



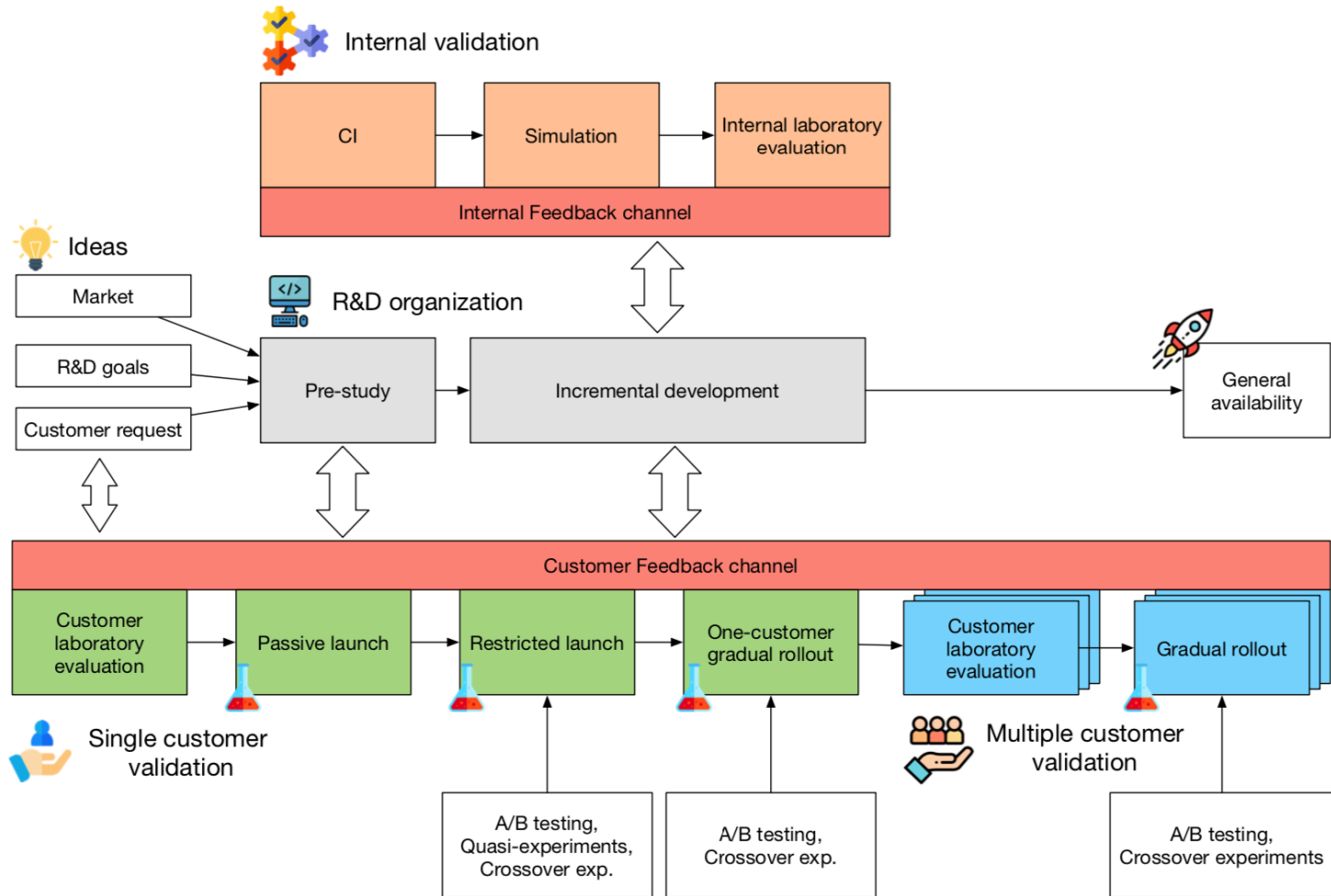
# **Data-Driven Continuous Evolution of Autonomous Systems of Systems (#19)**

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# Experiments in B2B and mission critical features?



# Experiments in B2B and mission critical features?

- Validation of a mission-critical feature in field
  - Faster than in-house validation
- Some customers see the value of faster field experiments that and are willing to cooperate!
  - Even if it breaks (because it will be fixed faster)
- Experimentation does not exclude internal validation, CI, laboratory experiments etc...
  - Those can happen in parallel or only with features that have evidence to work/deliver value

# Experiments in B2B and mission critical features?

- “The field deployment allowed us to identify a problem that none of our existing internal testing could identify”.
- “We are not more efficient when it comes to building time, but we know in advance that it works in the real world. That is the benefit of prototyping and experimenting”

# Continuous experimentation

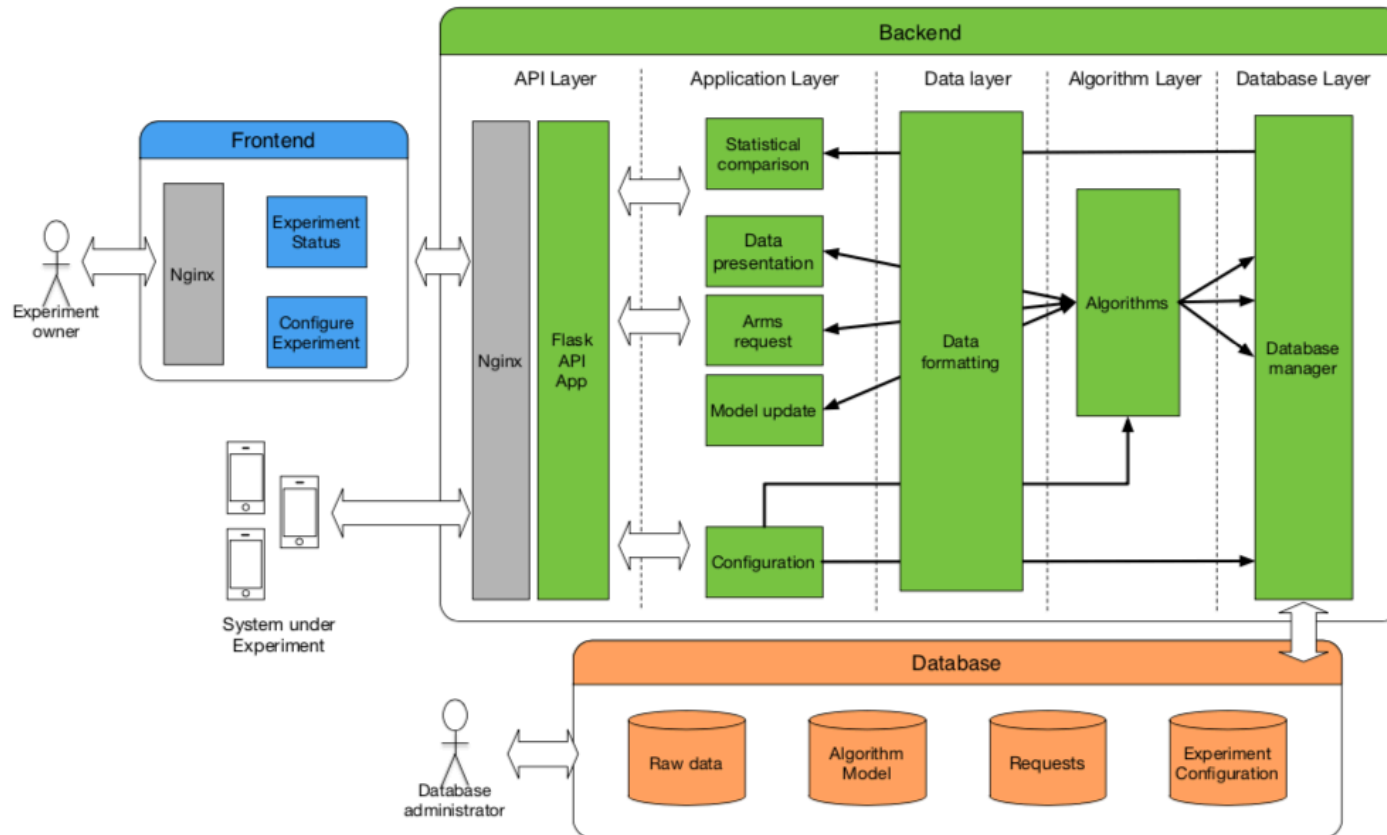
- A broader view of CE beyond A/B testing
- Different types of experiments
  - Business experiments
    - A/B testing, quasi experiments
  - Regression experiments
    - Dark launches, canary releases
  - Optimization experiments
    - Bayesian optimization, RL
  - Customer support experiments
    - Quasi-experiments, crossover
- Over 25 different techniques...

# How to select a CE technique?

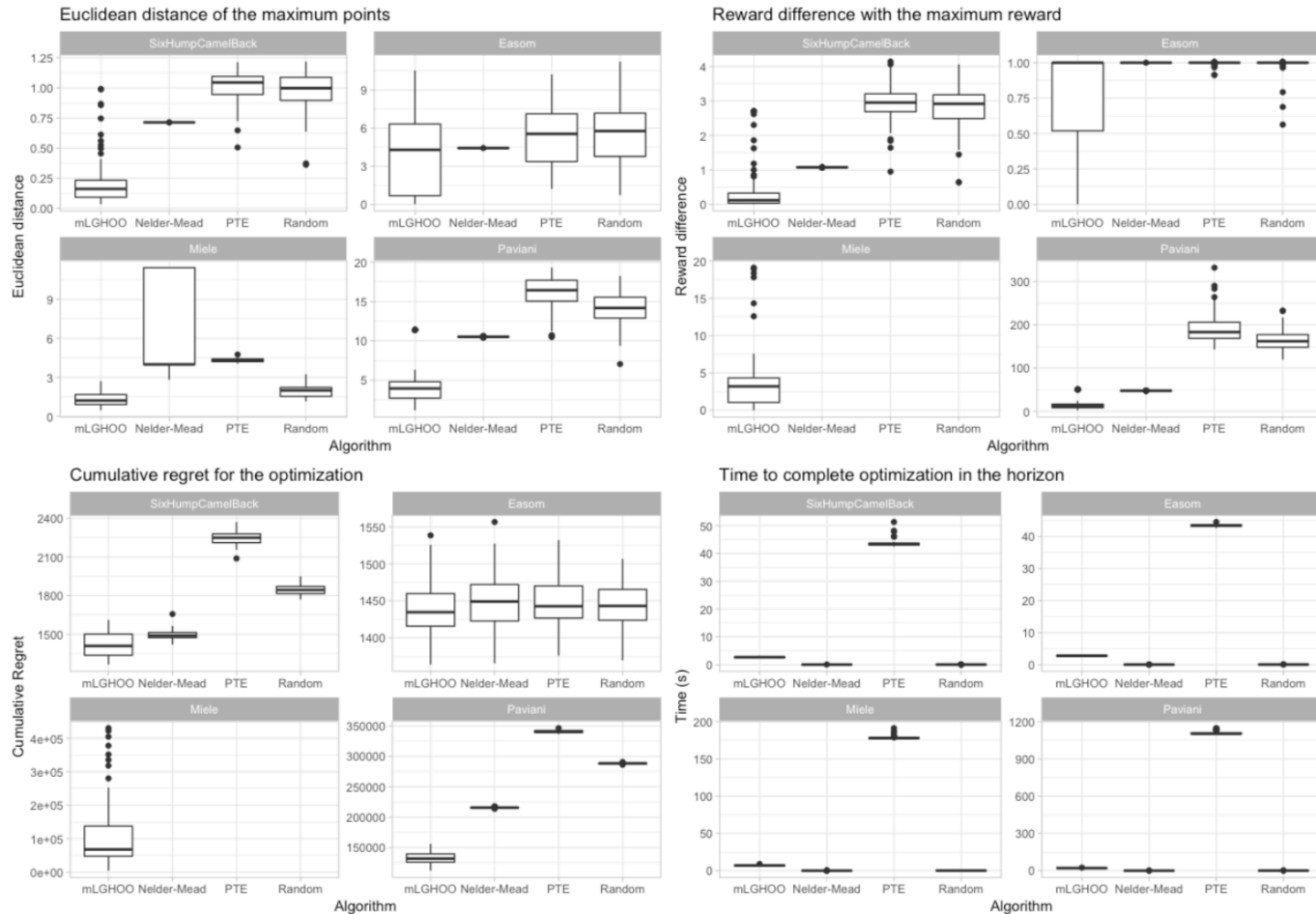
- Multiple case-study with 4 companies and 39 practitioners (with Ericsson from Software Center) + analysis of 108 papers
- When do I use each one of those?
- Why/when use one technique compared to other?
- Infrastructure to use

# Optimization experiments

- Automatically identify best configuration parameters



# How our system compares to other state-of-the-art systems?





# Statistical comparison of black box algorithms for optimization of expensive functions

- Sometimes it is not clear which to select in which case!
  - There are hundreds of different algorithms
- We are doing a large scale comparison between algorithms
  - Over 25 algorithms
  - Over 170 benchmark functions
  - Different measurements metrics
  - Different conditions (noise, budget etc...)
- Applications
  - Field optimization
  - Hyperparameter tuning of ML models
  - AutoML

# Customer aspects in CE

- Customers have different prioritization regarding experimentation
  - “I want to try the alpha 0.0.0.1a feature and help guide the development”
  - “I want the latest features in a stable release”
  - “I want the oldest supported software because I hate updates”
- What are the different profiles?
- How to plan experimentation activities based on customer profiles?
- Case study with Ericsson

**It would be great to understand and see how this works  
with different companies**

# CE in Software Product Lines

- Case study with Ericsson
- How to plan the experiment? Where to start? How to analyze the data?
- Which product do you select to start experimenting?
- Transfer conclusions to other products?
- Incorporate knowledge and experience in the experimentation process
- E.g. How do you experiment in a feature available in different products?
- **Investigate/validate in other companies**

# Sprint plans

- Understanding/adapting the HURRIER process in other companies
  - If you have interest contact me!
- Understand continuous experimentation from the software product line point of view.
  - Experiences from other companies
- Customer aspects in CE.
  - Experiences from other companies
- Comparison of algorithms for expensive black-box optimization