



AlOps: Anomaly detection with Prometheus

Marcel Hild, Red Hat

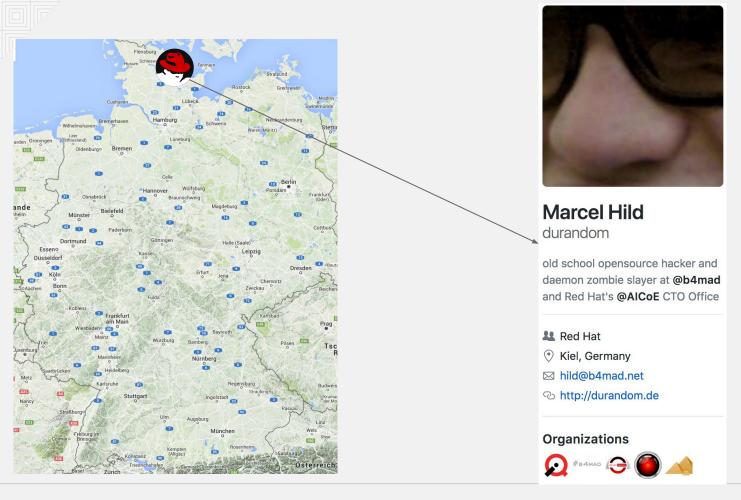
AlOps: Anomaly detection with Prometheus

Spice up your Monitoring with Al

Marcel Hild

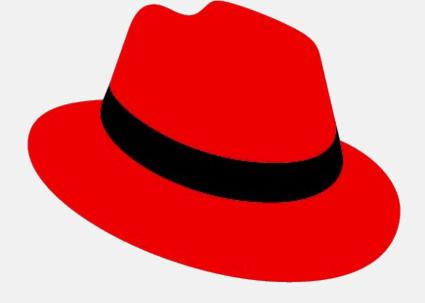
Principal Software Engineer @ Red Hat Al CoE / Office of the CTO











RedHat



Office of the CTO





Represents a workload requirement for our **platforms** across the hybrid cloud.



Applicable to Red Hat's existing core business in order to increase Open Source development and production efficiency.



Valuable to our customers as specific services and product capabilities, providing an Intelligent Platform experience.



Enable customers to build **Intelligent Apps** using Red Hat products as well as our broader partner ecosystem.





Project Thoth and Bots http://bit.ly/2zYfb6h

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OpenDataHub http://bit.ly/2y6Nh6m

110



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This Talk

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Agenda

Prometheus

Long term storage

Anatomy of an Anømål¥

Integration into monitoring setup



What's **not** in this talk



shiny product and the holy grail of monitoring

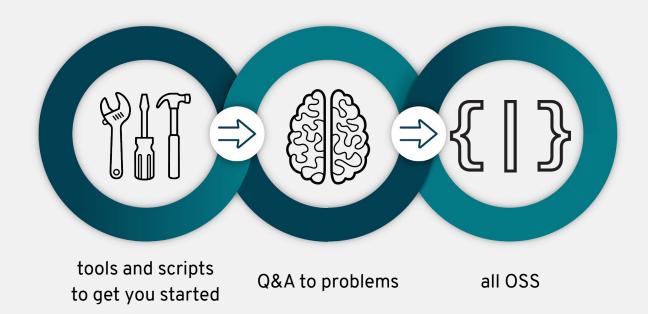


ready solution to turn your monitoring setup into spider demon



success story how we turned our messy monitoring into an advance ai monitoring

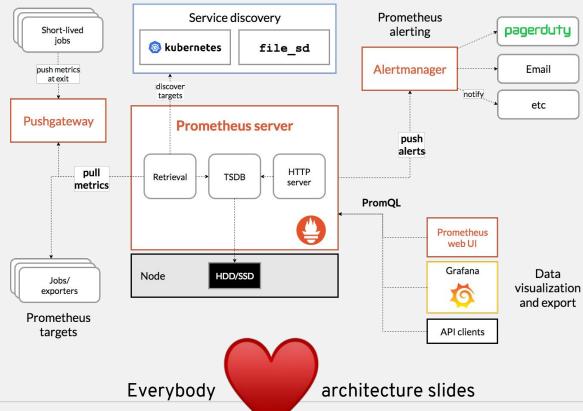
What **is** in this talk





What is prometheus?

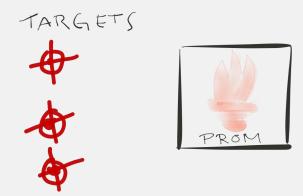




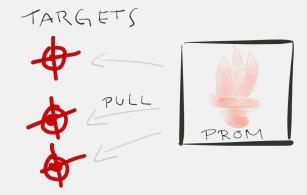




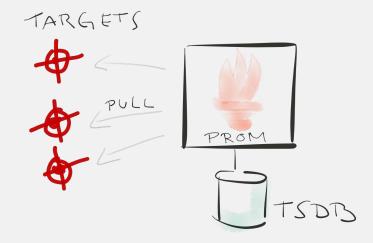




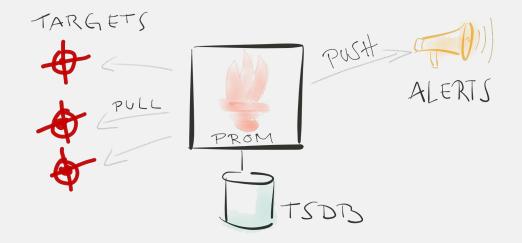














Prometheus is made for

MONITORING

ALERTING

SHORT TERM TIME SERIES DB



What do we need for machine learning?



Show me your DATA



Long term storage of Prometheus data



Too good to be true...



- Prometheus at scale
- Global query view
- Reliable historical data storage
- Unlimited retention
- Downsampling

thanos is in the making, but until then?



Works great, but...



gh/AlCoE/p-influx http://bit.ly/2y6CvwX

- easily hooked into prometheus with write and read endpoint
- Reliable historical data storage
- Great for data science
 - Pandas integration

Eats RAM for breakfast



Let's just store it...



prometheus scraper

- container can be configured to scrape any prometheus server
- can scrape all or a subset of the metrics
- stores data in ceph or S3 compliant storage
- can be queried with spark sql
- Future Proof: path to Thanos

gh/AICoE/p-Its http://bit.ly/2Qw9pho





Harness the power of spark to

- Query stored JSON files
- Distribute the workload
- Use spark library

notebook http://bit.ly/2PIZZVG

```
def get_stats(df):
    # calculate mean
    mean = df.agg(F.avg(F.col("values"))).head()[0]
   # calculate variance
    var = df.agg(F.variance(F.col("values"))).head()[0]
    # calculate standard deviation
    stddev = df.agg(F.stddev(F.col("values"))).head()[0]
    # calculate median
    median = float(df.approxQuantile("values", [0.5], 0.25)[0])
    return mean, var, stddev, median
mean, var, stddev, median = get_stats(data)
print("\tMean(values): ", mean)
print("\tVariance(values): ", var)
print("\tStddev(values): ", stddev)
print("\tMedian(values): ", median)
       Mean(values): 67087.9063346175
       Variance(values): 56691431555.4375
       Stddev(values): 238099.62527361838
       Median(values): 628.0
```



Things changed



- Prometheus at scale
- Global query view
- Reliable historical data storage
- Unlimited retention
- Downsampling

Success on OpenShift



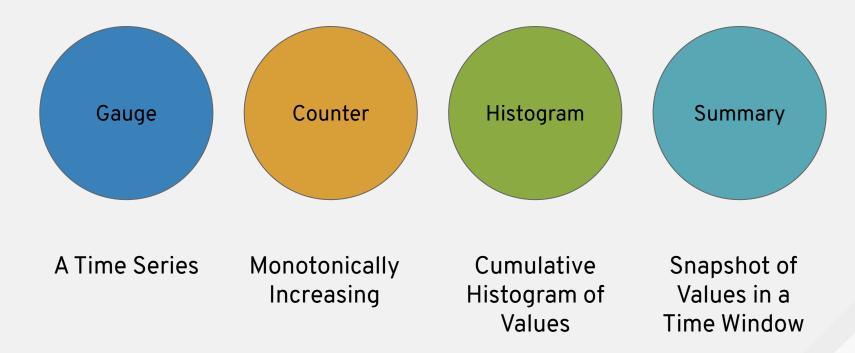
What do we REALLY need for machine learning?



Consistent DATA

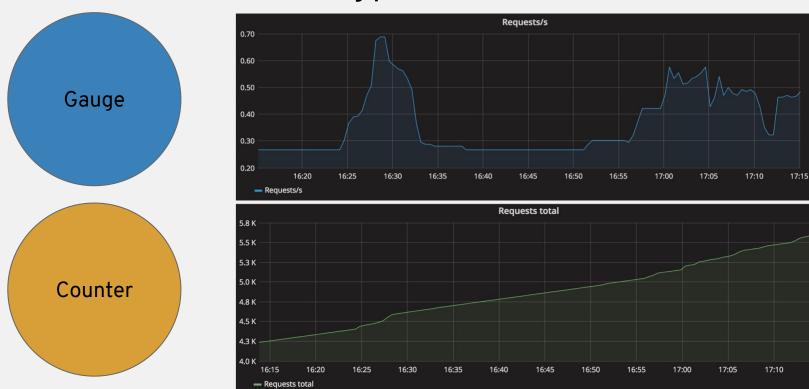


Prometheus Metric Types





Prometheus Metric Types



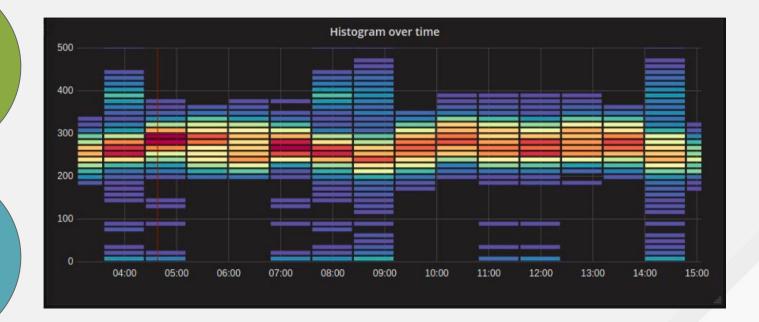


Prometheus Metric Types

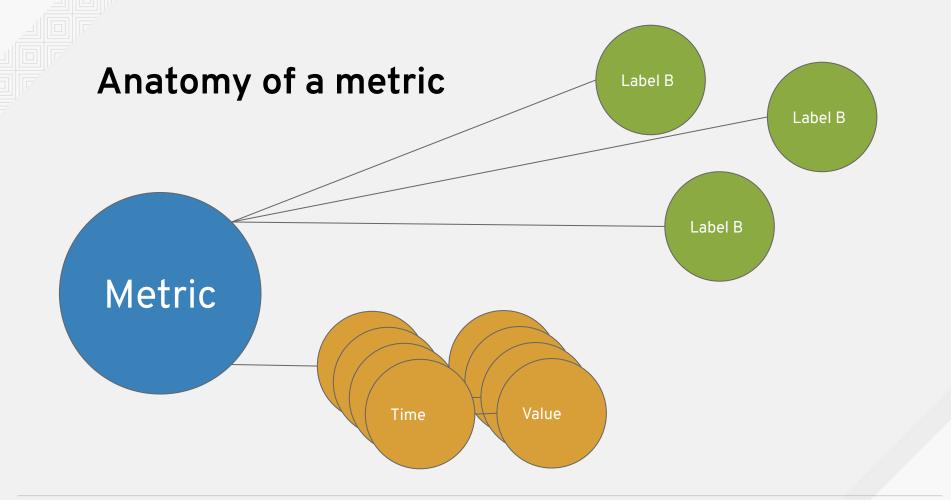
Histogram

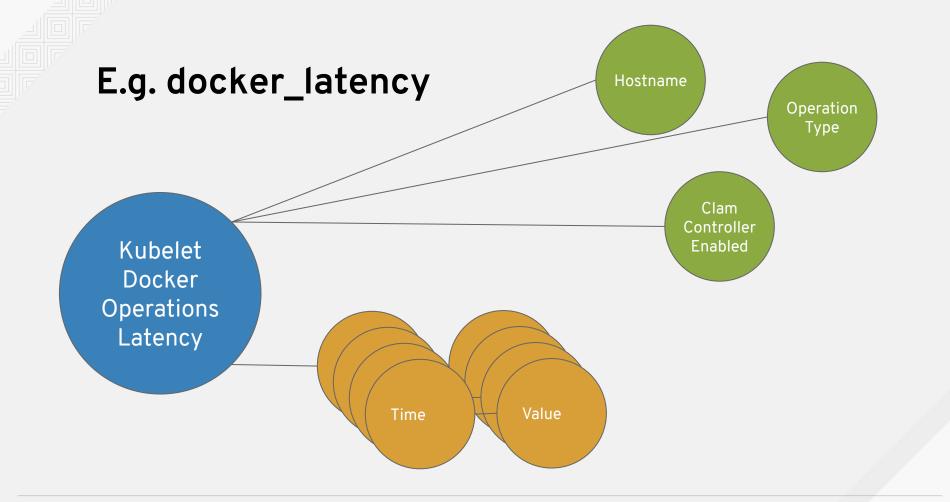
Cumulative

Summary
Time Window

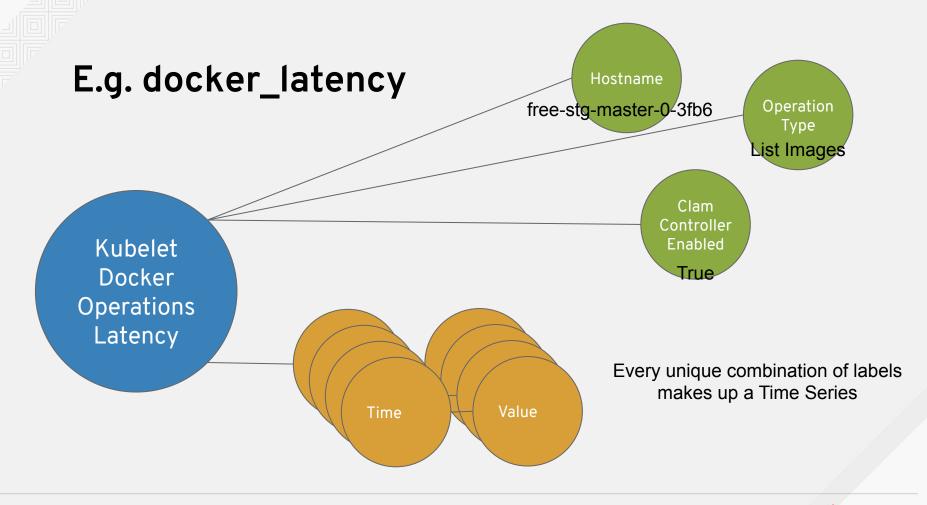












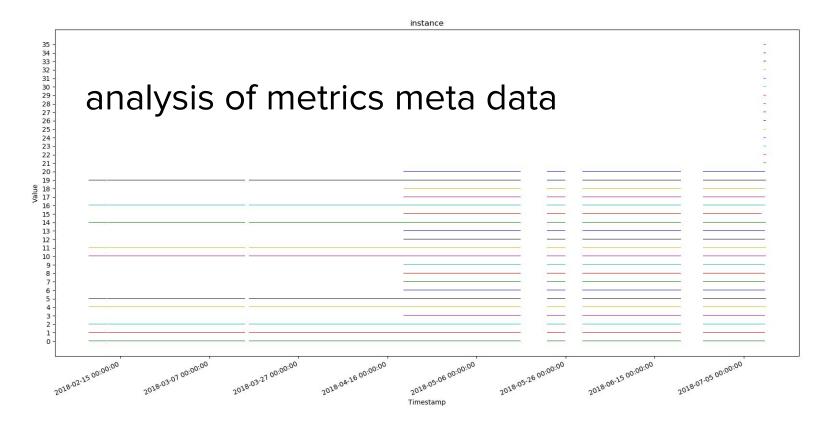
Monitoring is hard

GET/metrics

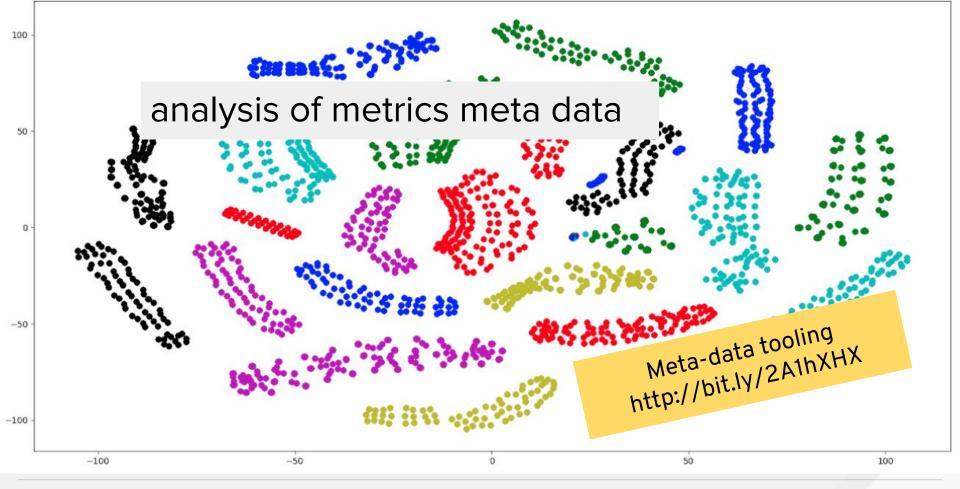
```
# HELP go gc duration seconds A summary of t
# TYPE go gc duration seconds summary
go gc duration seconds{quantile="0"} 9.7014e
go gc duration seconds{quantile="0.25"} 0.00
go gc duration seconds{quantile="0.5"} 0.000
go gc duration seconds{quantile="0.75"} 0.00
go gc duration seconds{quantile="1"} 0.10290
go gc duration seconds sum 0.239829369
go gc duration seconds count 196
# HELP go goroutines Number of goroutines th
# TYPE go goroutines gauge
go goroutines 144
# HELP go memstats alloc bytes Number of byt
# TYPE go memstats alloc bytes gauge
go memstats alloc bytes 4.5694928e+07
# HELP go memstats alloc bytes total Total n
# TYPE go memstats alloc bytes total counter
go memstats alloc bytes total 4.19435624e+09
```

- prometheus doesn't enforce a schema
 - /metrics can expose anything it wants
 - no control over what is being exposed by endpoints or targets
 - it can change if your endpoints change versions
- # of metrics to choose from
 - 1000+ for OpenShift
- State of the Art is Dashboards and Alerting
 - Dashboards and Alerting need domain knowledge
- No tools to explore meta-information in metrics









Anomaly Types



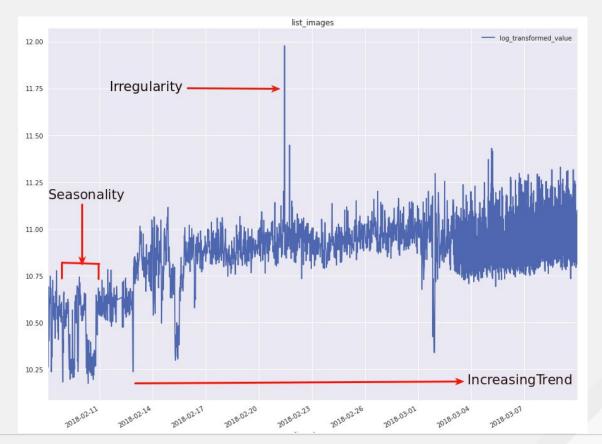
Components of Time Series

Trend

Increase or decrease in the series over a period of time.

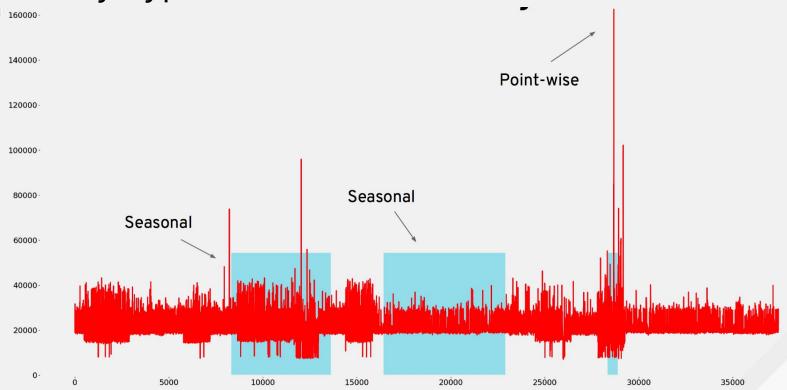
Seasonality

Regular pattern of up and down fluctuations. It is a short-term variation occurring due to seasonal factors.





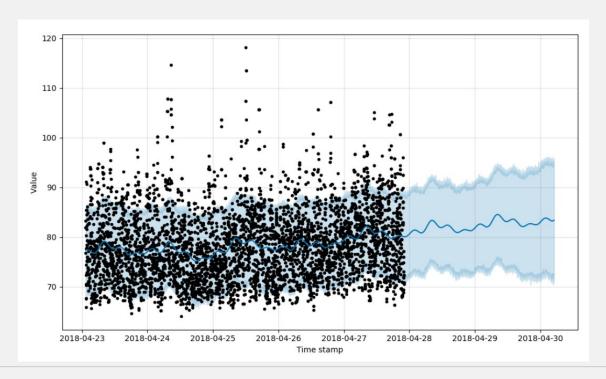
Anomaly Types





Anomaly Detection with Prophet

Predicting future data and dynamic thresholds

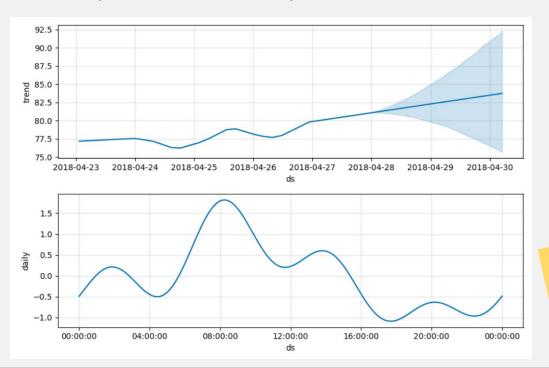


- list_images operation
- on OpenShift
- monitored by prometheus
- detecting outliers
- upper and lower bands



Anomaly Detection with Prophet

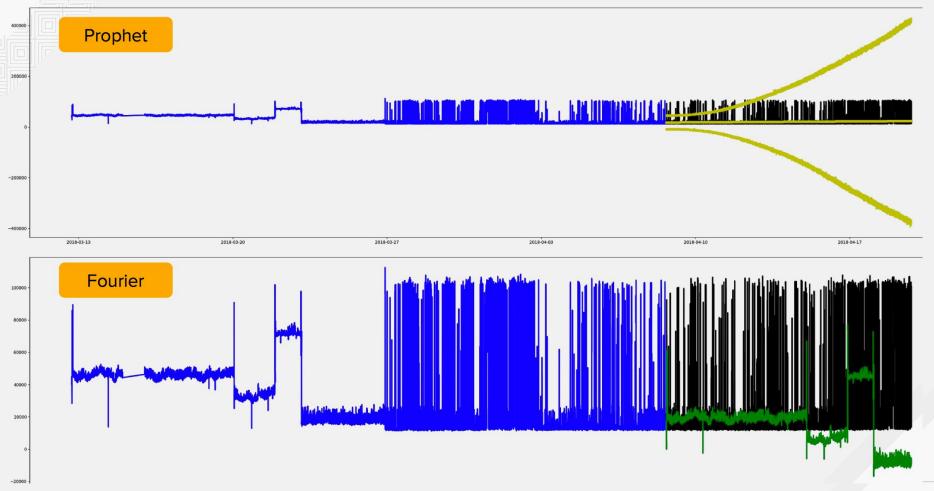
Extracting trends and seasonality



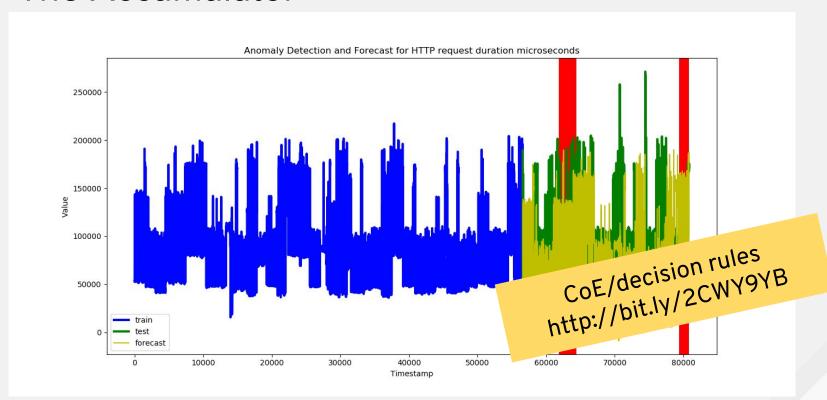
- list_images operation
- on OpenShift
- monitored by prometheus
- upward trends
- intraday seasonality

CoE/prophet http://bit.ly/2pLzGNj





The Accumulator



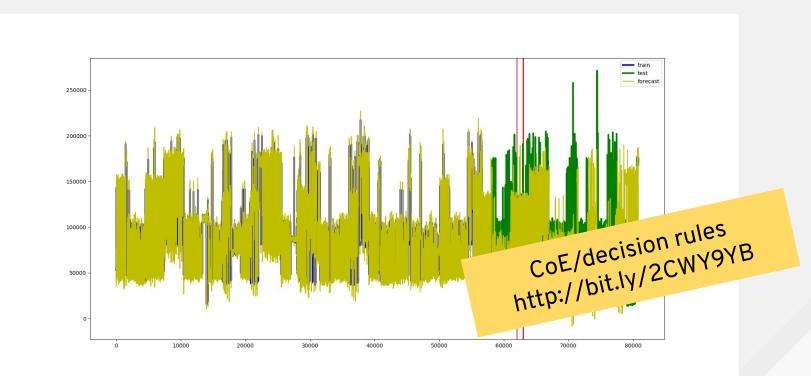


The Tail Probability





Combined



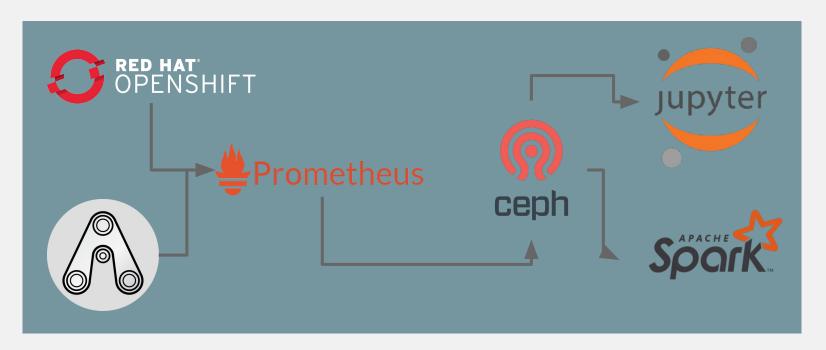


architecture setup so far



Research Setup

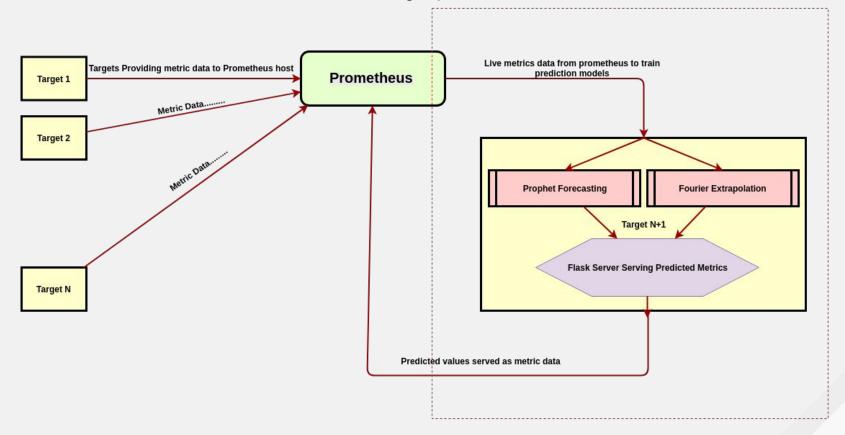
100% OpenSource Tooling

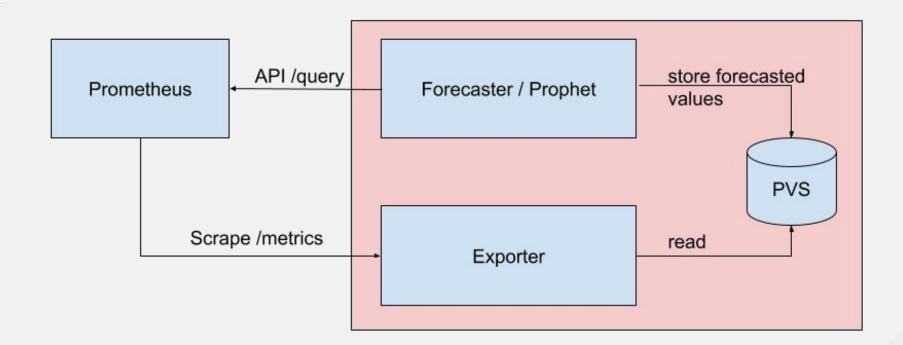


Now what? I want to



Prometheus Training Pipeline





GitHub, Inc. [US] https://github.com/AICoE/prometheus-anomaly-detector

Dockerfile	Update Dockerfil
Makefile	Add Makefile for ease of
README.md	Update README.md
app.py	Add more comments for
e ceph.py	Add functionality to retai
model.py	Make the live data query
prometheus.py	Make the live data query
requirements.txt	Update requirements.txt
train-prophet-deployment-templa	Add deployment templat

- Ready to use container
 - Local deployment
 - Kubernetes
 - OpenShift build config

CoE/prom-ad http://bit.ly/2yulCfh



Runtime configuration

```
29
     # Specific metric to run the model on
     metric_name = os.getenv('METRIC_NAME', 'kubelet_docker_operations_latency_microseconds')
```

Expose predictions via /metrics endpoint

```
# HELP predicted kubelet docker operations latency microseconds prophet anomaly Detected Anomaly using the Prophet model
 # PYPE predicted kubelet docker operations latency microseconds prophet anomaly gauge
predicted kubelet docker operations latency microseconds prophet anomal (beta kubernetes io arch="amd64", beta kubernetes io os="linux'
 0.00 prod upahift.eng.rdu2.redhat.com",operation_type="version",provider="rhos",quantile="0.5",region="compute",size="small"}
 # HELP predicted kubelet docker operations latency microseconds fourier anomaly Detected Anomaly using the Fourier model
 # TYPE predicted kubelet docker operations latency microseconds fewrier anomaly gauge
 predicted kubelet docker operations latency microseconds fourier anomal (beta kubernetes io arch="amd64",beta kubernetes io os="linux"
 0.00 ocp.prod.upshift.eng.rdu2.redhat.com",operation type="version",provider="rhos",quantile="0.5",region="compute",size="small" 0.0
```



Alerting Rules

```
Testing alert
alert: MetricOutofProphetBounds
     kubelet docker < ignoring(job, instance) predicted values prophet yhat lower or kubelet docker > ignoring(job, instance) predicted values prophet yhat upper
alert MetricOutofFourierBounds
     kubelet docker < ignoring(job, instance) predicted values fourier yhat lower or kubelet docker > ignoring(job, instance) predicted values fourier yhat upper
```

<u>Demo</u> <u>Time</u>















- predicted_node_load1_fourier_anomaly









I like what you got



notebooks http://bit.ly/2PIZZVG gh/AICoE/p-influx http://bit.ly/2y6CvwX

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QUESTIONS?

Meta-data tooling http://bit.ly/2A1hXHX

http://bit.ly/2CWY9YB

CoE/prom-ad http://bit.ly/2yulCfh

in linkedin-

/company/red-hat

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OpenDataHub http://bit.ly/2y6Nh6m Thanos Blog Post https://red.ht/2HpB8Az

gh/AICoE/p-lts http://bit.ly/2Qw9pho

