Microsoft Azure

Project Batcomputer

A working DevOps implementation for Machine Learning

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v0.0.3 (Beta)

Background

Motivation

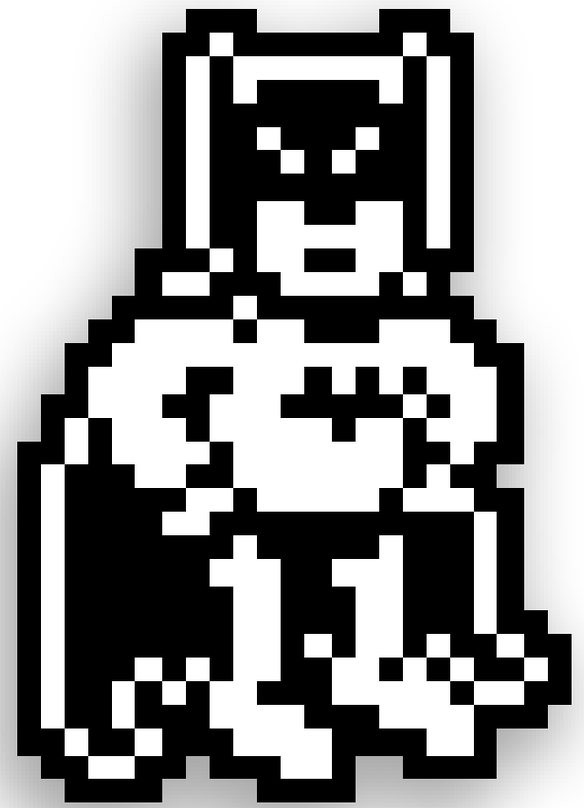
- Understand challenges in operationalisation of ML models
- Existing processes approaches deemed problematic
- “DevOps for AI”

Why Batcomputer?

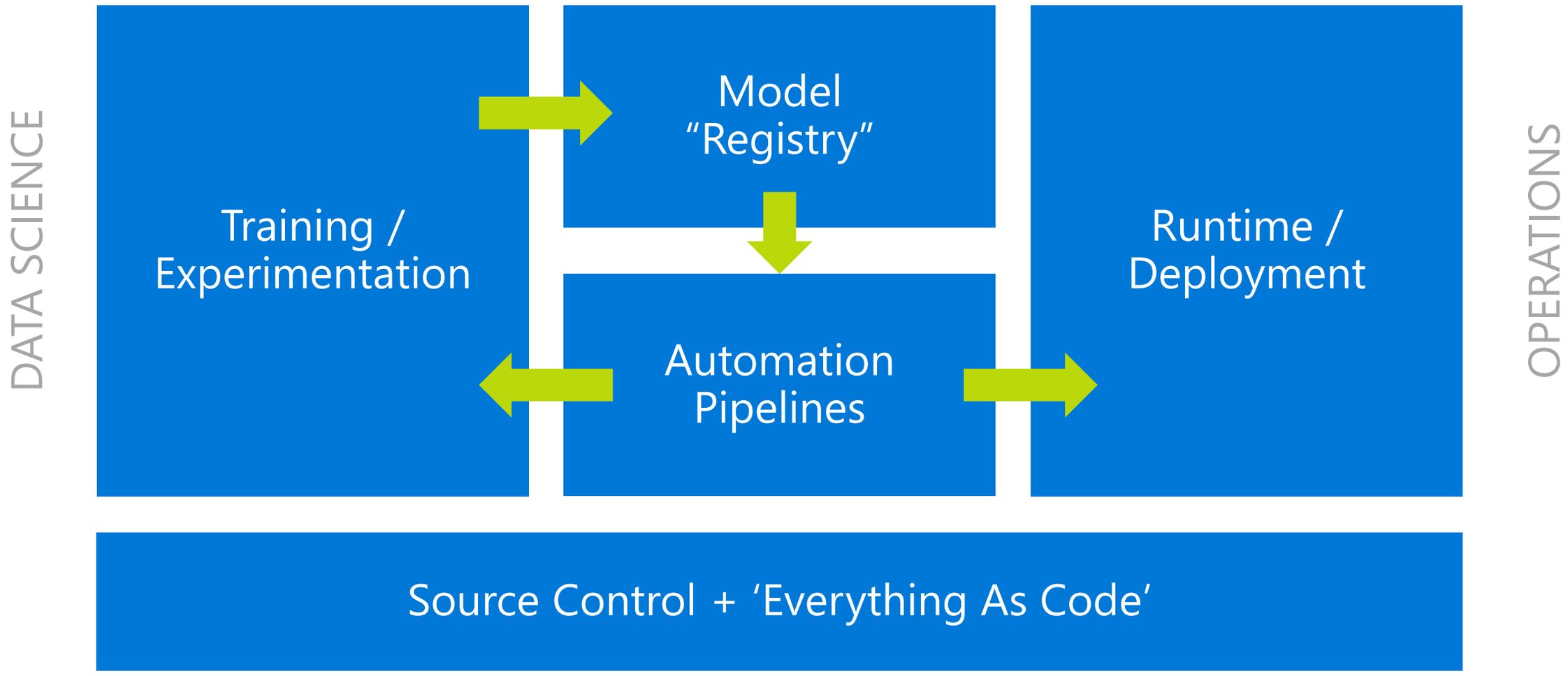
- Police recorded crime and outcomes data
- Source data as CSV - https://data.police.uk/data
- Build model of a given crime and region to predict – “Would you get caught?”

Core Principals & Benefits

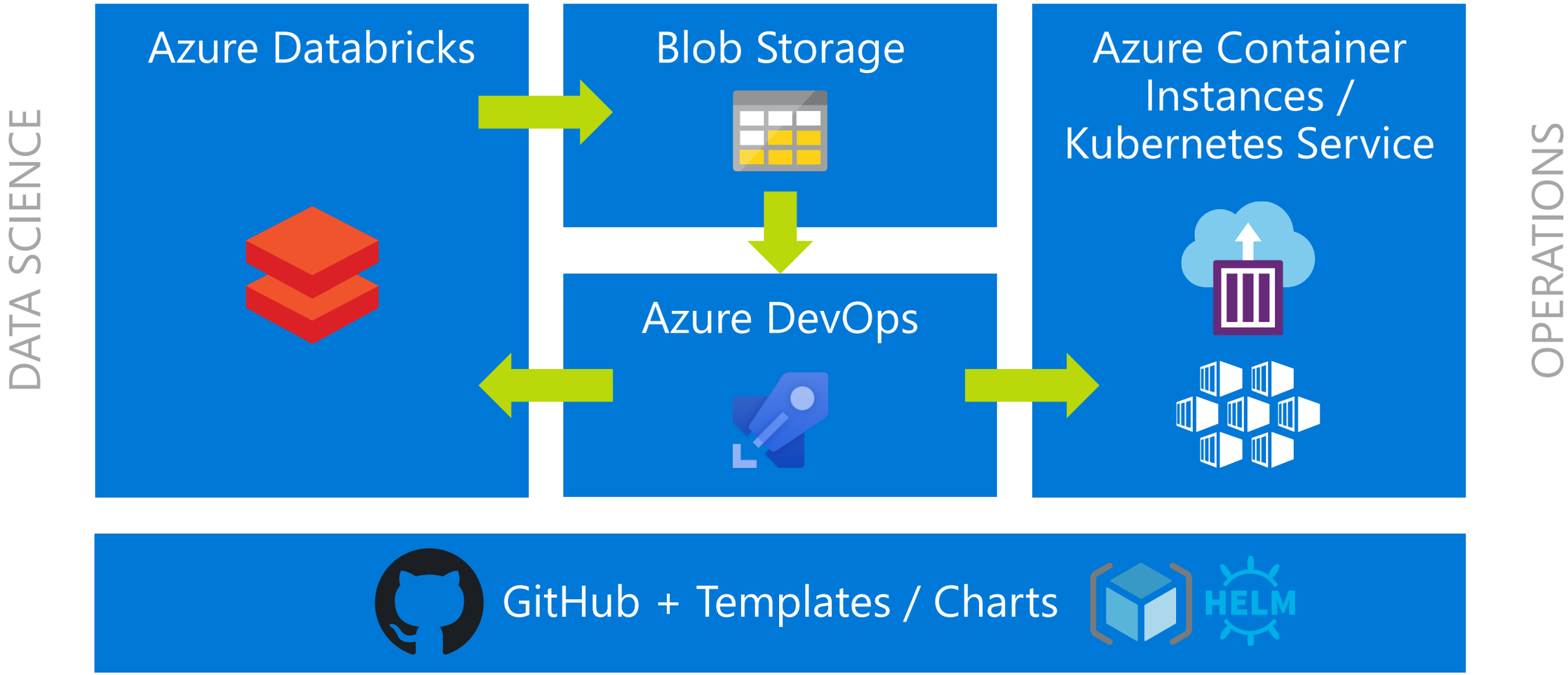
- Decouple model training experiments from operations/runtime
- Automated training, API builds & deployment
- Versioned models and API
- Config & infra as code
- Traceability



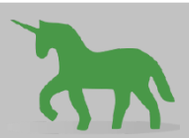
Conceptual Building Blocks



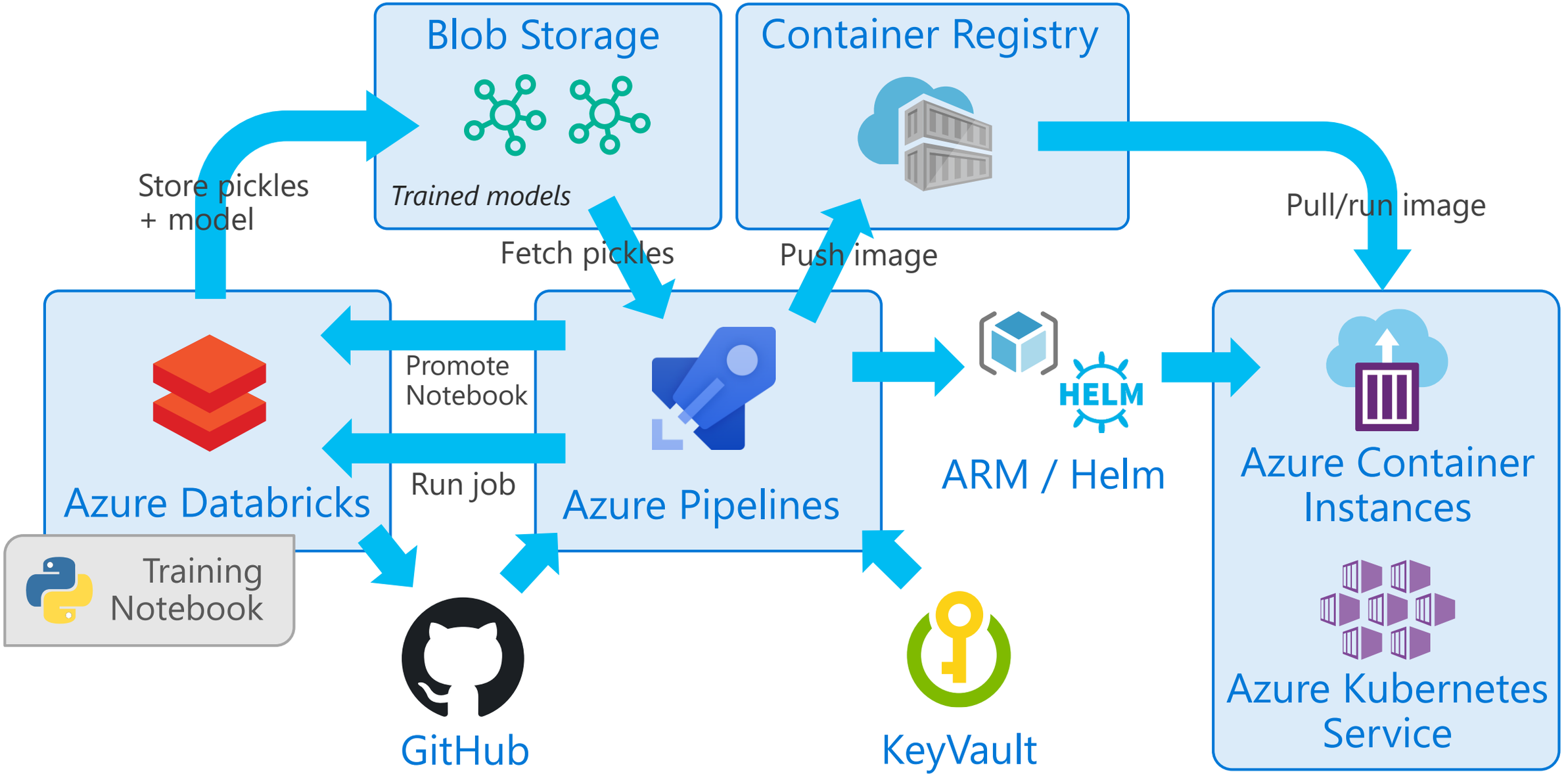
Conceptual Building Blocks – Project Batcomputer



Low Level Technology Stack

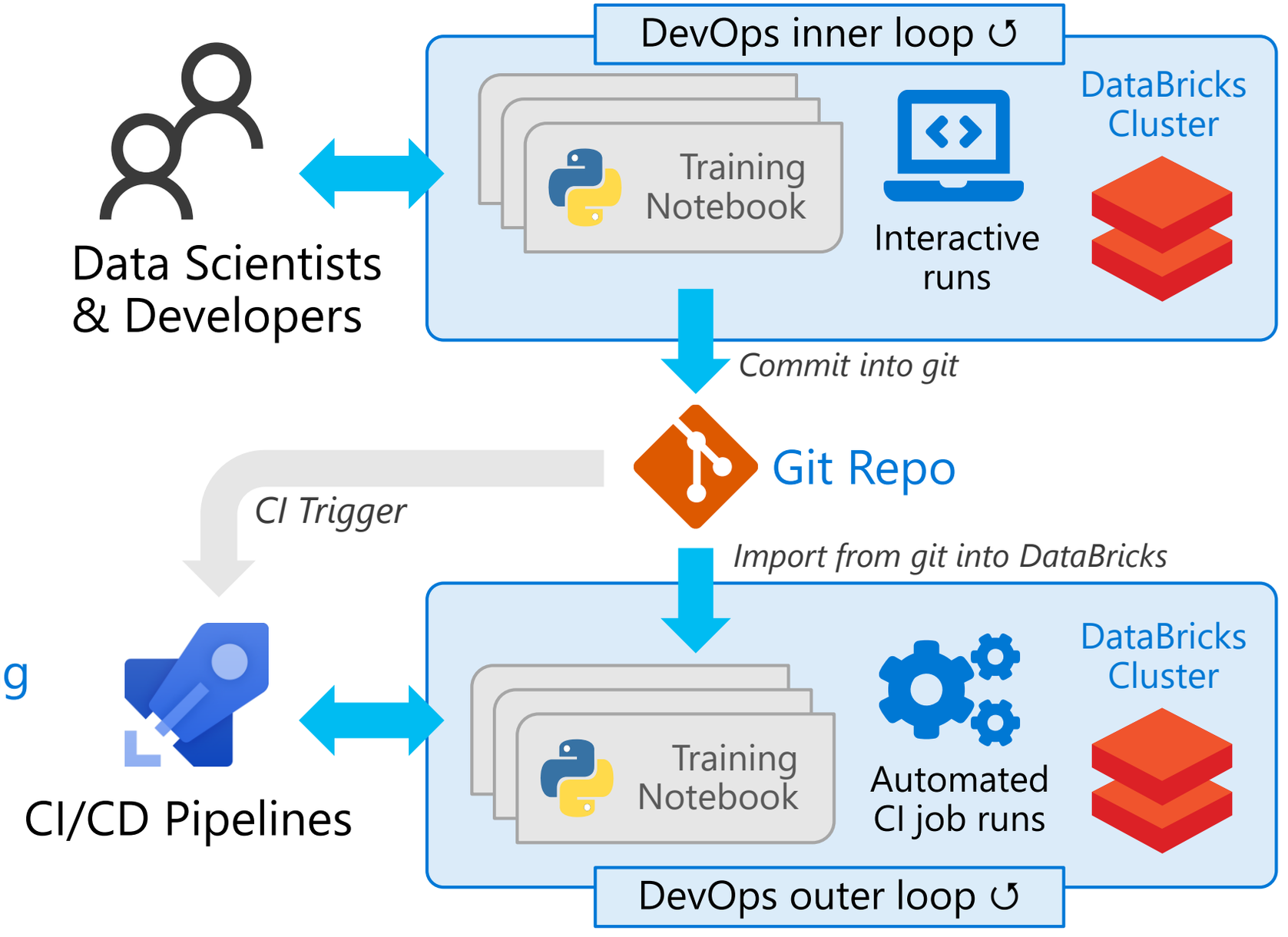
Swagger 	← API niceness
Gunicorn 	← HTTP Server
Flask 	← Web framework
Pickle 	← Serialisation
Scikit-Learn 	← Main ML framework
Python 	← Core language
Docker 	← Container Runtime

Training & Deployment – End To End Flow

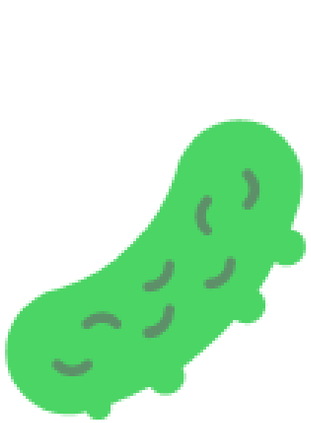


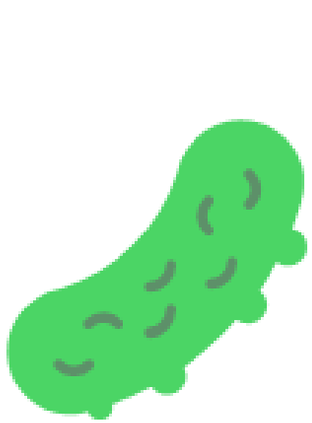
Core DevOps Practice - Continuous Integration

- Development & experimentation
- Central shared git repo
- CI triggered training & testing job runs

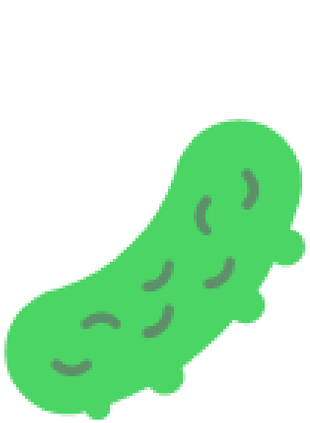


Pickling – Not Just The Model





```
{
“gender”: {
  “male”: 0,
  “female”: 1
}
```



```
[
  “conviction”,
  “dropped”,
  “settled”
]
```

model.pkl	lookup.pkl	flags.pkl
Scikit-learn model/classifier	Python dictionary of dictionaries	Python array
Standard object rehydration, version sensitive	Mapping parameters/strings to num for predict function	Maps output of prediction function to human readable strings or labels

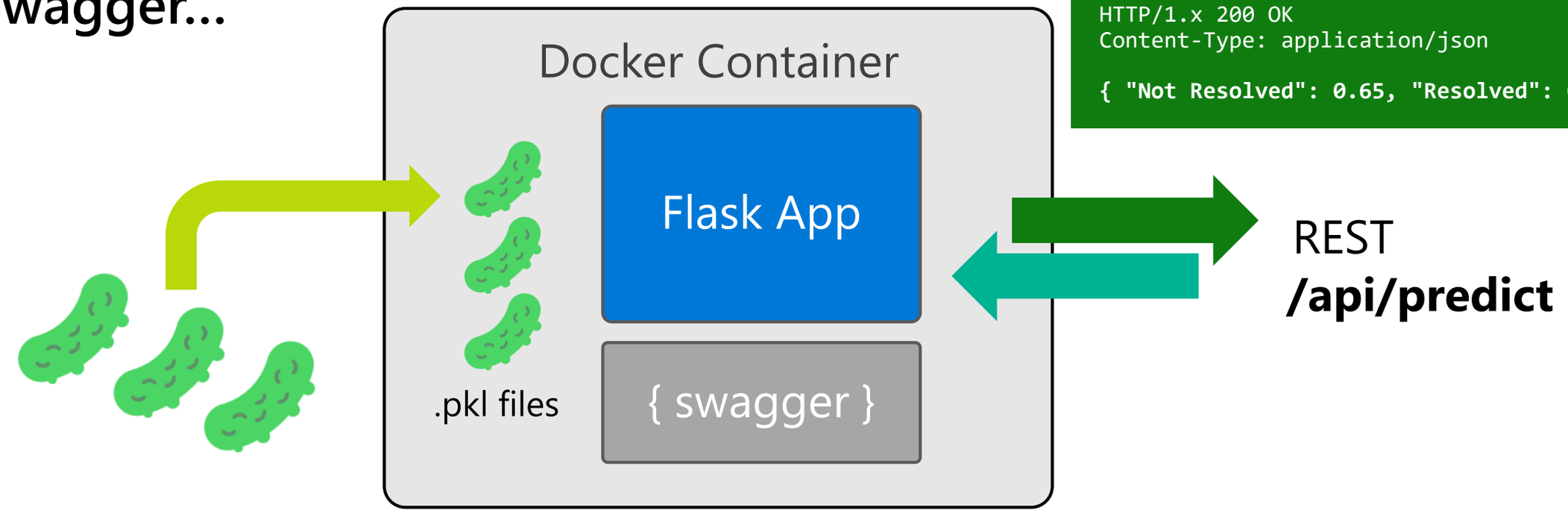
Wrapper App – Flask API

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

```
POST /api/predict
{
  "force": "Thames Valley Police",
  "crime": "Bicycle theft",
  "month": 10
}
```

```
HTTP/1.x 200 OK
Content-Type: application/json

{ "Not Resolved": 0.65, "Resolved": 0.35 }
```

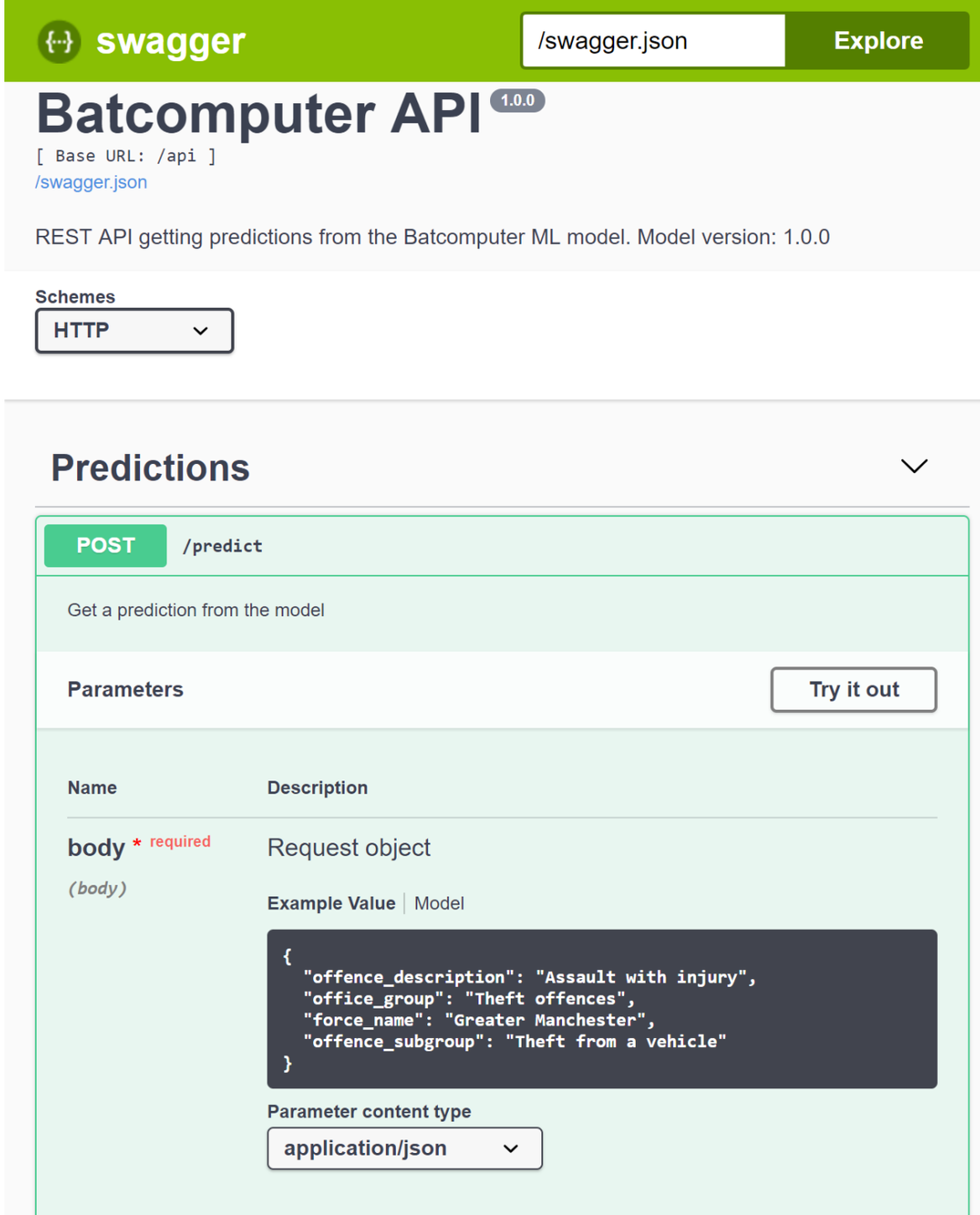


Some Decision Points

- Include model in container image or fetch at runtime?
- Wrapper app – make generic or tied into model?
- What are my API parameters?
- 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



Semantic Versioning

MAJOR Version

- Incompatible API or other breaking changes
- **Scoring inputs / feature 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 deployment process
- Writing your own wrapper isn't hard
- DataBricks has a great CLI & API
- DataBricks can be used with CI but it's not obvious

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