# Building Machine Learning Models Using AWS SageMaker



Jorge Vásquez SOFTWARE ENGINEER @jorvasquez2301



#### Overview



## Building a model for breast cancer detection

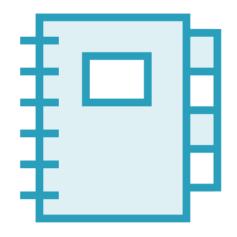
- Creating a notebook instance
- Creating the model using:
  - A built-in algorithm
  - Tensorflow
  - Apache MXNet



#### Creating and Configuring a Notebook Instance for Creating Breast Cancer Detection Models



#### AWS SageMaker Notebook Instance



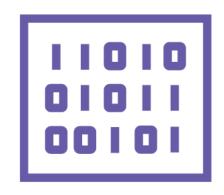
Fully managed ML compute instance running the Jupyter Notebook App, including related resources



#### AWS SageMaker Notebook Instance



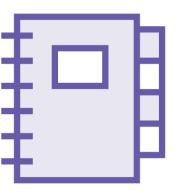
Anaconda packages



Tensorflow and Apache MXNet



Storage volume



Example Jupyter notebooks



#### AWS SageMaker Notebook Instance

Prepare and process data

Write code to train models

Deploy models to AWS SageMaker hosting

Test or validate your models





## Creating an AWS SageMaker Notebook Instance:

- Configuring instance name and type
- Assigning an IAM role for allowing access to S3
- Configuring storage volume size



# Building a Model in AWS SageMaker for Breast Cancer Detection Using a Built-in Algorithm



Supervised learning algorithm

Takes an image as input and classifies it into one of multiple output categories

Uses a
Convolutional
Neural Network
(ResNet)



#### Full training mode

Network is initialized with random weights

Network is trained from scratch

Needs a lot of input images

#### Transfer learning mode

Network is initialized with pre-trained weights

Just the top fully connected layer is initialized with random weights

Needs a smaller number of input images

The whole network is fine tuned with user images



Apache MXNet RecordIO (recommended)

Raw images (JPEG, PNG)



#### Using RecordIO format for input

Pipe Mode

Your training job streams data directly from S3

Faster start times

Better throughput

Reduced storage volume usage

#### Using Raw Images for input

File Mode

Loads all your training data from S3 to the training instance volumes

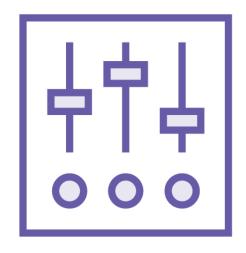
Slower start times

Lower throughput

Higher storage volume usage

Can work on Pipe Mode, if an augmented manifest file is provided





The Image Classification Algorithm has several hyperparameters that can be adjusted for better performance



### Hyperparameter

A hyperparameter is a parameter that is set before the learning process begins. These parameters are tunable and can directly affect how well a model trains (deepai.org)



**Number of classes** 

Number of training samples

Image Classification Required Hyperparameters



## Low-level AWS SDK for Python

High-level SageMaker Python library Available APIs for Building Models Using Built-in Algorithms





Creating a Jupyter notebook

Obtaining the histopathology images

**Exploring the images** 

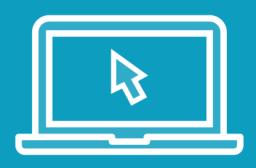
Converting images to the RecordIO format and upload to S3





Configuring the Image Classification Algorithm using the low-level AWS SDK for Python





Configuring the Image Classification Algorithm using the high-level SageMaker Python library



# Building a Model in SageMaker for Breast Cancer Detection Using Tensorflow



Latest supported version of Tensorflow is 1.12.0

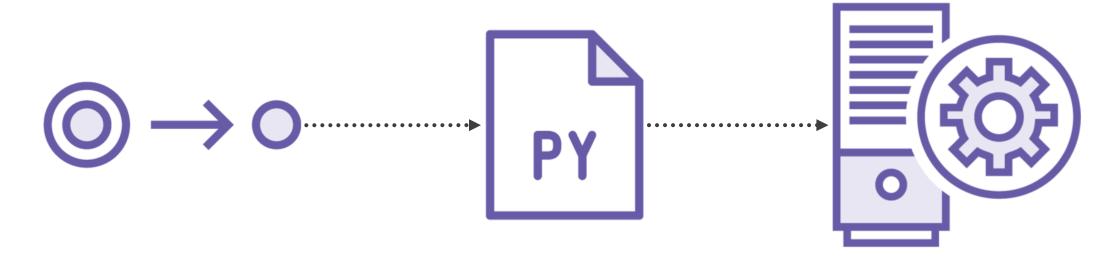
Using Tensorflow with the SageMaker Python SDK



It's recommended to use the latest supported version of Tensorflow, because that's where AWS focuses most of its development efforts



#### Building Tensorflow Models in SageMaker



Transform input images to the TFRecord format

Prepare the training script

Build a model using the Tensorflow estimator



#### Training Tensorflow Models

Tensorflow Training
Script

Since Tensorflow 1.11, Script mode is the training script format



#### Training Tensorflow Models

Tensorflow Training Script

#### Script mode supports training with a:

- Python script
- Python module
- Shell script



#### Training Tensorflow Models

Tensorflow Training
Script

SageMaker gives your script access to useful environment variables



#### Available Environment Variables

SM\_MODEL\_DIR

SM\_NUM\_GPUS

SM\_OUTPUT\_DATA\_DIR

SM\_CHANNEL\_XXXX





Converting images to the TFRecord format and upload to S3





Preparing a training Python script

Configuring a Tensorflow Estimator using the high-level SageMaker Python library



# Building a Model in SageMaker for Breast Cancer Detection Using Apache MXNet



Latest supported version of MXNet is 1.3.0

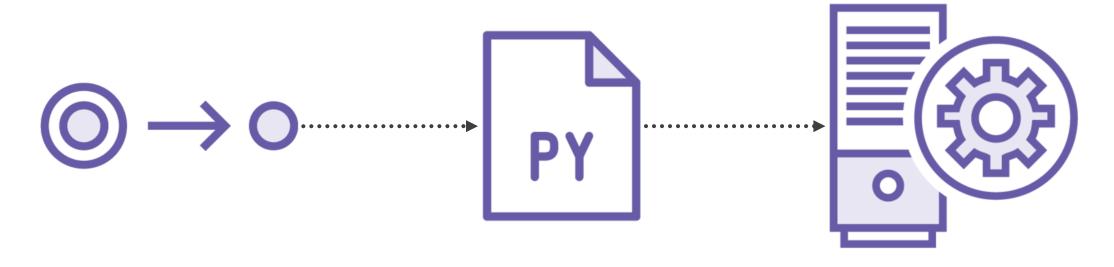
Using Apache MXNet with the SageMaker Python SDK



It's recommended to use the latest supported version of MXNet, because that's where AWS focuses most of its development efforts



#### Building MXNet Models in SageMaker



Transform input images to the RecordIO format

Prepare the training script

Build a model using the MXNet estimator



#### Training MXNet Models

MXNet Training Script The structure for training scripts changed with MXNet version 1.3, and that's the version we are going to use



#### Training MXNet Models

MXNet Training Script

SageMaker gives your script access to useful environment variables



#### Available Environment Variables

SM\_MODEL\_DIR

SM\_NUM\_GPUS

SM\_OUTPUT\_DATA\_DIR

SM\_CHANNEL\_XXXX





Preparing the training script

Configuring a MXNet Estimator using the high-level SageMaker Python library



#### Summary



How to create a notebook instance

Exploring and preparing input data with Jupyter

Configuring the built-in Image Classification Algorithm

- Low-level AWS SDK for Python
- High-level SageMaker Python Library

Creating a custom script with Tensorflow and MXNet using the high-level SageMaker Python Library

