Microsoft ML for Apache Spark

Unifying Machine Learning Ecosystems at Massive Scales

Mark Hamilton

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Overview

- Background
  - Spark + SparkML
  - MMLSpark
- Unifying ML Ecosystems
  - LightGBM, CNTK, Vowpal Wabbit
  - Multilingual Bindings
- Microservice Orchestration
  - Cognitive Services on Spark
- Model Deployment with Spark Serving
- Use Cases
  - The Snow Leopard Trust

APACHE Spark™

Vowpal Wabbit

LightGBM

CNTK

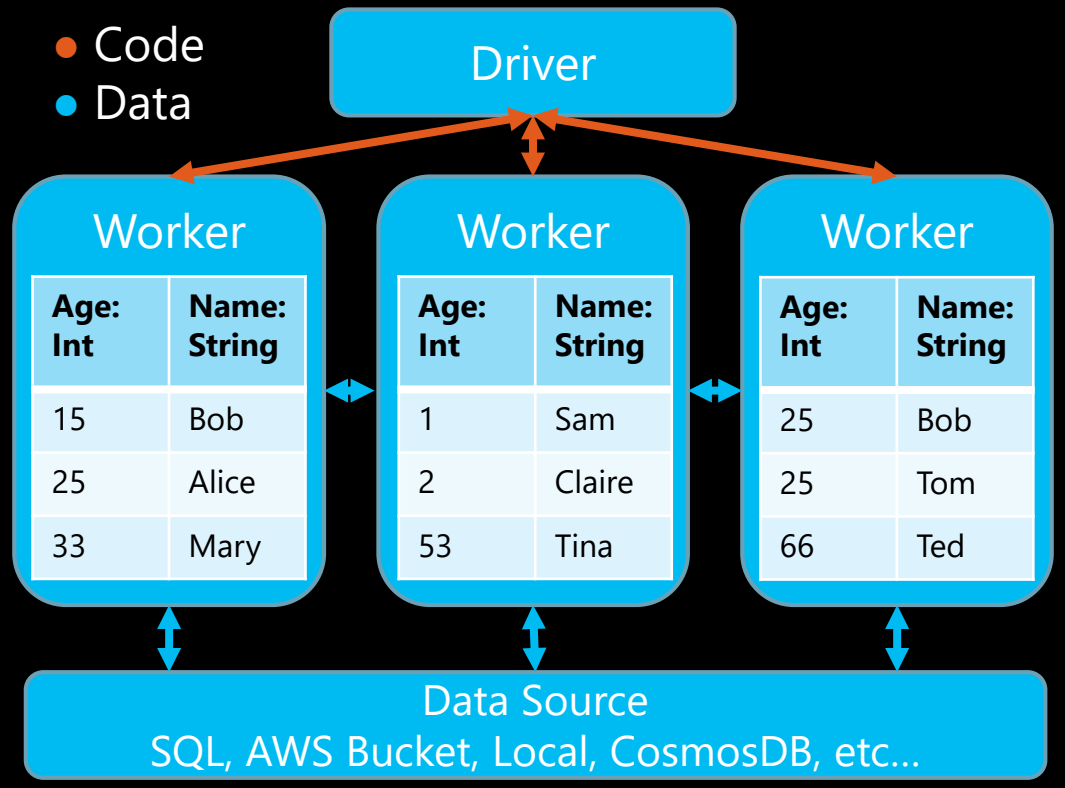
Cognitive Services

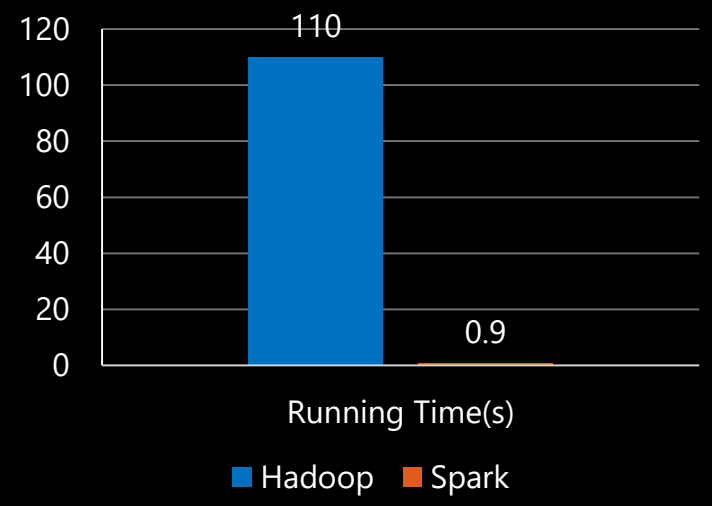
Kubernetes

Snow Leopard Trust

THE MET

- A **fault-tolerant distributed** computing framework
- Map Reduce + SQL
- Whole program optimization + query pushdown
- Elastic
- Scala, Python, R, Java, Julia
- ML, Graph Processing, Streaming

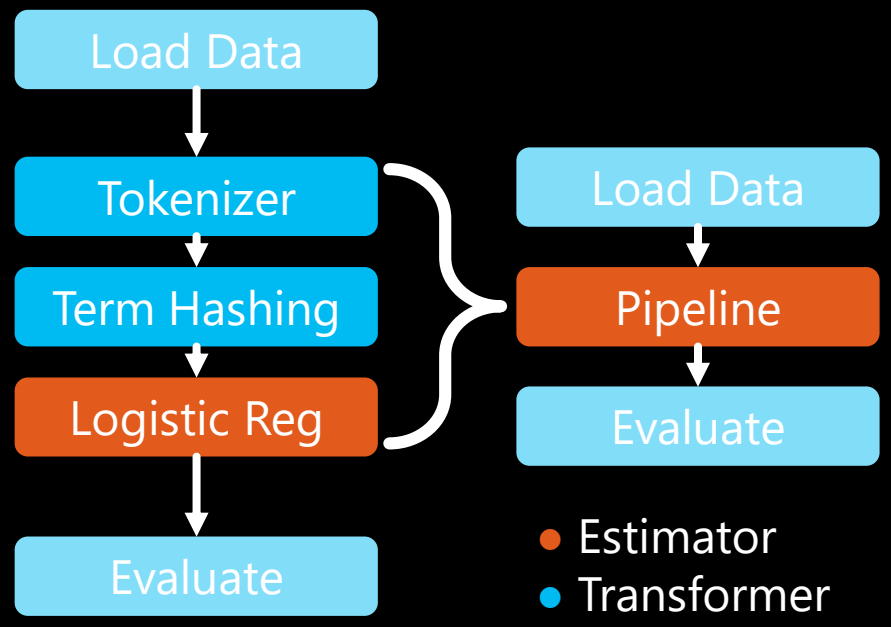




APACHE Spark ML

- High level library for distributed machine learning
- More general than SciKit-Learn
- All models have a uniform interface
  - Can compose models into complex pipelines
  - Can save, load, and transport models

```
data = spark.read.csv("hdfs://...")
train, test = data.randomSplit([.5,.5])
model = LogisticRegression().fit(train)
predictions = model.transform(test)
```



Microsoft Machine Learning for Apache Spark v0.18

Microsoft's Open Source Contributions to Apache Spark

Distributed Machine Learning

Fast Model Deployment

Microservice Orchestration

Multilingual Binding Generation

www.aka.ms/spark

Azure/mmlspark

Unifying Machine Learning Ecosystems

- Goals
  - Same API
  - Composable
  - Batch, Streaming, Serving
  - Elastically Distributed
  - Fault Tolerant
  - Multi-Language
  - Data Source Agnostic

Markus Cozowicz
marcozo@microsoft.com
Data Scientist

Image Processing with Open CV

Gradient Boosting with LightGBM

Deep Learning Pipelines (Databricks)

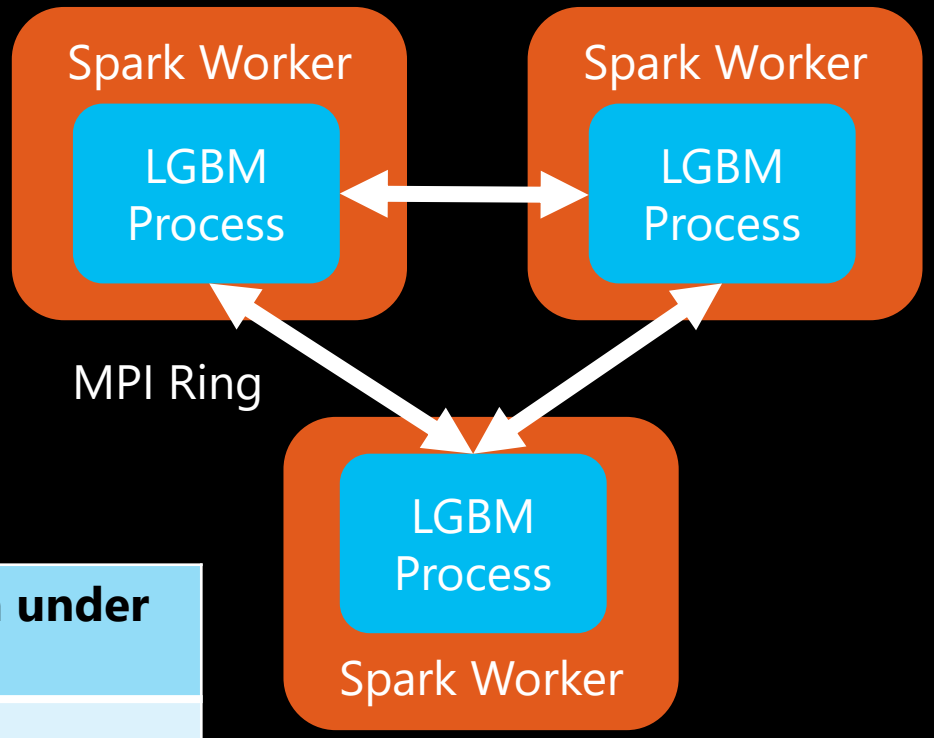
Text Analytics with Vowpal Wabbit

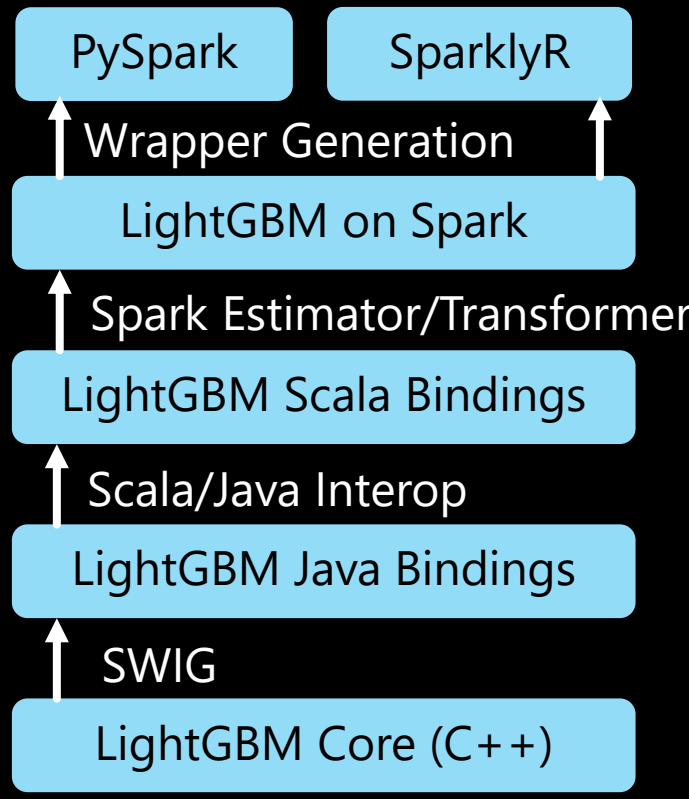
Distributed Model Interpretability with LIME

Deep Learning with CNTK

Example Backend: LightGBM on Spark

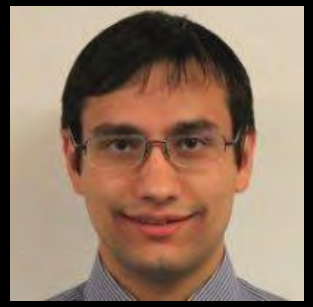
- Barrier Execution for Synchronizing Workers
- Fast Socket/MPI communication
- `mapPartitions` for Transformer





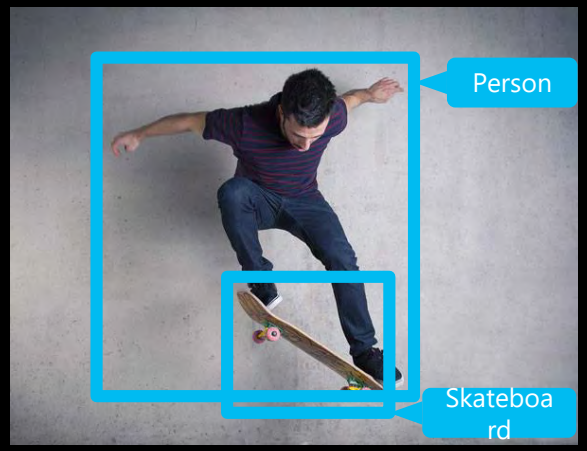
Framework	Time(s)	Area under ROC
XGBoost	52.60	.808
SparkML GBT	82.78	.788
LightGBM	**45.39**	**.812**

Ilya Matiach, ilmat@microsoft.com
Developer, Azure ML



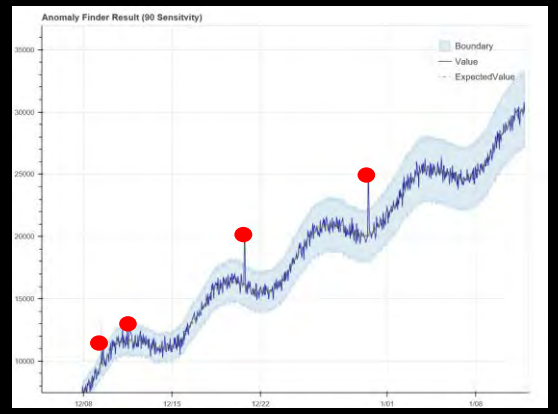
Cognitive Services

- High quality pre-built intelligent services
- No time intensive model training or deployment
- Leverage Microsoft Research and Azure ML
- **Available as Docker Containers**



Bing

"Hello World"



Place Time Range

I had a wonderful trip to Seattle last week and even visited the Space Needle 2 times!

Place

En-US | 84% positive

Vision

Object, scene, and activity detection

Face recognition and identification

Celebrity and landmark recognition

Emotion recognition

Text and handwriting recognition (OCR)

Customizable image recognition

Video metadata, audio, and keyframe extraction and analysis

Explicit or offensive content moderation

Speech

Speech transcription (speech-to-text)

Custom speech models for unique vocabularies or complex environment

Text-to-speech

Custom Voice

Real-time speech translation

Customizable speech transcription and translation

Speaker identification and verification

Language

Language detection

Named entity recognition

Key phrase extraction

Text sentiment analysis

Multilingual and contextual spell checking

Explicit or offensive text content moderation

PII detection for text moderation

Text translation

Customizable text translation

Contextual language understanding

Decision

Q&A extraction from unstructured text

Knowledge base creation from collections of Q&As

Semantic matching for knowledge bases

Customizable content personalization learning

Search

Ad-free web, news, image, and video search results

Trends for video, news

Image identification, classification and knowledge extraction

Identification of similar images and products

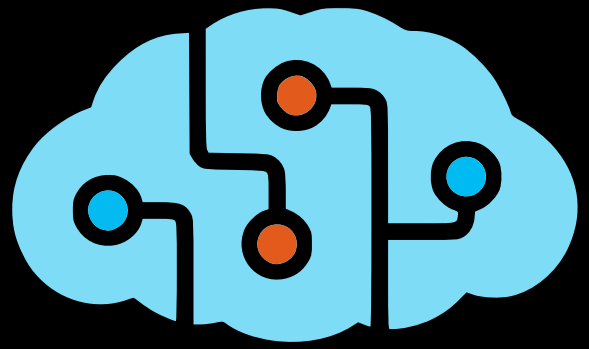
Named entity recognition and classification

Knowledge acquisition for named entities

Search query autosuggest

Ad-free custom search engine creation

Azure Cognitive Services on Spark



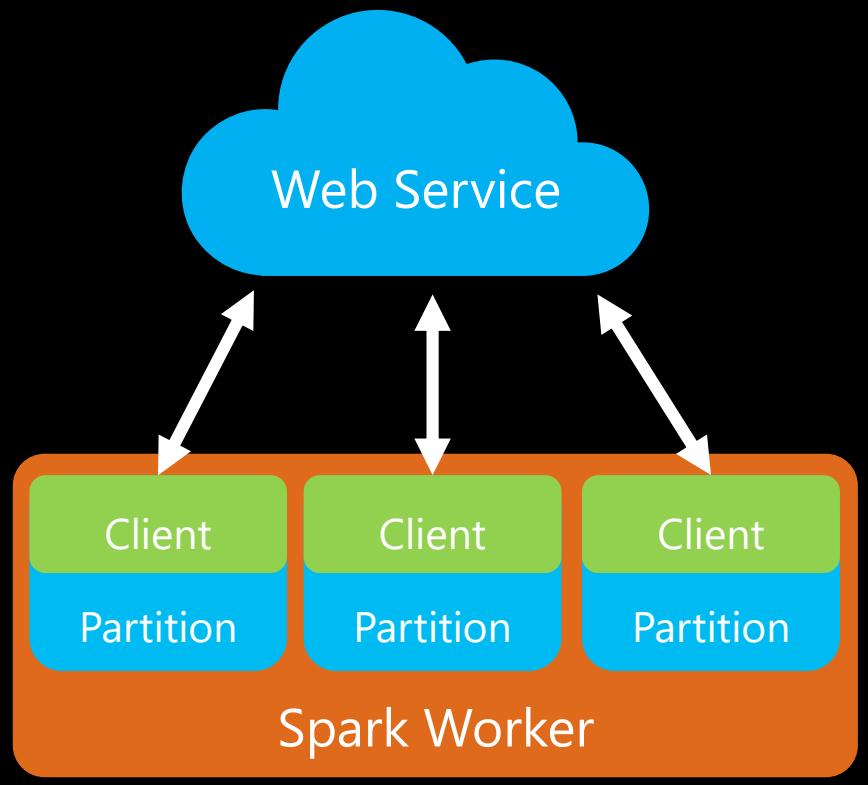
- Easy to use integration between Spark and the Azure Cognitive Services
- Composable and pipelinable with all other SparkML models!
- Exponential Backoffs, Backpressure, Batching, Async Parallelism
- Fully Fluent API

```
val df = new TextSentiment()
  .setTextCol("text")
  .setOutputCol("sentiment")
  .transform(inputs)
```

Features	Time (s)	Errors #
None	30.8	18993
EBO+BP	1163.0	0
EBO+BP+B	57.1	0
EBO+BP+B+P	**49.7**	**0**

HTTP on Spark

- Full Integration between HTTP Protocol and Spark SQL
- Spark as a Microservice Orchestrator
- Spark + X
- Support for all Spark Languages



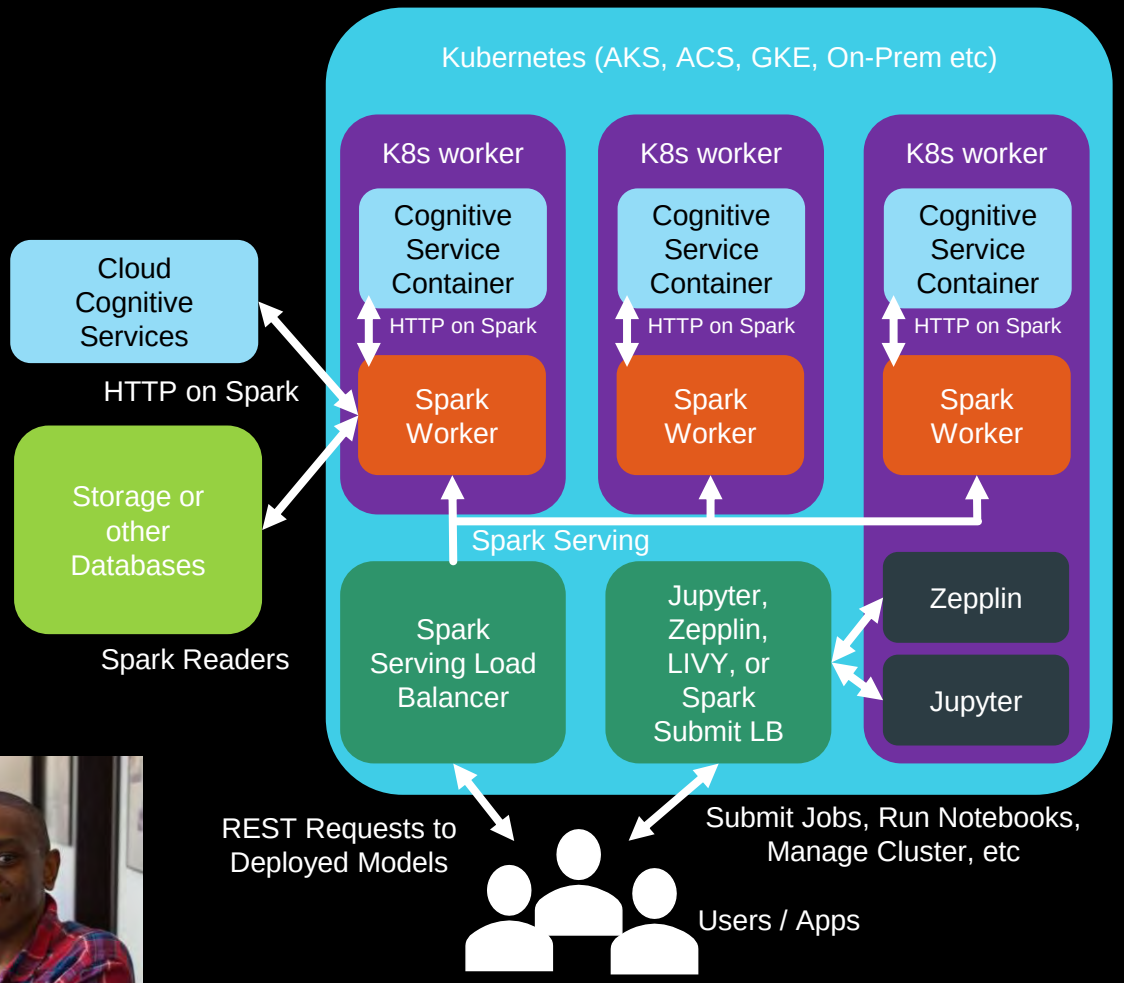

```
df = SimpleHTTPTransformer()
  .setInputParser(JSONInputParser())
  .setOutputParser(JSONOutputParser()
    .setDataType(schema))
  .setOutputCol("results")
  .setUrl(…)
```

Deploying on Kubernetes

- Works on any k8s cluster
- Helm: Package Manager for Kubernetes

```
helm repo add microsoft \
    https://microsoft.github.io/charts/repo
helm update

helm install microsoft/spark --version 1.0.0
```



Dalitso Banda, dbanda@microsoft.com
Microsoft AI Development Acceleration Program

Model Deployment with Spark Serving

- Sub-millisecond RESTful Model Deployment on Spark Clusters

Batch API:

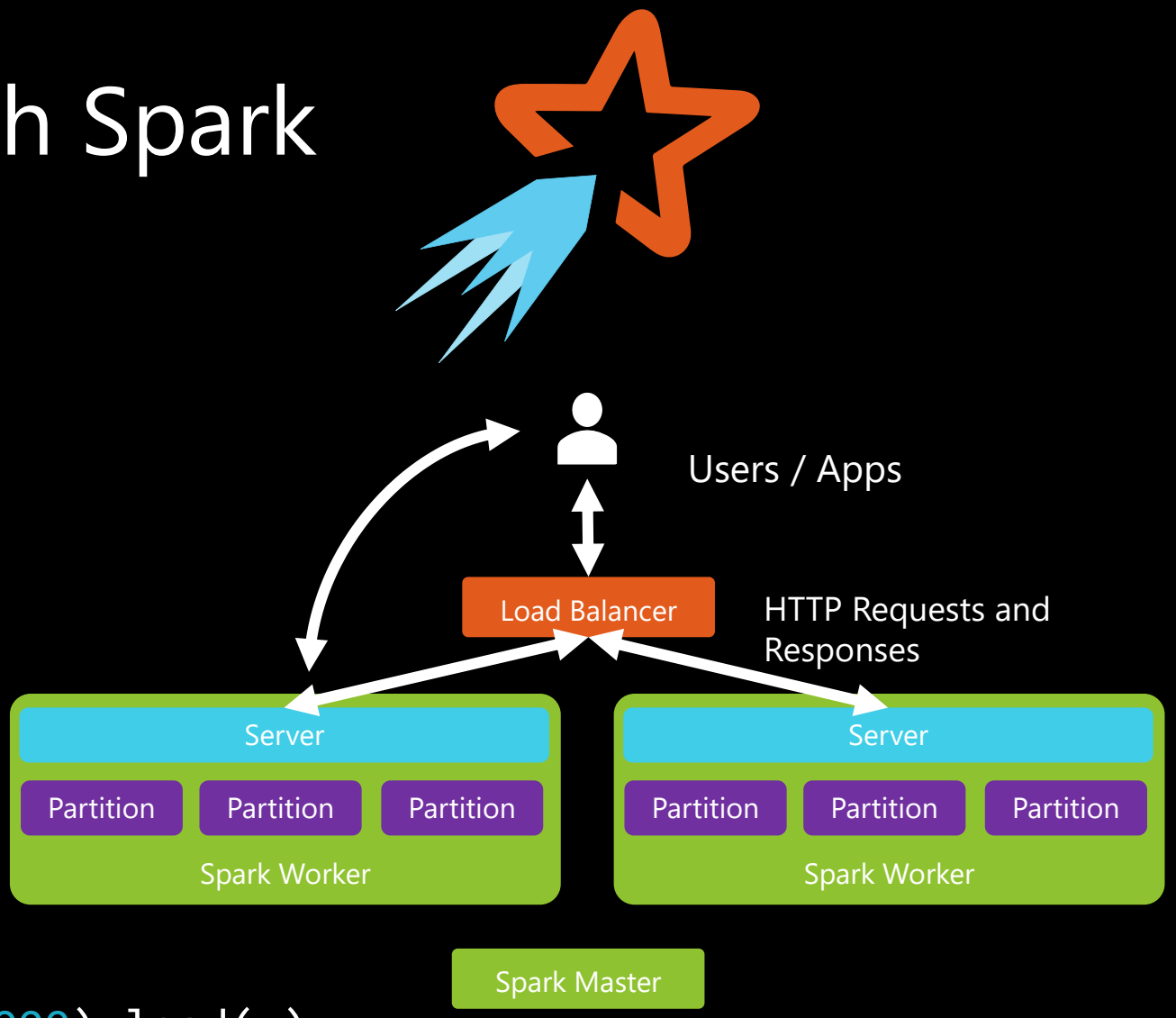
```
spark.read.parquet.load(…)
  .select(…)
```

Streaming API:

```
spark.readStream.kafka.load(…)
  .select(…)
```

Serving API:

```
spark.readStream.server("0.0.0.0", 5000).load(…)
  .select(…)
```



AI for Earth

Snow Leopard Trust

Endangered Status Matters

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Snow leopard no longer 'endangered'

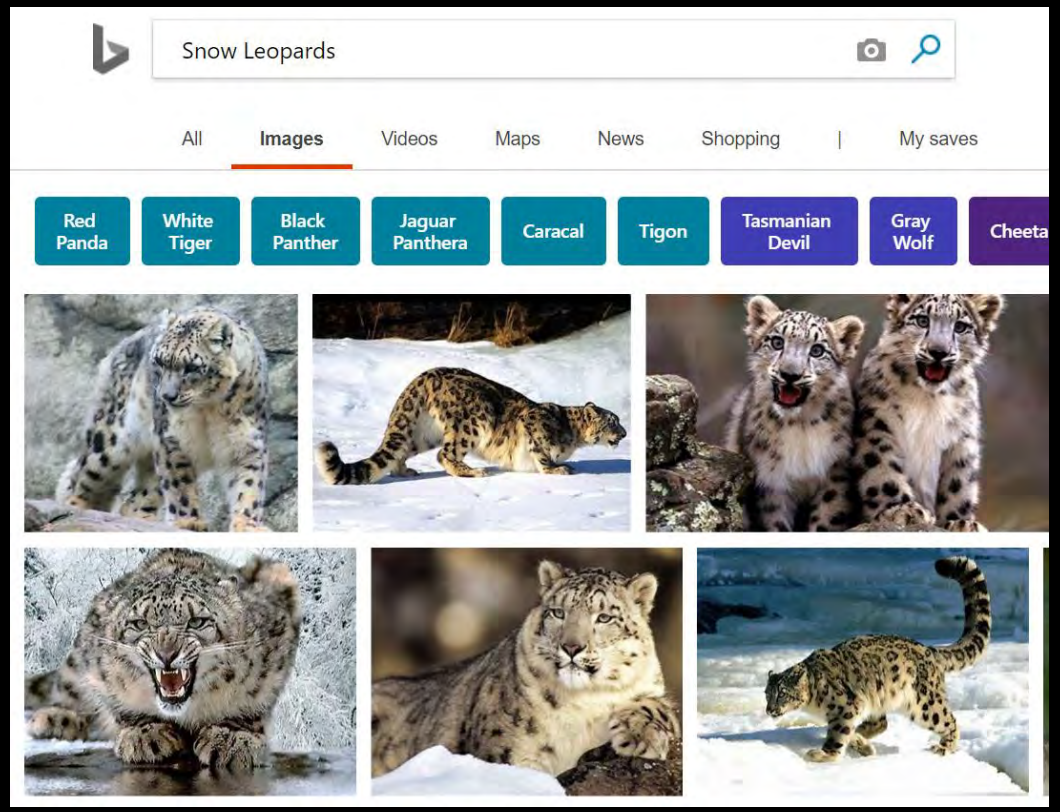
14 September 2017 Share

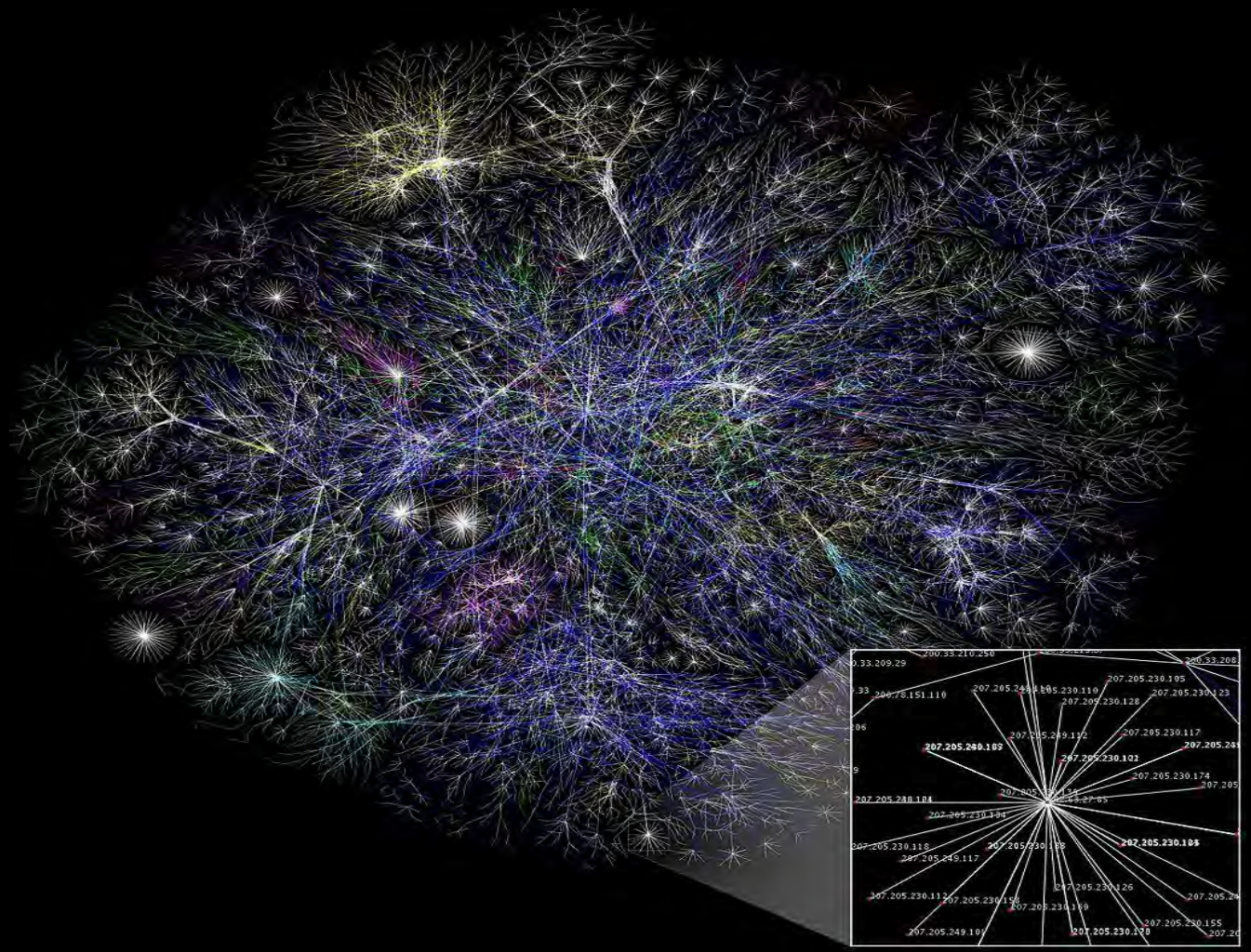
Statement on IUCN Red List Status Change of the Snow Leopard

The Snow Leopard Trust, one the leading conservation organizations working to protect this cat, opposes the IUCN's decision to change the snow leopard's Red List status from 'Endangered' to 'Vulnerable'.

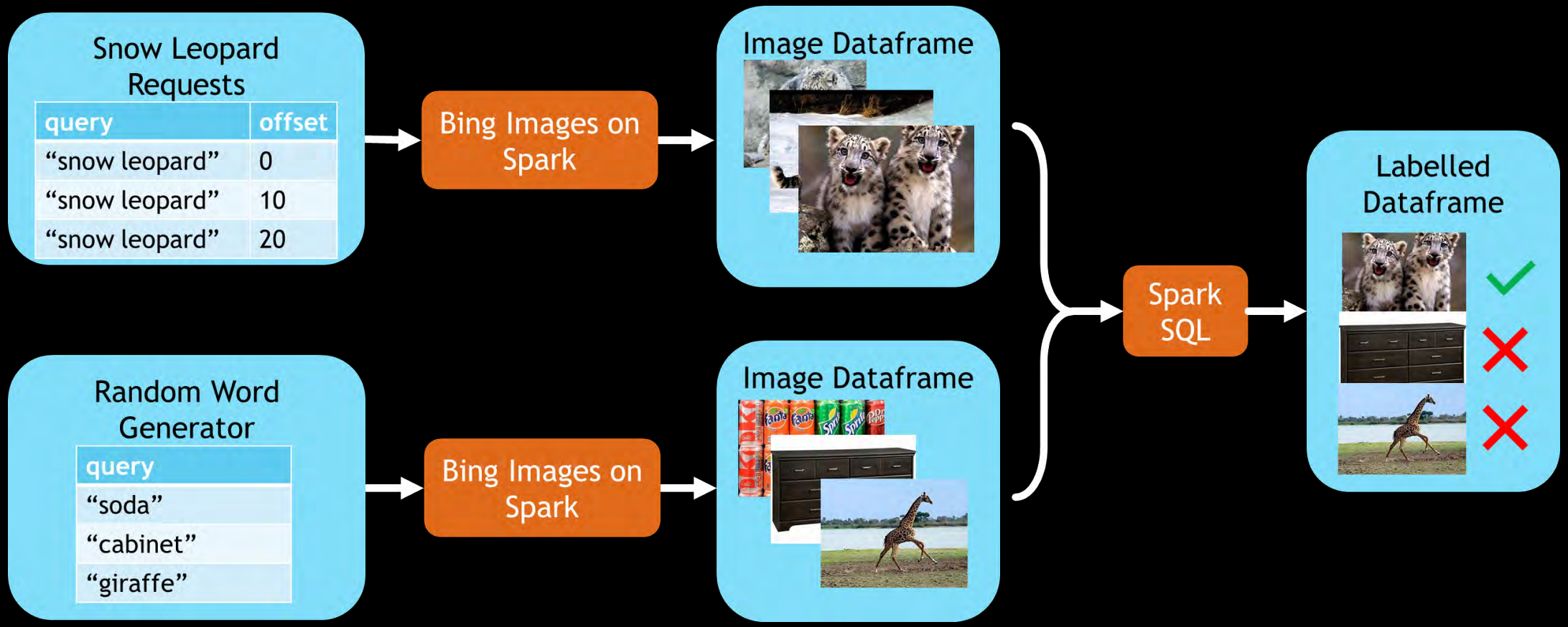
Remote Camera Trapping

Creating a labelled Training Dataset

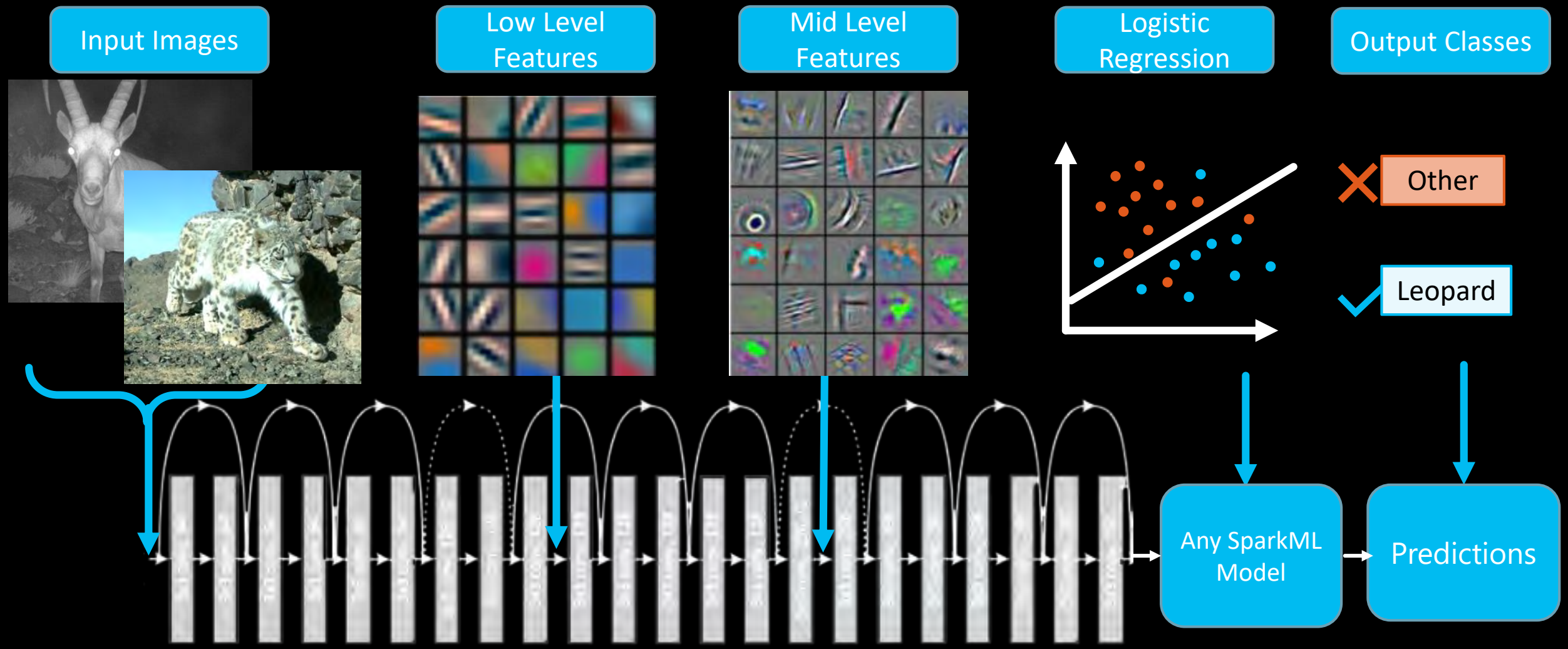




Creating a labelled Training Dataset



Transfer Learning with ResNet 50

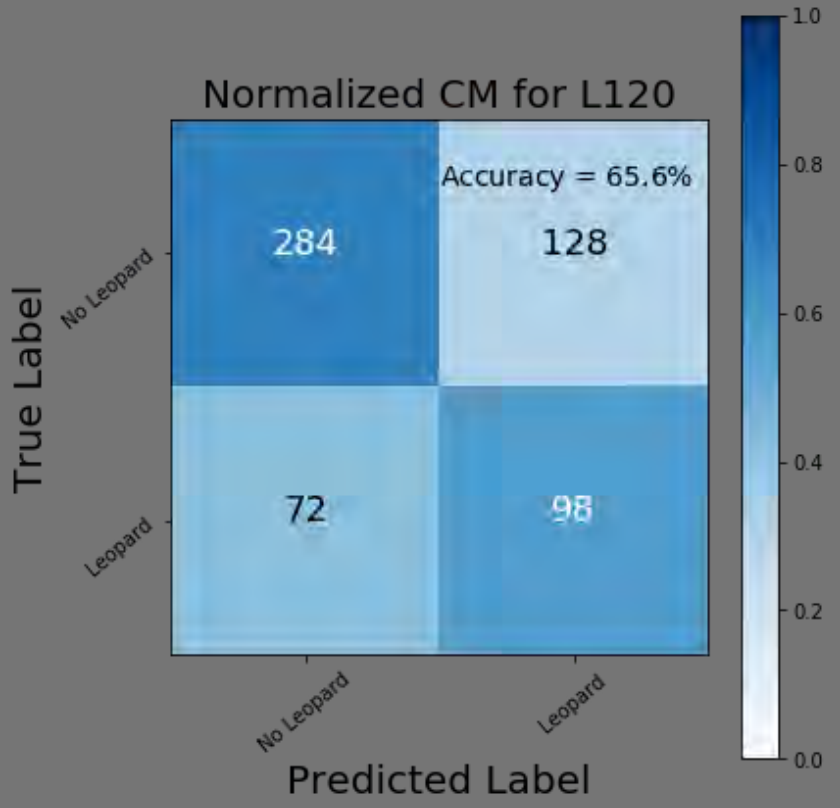


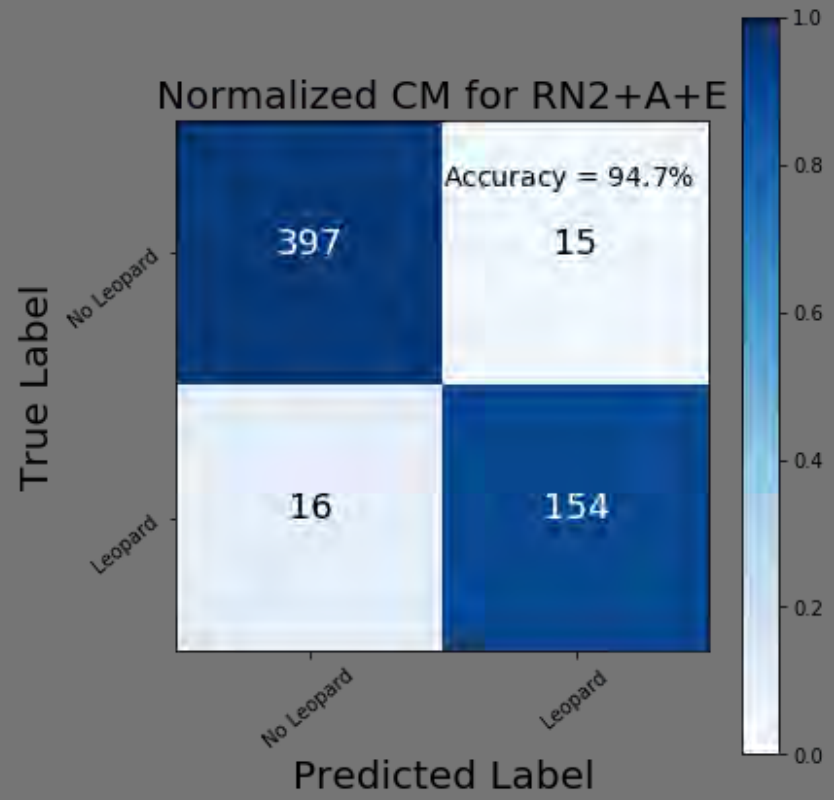
Filters from Zeiler + Fergus 2013

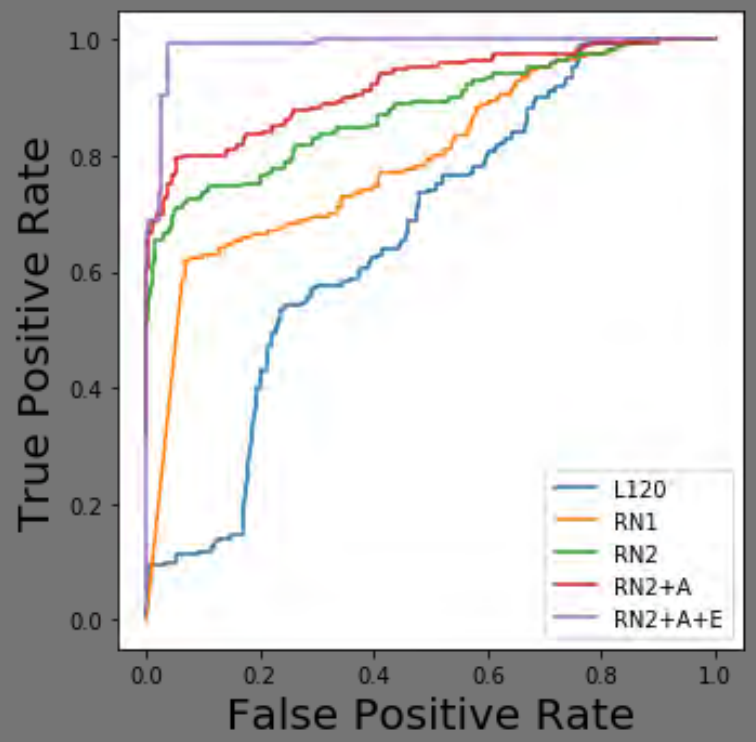
Performance

Without Deep Featurization

With Deep Featurization, Augmentation, and Temporal Ensembling



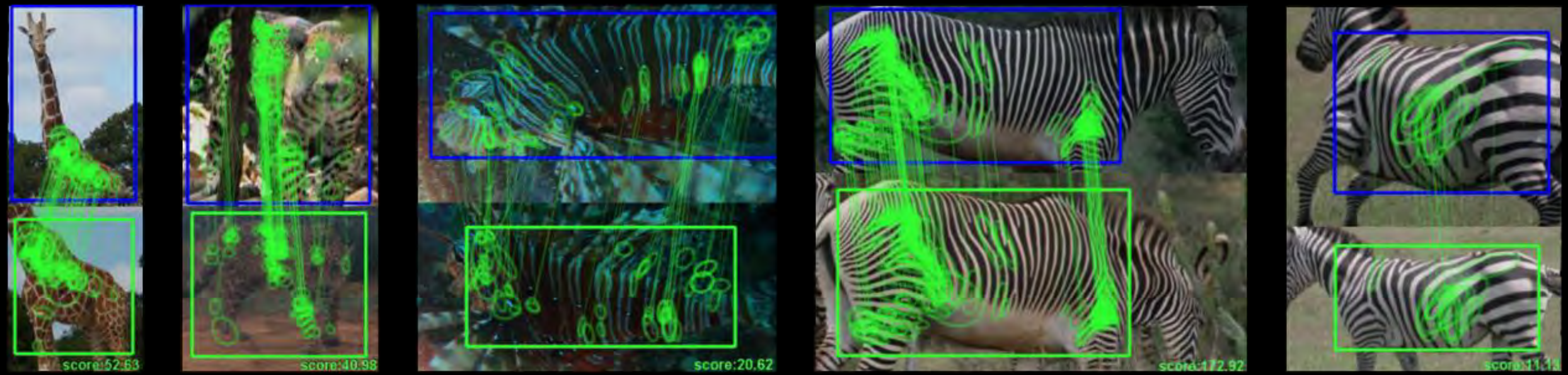




Accuracy 65.6%

Accuracy 94.7%

Goal: Identify Individual Leopards



Source: HotSpotter - Patterned Species Instance Recognition

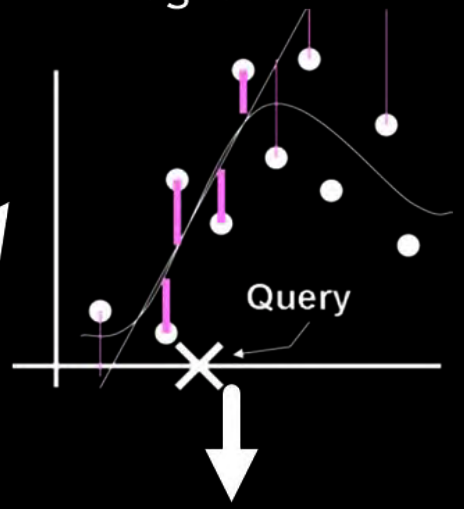
Automating Detection with LIME on Spark

Input Image

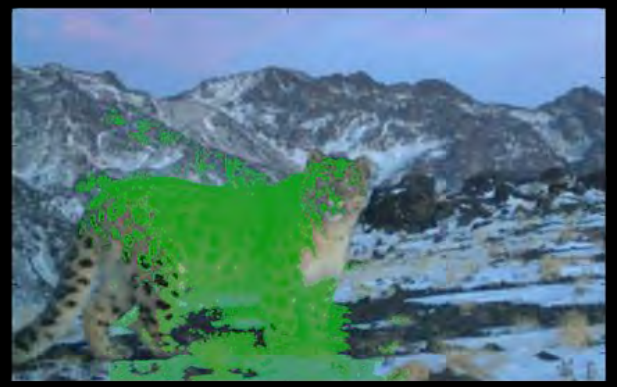
P(leopard) = .88

Perturbed Photos	P(leopard)
	.92
	.56
	.003

Locally Weighted Regression

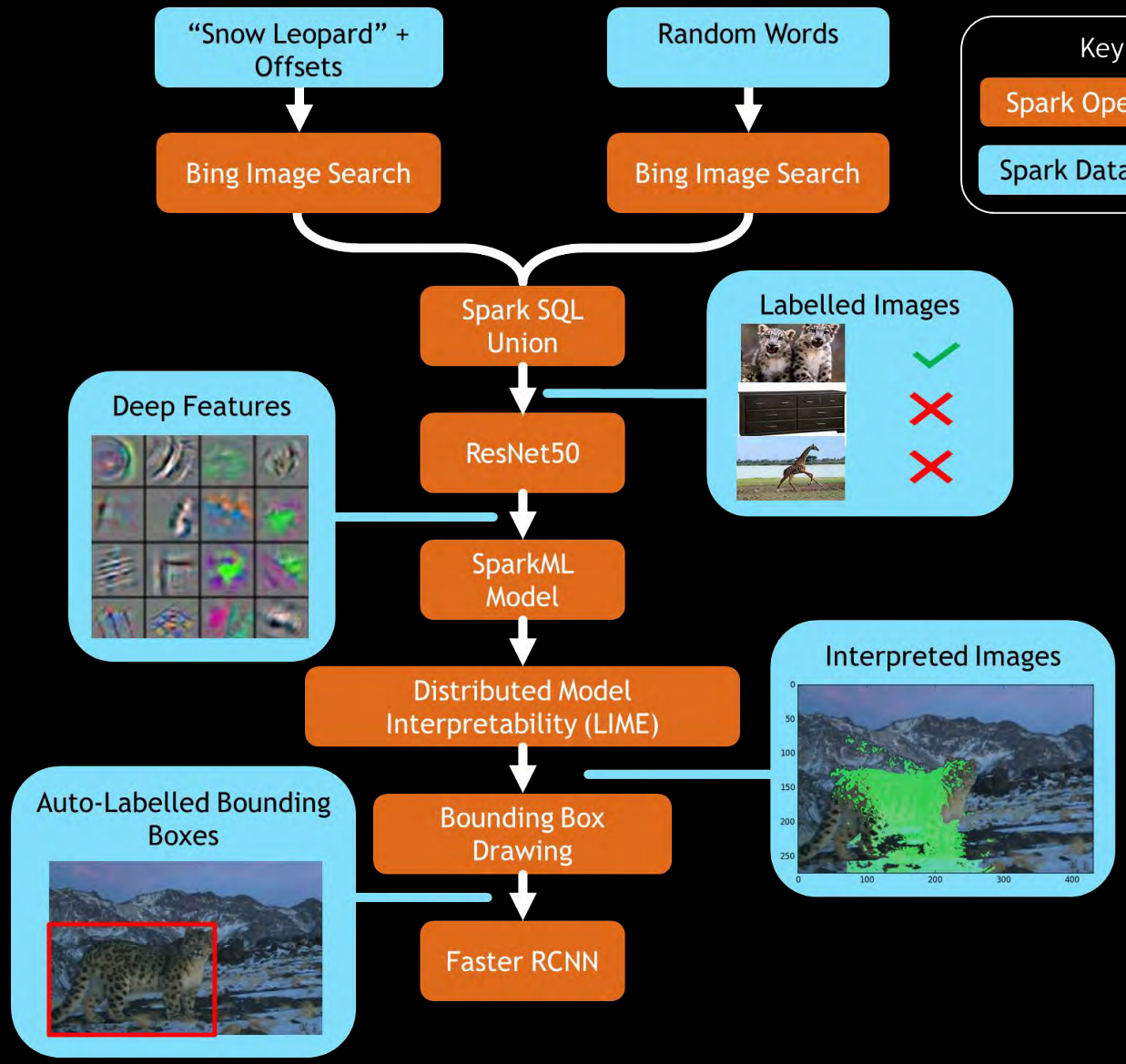


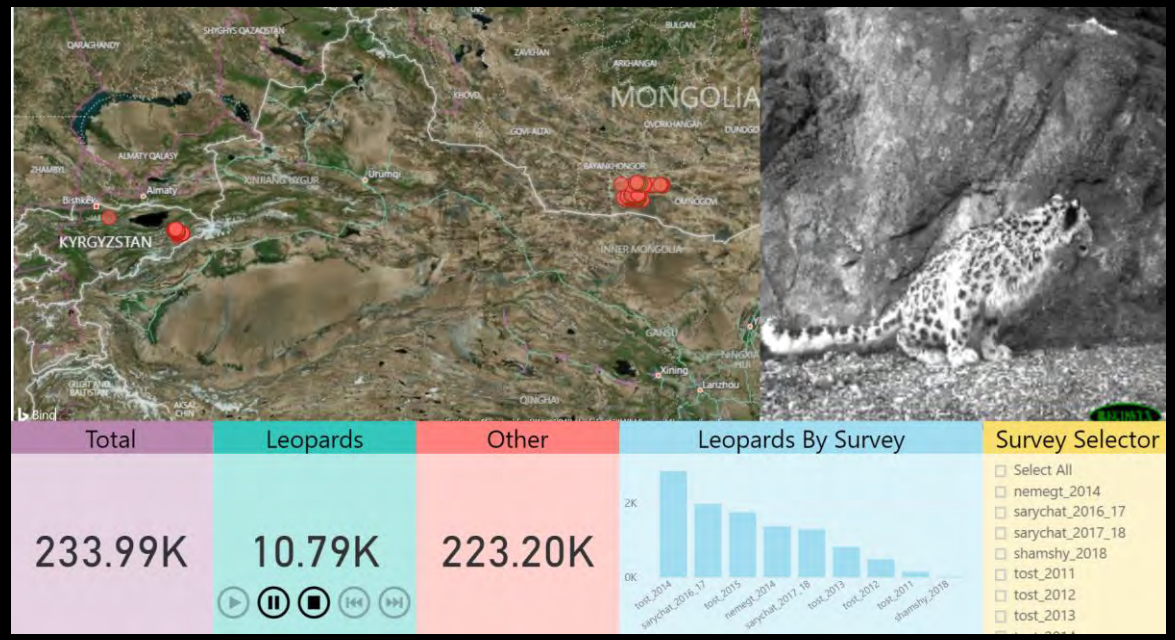
Explanation



LIME on Spark

End to End Architecture

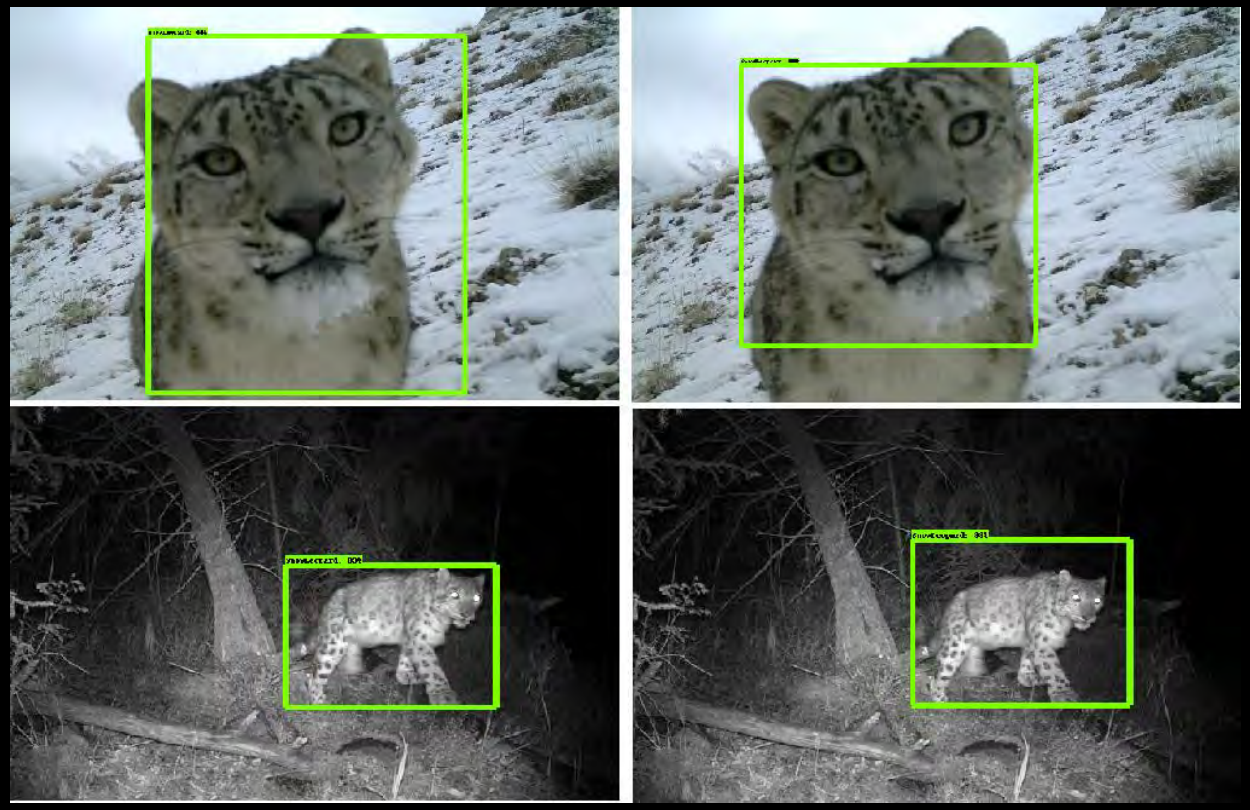




Results

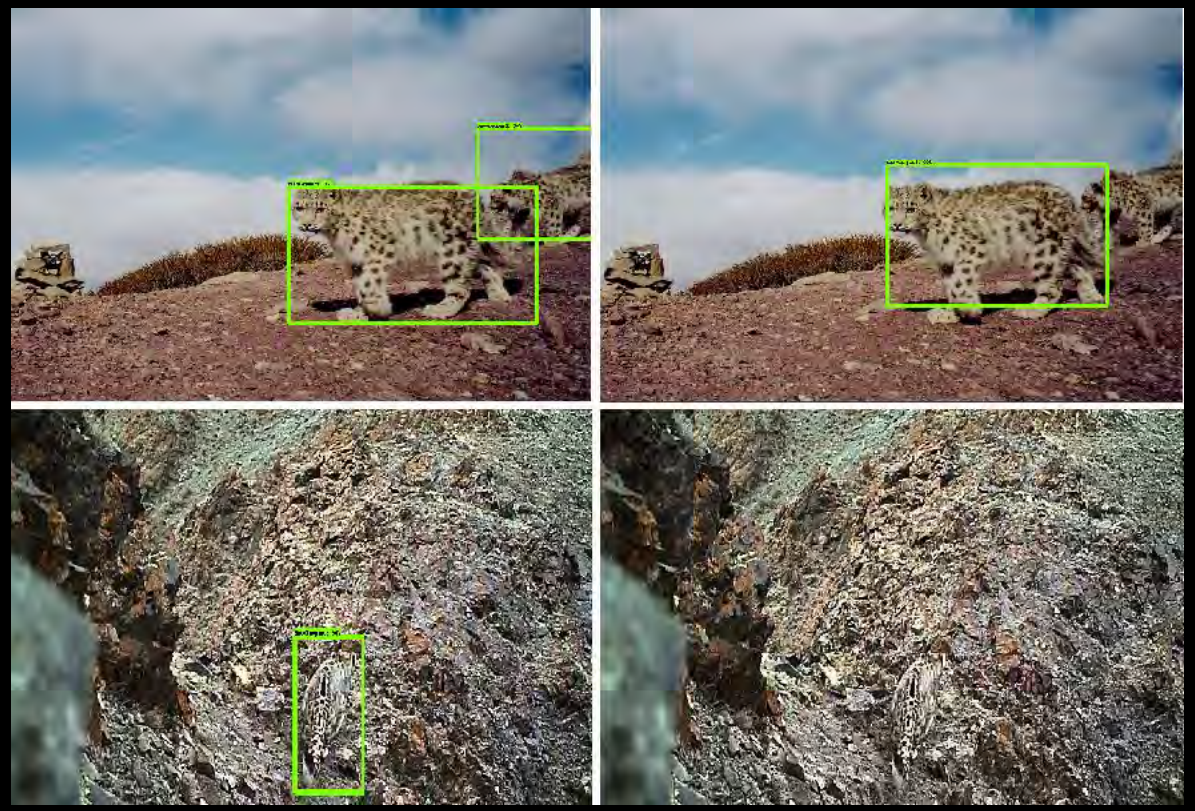
Human Labels

Unsupervised FRCNN Outputs



Human Labels

Unsupervised FRCNN Outputs



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Microservice Orchestration

Multilingual Binding Generation

www.aka.ms/spark

Azure/mmlspark

Thanks to

- You all!
- **Ilya Matiach: LightGBM on Spark**
- **Markus Cozowicz: VW on Spark**
- Sudarshan Raghunathan, Christina Lee, Daniel Ciborowski, Eli Barzilay, Tong Wen, Pablo Castro, Chris Hoder, Ryan Gaspar, Henrik Neilsen, Andrew Schonhoffer, Joseph Sirosh
- Microsoft NERD Garage Team + MIT Externship Program
- Snow Leopard Trust: Koustubh Sharma, Rhetick Sengupta, Jeff Brown, Michael Despines
- Microsoft Development Acceleration Team:
  - Dalitso Banda, Casey Hong, Karthik Rajendran, Manon Knoertzer, Tayo Amuneke, Alejandro Buendia
- Azure CAT, AzureML, and Azure Search Teams

Get in Touch

- Support: mmlspark-support@microsoft.com
- Me: marhamil@microsoft.com
- Github : Azure/mmlspark
- Website: www.aka.ms/spark
- Paper: www.aka.ms/spark-paper
- Contributions Welcome!
- Check out our MSR Podcast on Oct 2

Backup Slides

AI for Cultural Institutions

Microsoft

THE MET

MIT

Celebrating 2 years of Open Access at The MET

- In 2016 The MET Released 400k images under open access
- This past winter the MET released a new subject-keyword dataset of image annotations
- MIT, The MET, and Microsoft participated in a 3-day hackathon to create intelligent experiences using the new collection








OA OpenAccess

Goals:

Create new works of art

Use new work to explore existing art

Explore further with intelligent search

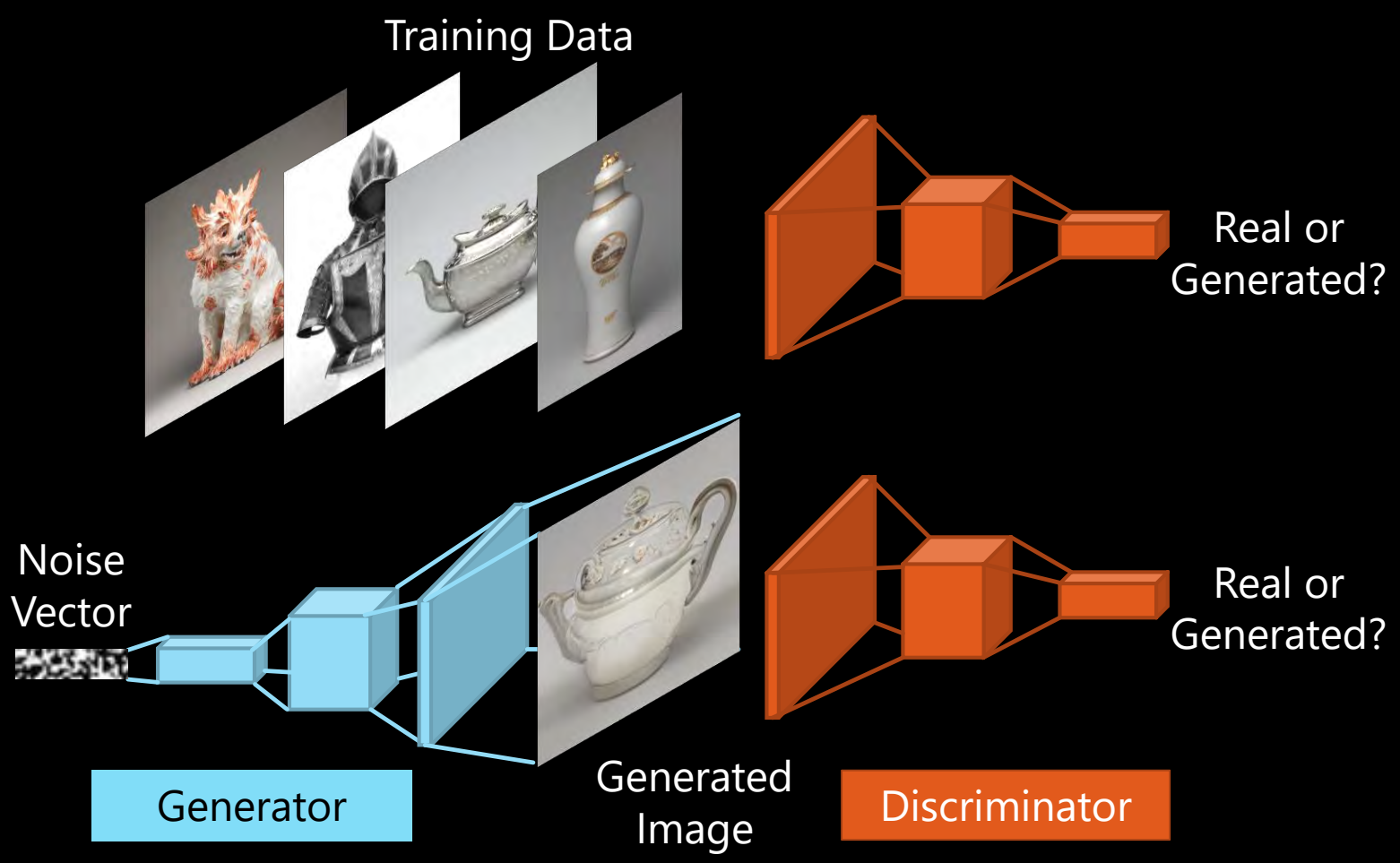
Needed Technologies:

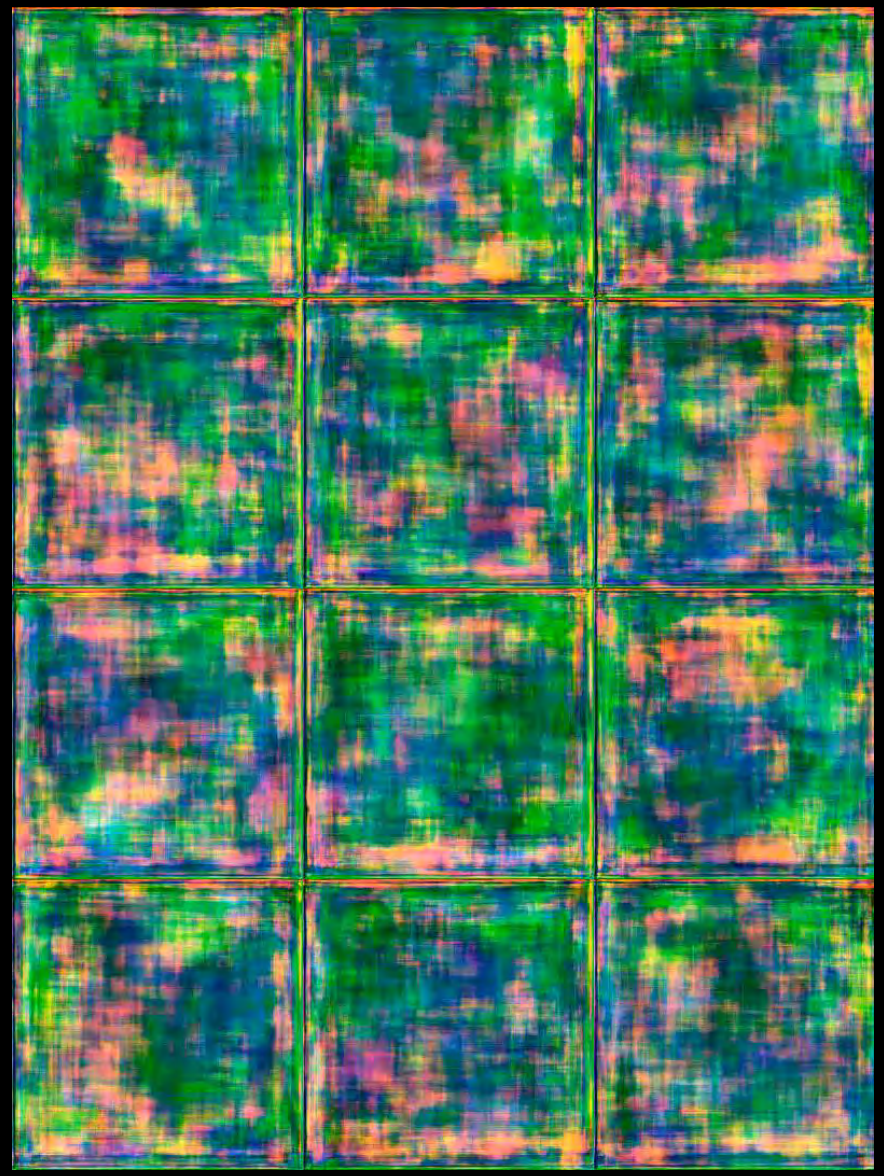
Generative Adversarial Networks

Reverse image search

Elasticsearch with Cognitive Services

Generative Adversarial Art





Custom Reverse Image Search

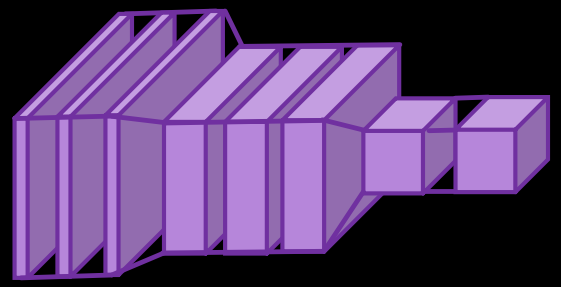
Query Image

ResNet Featurizer

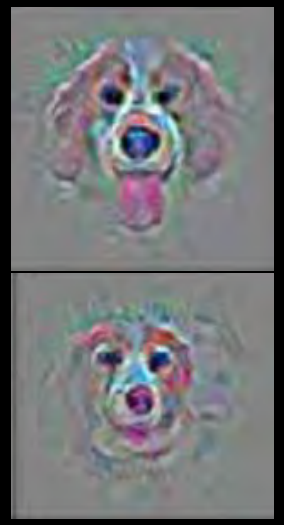
Deep Features

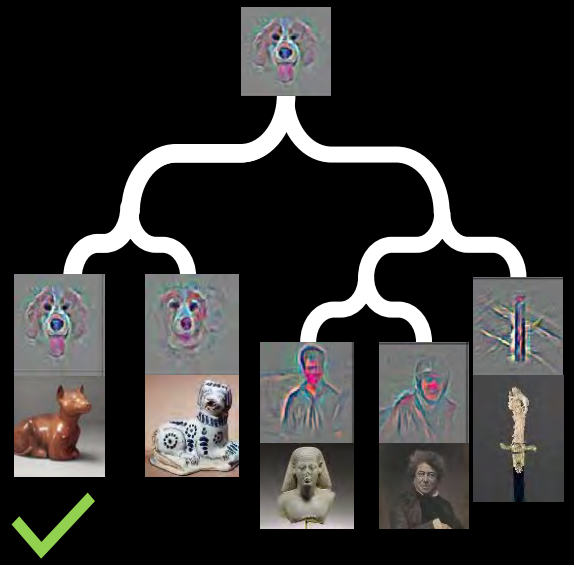
Fast Nearest Neighbor Lookup

Closest Match



MMLSpark

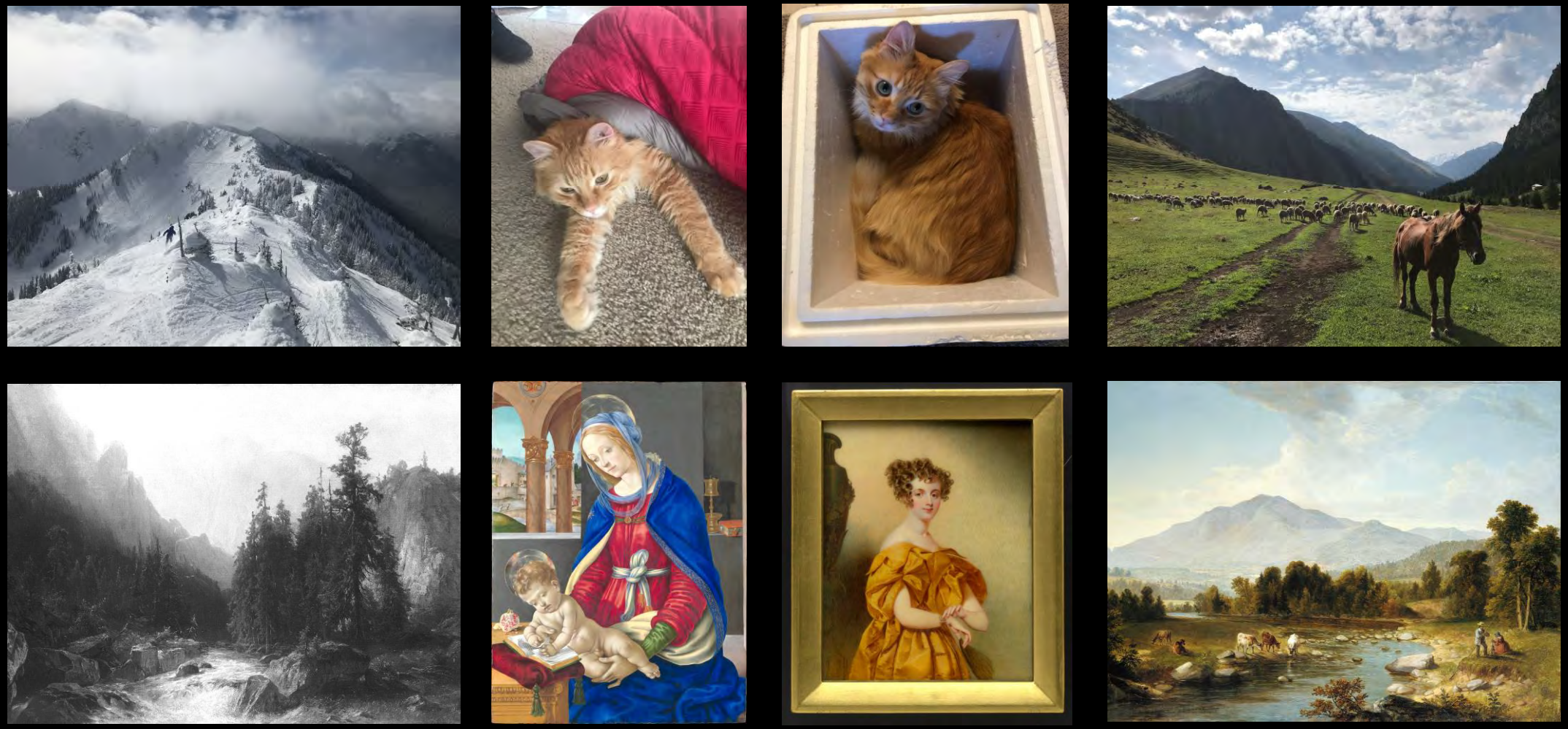




SparkML LSH or Annoy

Filters from Zeiler + Fergus 2013

Example Nearest Neighbors



Intelligent Search Index

- Pipe images through Computer Vision API to annotate image for searching
- Stream images and intelligent annotations to Azure Search

Query Image:

Describe Image Output:

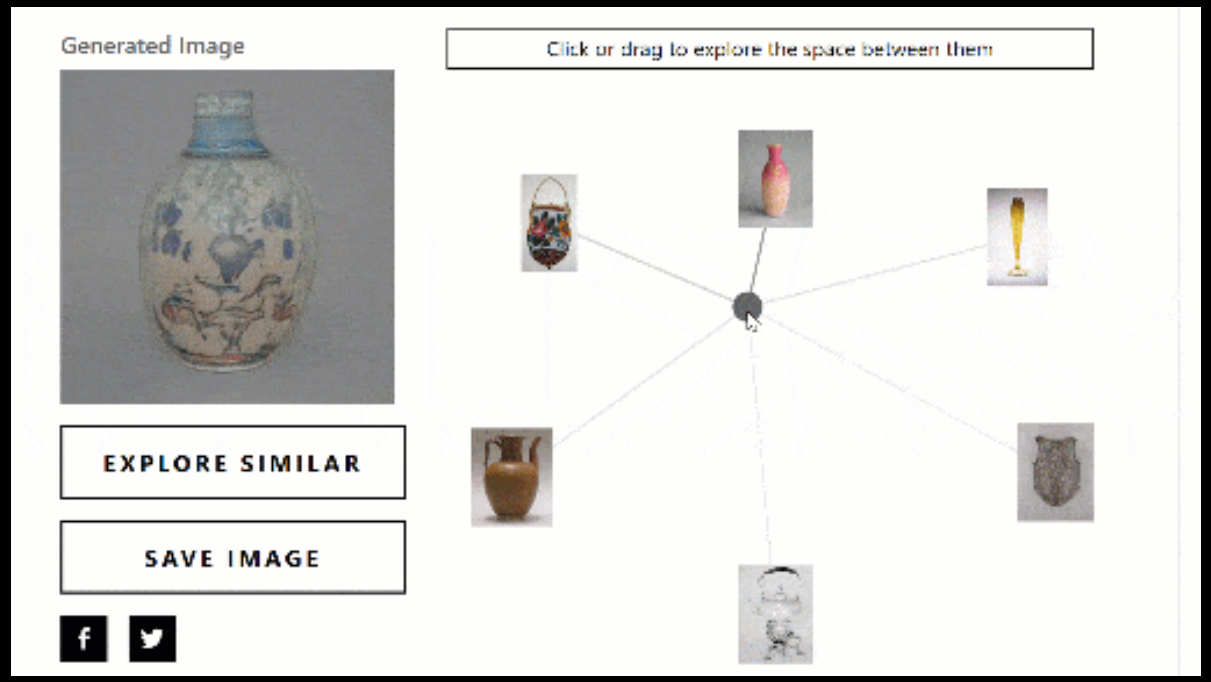
A picture containing a person

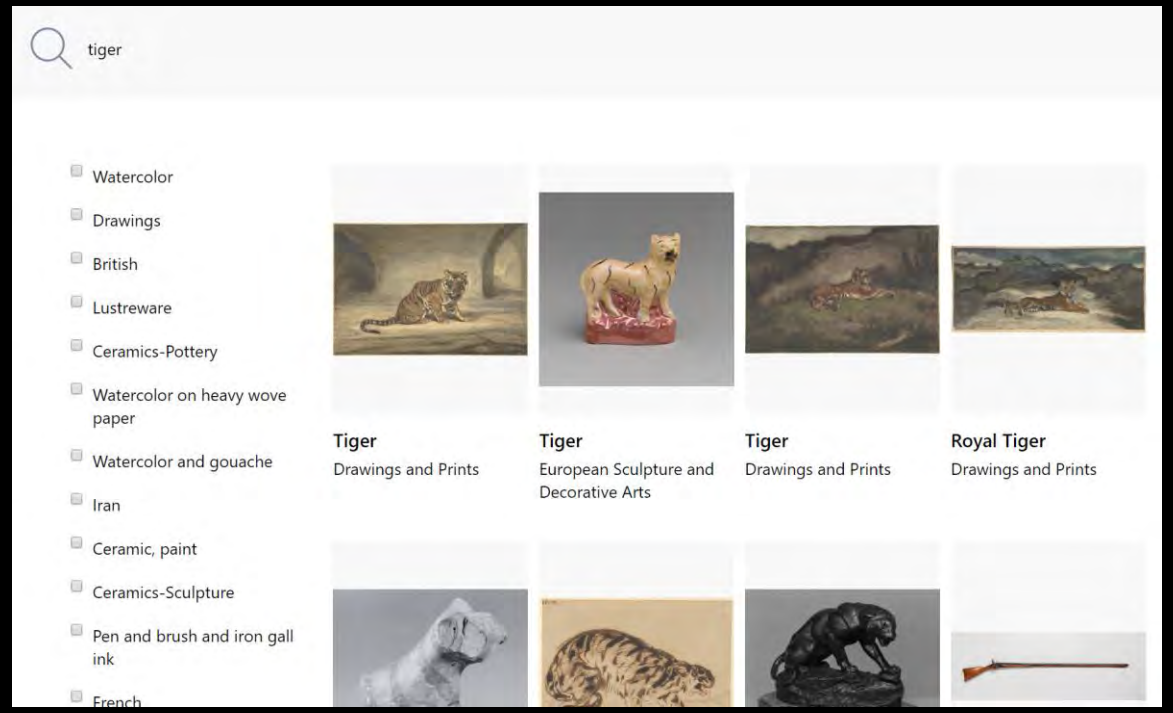
A picture containing a glass, cup

A fish swimming underwater

Deep Feature Nearest Neighbors:

End Application: Gen Studio

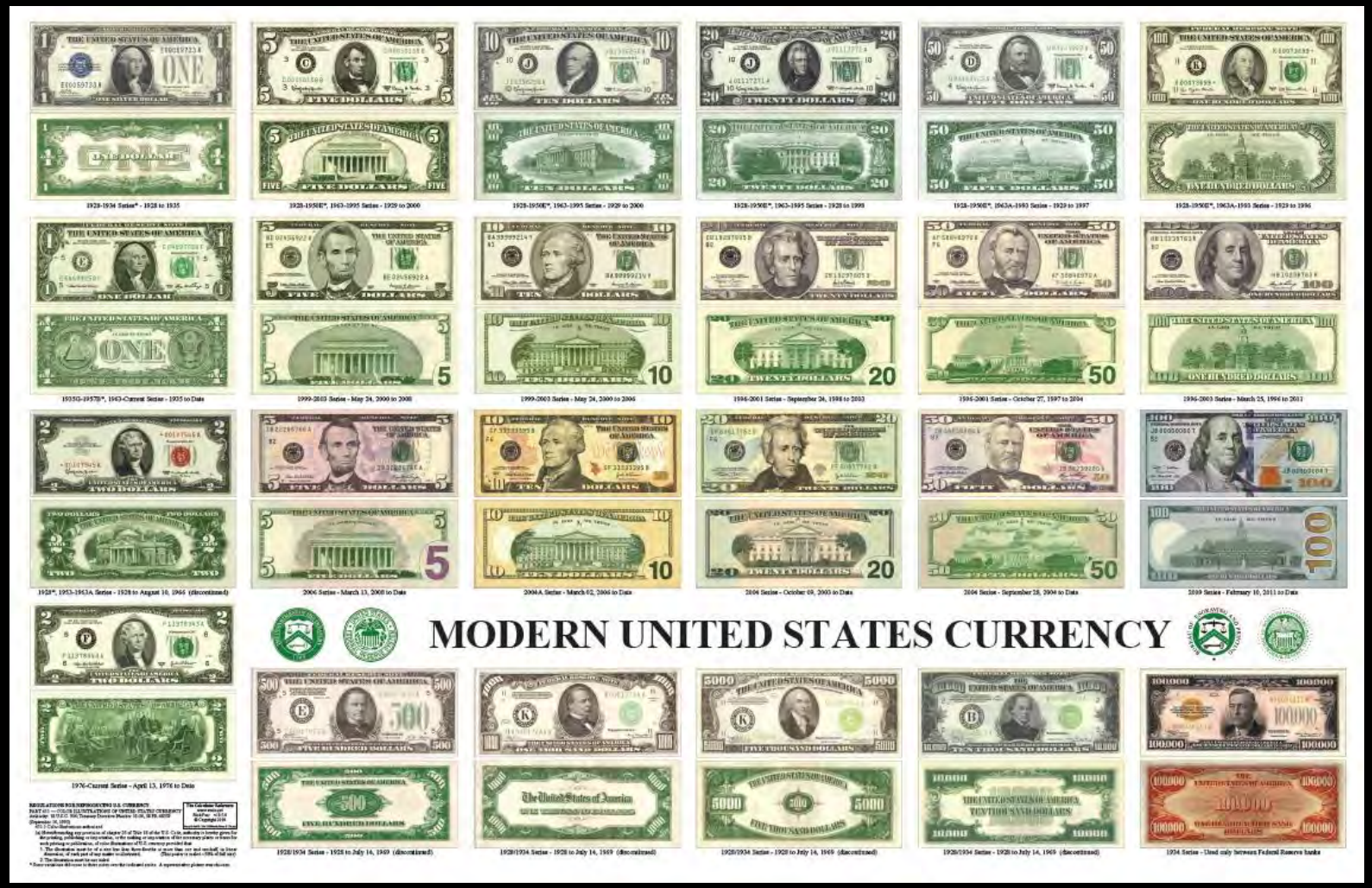




AI for Accessibility

Seeing AI

Currency Identification



A Familiar Architecture…