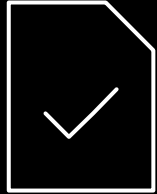


# Build, train and deploy Machine Learning models on Amazon SageMaker

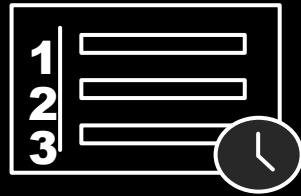
Julien Simon  
Global Evangelist, AI & Machine Learning  
@julsimon

Gàbor Stikkel  
Senior Data Scientist  
HID Global

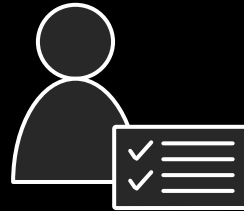
# Amazon SageMaker



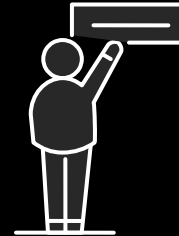
Collect and  
prepare training  
data



Choose and  
optimize your  
ML algorithm



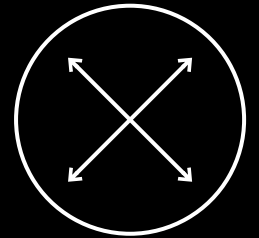
Set up and  
manage  
environments  
for training



Train and  
Tune ML Models



Deploy models  
in production



Scale and manage  
the production  
environment

Same service and SDK from experimentation to production

intuit



tinder



CONVOY

SIEMENS

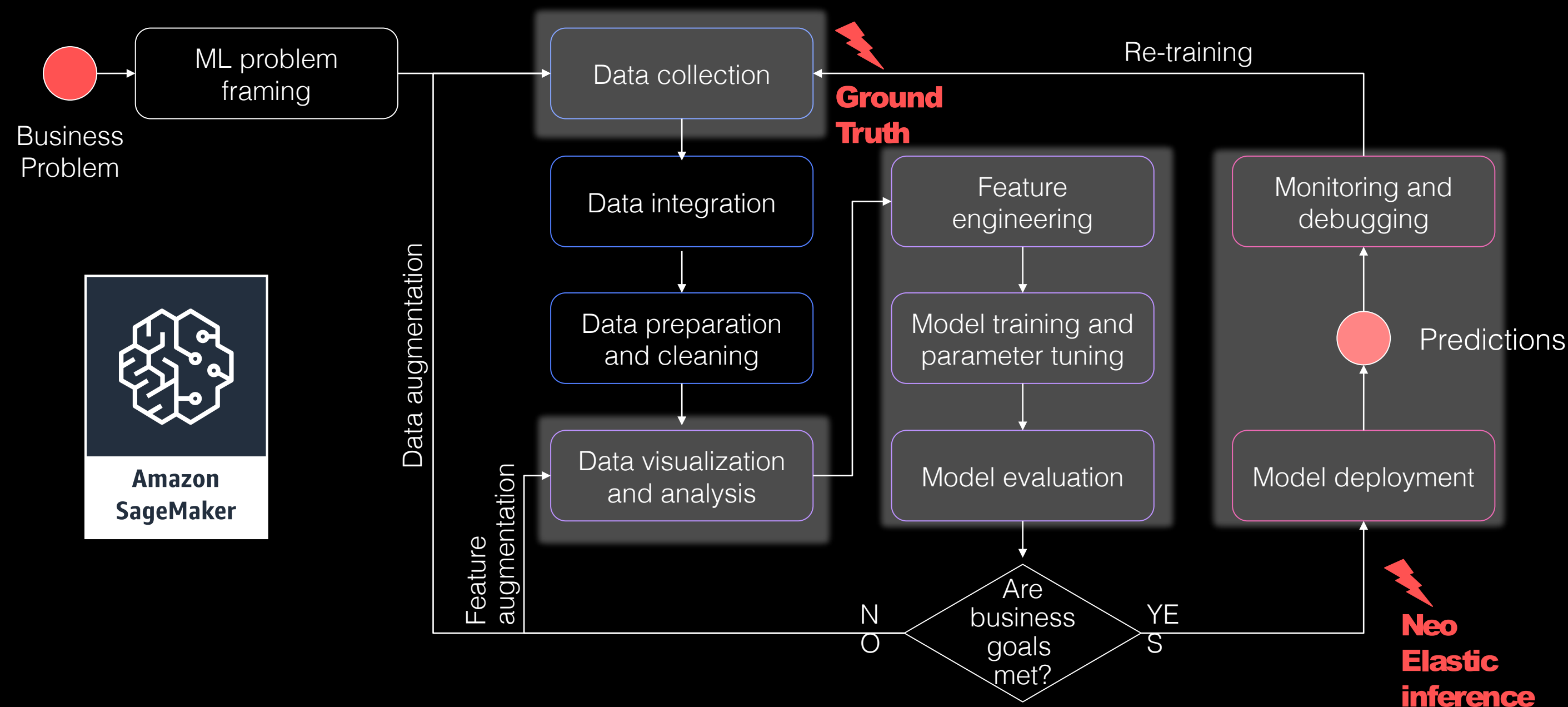


DOW JONES



SONY

# Build, train and deploy models using SageMaker





# The HID Global's Twist Contest

Gesture Recognition in Access Control

AWS Summit Stockholm, 22 May, 2019

Gábor Stikkel, Senior Data Scientist, HID Global





Every day millions of people in more than 100 countries use our products and services to securely access physical and digital places

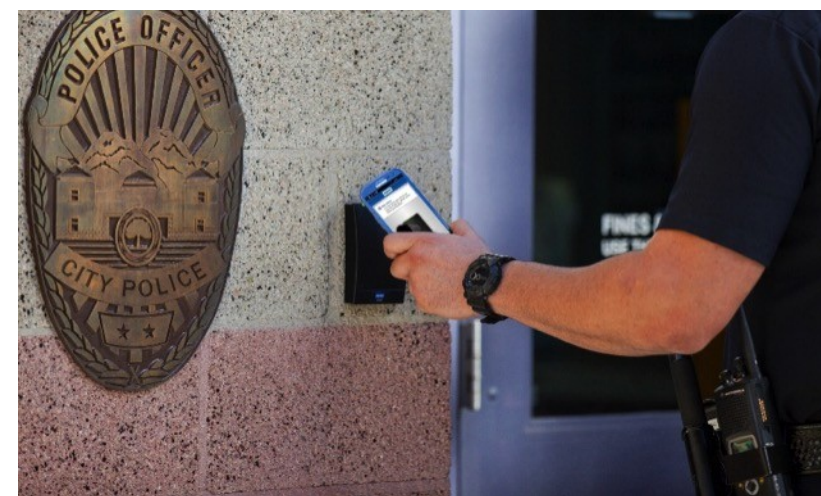


Over 2 billion things that need to be identified, verified and tracked are connected through HID's technology



3,200+ employees worldwide

Part of ASSA ABLOY: 47000+



ASSA ABLOY named in Forbes Top 100 of the **World's Most Innovative Companies**

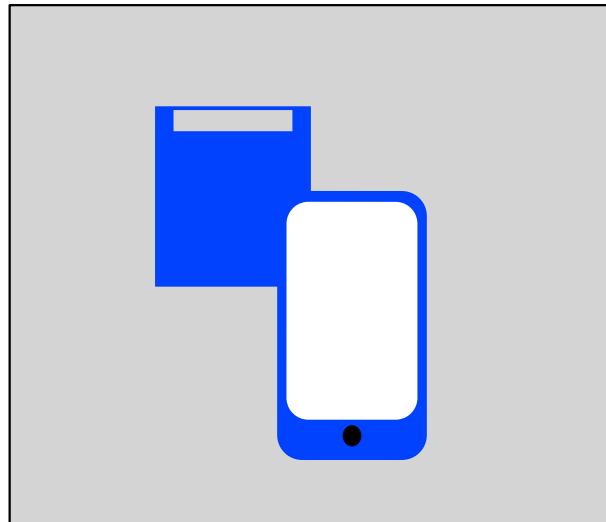


## GLASS Trailer (2019)

11 003 790 visningar

 132 TN  3,4 TN  DELA  SPARA ...

# Data-driven use cases in physical access control



Tap

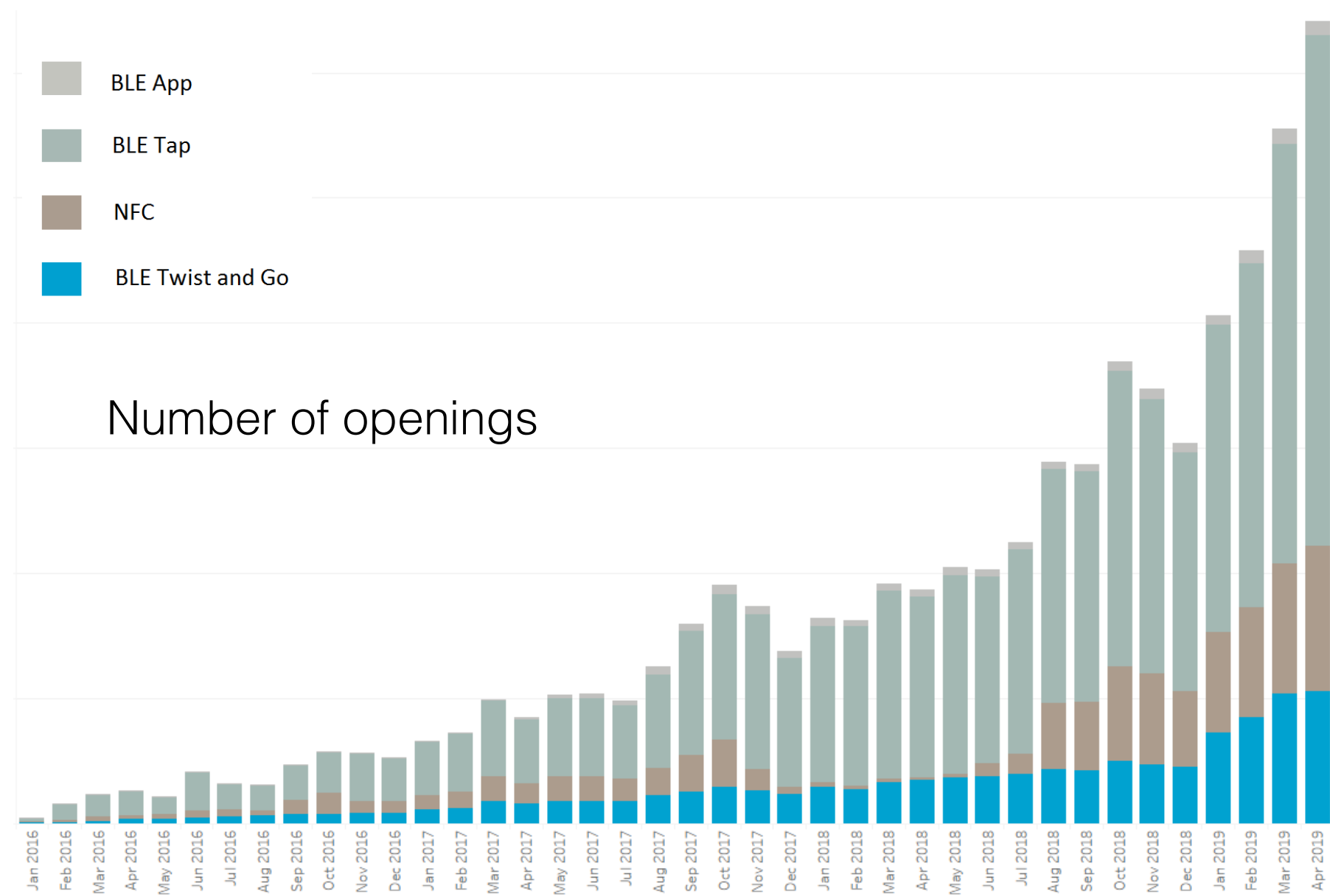


Twist and Go



Seamless  
Access

# Launch of Mobile Services was a success





# Data collection



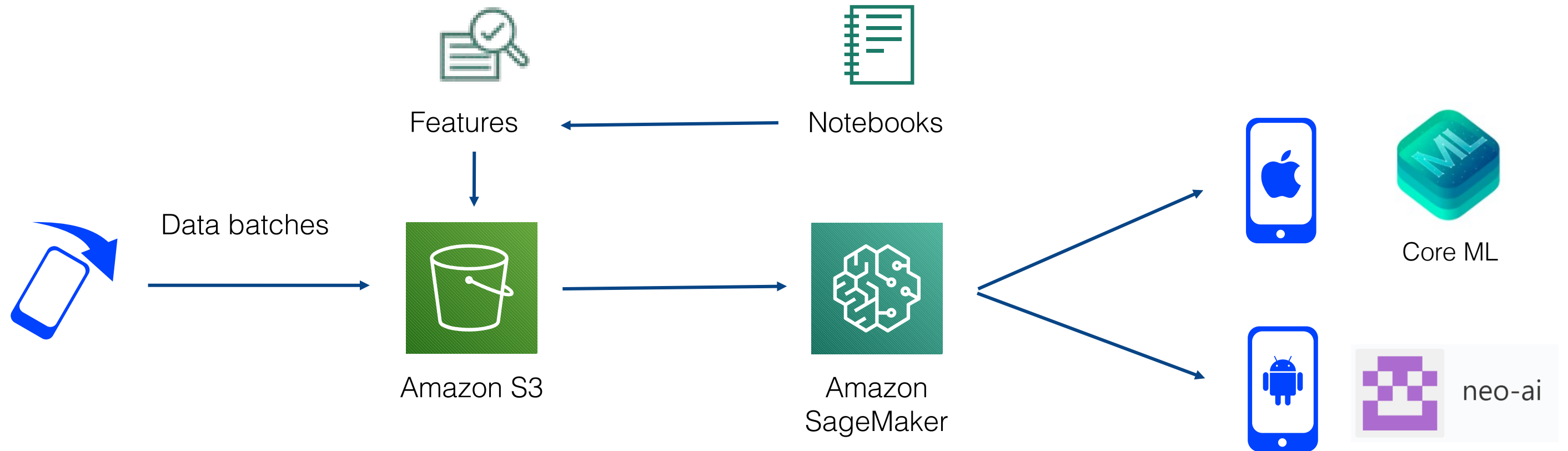
# Business problem

- Simple threshold based rule
- Many different behaviours
- Security issues

**Goal: reduce false positives whilst providing a delightful experience**

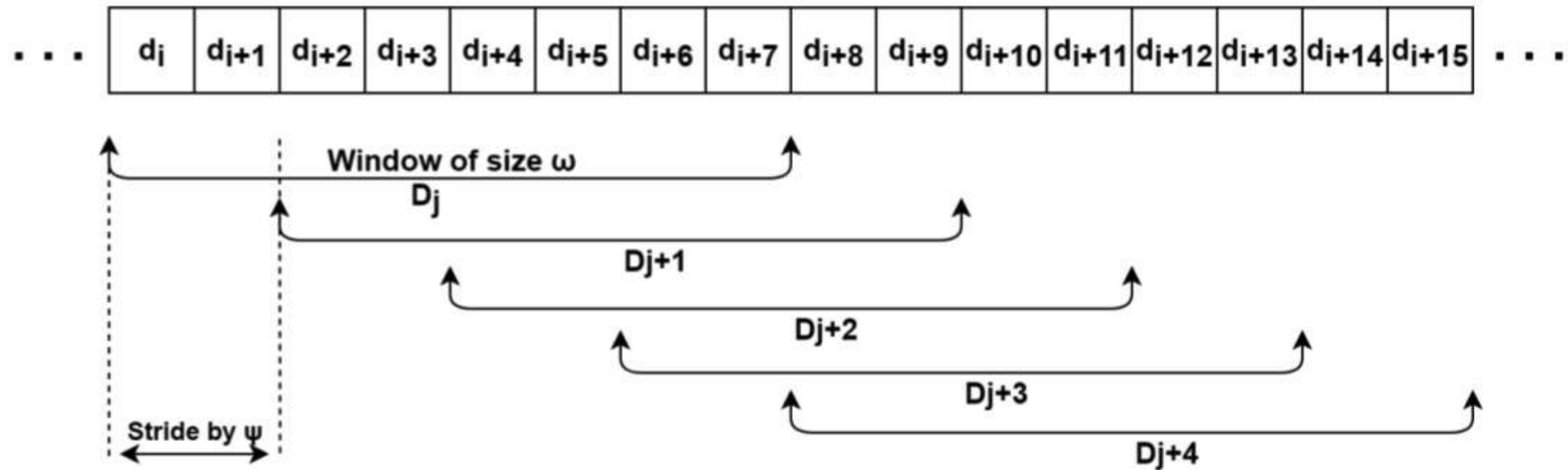


# Predictive Modeling Pipeline





# Feature calculation using sliding windows



Programmable Gesture Recognition for Augmenting Assistive Devices Sishir Patil et al. 2018 for details)

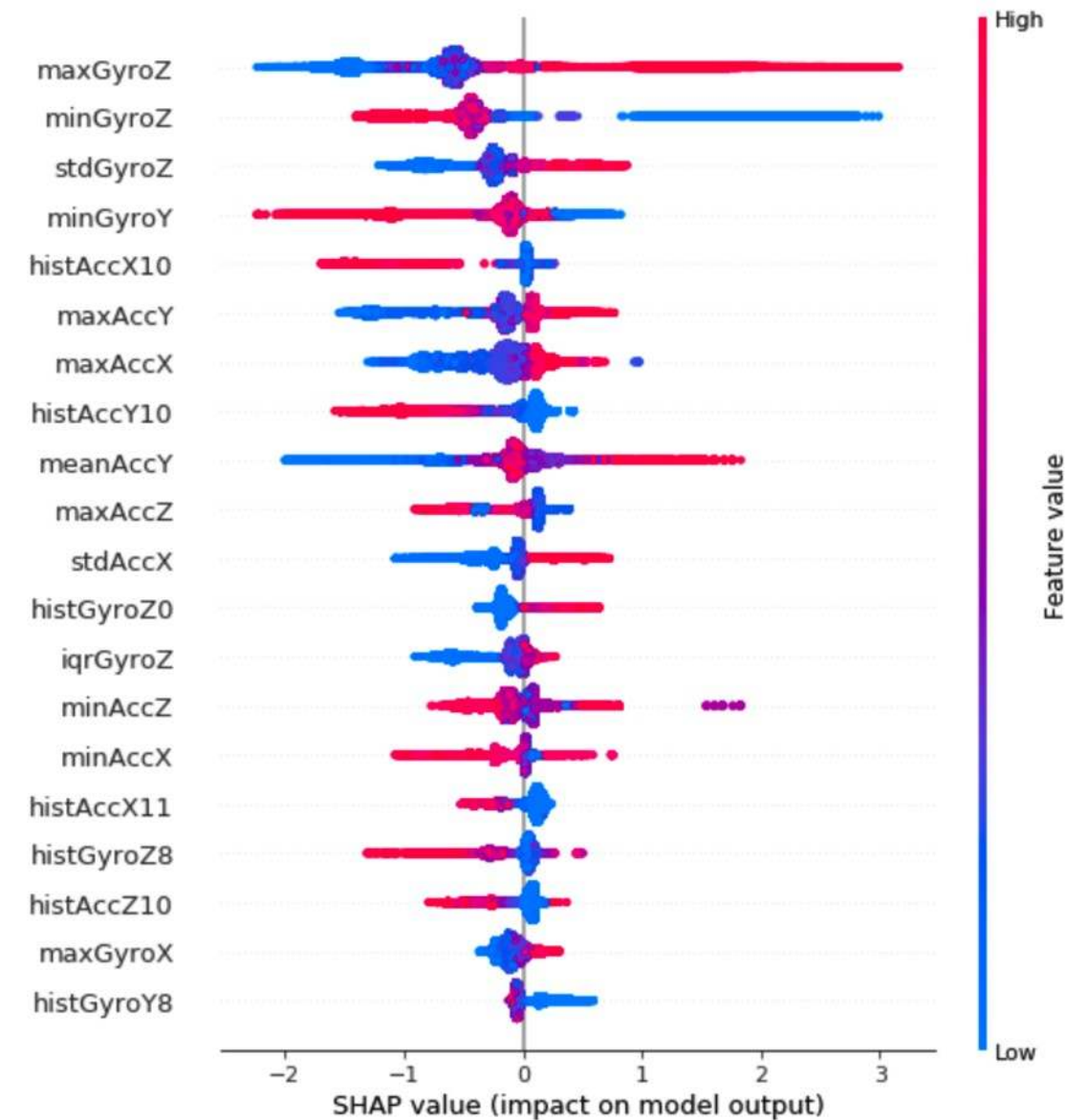
# Predictive modeling

## Neural networks

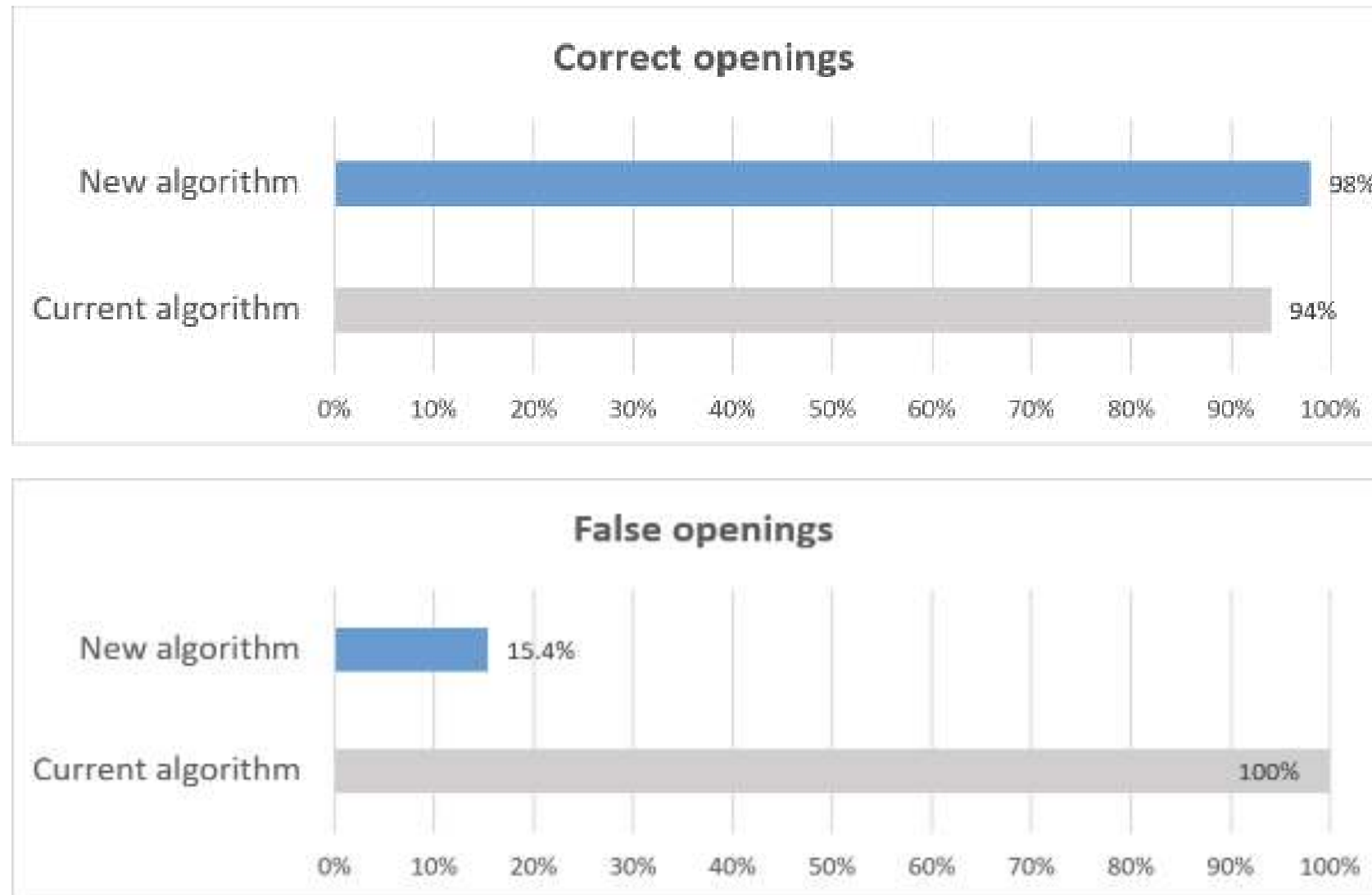
- reproducibility issues
- many parameters for even simple models

## Tree based ensembles

- better performance
- smaller footprint



# Reducing false positives



Extra improvement: twist is recognized ~275ms earlier



?



Phone picture is from <https://uae.souq.com/>

Tree picture is from <https://www.esat.kuleuven.be/>

# First opening based on a ML model!

2019-01-18 14:23:43.289 17694-17694/com.assaabloy.mobilekeys.android.v2 D/c.a.m.a.e.b.TwistAndGoUltraOpeningTrigger: [main] Prediction 0.0010217684

2019-01-18 14:23:43.306 17694-17694/com.assaabloy.mobilekeys.android.v2 D/c.a.m.a.e.b.TwistAndGoUltraOpeningTrigger: [main] Prediction 0.002618488

2019-01-18 14:23:43.326 17694-17694/com.assaabloy.mobilekeys.android.v2 D/c.a.m.a.e.b.TwistAndGoUltraOpeningTrigger: [main] Prediction 0.0030016804

2019-01-18 14:23:43.345 17694-17694/com.assaabloy.mobilekeys.android.v2 D/c.a.m.a.e.b.TwistAndGoUltraOpeningTrigger: [main] Prediction 0.0070577543

2019-01-18 14:23:43.363 17694-17694/com.assaabloy.mobilekeys.android.v2 D/c.a.m.a.e.b.TwistAndGoUltraOpeningTrigger: [main] Prediction 0.027276957

2019-01-18 14:23:43.384 17694-17694/com.assaabloy.mobilekeys.android.v2 D/c.a.m.a.e.b.TwistAndGoUltraOpeningTrigger: [main] Prediction 0.064488105

2019-01-18 14:23:43.404 17694-17694/com.assaabloy.mobilekeys.android.v2 D/c.a.m.a.e.b.TwistAndGoUltraOpeningTrigger: [main] Prediction 0.06967241

2019-01-18 14:23:43.421 17694-17694/com.assaabloy.mobilekeys.android.v2 D/c.a.m.a.e.b.TwistAndGoUltraOpeningTrigger: [main] Prediction 0.07492348

2019-01-18 14:23:43.443 17694-17694/com.assaabloy.mobilekeys.android.v2 D/c.a.m.a.e.b.TwistAndGoUltraOpeningTrigger: [main] Prediction 0.079435736

2019-01-18 14:23:43.461 17694-17694/com.assaabloy.mobilekeys.android.v2 D/c.a.m.a.e.b.TwistAndGoUltraOpeningTrigger: [main] Prediction 0.3511875

**2019-01-18 14:23:43.479 17694-17694/com.assaabloy.mobilekeys.android.v2 D/c.a.m.a.e.b.TwistAndGoUltraOpeningTrigger: [main] Prediction 0.5662894**

**2019-01-18 14:23:43.480 17694-17694/com.assaabloy.mobilekeys.android.v2 D/c.a.m.a.e.b.TwistAndGoUltraOpeningTrigger: [main] Opening!**

**2019-01-18 14:23:43.481 17694-17694/com.assaabloy.mobilekeys.android.v2 D/c.a.m.a.e.b.TwistAndGoUltraOpeningTrigger: [main] Twist and Go detected**

# Productification - Android

- [treelite](#) to compile the Python model into a C function
- AWS: [Neo AI](#) to accelerate model deployment to edge devices
- Keeping feature calculation in synch is [hard](#)

```
#include "header.h"

size_t get_num_output_group(void) {
    return 1;
}

size_t get_num_feature(void) {
    return 174;
}

static inline float pred_transform(float margin) {
    const float alpha = (float)1;
    return 1.0f / (1 + expf(-alpha * margin));
}

float predict(union Entry* data, int pred_margin) {
    float sum = 0.0f;
    unsigned int tmp;
    if (!(data[2].missing != -1) || (data[2].fvalue < 6.3081808)) {
        if (!(data[145].missing != -1) || (data[145].fvalue < -4.9956379)) {
            if (!(data[121].missing != -1) || (data[121].fvalue < -3.893259)) {
                if (!(data[59].missing != -1) || (data[59].fvalue < -5.1344109)) {
                    if (!(data[146].missing != -1) || (data[146].fvalue < 0.27744454)) {
                        if (!(data[156].missing != -1) || (data[156].fvalue < 6.5)) {
                            sum += (float)0.17714286;
                        } else {
                            sum += (float)-0.022222223;
                        }
                    } else {
                        if (!(data[8].missing != -1) || (data[8].fvalue < 6.449605)) {
                            sum += (float)-0.12;
                        } else {
                            sum += (float)-0;
                        }
                    }
                } else {
                    if (!(data[133].missing != -1) || (data[133].fvalue < 12.5)) {
                        if (!(data[46].missing != -1) || (data[46].fvalue < 9.5)) {
                            sum += (float)-0.14202899;
                        } else {
                            sum += (float)0.06666667;
                        }
                    } else {
                        if (!(data[0].missing != -1) || (data[0].fvalue < -11.438884)) {
                            sum += (float)-0;
                        } else {
                            sum += (float)0.1;
                        }
                    }
                }
            } else {
                if (!(data[75].missing != -1) || (data[75].fvalue < 10.5)) {
                    if (!(data[50].missing != -1) || (data[50].fvalue < 16.5)) {
                        if (!(data[117].missing != -1) || (data[117].fvalue < -0.98243845)) {
                            sum += (float)-0.022222223;
                        } else {
                            sum += (float)-0.17352942;
                        }
                    } else {
                        if (!(data[90].missing != -1) || (data[90].fvalue < 0.38042349)) {
                            sum += (float)-0.13846155;
                        } else {
                            sum += (float)0.10476191;
                        }
                    }
                } else {
                    sum += (float)0;
                }
            }
        }
    }
}
```



# Productification - iOS



- More black-box than Android
- Same challenges with feature calculation

## ▼ Machine Learning Model

Name GaborModel

Type

Size 550 KB

Author unknown

Description description not included

License unknown

## ▼ Model Class

 GaborModel

Model class has not been generated yet.

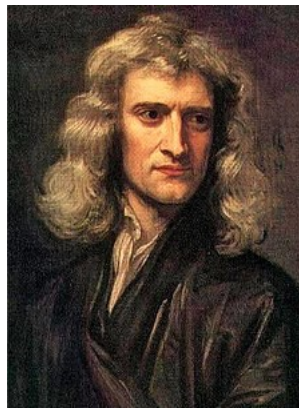
# Conclusions

Data is the new water – it comes from every tap

People are unpredictable – they invent all sorts of gestures

Lowest hanging fruits are grown on decision trees

**”ML tool support from AWS making data scientists' life easier”**



Isaac Newton  
1643 - 1727



C-A. de Coulomb  
1736 - 1806



Alessandro Volta  
1745 - 1827



André-Marie Ampère  
1775 - 1836



Georg Ohm  
1789 - 1854



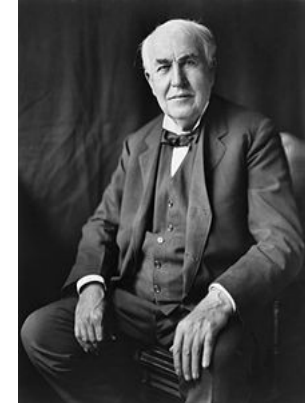
Michael Faraday  
1791 - 1867



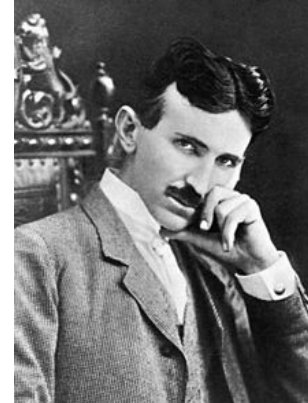
Joseph Henry  
1797 - 1878



Ada Lovelace  
1815 - 1852



Thomas Edison  
1847 - 1931



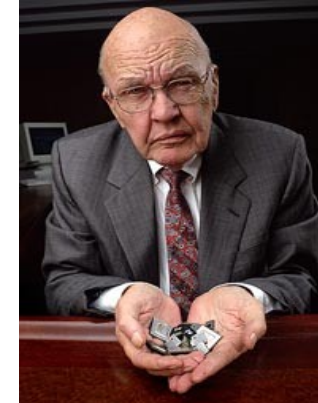
Nikolas Tesla  
1856 - 1943



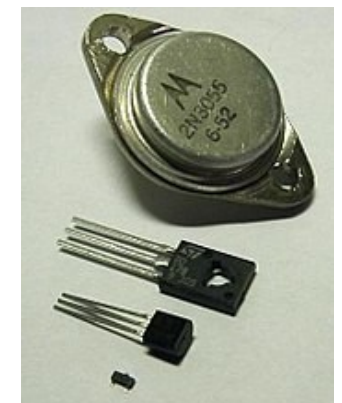
Arthur Samuel  
1901 - 1990



János Neumann  
1903 - 1957



Jack Kilby  
1923 - 2005



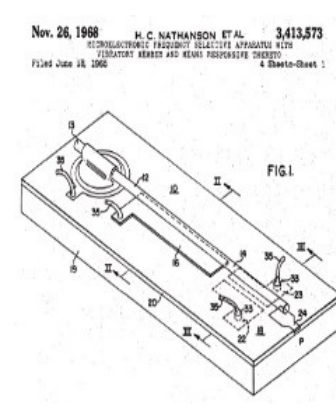
Transistor  
1926 -



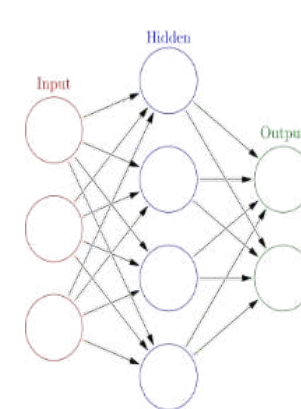
Robert Noyce  
1927 - 1990



Leo Breiman  
1928 - 2005



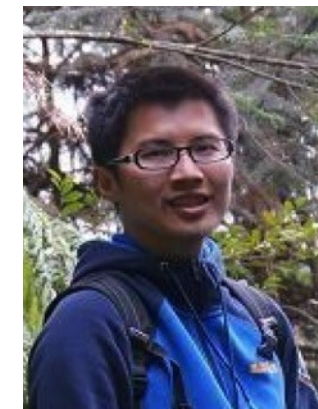
Harvey Nathanson  
1936 -



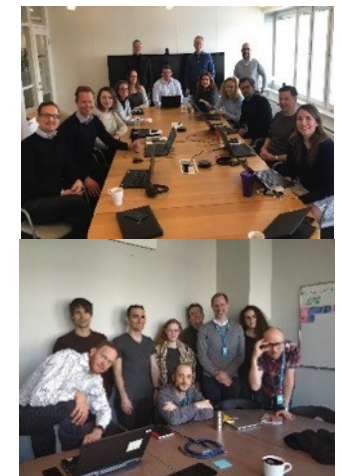
McCulloch - Pitts  
1943



Steve Jobs  
1955 - 2011



Tianqi Chen





# The five beer team



# Deep Learning on Amazon SageMaker

# Model options



Training code

**AWS Machine Learning Marketplace: 150+ off-the-shelf models**

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



# Built-in Deep Learning frameworks: just add your code



- Built-in containers for **training** and **prediction**.
  - Code available on Github, e.g. <https://github.com/aws/sagemaker-tensorflow-containers>
  - Build them, run them on your own machine, customize them, etc.
- **Script mode**: use the **same code** as on your laptop

---

**No infrastructure work required**: simply define instance type and instance count

**Distributed training** out of the box: zero setup

**Pipe mode**: stream infinitely large datasets directly from Amazon S3



# AWS: The platform of choice to run TensorFlow



**85%** of all  
TensorFlow  
workloads in the  
cloud runs on AWS

Source: Nucleus Research, November 2018

# Optimizing Tensorflow for Amazon EC2 instances

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 **11x faster** than training on the stock binaries.

P3 instances (NVIDIA V100)

Tensorflow scaling efficiency with 256 GPUs

**65**

Stock version



**90**  
**%**

AWS-optimized version

# Apache MXNet: Deep Learning for enterprise developers



Start with off-the-shelf models

- Gluon CV and Gluon NLP
- 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/Scala APIs
- Model Server

# Demo: Keras+Tensorflow

Script mode

Automatic model tuning

Elastic inference

<https://gitlab.com/juliensimon/dlnotebooks/tree/master/keras/04-fashion-mnist-sagemaker-advanced>



# Getting started

<http://aws.amazon.com/free>

<https://aws.amazon.com/sagemaker>

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

<https://github.com/aws-labs/amazon-sagemaker-examples>

<https://medium.com/@julsimon>

<https://gitlab.com/juliensimon/dlnotebooks>

# Thank you!

Julien Simon  
Global Evangelist, AI & Machine Learning  
@julsimon

Gàbor Stikkel  
Senior Data Scientist  
HID Global



Please complete the  
session survey.