





### Bifrost Toolkit: Data-Driven Release Strategies

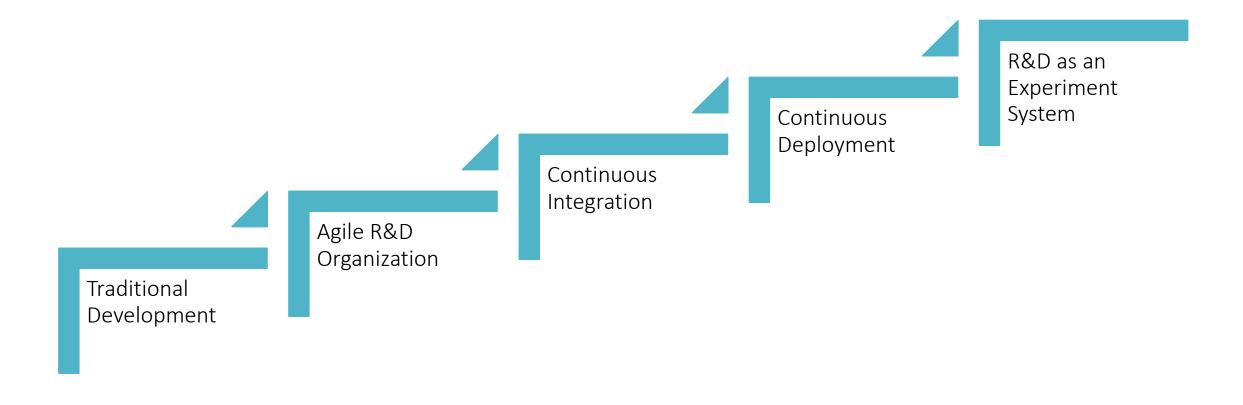
Formalize and automate real-time, data-driven live-testing methods using a DSL

Master Thesis Defense – 28.04.2016 Dominik Schöni





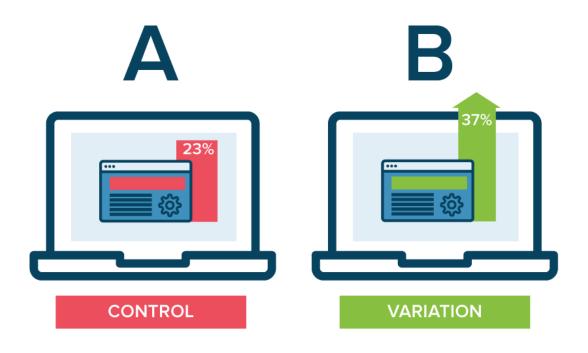
# The Stairway to Heaven



[OAB12]

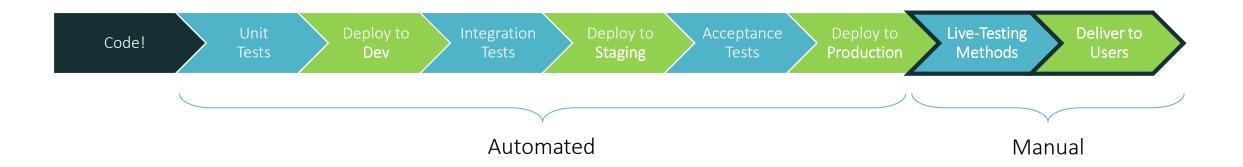
# Live-Testing Methods

A/B Testing



Source: http://www.optimizely.com

# Releasing Software



## Automation is key

Automation allows for transparency and traceability

Gives developers confidence when releasing

Time to market gets accelerated

Computers are better at repeatable tasks than humans

### Research Question #1

"How can we formalize a (generic) model for data-driven release and deployment strategies?"

## Methods of Live-Testing

A/B Testing:

A way to compare two versions of software with each other, often only differentiated in one aspect

Canary Launches:

Introducing a new version into a stable environment

Shadow Launches:

Deploying features that are not visible to users

Phased Rollouts:

Expose users gradually to a new software version

Blue/Green Deployments:

Having two identical production environments, allowing to switch seamlessly from one to another

# Identifying characteristics

A/B Testing

Canary Launches

**Shadow Launches** 

**Phased Rollouts** 

Blue/Green Deployments

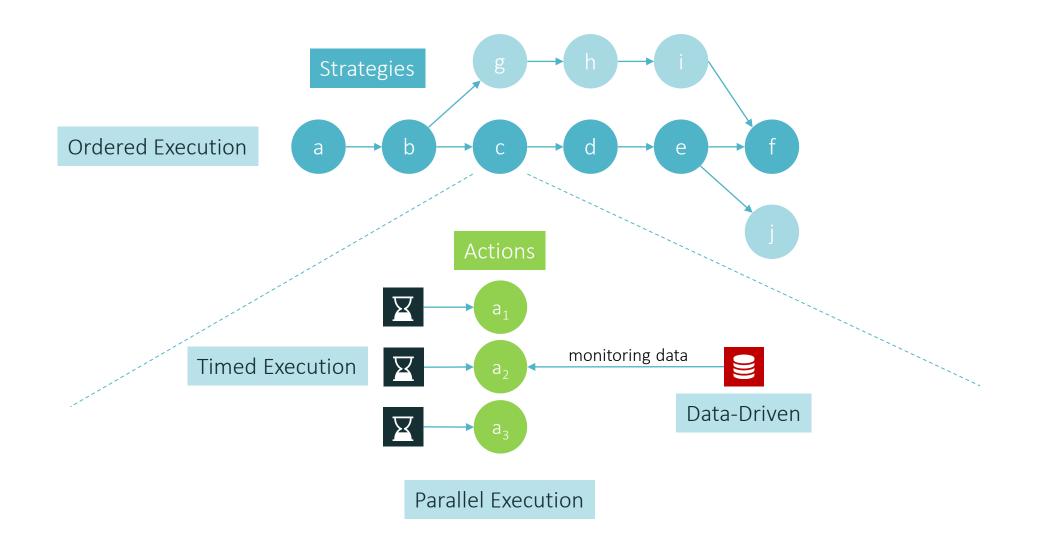
Ordered Execution

Parallel Execution

**Timed Execution** 

Data-Driven

### Bifrost Release Model



### Bifrost Release Model

The model was formalized using a mathematical representation:

```
Release: \{\{c_1, ..., c_n\}, (s_1, ..., s_n)\}
```

 $Strategy: \{Parallel_A, Success_A, Failure_A\}$ 

Action:  $A = \{\Theta, \Delta, \Omega\}$ 

 $Parallel_A: \{a_1, \ldots, a_n \mid a_i \in Action\}$ 

 $Success_A: \{x \mid x \in Action\}$ 

 $Failure_A: \{x \mid x \in Action\}$ 

....

### Research Question #2

"How can we build a tool that supports and automates datadriven release and deployment strategies for microservicesbased architectures in a non-intrusive way?"

## Leveraging the Bifrost Release Model

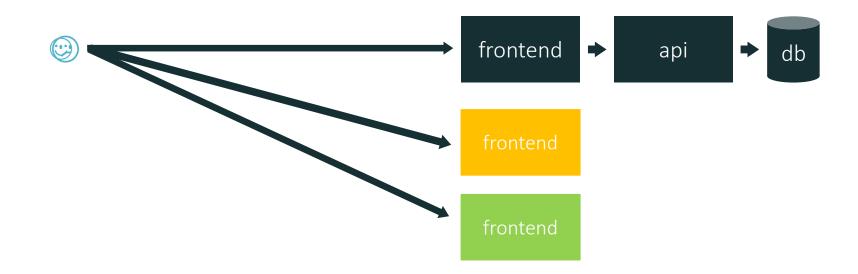
The model identifies abstract concepts of releases, strategies and actions.

→ Translated for developers into a Domain Specific Language

Bifrost DSL: YAML-based DSL

- Specifies what services to modify during a release, and how.
- Includes concepts of releases, strategies and actions as in the Bifrost Release Model.
- Enables transparency and traceability as it is file-based and can be version-controlled.

# Approach: Implementation Techniques

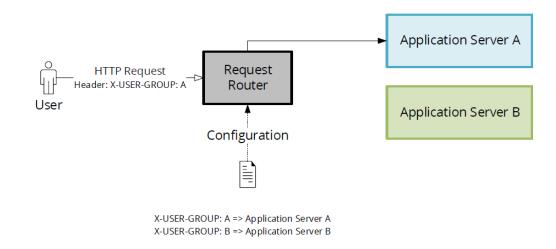


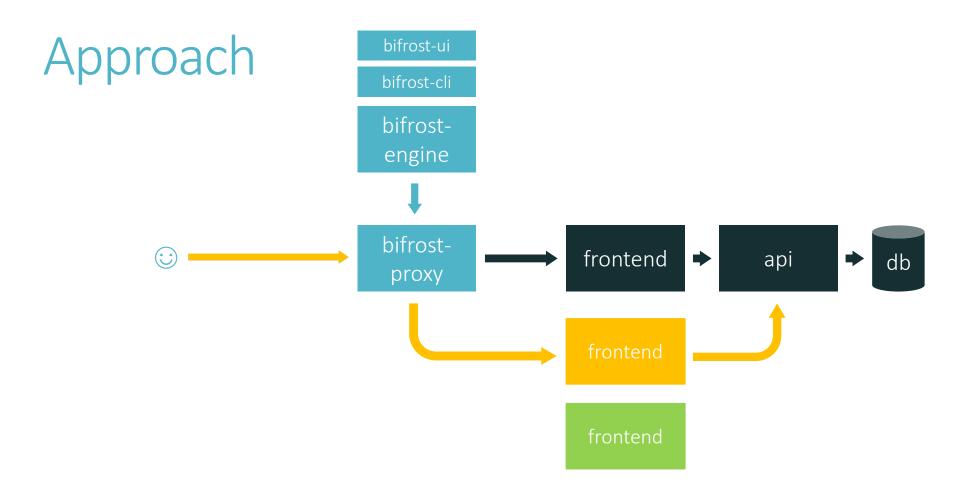
# Approach: Implementation Techniques

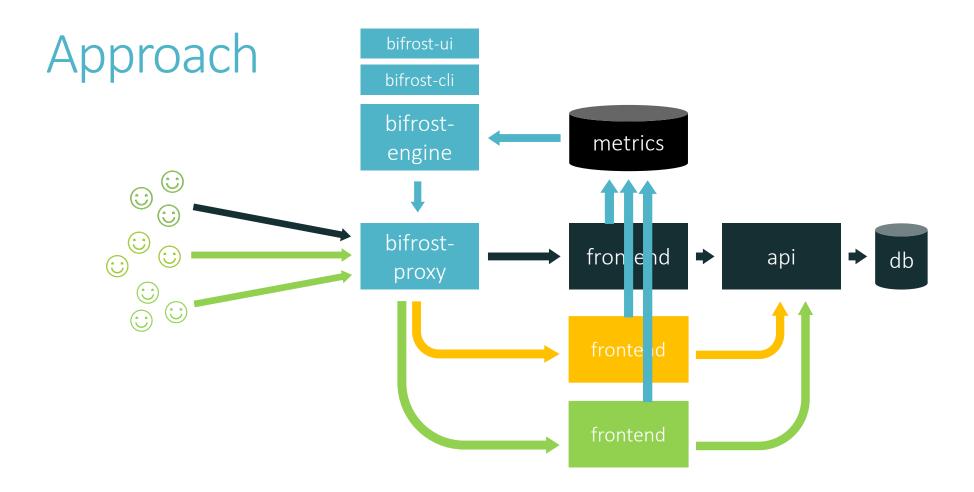
#### Feature Toggles

```
1  // Get all of a user's enabled features
2  var Features = fflip.userFeatures(someFreeUser);
3  if(Features.closedBeta) {
4   console.log('Welcome to the Closed Beta!');
5  }
6  // Or, just get a single one
7  if (fflip.userHasFeature(someFreeUser, 'closedBeta')) {
8   console.log('Welcome to the Closed Beta!');
9  }
```

#### Dynamic Request Routing

















Demo











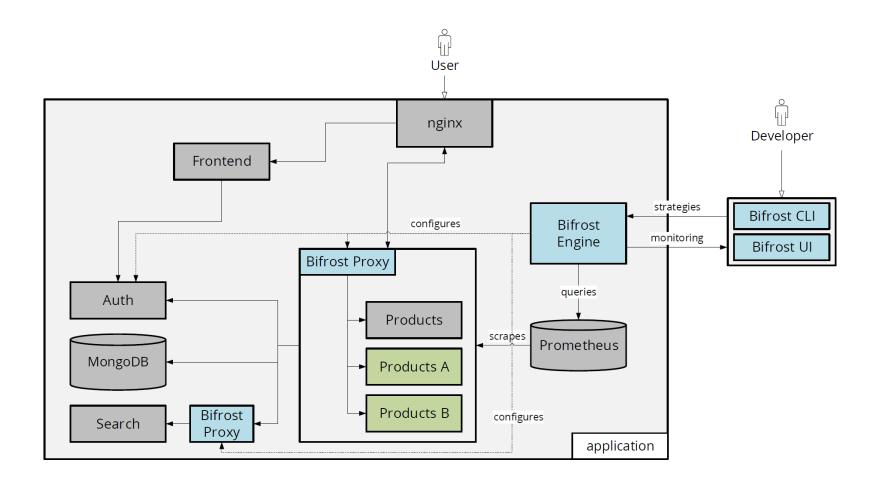
# Evaluation

### Qualitative Evaluation

Compared the Bifrost Toolkit to 5 similar tools using a set of defined dimensions.

- Tools exist, but mostly focus on specific use cases
- Most tools are either closed-source, tightly coupled with existing PaaS or locked-in to specific development platforms and architectures
- Bifrost Toolkit allows for more complex release procedures
- Only one additional tool allows flexible data-driven decision making

## Quantitative Evaluation



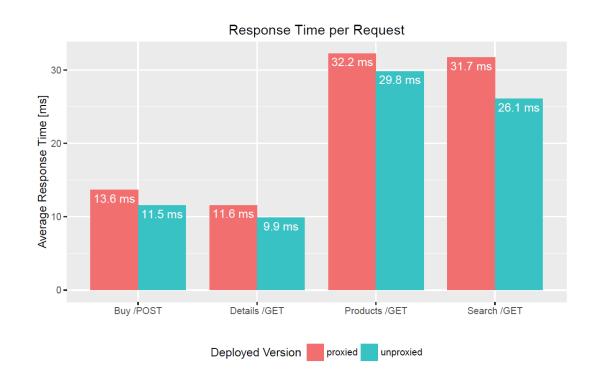
# Request Performance

JMeter Testsuite fires 4 requests simulating normal user traffic to the application. Conducted with the Bifrost Toolkit deployed (**proxied**) and without (**unxproxied**), to measure the raw proxy overhead:

- Buying a product
- Querying product-details
- Wildcard-search for a product
- Retrieving list of all products

# Request Performance

- One proxy instance adds approximately 2.4ms of delay
- The size of request and response can influence the added delay
- When deploying multiple proxies into the application landscape, a linear increase of the delay is to be expected



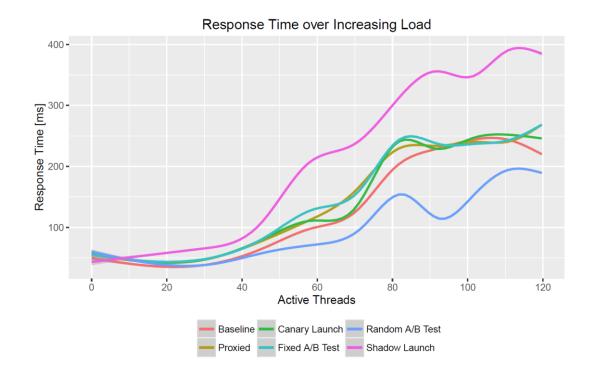
### Filter Performance

4 different strategies/actions that simulate various live-testing methods under increasing load:

- A/B testing with fixed user allocation
- A/B testing with random user allocation
- Canary launching using header-filtering
- Shadow launching with traffic duplication

### Filter Performance

- Fixed A/B testing adds a performance overhead
- Canary launches (and thus header filtering) show similar performance as running the proxy without filter
- Shadow launches increased the response time noticeably.



#### Release Performance

Simulated a complete release using multiple live-testing methods:

#### Canary Launch

• 5% Traffic Redirection each to Product A and Product B for 60 seconds.

#### Shadow Launch

Duplicate Traffic to Product A and Product B for 60 seconds.

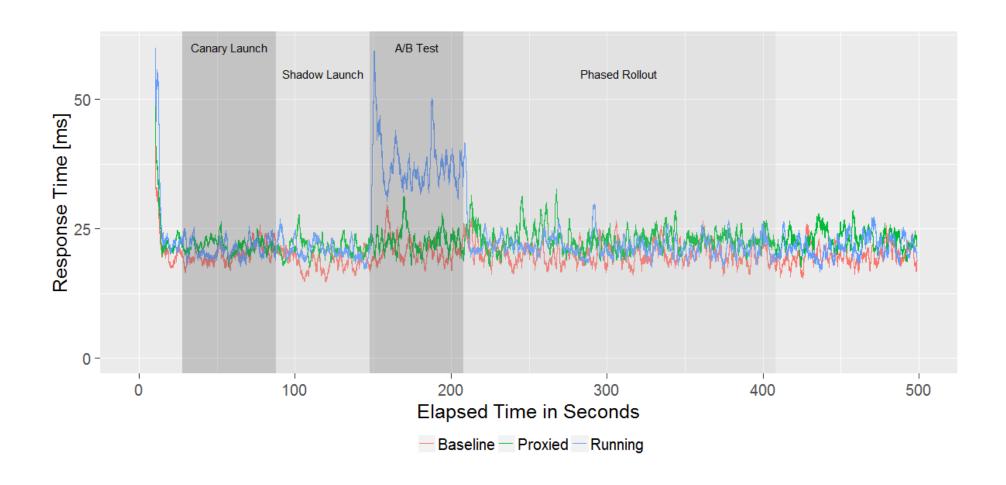
#### A/B Test

Monitor sales performance of Product A and Product B using sticky sessions for 60 seconds.

#### Phased Rollout

Rollout winner of A/B test over 200 seconds.

## Release Performance



## Summary and Conclusion

«How can we formalize a (generic) model for data-driven release and deployment strategies?»

Introduces a generic model for data-driven release and deployment strategies.

- Formalized using a mathematical model.
- Built using a usage scenario, incorporating identified characteristics of live-testing methods.

## Summary and Conclusion

«How can we build a tool that supports and automates data-driven release and deployment strategies for microservices-based architectures in a non-intrusive way?»

Implemented a prototype using dynamic request routing, that makes use of the Bifrost Release Model.

- Provides a wider range of functionality than existing tools and approaches.
- Is platform-independent and does not require any modification of existing source code.
- Introduces a measurable but small delay into applications, and thus could profit from further optimizations.

### Future Work

#### Possible improvements:

- Introduce formal model verification
- Add additional metric-providers
- Explore detailed integration in deployment pipeline
- Support for feature toggles
- Provide a deeper integration with existing laaS- or PaaS-providers









Questions?



### Bifrost Release Model

```
\{\{c_1,\ldots,c_n\},(s_1,\ldots,s_n)\}
Release:
                            \{Parallel_A, Success_A, Failure_A\}
Strategy:
                            \{a_1,\ldots,a_n\mid a_i\in Action\}
Parallel_A:
                           \{x \mid x \in Action\}
Success_A:
                           \{x \mid x \in Action\}
Failure_A:
                                                               f_a(a_i) = result \mid result \in \{True, False\}
                           A = \{\Theta, \Delta, \Omega\}
Action:
                                 f(c_i) \to c_i'
      \Theta:
```

### Bifrost Release Model

Evaluation of Action using a Timer

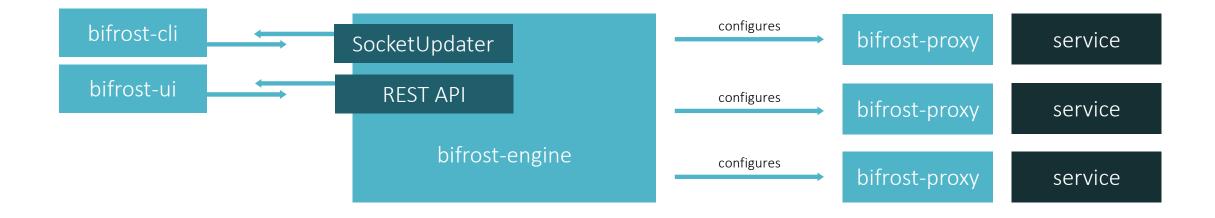
$$f_{\Delta}(a_i) = \{f_a(a_i)_1...f_a(a_i)_n\} = \begin{cases} True & |f_a(a_i) \in True| \ge \Delta_{Threshold} \\ False & |f_a(a_i) \in True| < \Delta_{Threshold} \end{cases}$$

Evaluation of Strategy:

$$f_s(\{a_1,..,a_n\}) \to f_a(a_1) \land ... \land f_a(a_n) \to result \mid result \in \{True, False\}$$

## Architecture and Technology

Written in ECMAScript 6 (JavaScript)



#### Actions

#### Route

- %-Traffic Filter
- Header-Field Filter
- Sticky Sessions

#### Request

- HTTP-Request
- Checks for statuscode
- Healthcheck

#### **Nested Actions**

- AND/OR Actions
- Nesting allows to model complex rulesets

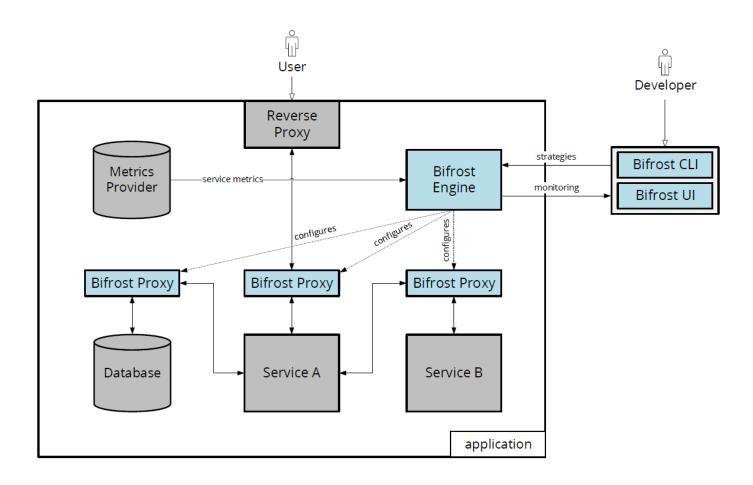
#### Stop

Allows for manual confirmation after strategy

#### Metric

- Various providers possible.
- Simple inline-validation or custom validation using HTTP-Requests

### Architecture



## Challenges

Docker is changing their **network** feature. No support for new alias-feature on API yet.

• https://github.com/docker/docker/issues/18699

# Compared Tools

	A/B Testine	Canary Launs	jies Shadow Laur	dhes Phased Rollo	its Blue Green	Jeployments Combination
GateKeeper	Yes	Yes	Yes	Partial	No	Yes
CanaryAdvisor	No	Yes	No	No	No	Yes
Vamp	Yes	Yes	No	No	Yes	No
Scientist!	Partial	Partial	Yes	Partial	No	Partial
ION-Roller	No	Partial	No	Yes	Yes	No
Bifrost	Yes	Yes	Yes	Yes	Yes	Yes

Table 5.1: Feature-comparison of analyzed tools

## Qualitative Evaluation - Results

	Platform	ghosic Coder <sup>M</sup>	guttal Performans	e Neutrality	hity Automated	Data Driver	Releases Open-Source
GateKeeper	Partial	No	Yes	Yes	Yes	Yes	No
CanaryAdvisor	Yes	Yes	Yes	No	Partial	No	No
Vamp	Partial	Yes	No	Yes	No	No	Yes
Scientist!	No	No	Partial	No	No	No	Yes
ION-Roller	Partial	Yes	Yes	Yes	No	No	Yes
Bifrost	Yes	Yes	No	Yes	Yes	Yes	Yes

Table 5.2: Comparison of live-testing tools

# Request Performance

Request	Туре	Mean	+/-	SD	Min	Max	Median
Buy /POST	Proxied	13.64	+2.5	16.82	5	252	8
	Unproxied	11.14		16.11	4	201	6
Details /GET	Proxied	11.58	+1.71	16.33	5	269	7
	Unproxied	9.9		16.33	4	215	6
Products /GET	Proxied	32.25	+2.45	20.82	14	281	27
	Unproxied	29.80		21.45	12	339	24
Search /GET	Proxied	31.69	+5.63	25.23	10	340	25
	Unproxied	26.06		23.89	7	267	19

Table 5.3: Results of request performance test in milliseconds

## Example: Deployment/Services

```
name: Webshop Redesign A/B Test
deployment:
 orchestrator:
   proxy:
     mapping:
       frontend: bifrost frontend proxy
 services:
  - name: ClassicShop
    host: frontend
    port: 3000
  - name: RedesignedShop
    host: frontend redesigned
    port: 3000
```

# Example: Strategies

```
strategies:
   name: 50_50_traffic
   actions:
      - route:
          from: ClassicShop
          to: RedesignedShop
          filters:
            - traffic:
               percentage: 50
          intervalTime: 60
   next: check metrics
- name: check metrics
   actions:
```

# Example: Strategies

```
AND:
 actions:
  - OR:
      - metric:
          providers:
            - prometheus:
                query: "avg_over_time(container_cpu_usage_seconds_total{instance="workerA_1"}[60s])"
          validator: ">0.5"
      - metric:
          providers:
            - prometheus:
                query: "avg_over_time(container_cpu_usage_seconds_total{instance="workerA 2"}[60s])"
          validator: ">0.5"
   - OR:
      - metric:
          providers:
            - prometheus:
                query: "avg over time(container cpu usage seconds total{instance="workerB 1"}[60s])"
          validator: ">0.5"
      - metric:
          providers:
            - prometheus:
                query: "avg over time(container cpu usage seconds total{instance="workerB 2"}[60s])"
          validator: ">0.5"
onTrue: "highCPU"
onFalse: "lowCPU"
```

# Approach: Specifying Release Strategies

```
name: Webshop Redesign A/B Test
deployment:
 orchestrator:
   proxy:
     mapping:
       frontend: bifrost_frontend proxy
 services:
  - name: ClassicShop
    host: frontend
    port: 3000
  - name: RedesignedShop
    host: frontend redesigned
    port: 3000
```

# Approach: Specifying Release Strategies

```
strategies:
 - name: 50 50 traffic
   actions:
      - route:
          from: ClassicShop
          to: RedesignedShop
          filters:
            - traffic:
                percentage: 50
          intervalTime: 60
   next: check metrics
   name: check metrics
   actions:
```