



Machine Learning on OpenShift

From Data Scientist to Application Developer

Michael McCune
Principal Software Engineer



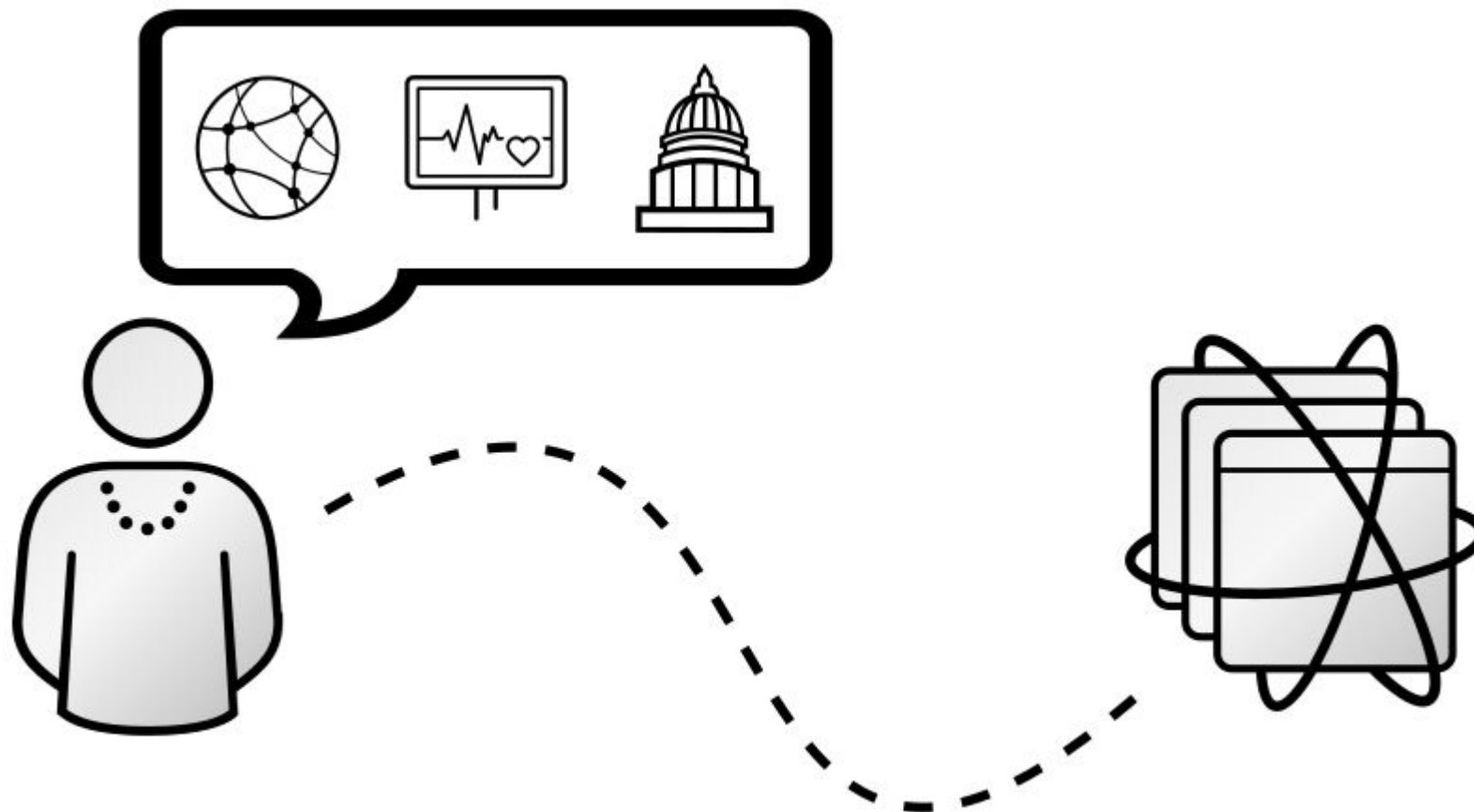
Forecast

- ▶ Introduction
- ▶ Technology Review
- ▶ Building Intelligent Applications
- ▶ Lessons Learned
- ▶ Next Steps

Who is this guy?

- ▶ Joined Red Hat 6 years ago
- ▶ Full stack, from embedded to orchestration
- ▶ Emerging technology at Red Hat
- ▶ Big Data on OpenStack and OpenShift

We are talking about a journey



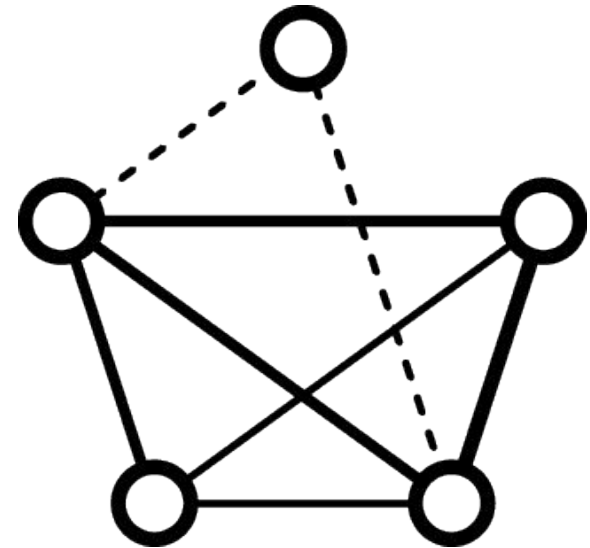
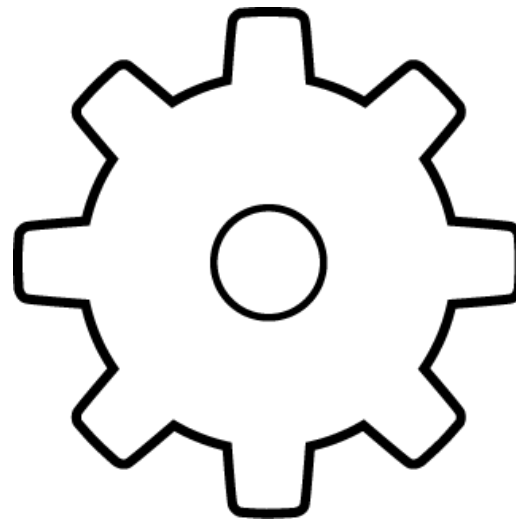
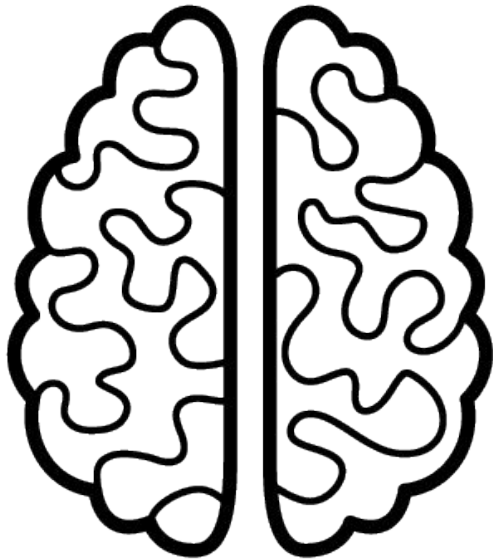




Level Set

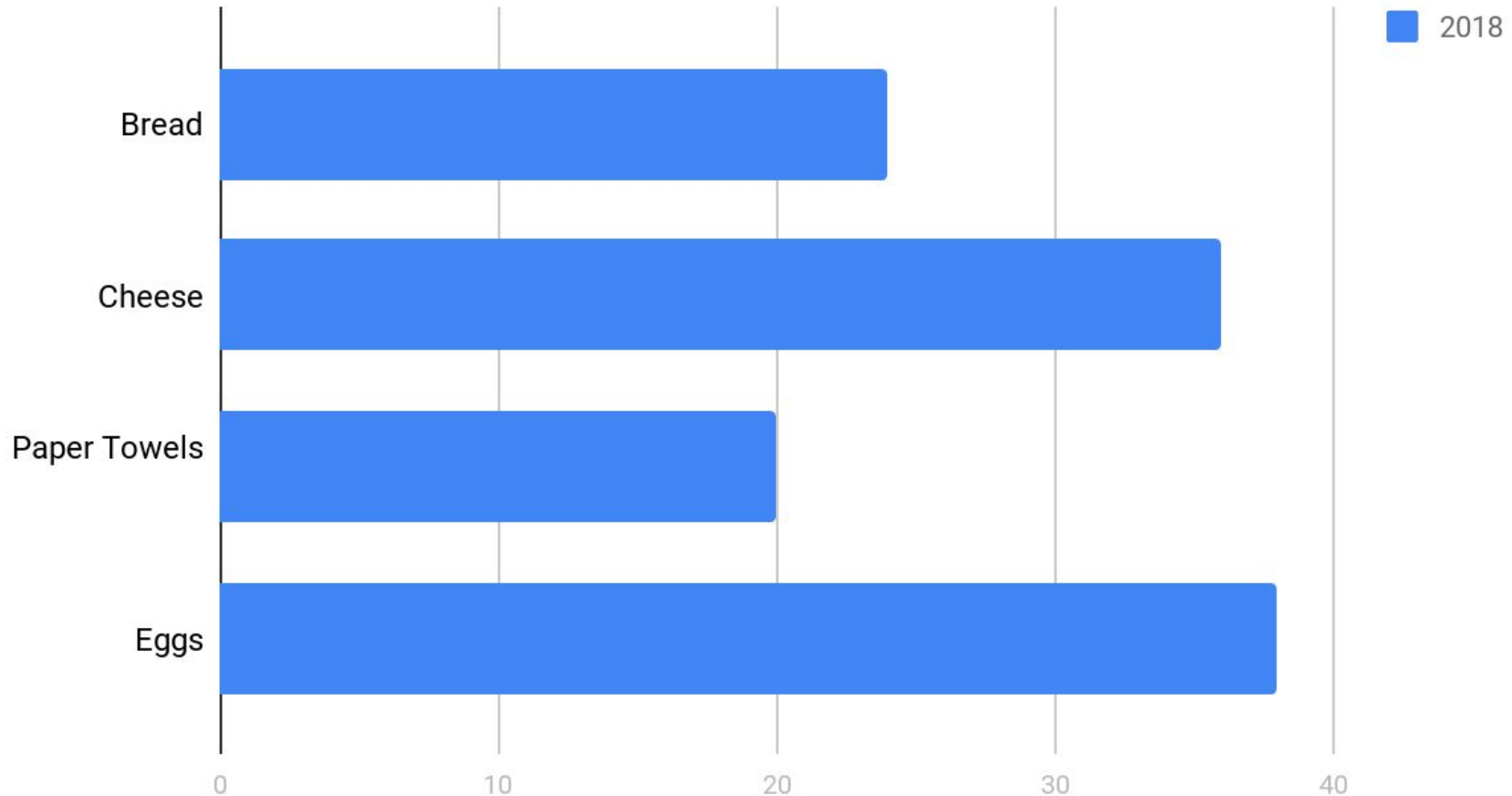


Machine Learning



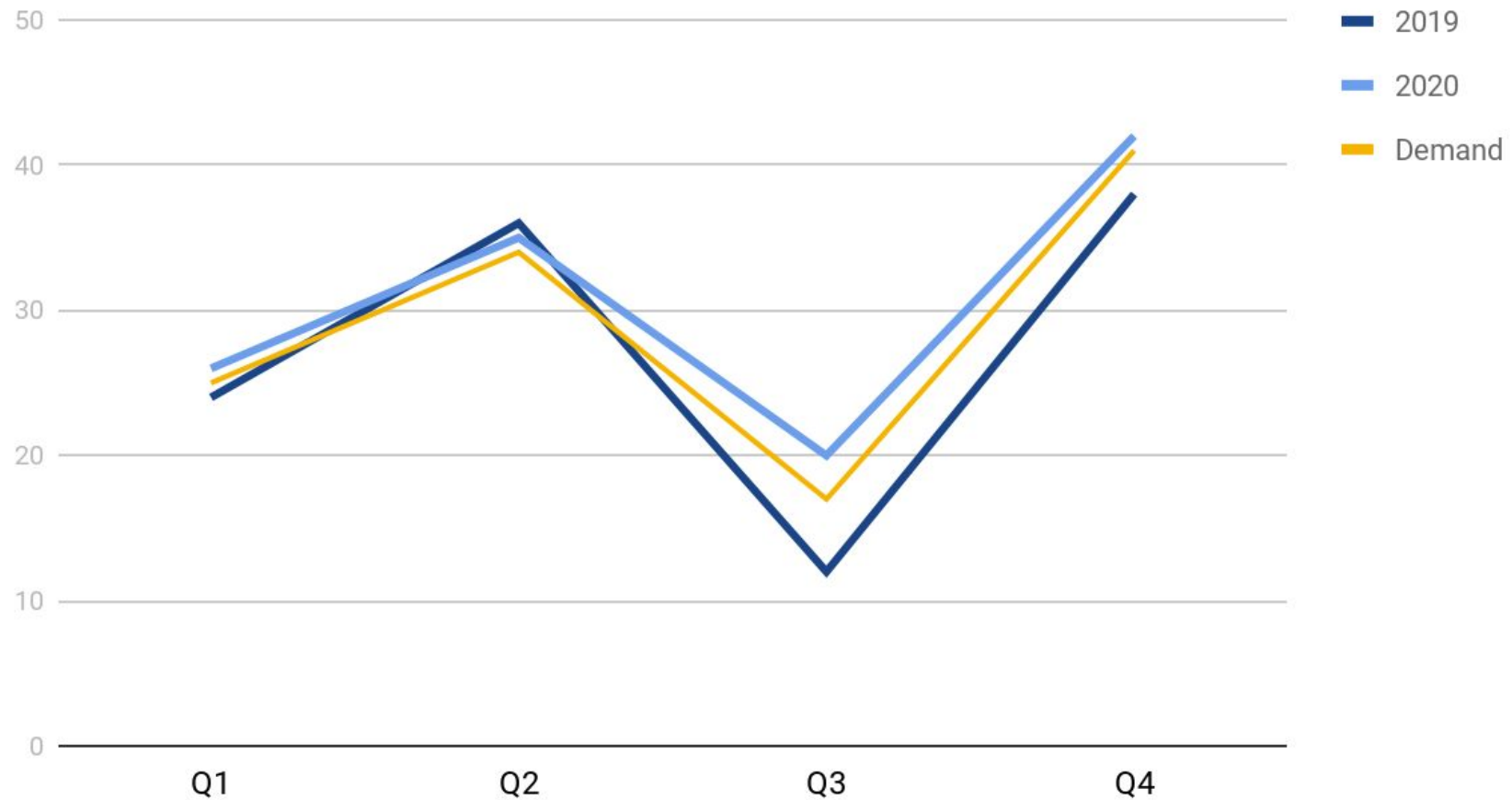
Identifying Trends

Customers who bought Spam also bought



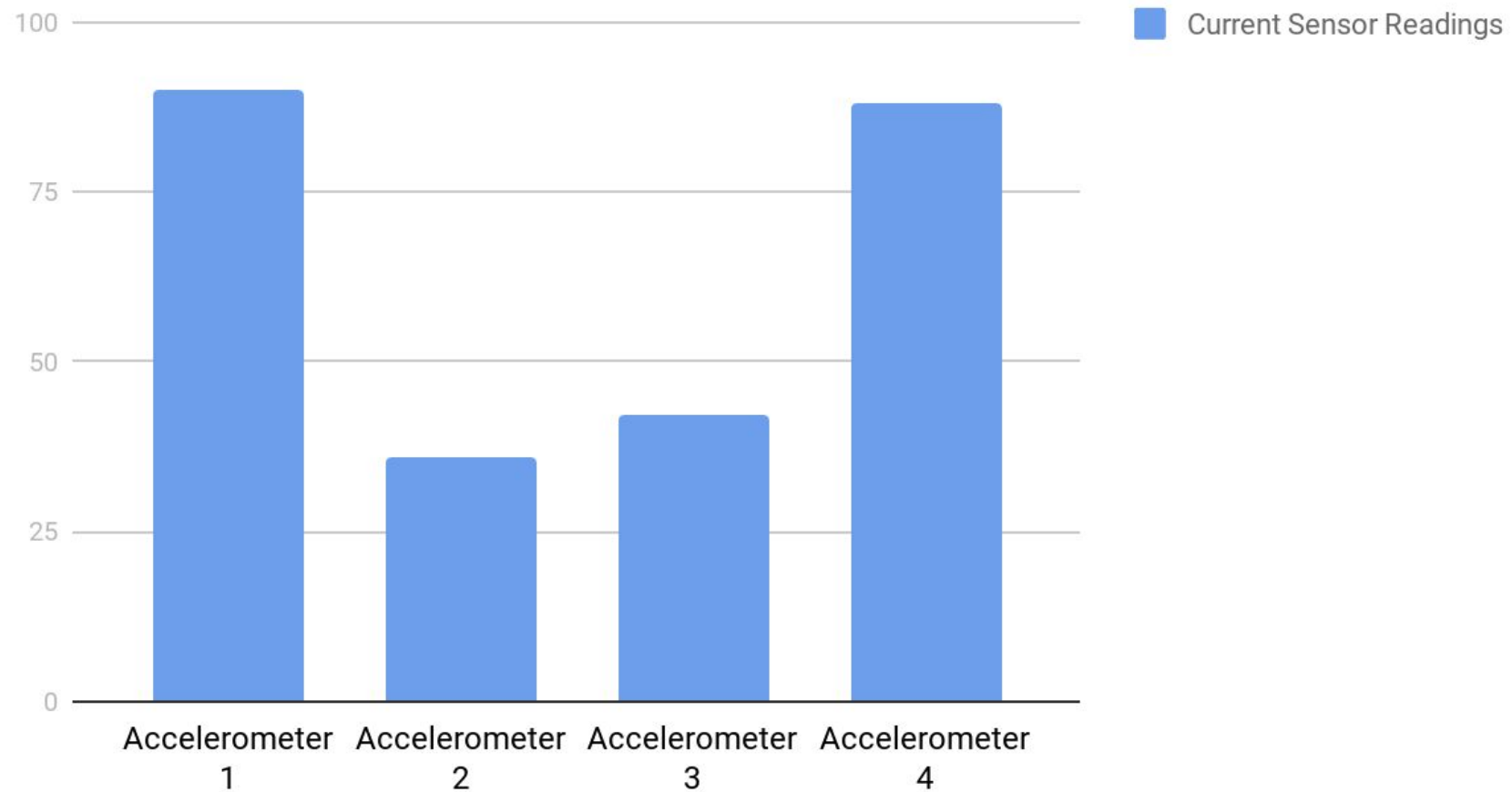
Predicting Results

Resource consumption by quarter



Estimating Values

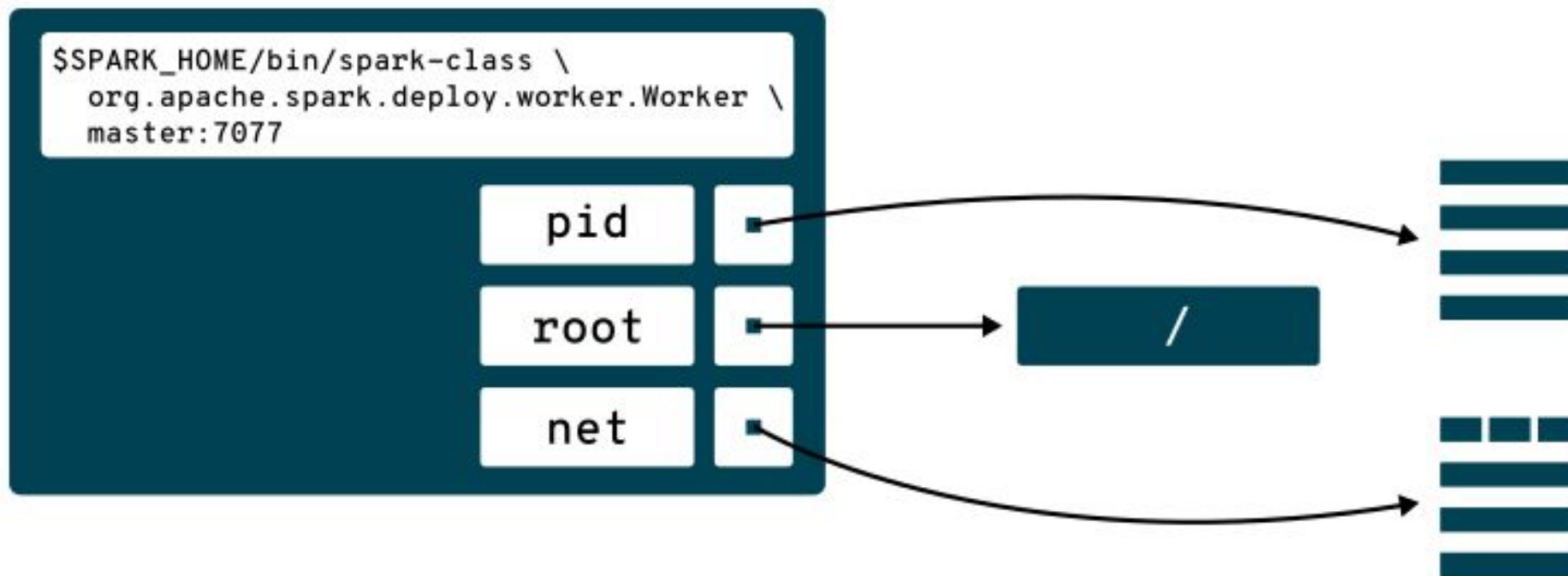
Estimated speed 88mph



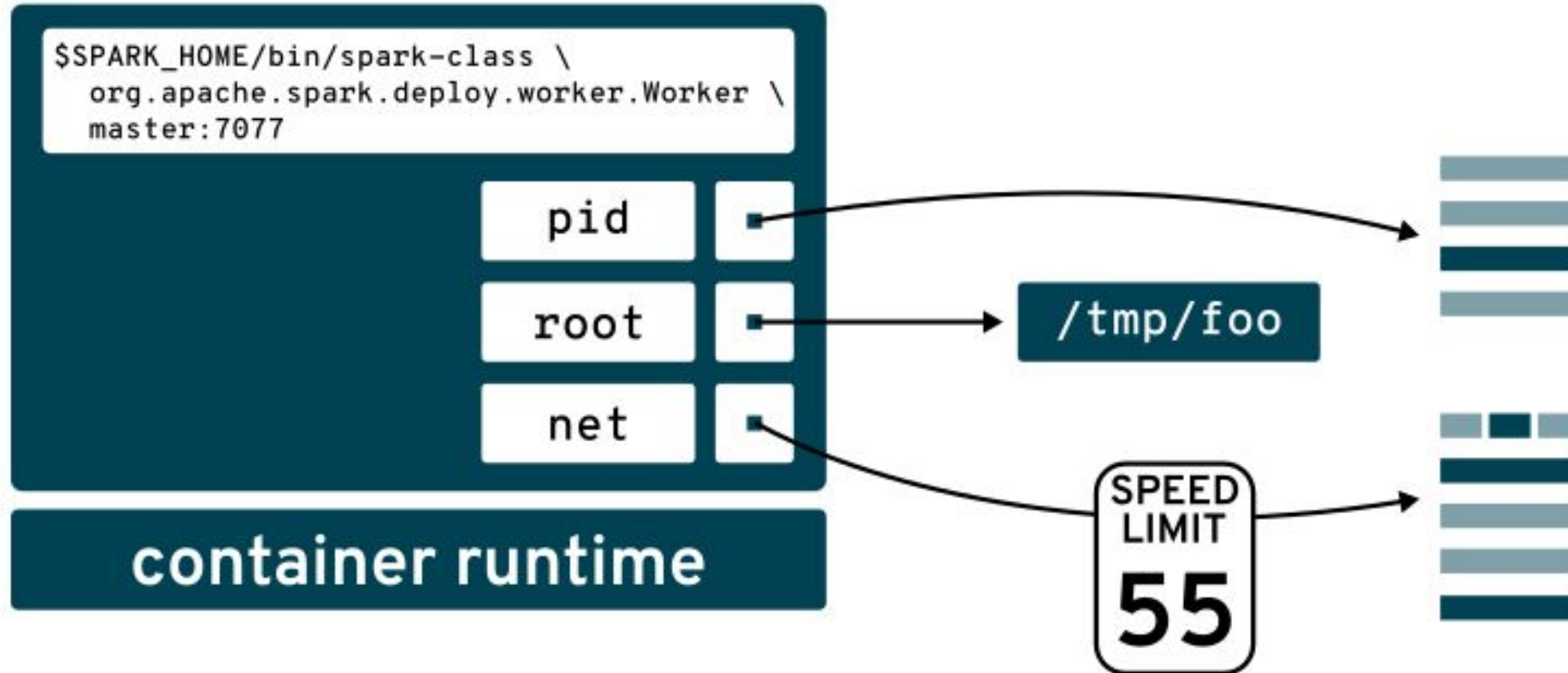
Kubernetes

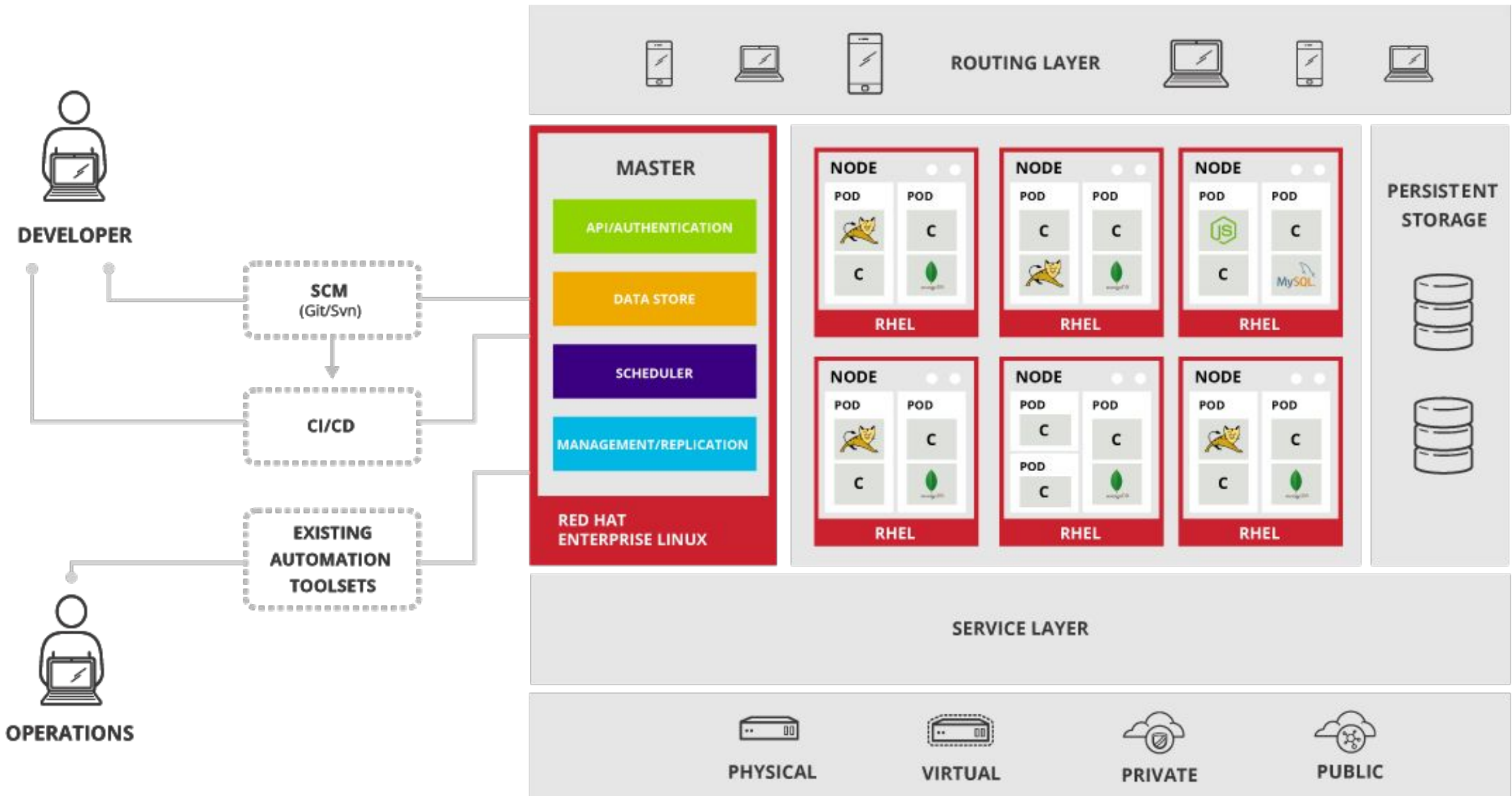


Let's talk containers

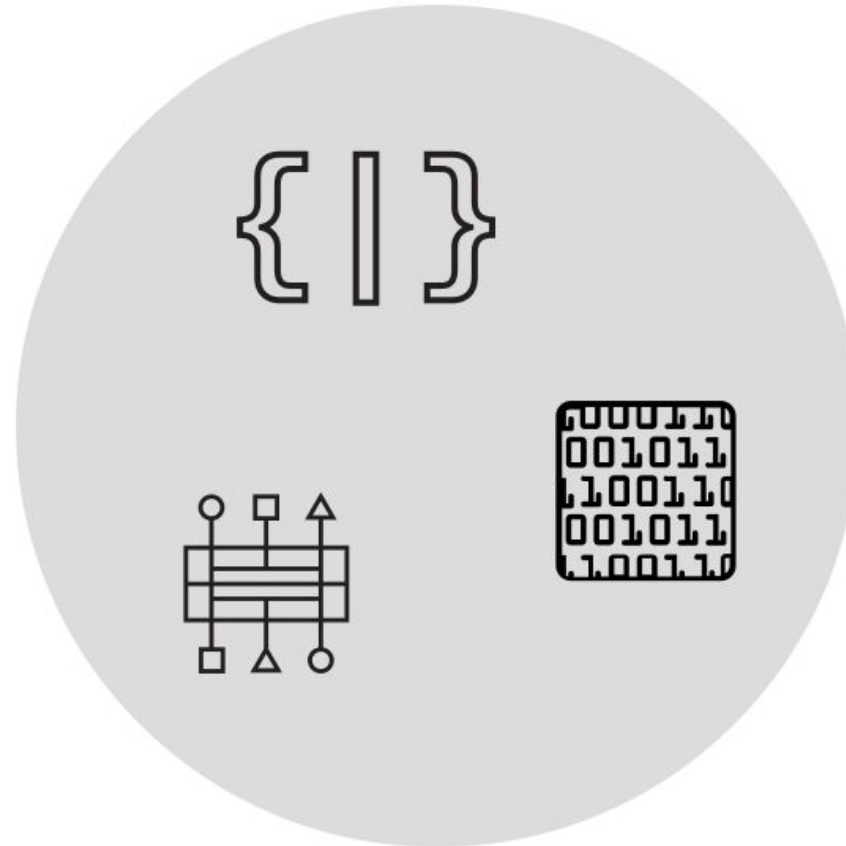


Let's talk containers

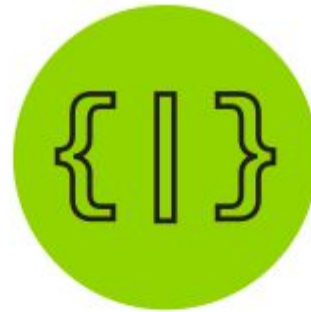
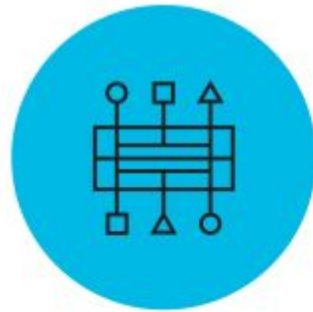




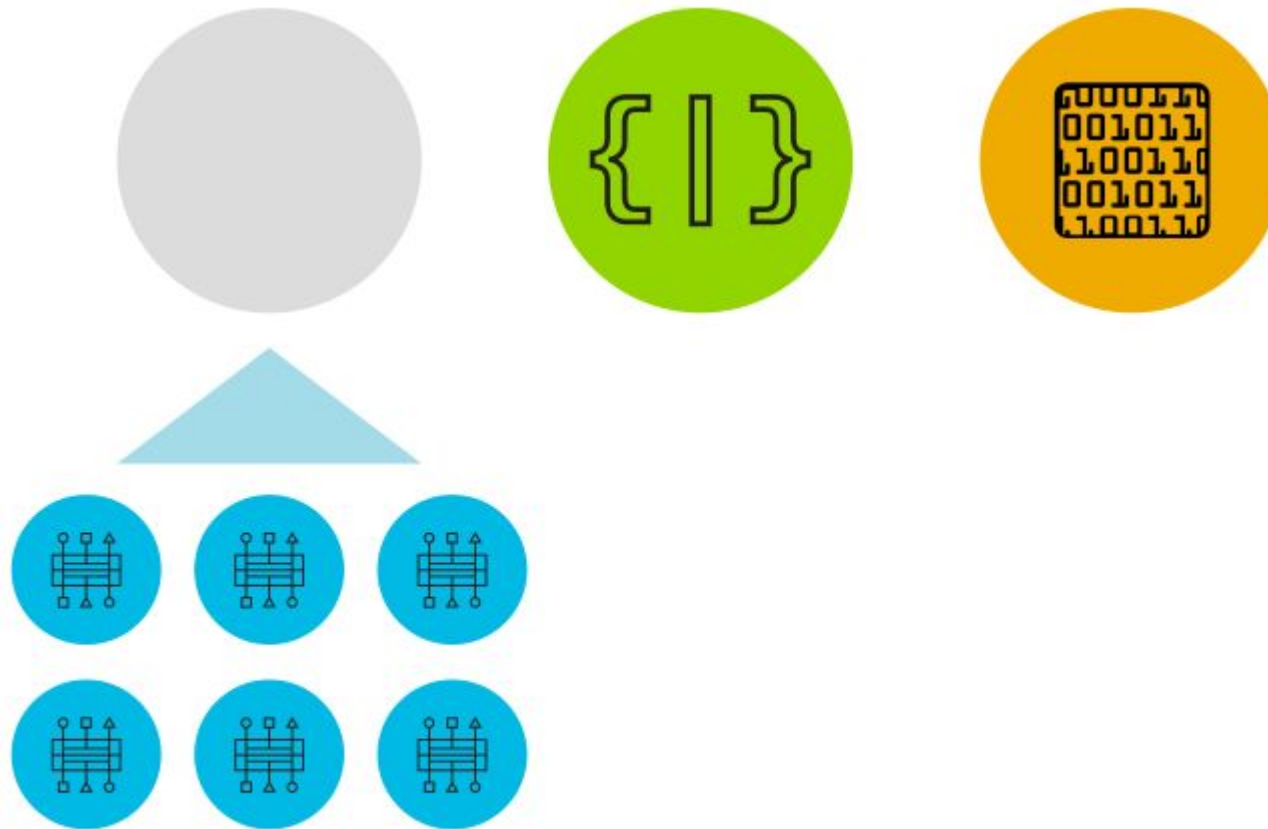
Microservice Architectures



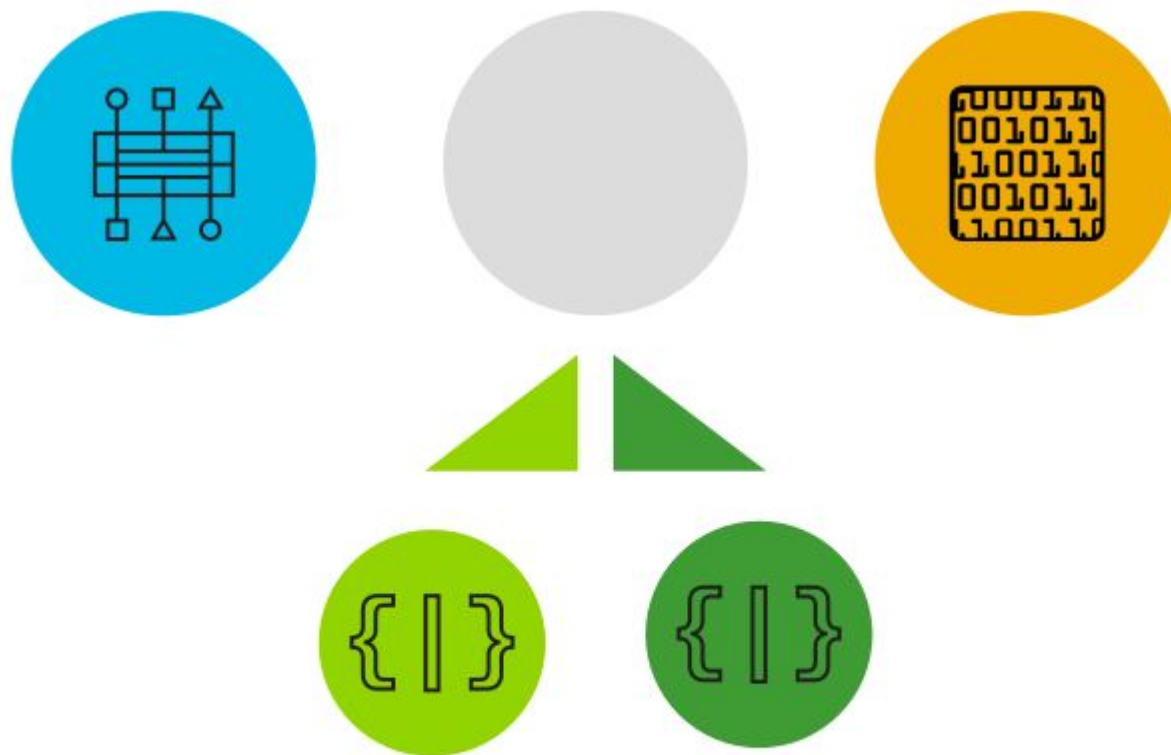
Composable



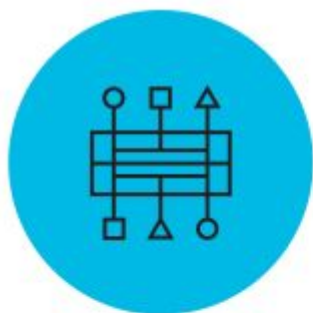
Scalable



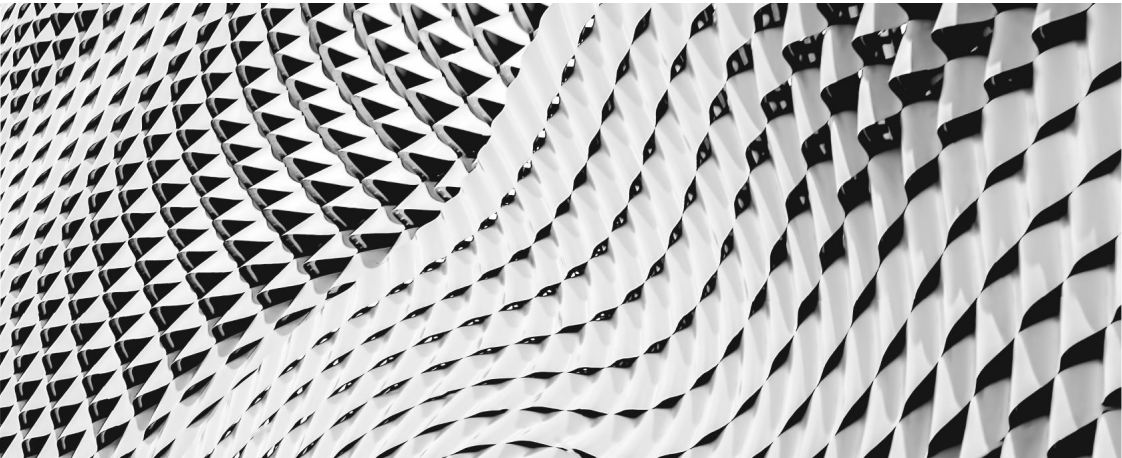
Flexible



Resilient

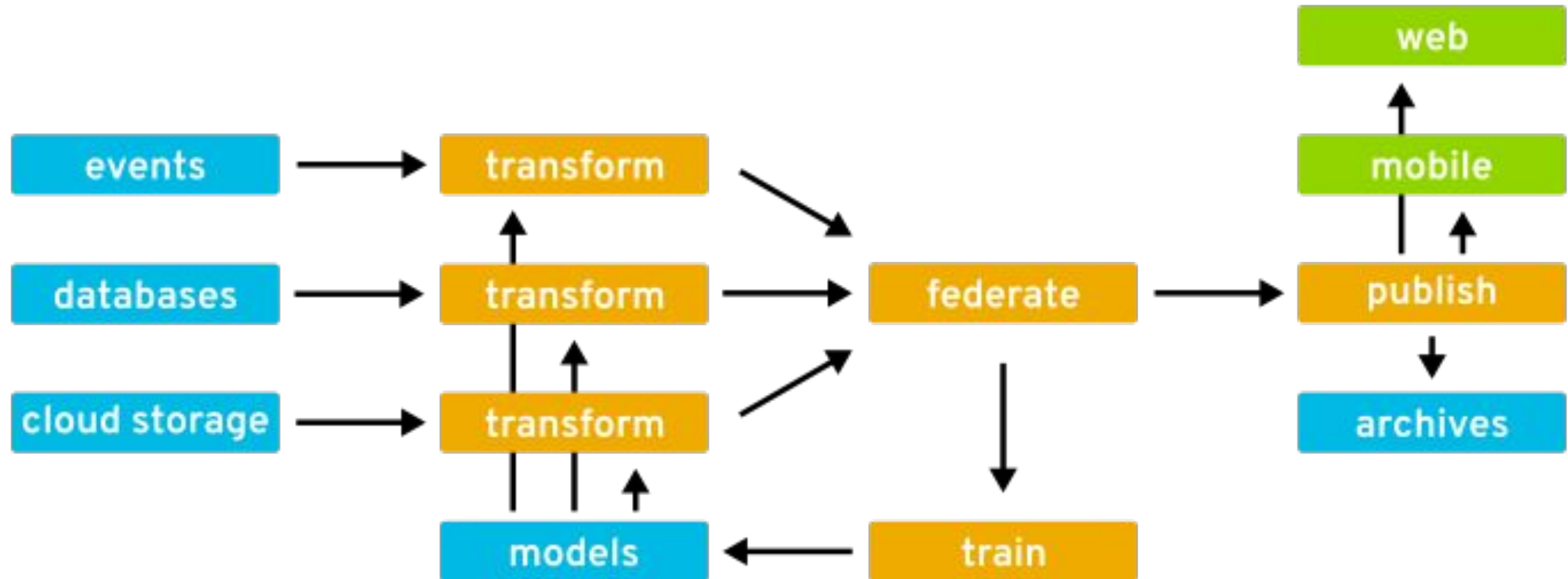


Intelligent Applications

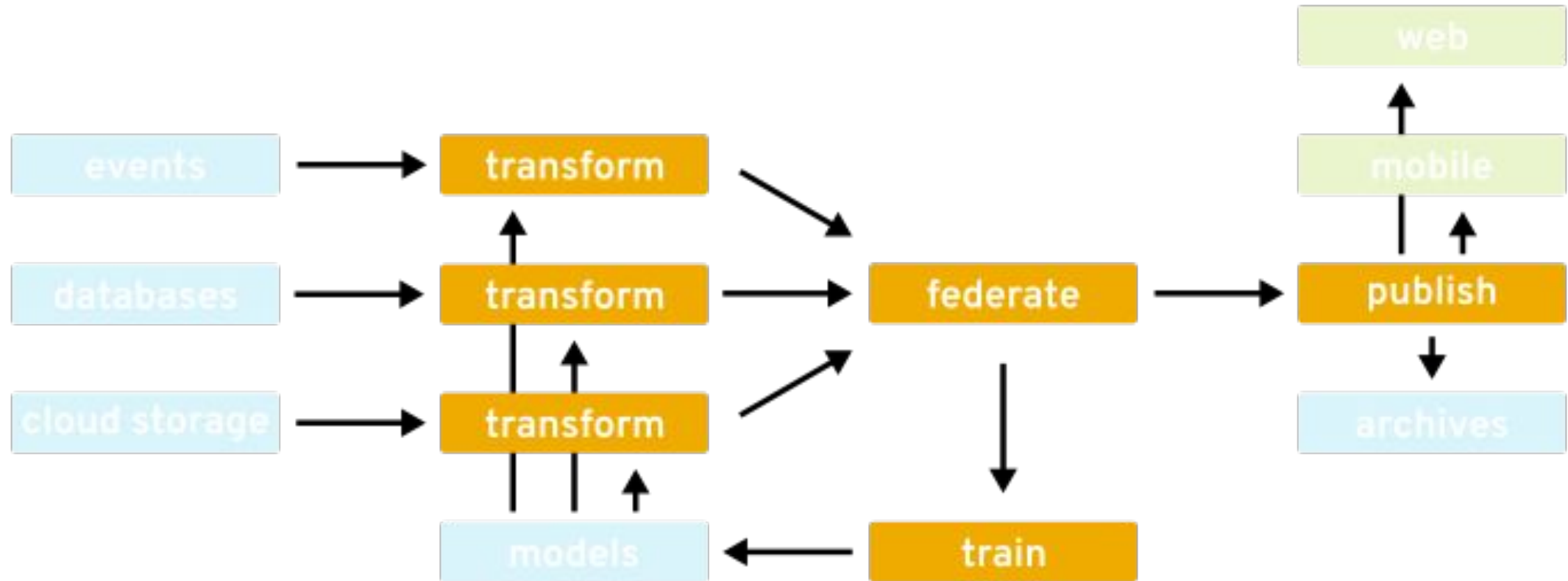


Intelligent applications
collect and learn from
data to provide **improved**
functionality with
longevity and popularity.

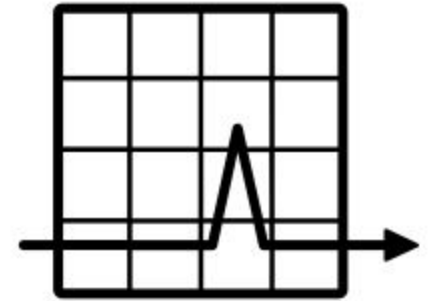
Intelligent Application Pipeline



Service Based Components

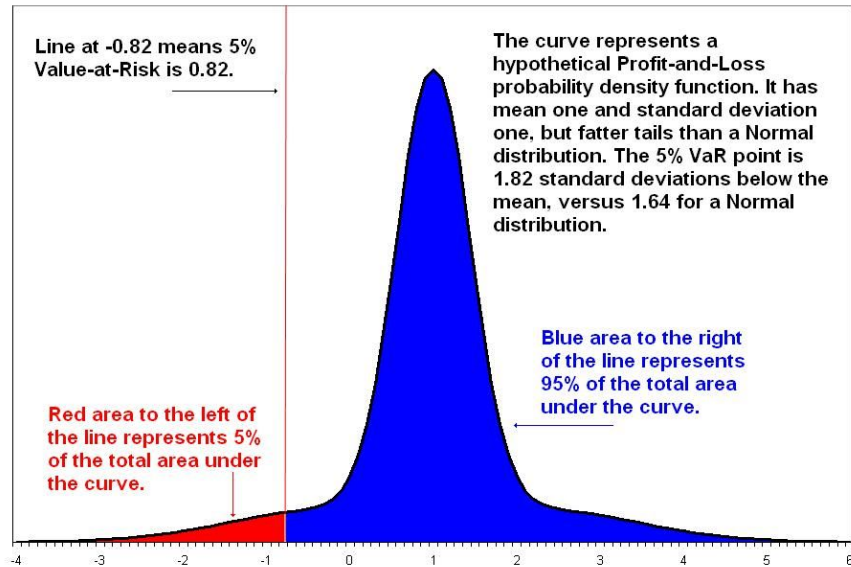


Intelligent Application Lifecycle

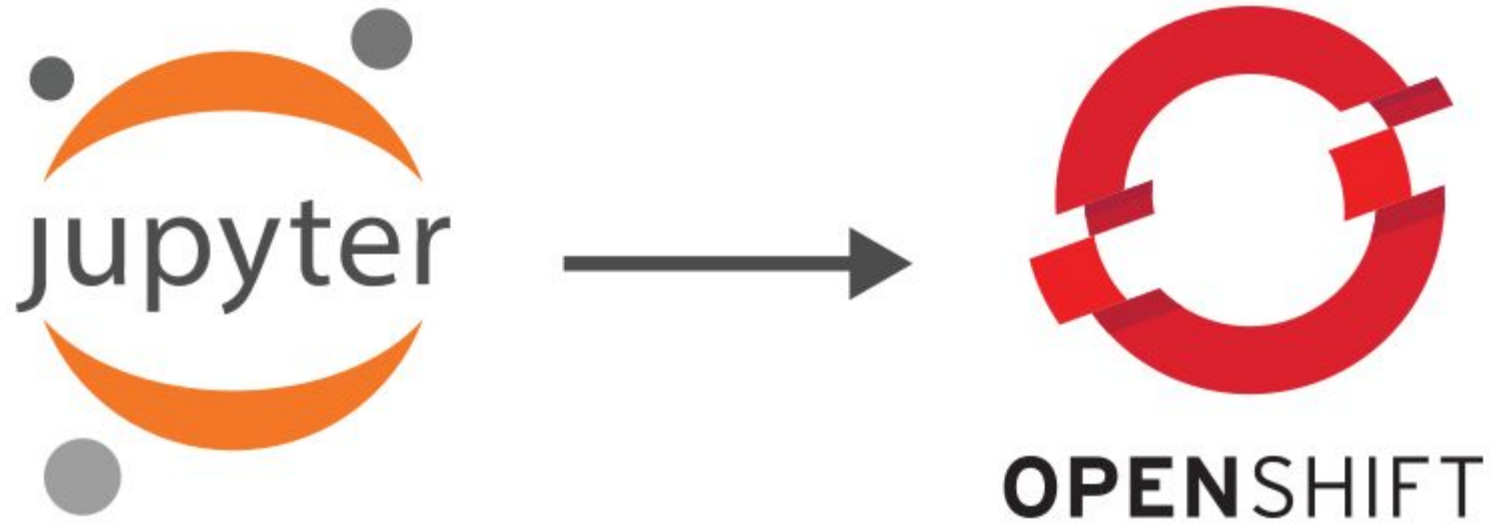


Case Study

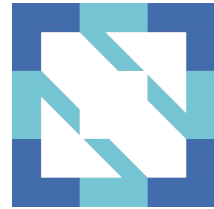
Value at Risk



Going Cloud Native



What is Cloud Native?



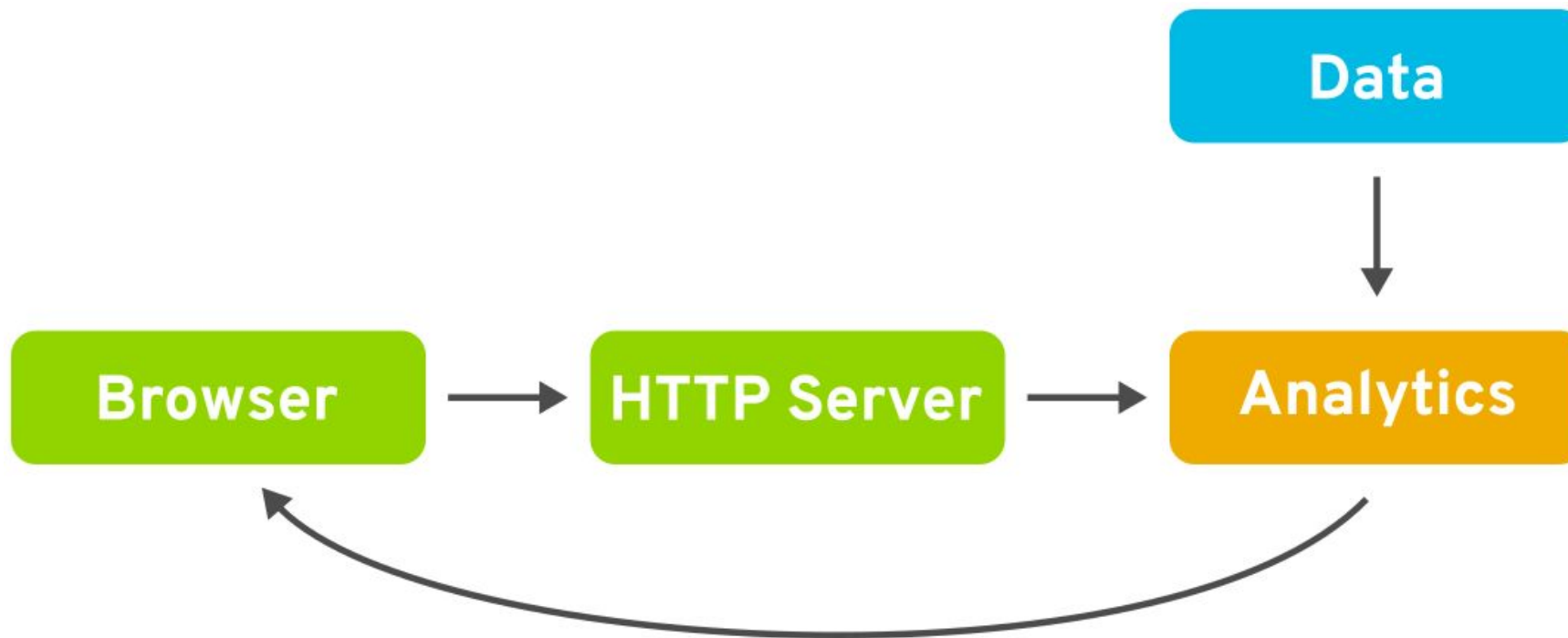
CLOUD NATIVE
COMPUTING FOUNDATION

- ▶ Containerized
- ▶ Dynamically orchestrated
- ▶ Microservice architectures
- ▶ learn more at **cncf.io/about/faq**

What will your application do?





Storyboard your architecture




Choose wisely



VAR Demo

**Red Hat**
OpenShift Container Platform

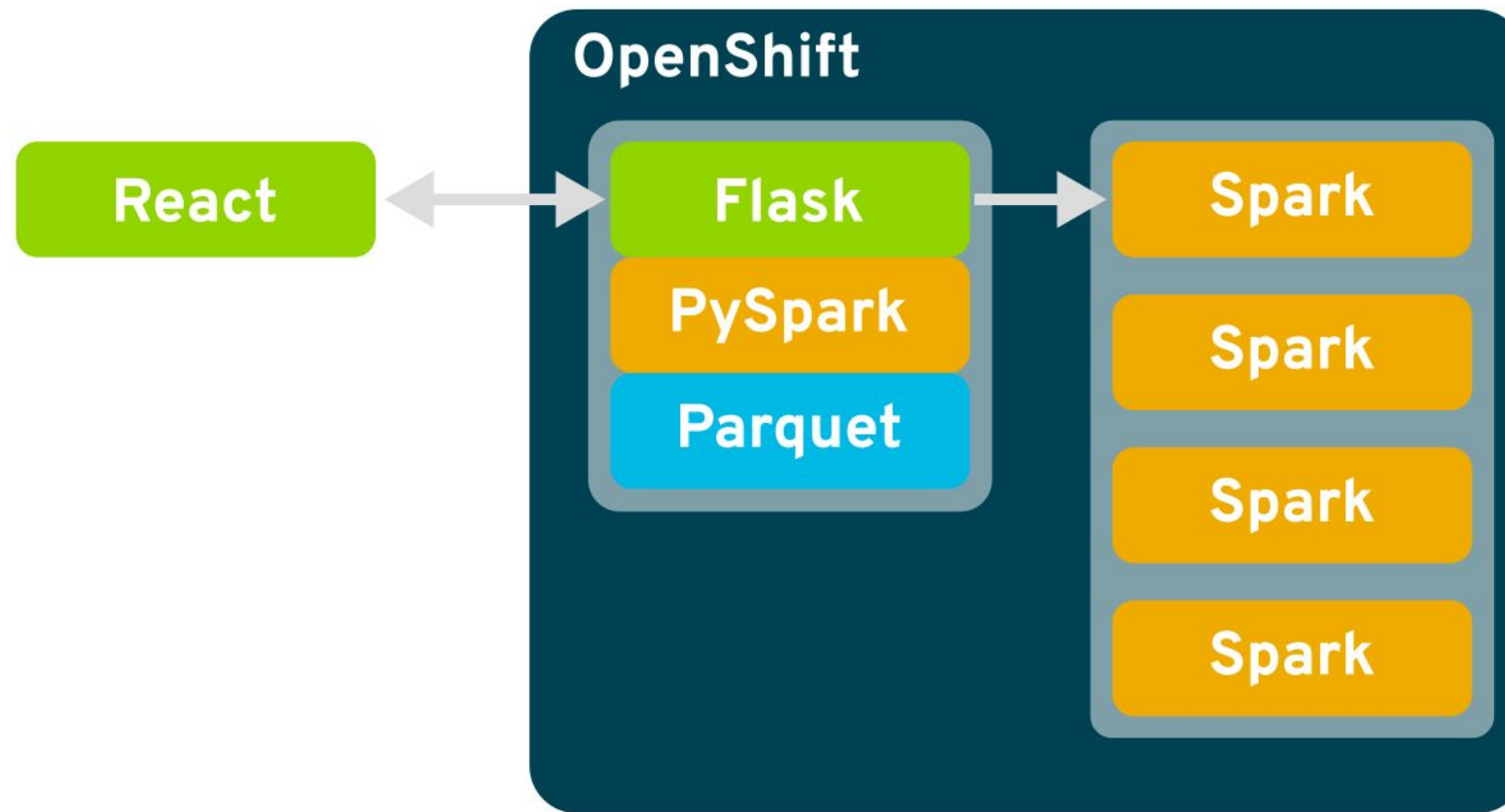
 Administrator ▼

Home ▼

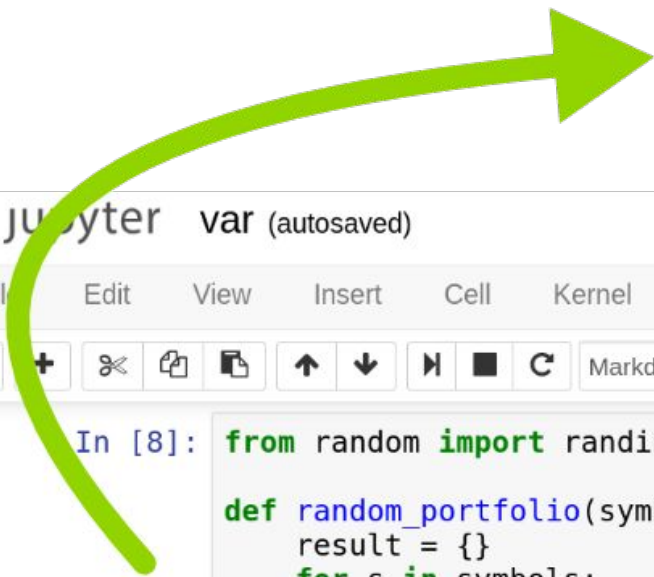
Projects

Create Project

General architecture



How was it built?



The image shows two code snippets. On the left is a Jupyter notebook interface with a menu bar (File, Edit, View, Insert, Cell, Kernel) and a toolbar. The code in the notebook is:

```
In [8]: from random import randint, seed

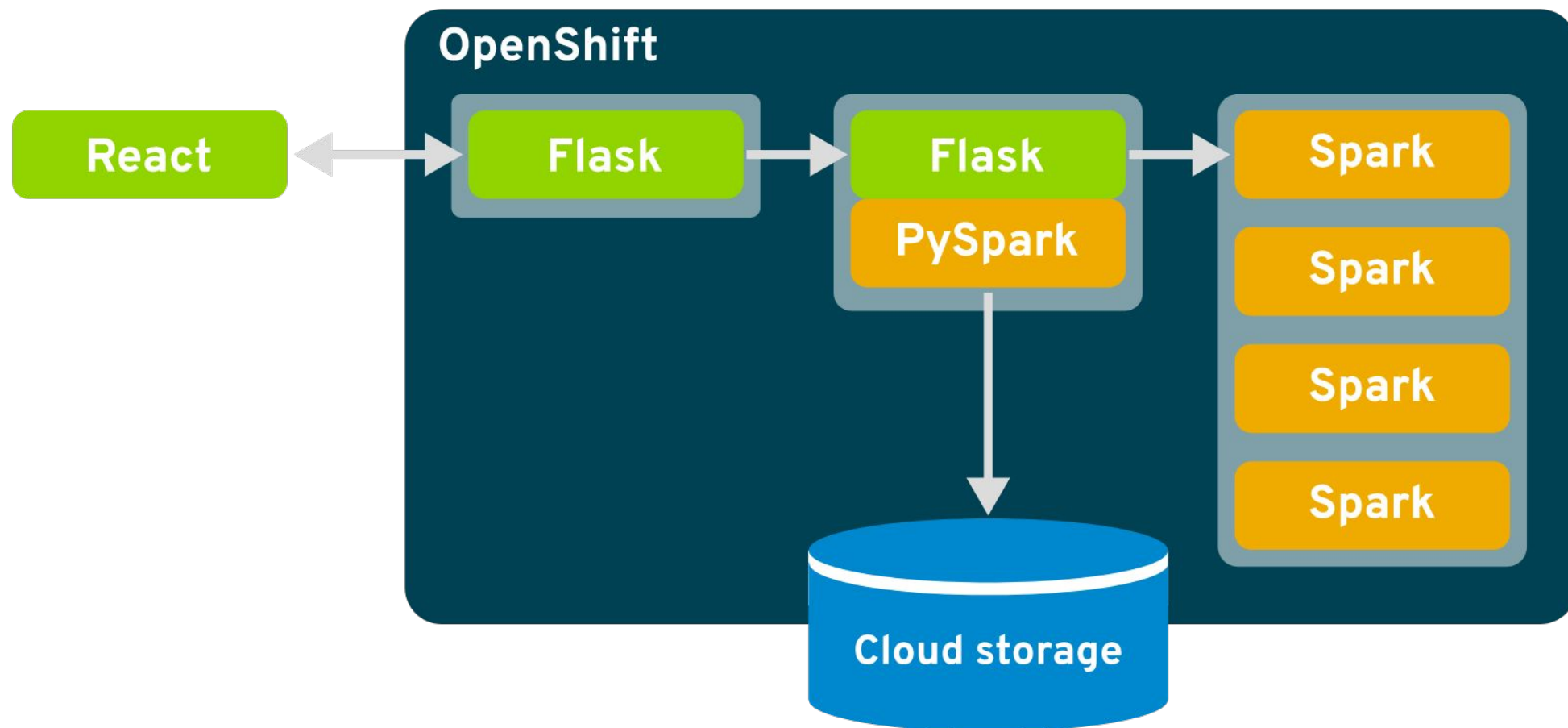
def random_portfolio(symbols):
    result = {}
    for s in symbols:
        result[s] = prices[s] * (randint(1, 1000) * 11)
    return result

def portfolio_value(pf):
    return sum([v for v in pf.values()])
```

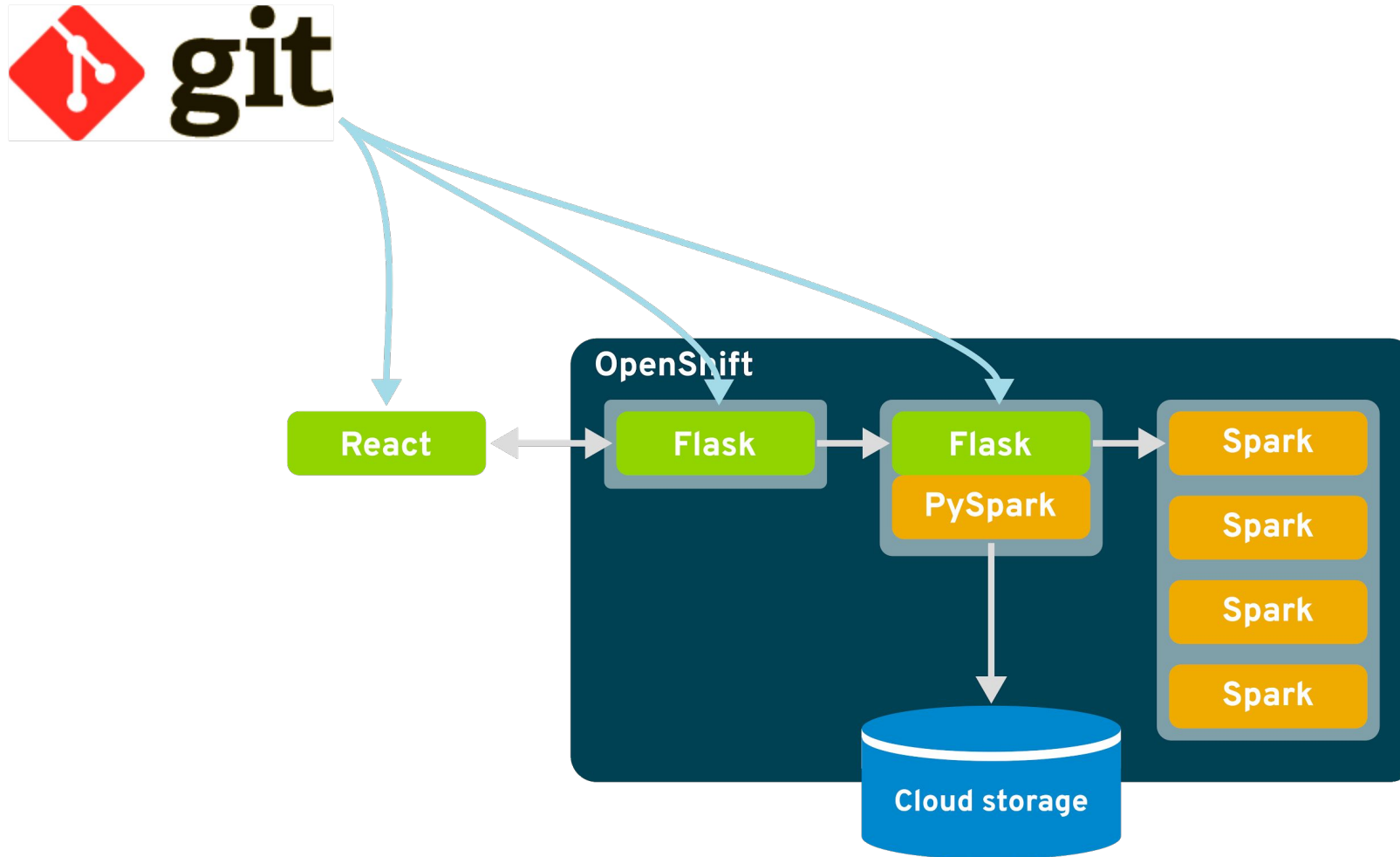
On the right is a dark-themed code editor window titled 'app.py' showing the following code:

```
26
27
28 def portfolio_value(pf):
29     """Given a dictionary of stock values, return the total value."""
30     return sum([v for v in pf.values()])
31
32
33 def seeds(count):
34     """Return a list of random values of the specified length."""
35     return [random.randint(0, 1 << 32 - 1) for i in range(count)]
36
37
```

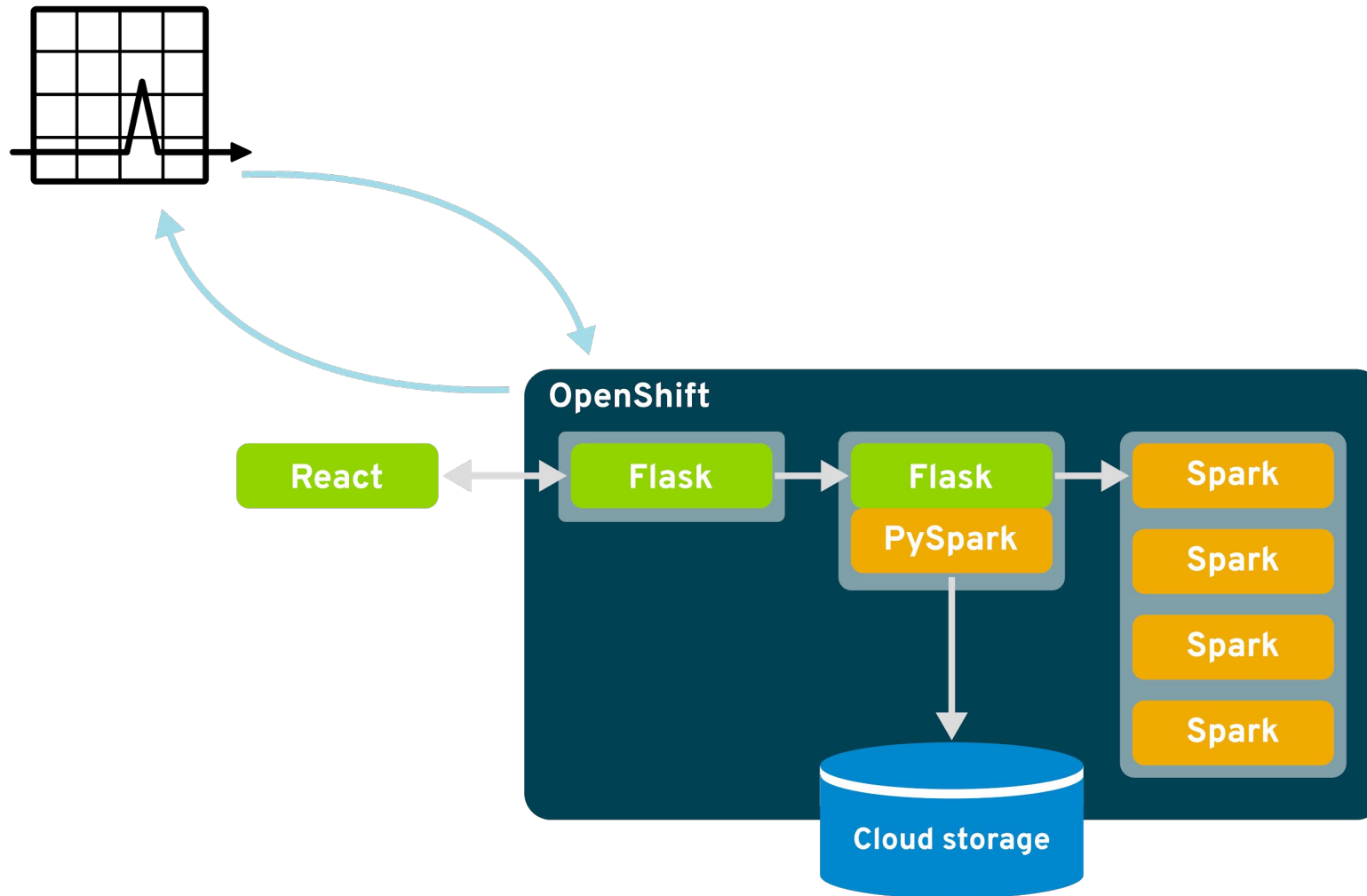
Moving out of alpha



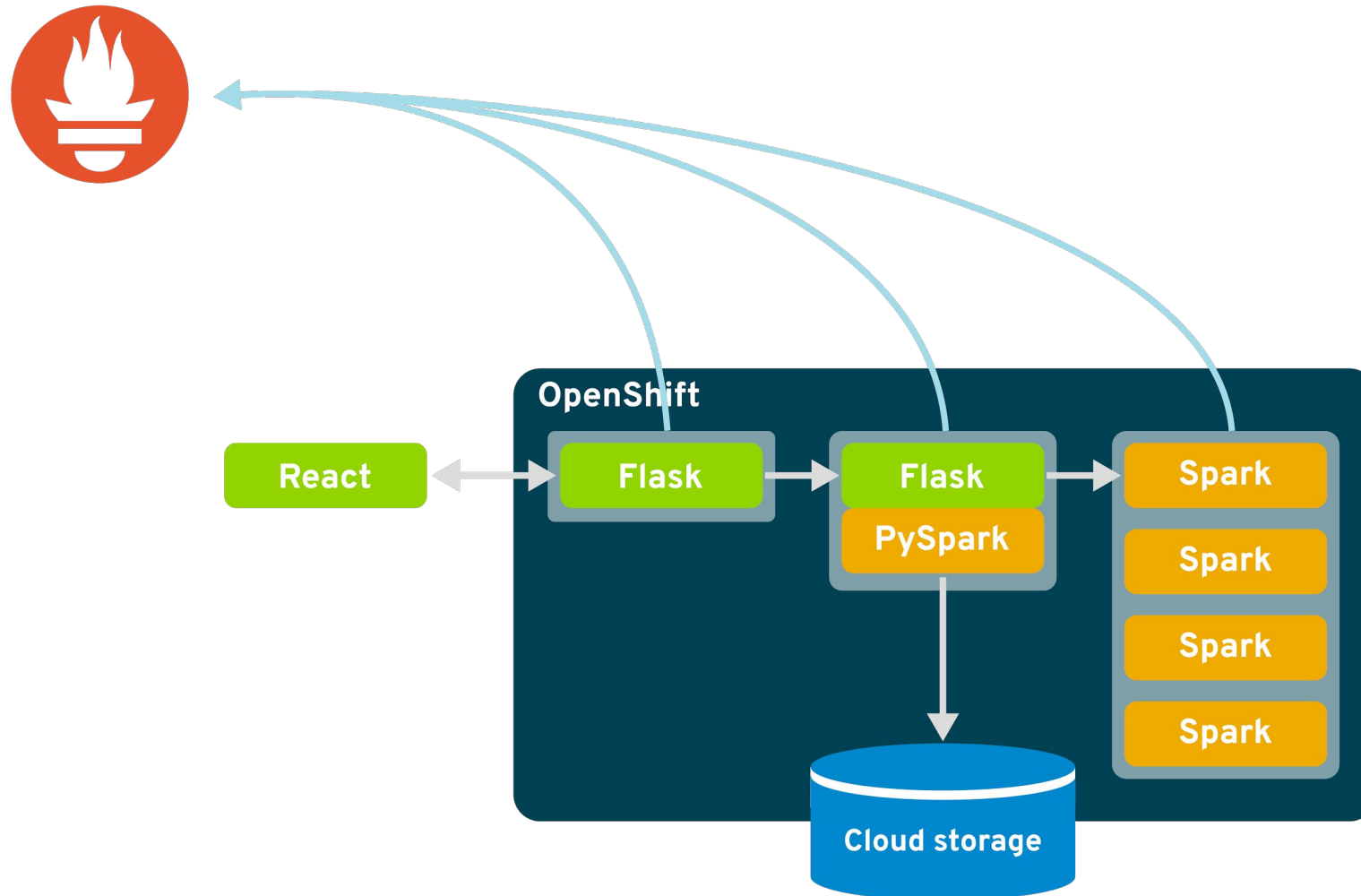
Layer in DevOps



Augment with lifecycle monitoring



Harvest customized metrics



Lessons Learned



Code Comments 3.0

Value-at-risk calculations

The basic idea behind the value-at-risk calculation is that we're going to look at the historical returns of a portfolio of securities and run many simulations to determine the range of returns we can expect from these. We can then predict, over a given time horizon, what our expected loss is at a given probability, e.g., we might say that there is less than a 10% chance that the portfolio will lose more than \$1,000,000.

Note that this is a didactic example and consequently makes some simplifying assumptions about the composition of the portfolio (i.e., only long positions in common stocks, so no options, dividends, or short selling) and the behavior of the market (i.e., day-to-day return percentages are normally-distributed and independent). Do not use this code to guide actual investment decisions!

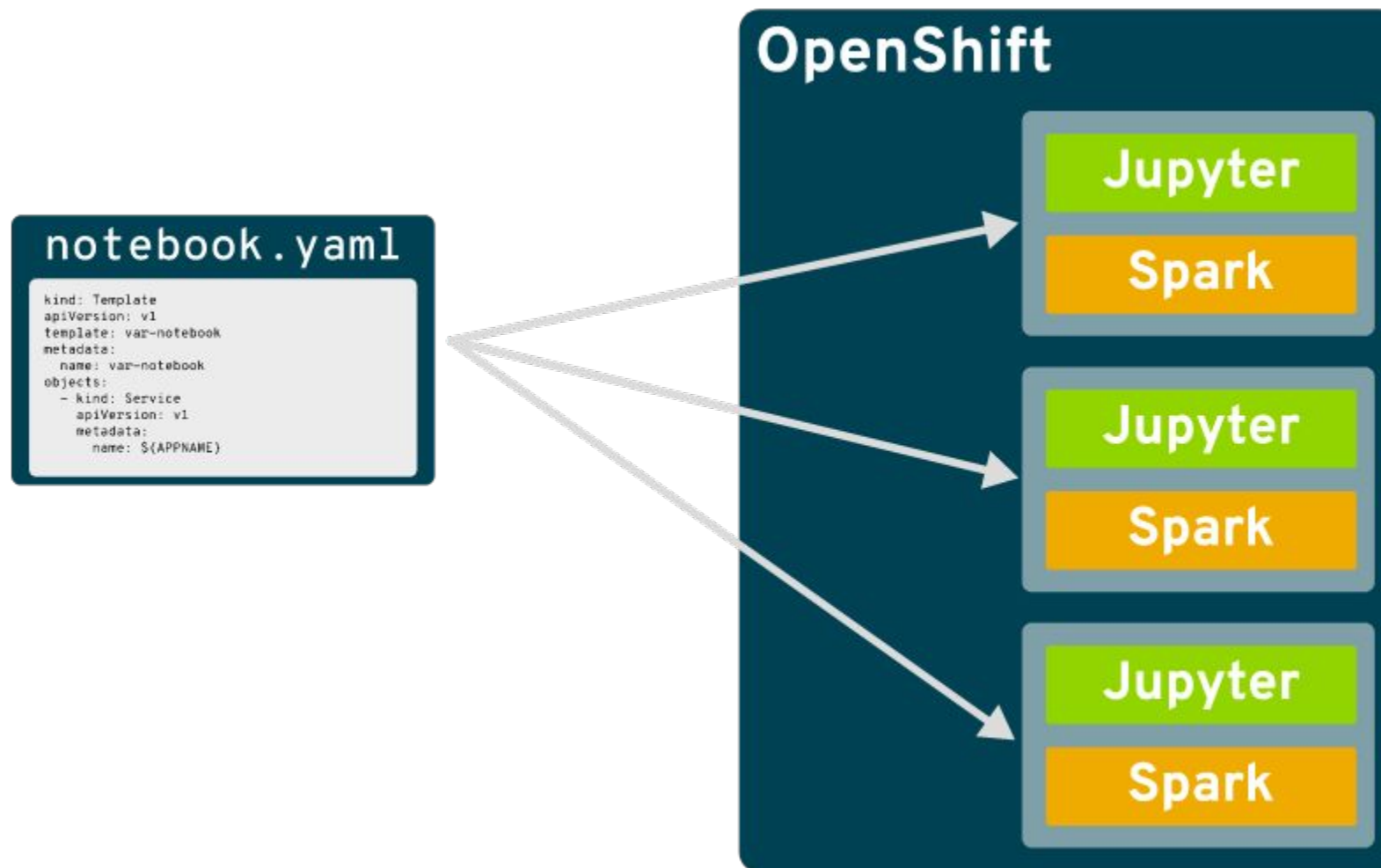
Basic setup

Here we import the pyspark module and set up a SparkSession.

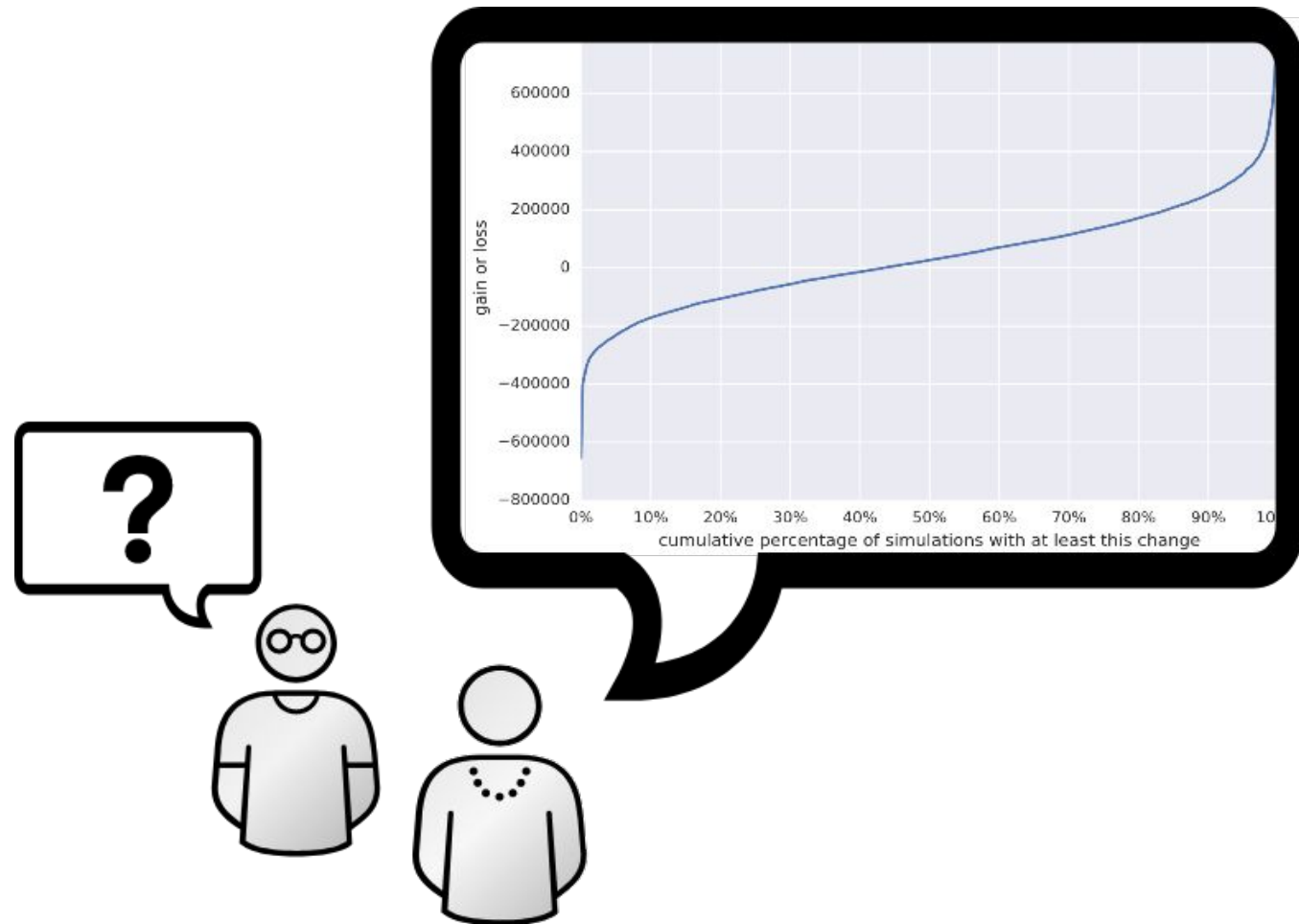
```
In [1]: import pyspark
        from pyspark.context import SparkContext
        from pyspark.sql import SparkSession, SQLContext

        spark = SparkSession.builder.master("local[*]").getOrCreate()
```

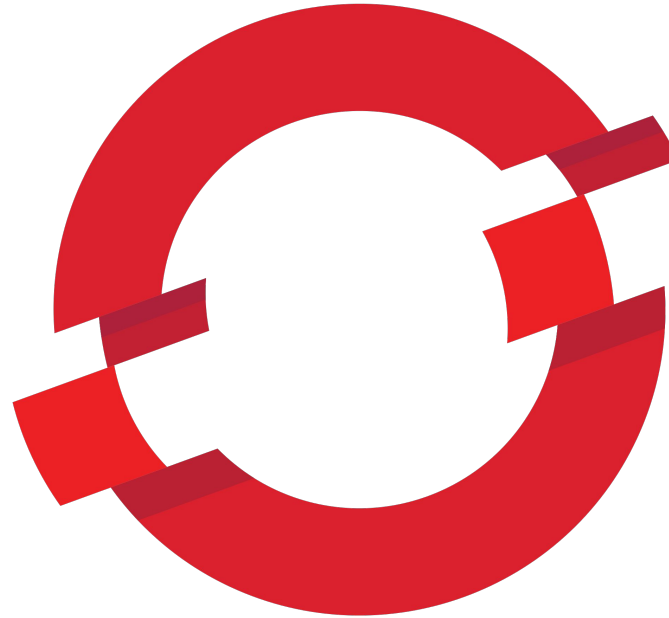
Templated Repeatability



Your greatest tool on the journey



Second greatest tool?



OPENSHIFT

What Next?

AI/ML on OpenShift



www.openshift.com/learn/topics/ai-ml

Open Data Hub



OPEN DATA HUB
AI Platform powered by Open Source

opendatahub.io

Radanalytics



radanalytics.io

Thank you

Keep in touch

elmiko@redhat.com

@elmiko@mastodon.technology

github.com/elmiko/rhug-artifacts

