer.
- Regression and classification are examples of supervised learning.

Unsupervised learning

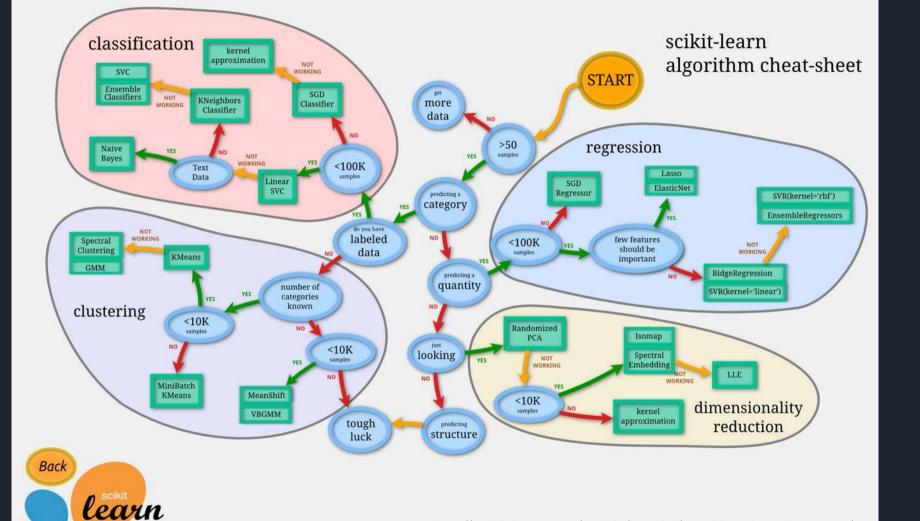
- Run an algorithm on an unlabeled data set.
- The model learns patterns and organizes samples accordingly.
- Clustering and topic modeling are examples of unsupervised learning.



Types of Machine Learning

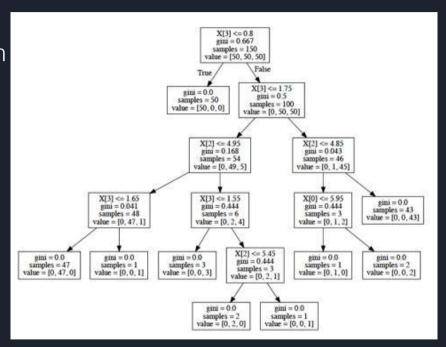






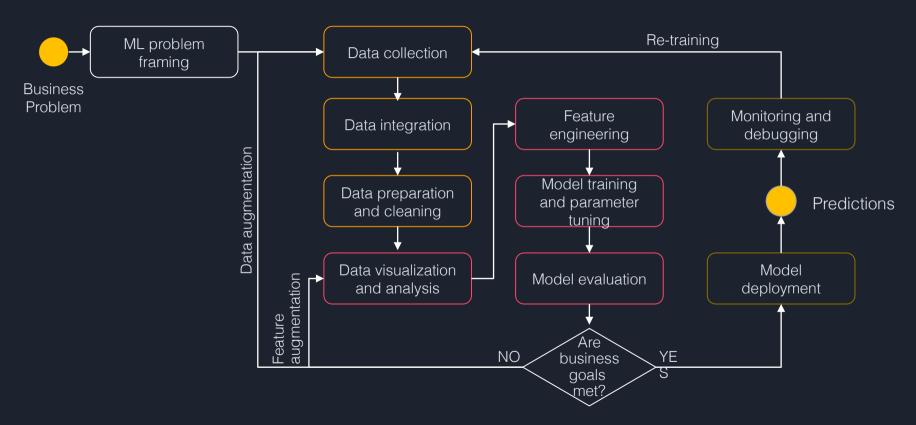
Algorithm example: decision trees

- Supervised learning algorithm
- Goal: build a decision tree for regression or classification
- Data set : features + target attribute (value or class identifier)
- Intuition: find the "best" feature thresholds to go left or right
- "Easy" to interpret
- Advanced variants with multiple trees: Random Forests, XGBoost, etc.





The Machine Learning cycle





Putting your Machine Learning Projects on the right track



1 - Set expectations

- What is the business question you're trying to answer?
 - One sentence on the whiteboard
 - Must be quantifiable
- Do you have (enough) data that could help?
- Involve everyone and come to a common understanding
 - Business, IT, Data Engineering, Data Science, Ops, etc.

- « We want to see what this technology can do for us »
- « We have tons of relational data, surely we can do something with it »
- « I read this cool article about FooBar ML, we ought to try it »





2 - Define clear metrics

- What is the business metric showing success?
- What's the baseline (human and IT)?
- What would be a significant and reasonable improvement?
- What would be reasonable further improvements?
- « The confusion matrix for our support ticket classifier has significantly improved ». Huh?
- « P90 time-to-resolution is now under 24 hours ». Err....
- « Misclassified emails have gone down 5.3% using the latest model ». So?
- « The latest survey shows that 'very happy' customers are up 9.2% ». Woohoo!



3 - Assess needs (not wants) and skills

- Building a data set describing the problem?
- Cleaning, preparing and curating it?
- Writing and tweaking ML algorithms?
- Managing ML infrastructure?

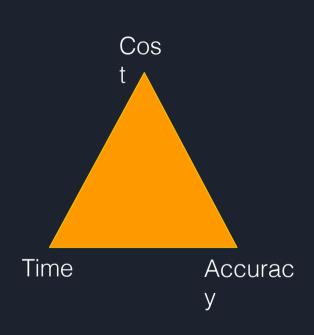




4 - Pick the best tool for the job

- Cost, time to market, accuracy: pick two
- The least expensive and fastest option won't probably be the most accurate.
 - Maybe enough to get started, and learn more about the problem.
- Improving accuracy will take increasingly more time and money.
 - Diminishing returns! Know when to stop.
- Keep an eye on actionable state of the art advances, ignore the rest
 - Transfer learning
 - AutoML





5 - Use proven best practices

- No, things are not different this time.
- AI / ML is software engineering
 - Dev, test, QA, documentation, Agile, versioning, etc.
 - Involve all teams
- Sandbox tests are nice, but truth is in production
 - Get there fast, as often as needed
 - CI / CD and automation are required
 - Devops for ML



Universal Pictures



6 - Iterate, iterate, iterate aka Boyd's Law (1960)

- Start small
- Try the simple things first
- Go to production quickly
- Observe prediction errors
- Act: fix data set? Add more data? Tweak the algo? Try another algo?
- Repeat until accuracy gains become irrelevant
- Move to the next project



Machine Learning at Amazon



Machine Learning innovation at Amazon

Recommended for You Based on Kindle Paper Display w...



MoKo Case for Kindle Paperwhite, Premium Thinnest and Lightest Leather Cover with... \$28 \$98 \$9.99 \(\begin{align*} Prime \)



Case Cover for All-New Kindle (Both 2012...

★★★☆☆:
\$3.99 \(Prime\)









Our mission at AWS

Put machine learning in the hands of every developer



MACHINE LEARNING IS HAPPENING IN COMPANIES OF EVERY SIZE AND INDUSTRY

Tens of thousands customers have chosen AWS for their ML workloads | More than twice as many customers using ML than any other cloud provider





The AWS ML Stack

Broadest and most complete set of Machine Learning capabilities

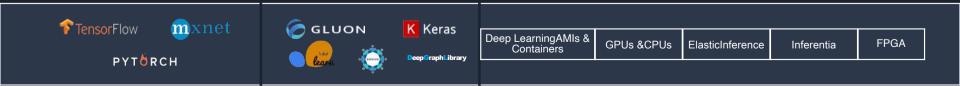
AI SERVICES

VISION	SPEECH		TEXT		SEARCH	СНАТВОТЅ	PERSONALIZATION	FORECASTING	FRAUD		CONTACT CENTERS		
@				A > 文	•			®	a	ŶŸ	<u>.</u>	(A)	
Amazon Rekognition		Amazon Transcribe	Amazon Comprehend	Amazon Translate	Amazon Textract	Amazon Kendra	Amazon Lex	Amazon Personalize	Amazon Forecast	Amazon Fraud Detector	Amazon CodeGuru	Contact Lens	
		+Medical	+Medical									For Amazon Connect	

ML SERVICES



ML FRAMEWORKS & INFRASTRUCTURE





Al Services



The AWS ML Stack

Broadest and most complete set of Machine Learning capabilities

AI SERVICES

VISION	SPEECH		TEXT			SEARCH	СНАТВОТЅ	PERSONALIZATION	FORECASTING	FRAUD		CONTACT CENTERS	
Ø				A > 文	• © =		\ (0)	®	a	\$ \\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\	<u>.</u>	(A)	
Amazon Rekognition	Amazon Polly	Amazon Transcribe	Amazon Comprehend	Amazon Translate	Amazon Textract	Amazon Kendra	Amazon Lex	Amazon Personalize	Amazon Forecast	Amazon Fraud Detector	Amazon CodeGuru	Contact Lens	
		+Medical	+Medical									For Amazon Connect	

MI SERVICES



ML FRAMEWORKS & INFRASTRUCTURE





ML Services



The AWS ML Stack

Broadest and most complete set of Machine Learning capabilities

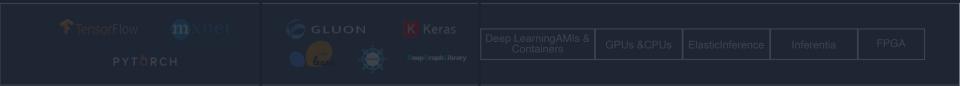
AI SERVICES



ML SERVICES



ML FRAMEWORKS & INFRASTRUCTURE





The machine learning workflow is iterative and complex

Train & Tune Prepare₁ Deploy & Manage Build 101011010 010101010 000011110 Collect and Set up and manage Train, debug, and Deploy Choose or build an Scale and manage Monitor Validate Manage training runs prepare environments tune models model in ML algorithm the production models predictions training data for training production environment



Amazon SageMaker helps you build, train, and deploy models

Train & Tune Prepare Build Deploy & Manage

Web-based IDE for machine learning

Automatically build and train models

Fully managed data processing jobs and data labeling workflows

101011010 010101010 000011110

Collect and prepare training data One-click collaborative notebooks and built-in. high performance algorithms and models



Choose or build an ML algorithm

Debugging and optimization



Set up and manage Train, debug, and environments tune models for training

Manage training runs

Visually track and

compare experiments

deployment and

data drift

Add human review of predictions

Fully managed with auto-scaling for 75% less









Deploy model in production

Monitor models

Validate predictions Scale and manage the production environment

Model options



AWS Marketplace for Machine Learning



Training code



Amazon SageMaker
AutoPilot

Factorization Machines

Linear Learner

Principal Component

Analysis

K-Means Clustering

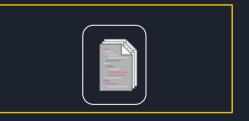
XGBoost

Built-in Algorithms (17)
No ML coding required



Built-in Frameworks

Bring your own code Use open source containers



Bring Your Own
Full control, run your

container
R. C++. etc.

Fully managed training, spot instances included

Frameworks and Infrastructure



The AWS ML Stack

Broadest and most complete set of Machine Learning capabilities

AI SERVICES

ML SERVICES



ML FRAMEWORKS & INFRASTRUCTURE





AWS: The platform of choice for TensorFlow

https://aws.amazon.com/tensorflow



89% of all deep learning workloads in the cloud run on AWS

85% of all TensorFlow workloads in the cloud run on AWS

Source: Nucleus Research, T147, October 2019



Amazon Elastic Inference

https://aws.amazon.com/blogs/machine-learning/optimizing-costs-in-amazon-elastic-inference-with-amazon-tensorflow/



Lower inference costs up to 75%



Match capacity to demand



Available between 1 to 32 TFLOPS per accelerator

Integrated with Amazon EC2 and Amazon SageMaker Support for TensorFlow, Apache MXNet (Incubating) —PyTorch coming soon

Single and mixedprecision operations



Amazon EC2 Inferentia

- Fast, low-latency inferencing at a very low cost
 - 64 TeraOPS on 16-bit floating point (FP16 and BF16) and mixedprecision data.
 - 128 TeraOPS on 8-bit integer (INT8) data.
- Neuron SDK: https://github.com/aws/aws-neuron-sdk
 - Available in Deep Learning AMIs and Deep Learning Containers
 - TensorFlow and Apache MXNet, PyTorch coming soon



https://ml.aws

https://aws.amazon.com/sagemaker

https://github.com/aws/sagemaker-python-sdk.

https://github.com/awslabs/amazon-sagemaker-examples

https://youtube.com/juliensimonfr https://medium.com/@julsimon

Published August 2020

Discount link for the paper edition on Amazon (US only) https://www.amazon.com/gp/mpc/AOHJSZC7A0AV5

Discount code for the e-book edition on Packt <u>20SAGEMAKER</u>

https://www.packtpub.com/product/learn-amazon-sagemaker/9781800208919

Valid until November 11th



