

# Project Batcomputer

A working DevOps  
implementation for Machine  
Learning

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*v0.0.2 (Alpha)*

# Background

## Motivation

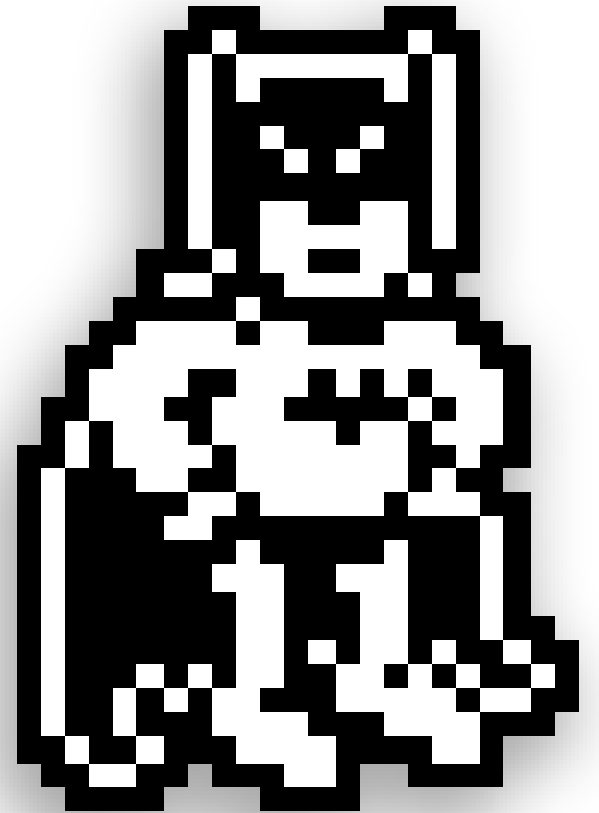
- Understand challenges in operationalisation of ML models
- Existing processes (e.g. Azure Machine Learning Service) deemed problematic
- “DevOps for AI”

## Why Batcomputer?

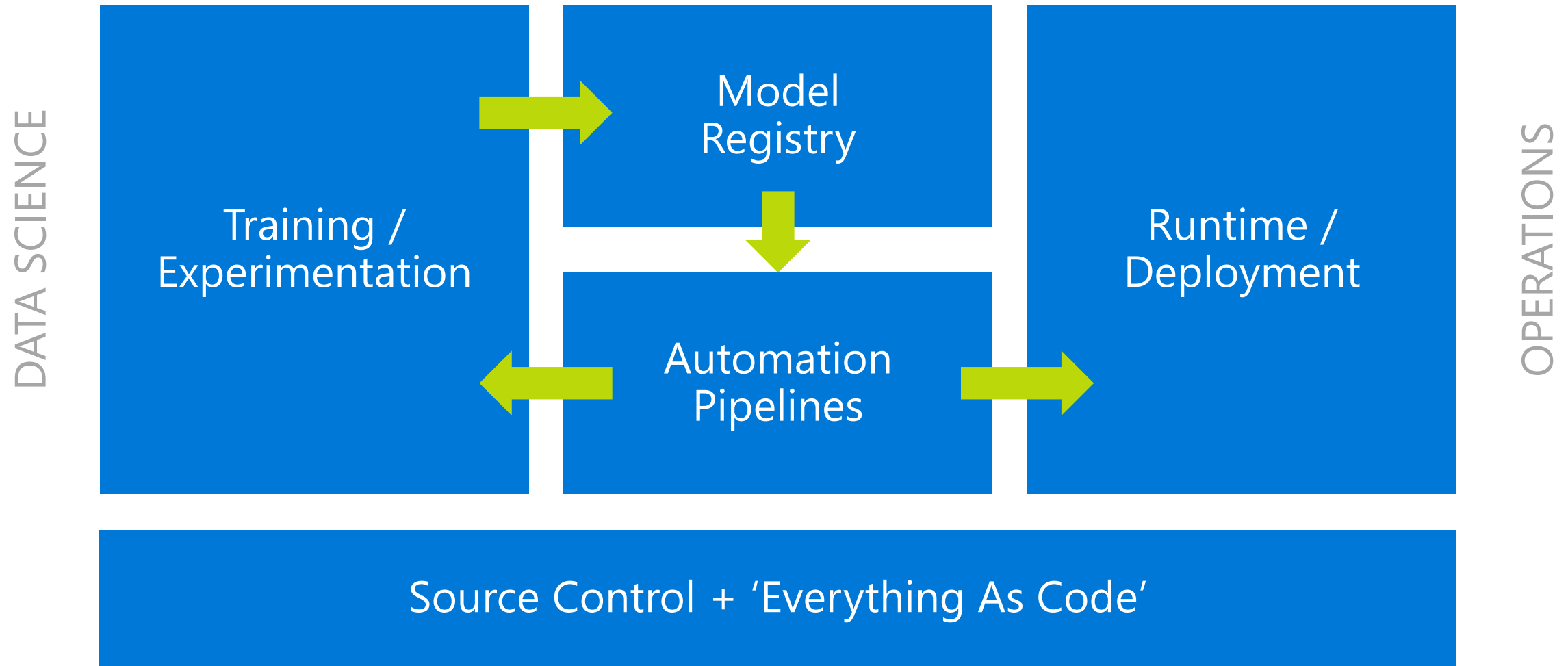
- Police recorded crime and outcomes open data tables
- <https://www.gov.uk/government/statistics/police-recorded-crime-open-data-tables>
- Build model of a given crime and/or region to predict outcome

# Core Principals & Benefits

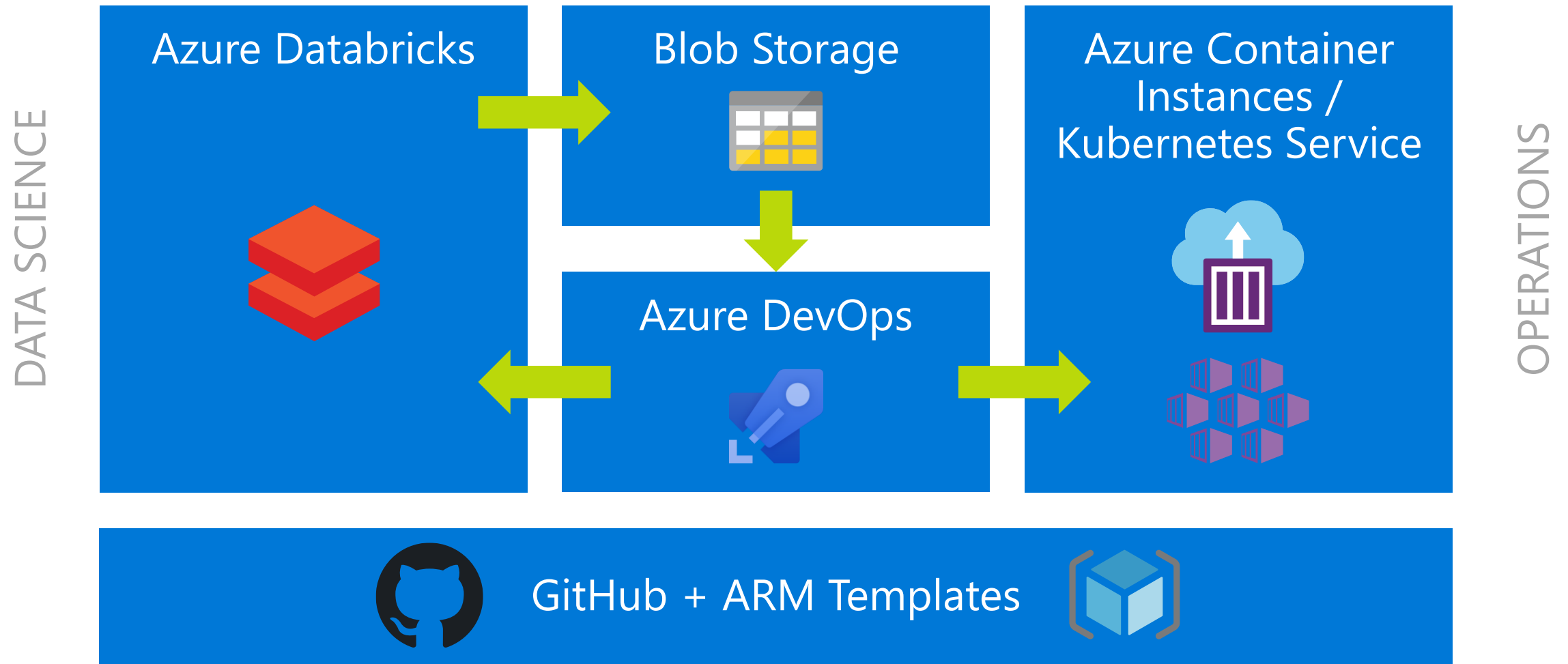
- Decouple model training experiments from operations/runtime
- Automated training, API builds & deployment
- Versioned models and API
- Config & infra as code
- Traceability



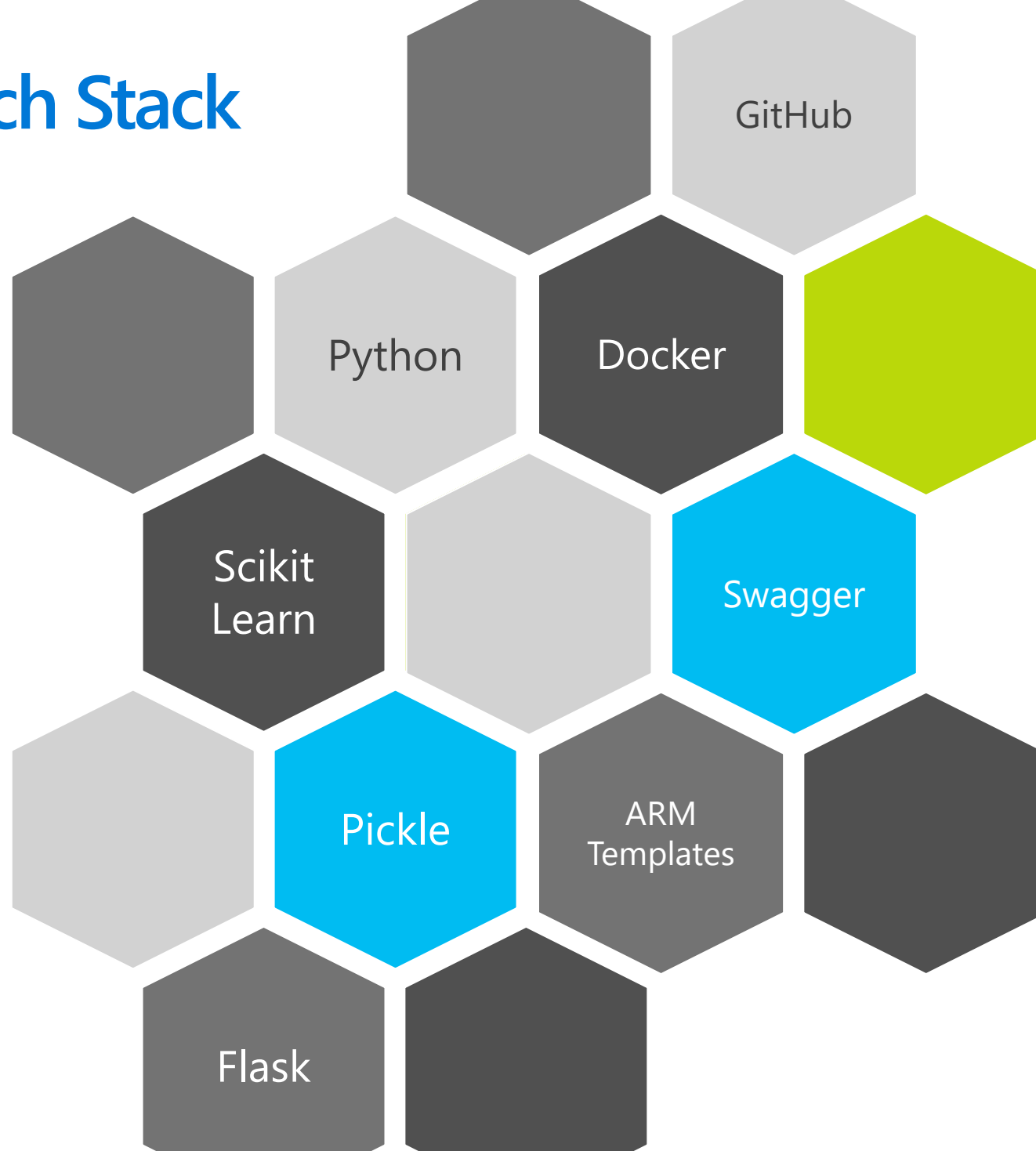
# Conceptual Building Blocks



# Conceptual Building Blocks – Project Batcomputer



# Low Level Tech Stack



# Low Level Technology Stack

Swagger



API niceness

Flask



Web framework / server

Pickle



Serialisation

Scikit-Learn



Main ML framework

Python



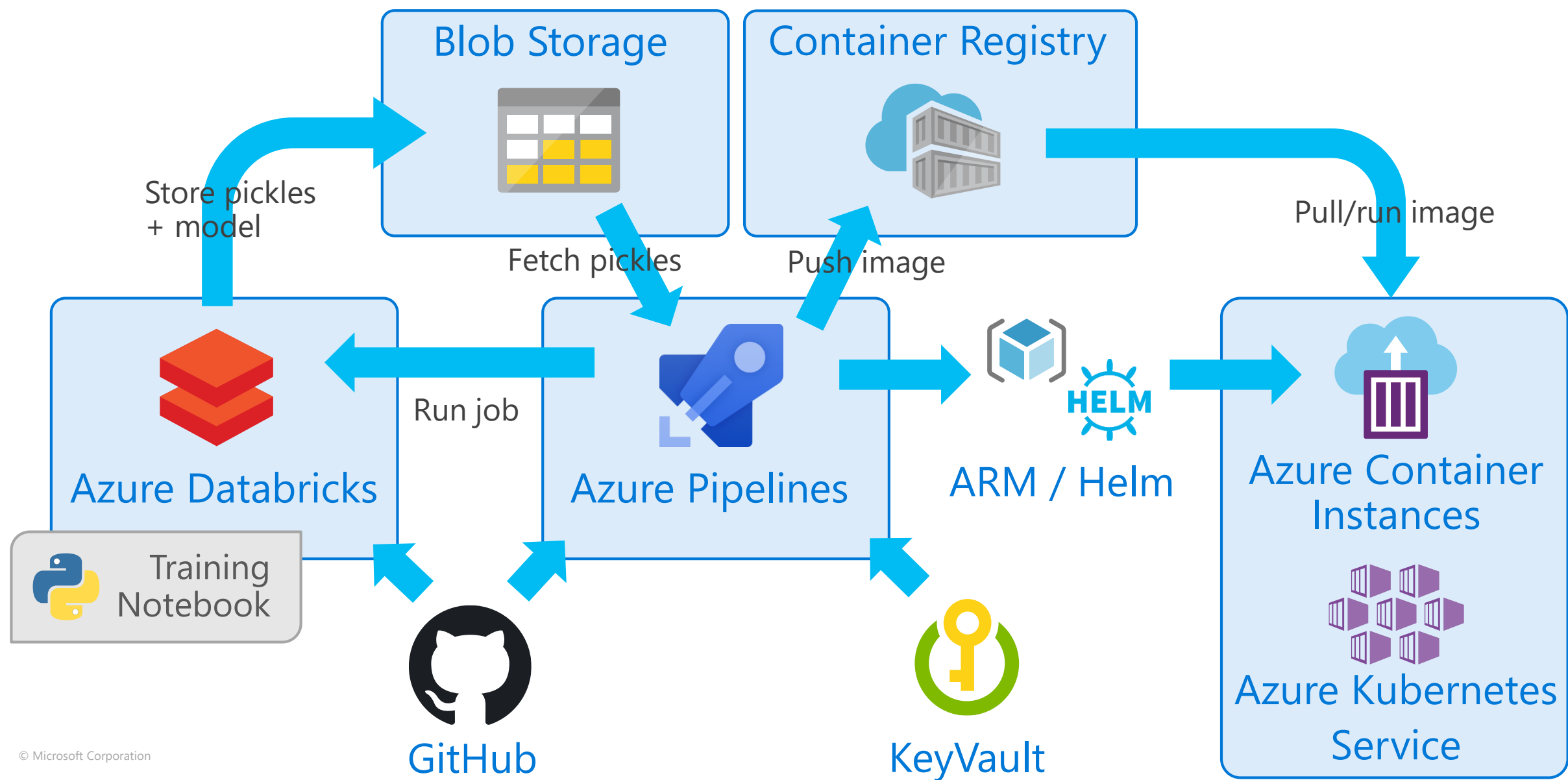
Core language

Docker



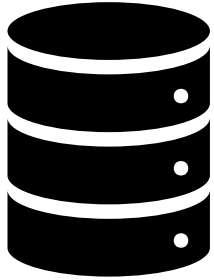
Container Runtime

# Automation – End To End Flow





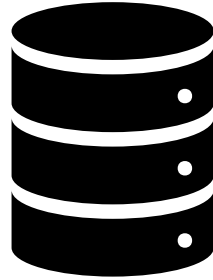
# Pickling – Not Just The Model



**model.pkl**

Scikit-learn model/classifier

Standard object rehydration,  
version sensitive

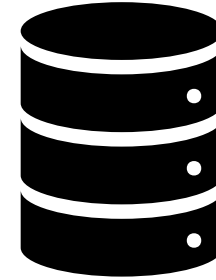


**lookup.pkl**

Python dictionary of  
dictionaries

Mapping  
parameters/strings to num  
for predict function

```
{  
  "gender": {  
    "male": 0,  
    "female": 1  
  }  
}
```



**flags.pkl**

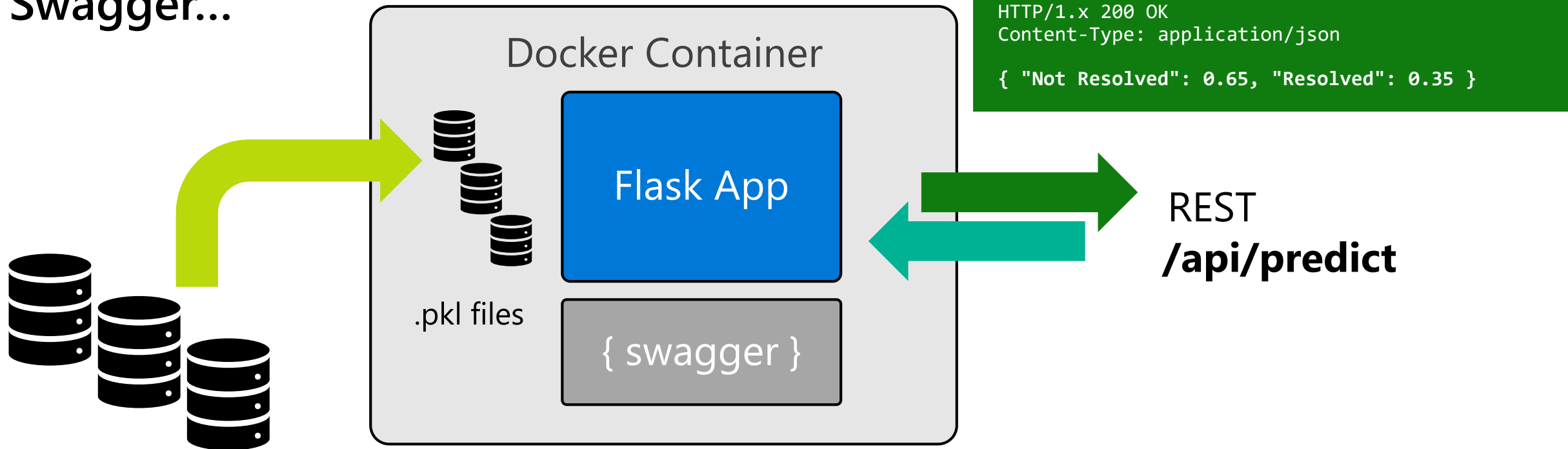
Python array

Maps output of prediction  
function to human readable  
strings or labels

```
[  
  "conviction",  
  "dropped",  
  "settled"  
]
```

# Wrapper App – Flask API

- Uses Flask web framework
- Creates RESTful API for model parameters
- Consumes .pkl files
- Swagger...



# Some Decision Points


- Include model in container image or fetch at runtime?
- Wrapper app – make generic or tied into model?
- Convention based model registry – good enough?
- Use more robust web server than Flask?
- Automate training?



# Swagger

- We want to be RESTful
- Dynamic
  - Generated from lookup & flags pickles at runtime
- Swagger UI
  - For testing & eye candy



 **swagger**

/swagger.json

Explore

## Batcomputer API 1.0.0

[ Base URL: /api ]  
</swagger.json>

REST API getting predictions from the Batcomputer ML model. Model version: 1.0.0

Schemes

HTTP

### Predictions

POST /predict

Get a prediction from the model

Parameters

Try it out

Name	Description
<b>body</b> * required (body)	Request object

Example Value | Model

```
{  "offence_description": "Assault with injury",  "offence_group": "Theft offences",  "force_name": "Greater Manchester",  "offence_subgroup": "Theft from a vehicle"}
```

Parameter content type

application/json

# Semantic Versioning

## MAJOR Version

- Incompatible API or other breaking changes
- **Scoring parameter changes i.e. API change**

## MINOR version

- Add functionality in a backwards-compatible manner
- **Trained using different parameters/classifiers, but API same**

## PATCH version

- Backwards-compatible bug fixes
- **Trained the model on different data**

# Versioning – Touches Everything



Also...

- Resource names in Azure controlled via ARM templates
- ACI DNS names & prefixes,  
e.g. batcomputer-2-0-8.westeurope.azurecontainer.io
- Object names in Kubernetes (pods, services), controlled via Helm chart

# Some Learnings

- Pickled Scikit-learn models are version sensitive
- Keep Python version in sync with DataBricks
- Installing numpy, scipy and scikit-learn is SLOW, pre-build base image
- Version number is the key parameter for the whole process
- Writing your own wrapper isn't hard
- DataBricks is easy to use & has a great CLI



