



# Build, train, and deploy machine learning models with Amazon SageMaker

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### Amazon SageMaker



Collect and prepare training data



Choose and optimize your ML algorithm



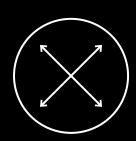
Set up and manage environments for training



Train and tune ML models



Deploy models in production



Scale and manage the production environment

Same service and APIs from experimentation to production



























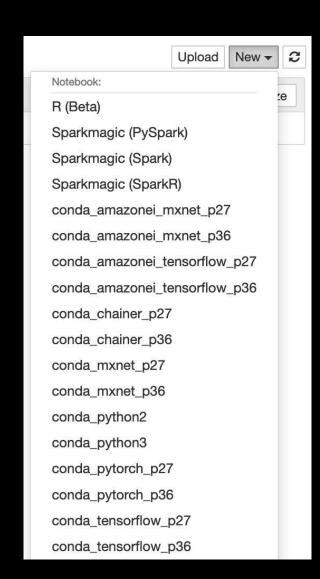


## Building models



#### Notebook instances

- Fully managed instances, from *ml.t2.medium* to *p3.16xlarge*
- Pre-installed with Jupyter and Conda environments
  - Python 2.7 & 3.6
  - Open-source libraries (TensorFlow, Apache MXNet, etc.)
  - Beta support for R NEW!
  - Amazon Elastic Inference for cost-effective GPU acceleration
- Lifecycle configurations
- VPC, encryption, etc.
- Get to work in minutes, zero setup





# Demo: simple R notebook

https://github.com/awslabs/amazon-sagemaker-examples/blob/master/advanced\_functionality/r\_kernel/example\_r\_notebook.ipynb

Using SageMaker APIs from R: <a href="https://github.com/awslabs/amazon-sagemaker-examples/blob/master/advanced-functionality/r-kernel/using-r-with-amazon-sagemaker.ipynb">https://github.com/awslabs/amazon-sagemaker-examples/blob/master/advanced-functionality/r-kernel/using-r-with-amazon-sagemaker.ipynb</a>



### Model options

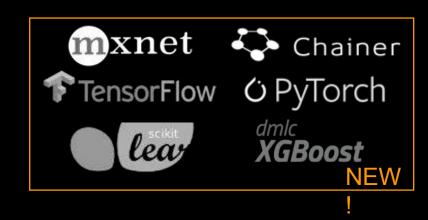


Training code

Factorization Machines
Linear Learner
Principal Component Analysis
K-Means Clustering
XGBoost
And more

Built-in Algorithms (17)

No ML coding required
No infrastructure work required
Distributed training
Pipe mode



**Built-in Frameworks** 

Bring your own code: Script mode
Open-source containers
No infrastructure work required
Distributed training
Pipe mode



Bring Your Own Container

Full control, run anything! R, C++, etc. No infrastructure work required

### The Amazon SageMaker API

- Python SDK orchestrating all Amazon SageMaker activity
  - High-level objects for algorithm selection, training, deploying, automatic model tuning, etc.

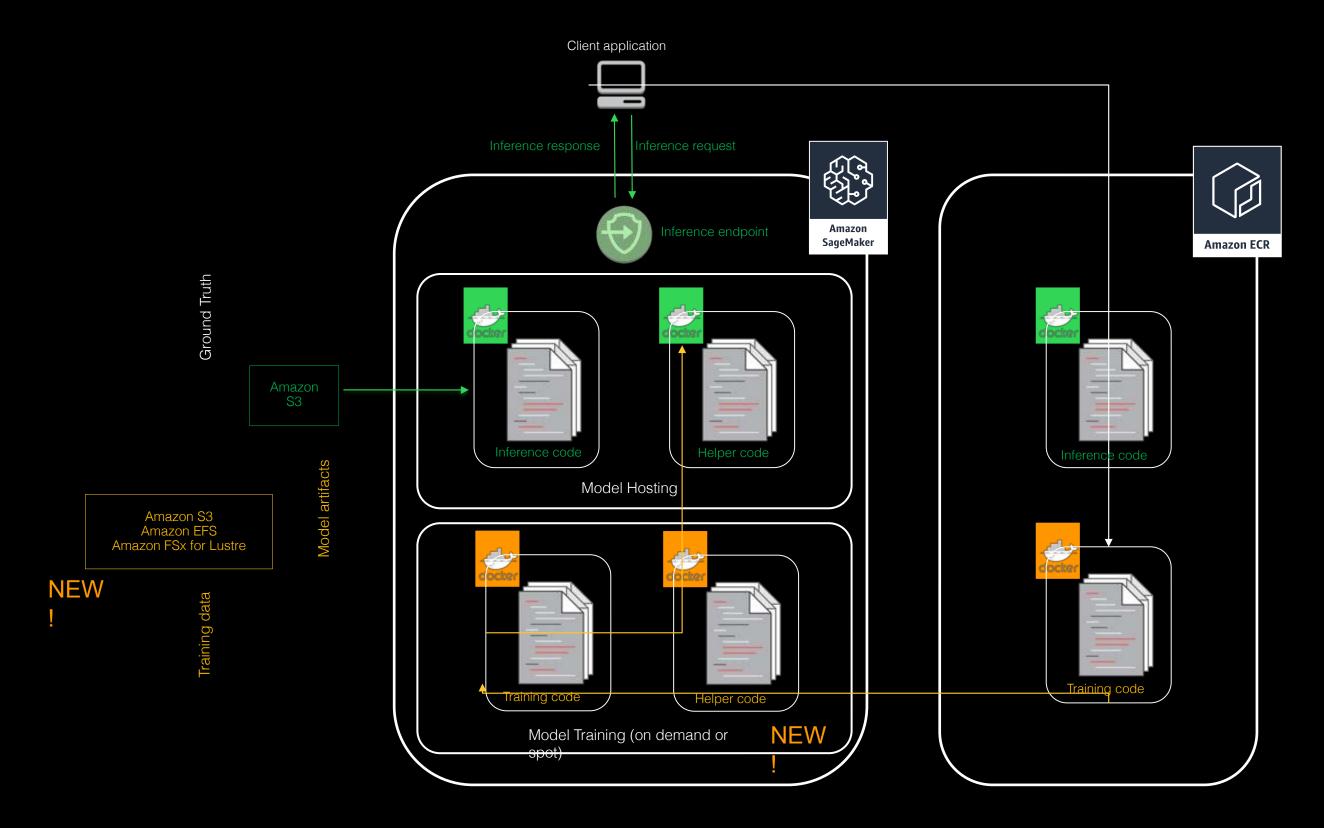
https://github.com/aws/sagemaker-python-sdk

Spark SDK (Python & Scala)
 https://github.com/aws/sagemaker-spark/tree/master/sagemaker-spark-sdk

#### AWS SDK

- Service-level APIs for scripting and automation
- CLI: 'aws sagemaker'
- Language SDKs: boto3, etc.







## Built-in algorithms



# Built-in algorithms Orange: supervised, yellow: unsupervised

Linear Learner: Regression, classification	Image Classification: Deep learning (ResNet)
Factorization Machines: Regression, classification, recommendation	Object Detection (SSD): Deep learning (VGG or ResNet)
<b>K-Nearest Neighbors</b> : Non-parametric regression and classification	Neural Topic Model: Topic modeling
XGBoost: Regression, classification, ranking <a href="https://github.com/dmlc/xgboost">https://github.com/dmlc/xgboost</a>	Latent Dirichlet Allocation: Topic modeling (mostly)
K-Means: Clustering	BlazingText: GPU-based Word2Vec, and text classification
Principal Component Analysis: Dimensionality reduction	Sequence to Sequence: Machine translation, speech to text and more
Random Cut Forest: Anomaly detection	DeepAR: Time-series forecasting (RNN)
Object2Vec: General-purpose embedding	IP Insights: Usage patterns for IP addresses
Semantic Segmentation: Deep learning	a

### Demo:

# Sentence classification with BlazingText

<u>https://github.com/awslabs/amazon-sagemaker-examples/tree/master/introduction to amazon algor ithms/blazingtext text classification dbpedia</u>



#### Demo:

# Built-in image classification with transfer learning

https://gitlab.com/juliensimon/dlnotebooks/blob/master/sagemaker/06-Image-classification-deeplens.ipynb



## Built-in frameworks



### Built-in frameworks: Just add your code



- Built-in containers for training and prediction
  - Open-source, e.g., <a href="https://github.com/aws/sagemaker-tensorflow-containers">https://github.com/aws/sagemaker-tensorflow-containers</a>
  - Build them, run them on your own machine, customize them, etc.
- Local mode: Train and predict on your notebook instance, or on your local machine
- Script mode: Reuse existing code with minimal changes



### TensorFlow on AWS

C5 instances (Intel Skylake)

Training ResNet-50 with the ImageNet dataset using our optimized build of TensorFlow 1.11 on a c5.18xlarge instance type is designed to be 11x faster than training on the stock binaries

P3 instances (NVIDIA V100)

TensorFlow scaling efficiency with 256 GPUs

**65** 

Stock version



90

9/0

AWS-optimized version



### Apache MXNet: Deep learning for enterprise developers









































#### Start with off-the-shelf models

- Gluon CV, Gluon NLP, Gluon TS
- ONNX compatibility

#### Fast and scalable training

- Keras-MXNet up to 2x faster than Keras-TensorFlow
- Near-linear scalability up to 256 GPUs
- Dynamic training

#### Easy deployment

- Java and Scala APIs
- Model Server



#### Demo:

# Fashion-MNIST classification with Keras/TensorFlow

- + Script Mode
- (+ Managed Spot Training)
- (+ Elastic Inference)

https://aws.amazon.com/blogs/machine-learning/train-and-deploy-keras-models-with-tensorflow-and-apache-mxnet-on-amazon-sagemaker/

https://gitlab.com/juliensimon/dlnotebooks/tree/master/keras/05-keras-blog-post



### Getting started

http://aws.amazon.com/free

https://ml.aws

https://aws.amazon.com/sagemaker

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

https://github.com/aws/sagemaker-spark

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

https://gitlab.com/juliensimon/dlnotebooks



# Thank you!

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