INFRASTRUCTURE QUALITY, DEPLOYMENT, AND OPERATIONS

Christian Kaestner

Required reading: Eric Breck, Shanqing Cai, Eric Nielsen, Michael Salib, D. Sculley. The ML Test Score: A Rubric for ML Production Readiness and Technical Debt Reduction. Proceedings of IEEE Big Data (2017)

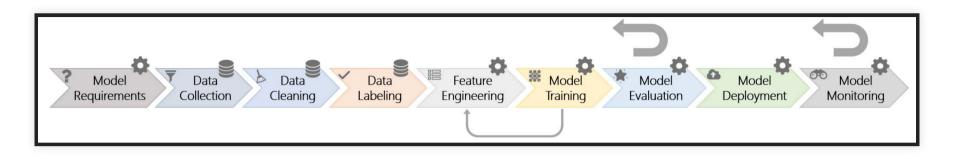
Recommended readings: Larysa Visengeriyeva. Machine Learning Operations - A Reading List, InnoQ 2020

LEARNING GOALS

- Implement and automate tests for all parts of the ML pipeline
- Understand testing opportunities beyond functional correctness
- Automate test execution with continuous integration
- Deploy a service for models using container infrastructure
- Automate common configuration management tasks
- Devise a monitoring strategy and suggest suitable components for implementing it
- Diagnose common operations problems
- Understand the typical concerns and concepts of MLOps

BEYOND MODEL AND DATA QUALITY

POSSIBLE MISTAKES IN ML PIPELINES



Danger of "silent" mistakes in many phases



POSSIBLE MISTAKES IN ML PIPELINES

Danger of "silent" mistakes in many phases:

- Dropped data after format changes
- Failure to push updated model into production
- Incorrect feature extraction
- Use of stale dataset, wrong data source
- Data source no longer available (e.g web API)
- Telemetry server overloaded
- Negative feedback (telemtr.) no longer sent from app
- Use of old model learning code, stale hyperparameter
- Data format changes between ML pipeline steps

EVERYTHING CAN BE TESTED?



Speaker notes

Many qualities can be tested beyond just functional correctness (for a specification). Examples: Performance, model quality, data quality, usability, robustness, ... not all tests are equality easy to automate

TESTING STRATEGIES

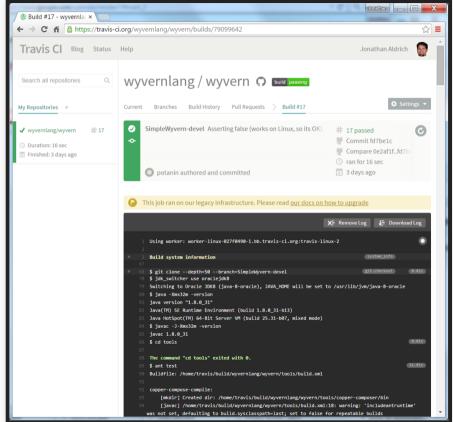
- Performance
- Scalability
- Robustness
- Safety
- Security
- Extensibility
- Maintainability
- Usability

How to test for these? How automatable?

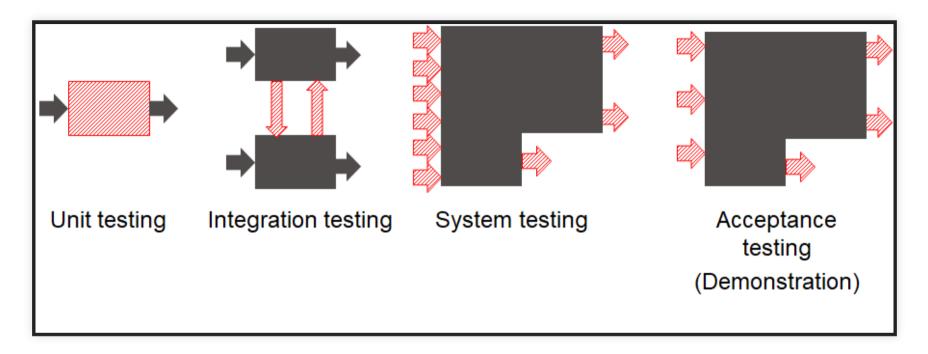
TEST AUTOMATION

FROM MANUAL TESTING TO CONTINUOUS INTEGRATION





UNIT TEST, INTEGRATION TESTS, SYSTEM TESTS



Speaker notes

Software is developed in units that are later assembled. Accordingly we can distinguish different levels of testing.

Unit Testing - A unit is the "smallest" piece of software that a developer creates. It is typically the work of one programmer and is stored in a single file. Different programming languages have different units: In C++ and Java the unit is the class; in C the unit is the function; in less structured languages like Basic and COBOL the unit may be the entire program.

Integration Testing - In integration we assemble units together into subsystems and finally into systems. It is possible for units to function perfectly in isolation but to fail when integrated. For example because they share an area of the computer memory or because the order of invocation of the different methods is not the one anticipated by the different programmers or because there is a mismatch in the data types. Etc.

System Testing - A system consists of all of the software (and possibly hardware, user manuals, training materials, etc.) that make up the product delivered to the customer. System testing focuses on defects that arise at this highest level of integration. Typically system testing includes many types of testing: functionality, usability, security, internationalization and localization, reliability and availability, capacity, performance, backup and recovery, portability, and many more.

Acceptance Testing - Acceptance testing is defined as that testing, which when completed successfully, will result in the customer accepting the software and giving us their money. From the customer's point of view, they would generally like the most exhaustive acceptance testing possible (equivalent to the level of system testing). From the vendor's point of view, we would generally like the minimum level of testing possible that would result in money changing hands. Typical strategic questions that should be addressed before acceptance testing are: Who defines the level of the acceptance testing? Who creates the test scripts? Who executes the tests? What is the pass/fail criteria for the acceptance test? When and how do we get paid?

ANATOMY OF A UNIT TEST

```
import org.junit.Test;
import static org.junit.Assert.assertEquals;
public class AdjacencyListTest {
    @Test
    public void testSanityTest(){
        // set up
        Graph g1 = new AdjacencyListGraph(10);
        Vertex s1 = new Vertex("A");
        Vertex s2 = new Vertex("B");
        // check expected results (oracle)
        assertEquals(true, g1.addVertex(s1));
        assertEquals(true, g1.addVertex(s2));
        assertEquals(true, g1.addEdge(s1, s2));
        assertEquals(s2, g1.getNeighbors(s1)[0]);
```

INGREDIENTS TO A TEST

- Specification
- Controlled environment
- Test inputs (calls and parameters)
- Expected outputs/behavior (oracle)

UNIT TESTING PITFALLS

- Working code, failing tests
- Smoke tests pass
- Works on my (some) machine(s)
- Tests break frequently

How to avoid?

HOW TO UNIT TEST COMPONENT WITH DEPENDENCY ON OTHER CODE?



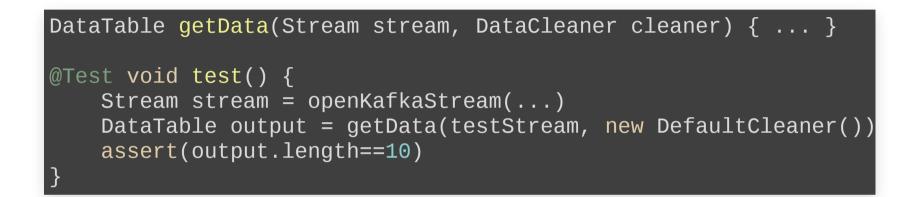
EXAMPLE: TESTING PARTS OF A SYSTEM





EXAMPLE: USING TEST DATA





EXAMPLE: USING TEST DATA



```
DataTable getData(Stream stream, DataCleaner cleaner) { ... }
@Test void test() {
    Stream testStream = new Stream() {
        int idx = 0;
        // hardcoded or read from test file
        <u>String[]</u> data = [ ... ]
        public void connect() { }
        public String getNext() {
            return data[++idx];
    DataTable output = getData(testStream, new DefaultCleaner())
    assert(output.length==10)
```

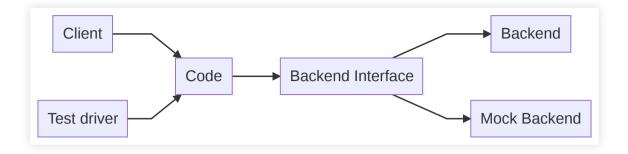
EXAMPLE: MOCKING A DATACLEANER OBJECT

```
DataTable getData(KafkaStream stream, DataCleaner cleaner) { ...
@Test void test() {
    DataCleaner dummyCleaner = new DataCleaner() {
        boolean isValid(String row) { return true; }
        ...
    }
    DataTable output = getData(testStream, dummyCleaner);
    assert(output.length==10)
}
```

EXAMPLE: MOCKING A DATACLEANER OBJECT



Mocking frameworks provide infrastructure for expressing such tests compactly.



TEST ERROR HANDLING

```
@Test void test() {
    DataTable data = new DataTable();
    try {
        Model m = learn(data);
        Assert.fail();
    } catch (NoDataException e) { /* correctly thrown */ }
}
```

Speaker notes

Code to test that the right exception is thrown

TESTING FOR ROBUSTNESS

manipulating the (controlled) environment: injecting errors into backend to test error handling

TEST LOCAL ERROR HANDLING (MODULAR PROTECTION)

```
@Test void test() {
   Stream testStream = new Stream() {
      int idx = 0;
      public void connect() {
        if (++idx < 3)
            throw new IOException("cannot establish connecti
        }
        public String getNext() { ... }
   }
   DataLoader loader = new DataLoader(testStream, new DefaultCl
   ModelBuilder model = new ModelBuilder(loader, ...);
   // assume all exceptions are handled correctly internally
   assert(model.accuracy > .91)
```

Speaker notes

Test that errors are correctly handled within a module and do not leak

Packages

Coverage Report - All Packages

All

 net.sourceforge.cobertura.check

 net.sourceforge.cobertura.coveragedat

 net.sourceforge.cobertura.instrument

 net.sourceforge.cobertura.instrument

 net.sourceforge.cobertura.reporting

 net.sourceforge.cobertura.reporting

 net.sourceforge.cobertura.reporting.htr

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>

All Packages

Classes

AntUtil (88%) Archive (100%) ArchiveUtil (80%) BranchCoverageData (N/A) CheckTask (0%) ClassData (N/A) ClassInstrumenter (94%) ClassPattern (100%) CoberturaFile (73%) CommandLineBuilder (96%) CommonMatchingTask (88%) ComplexityCalculator (100%) ConfigurationUtil (50%) CopyFiles (87%) CoverageData (N/A) CoverageDataContainer (N/A) CoverageDataFileHandler (N/A) CoverageRate (0%) ExcludeClasses (100%) FileFinder (96%) FileLocker (0%) FirstPassMethodInstrumenter (100%) HTMLReport (94%) HasBeenInstrumented (N/A) Header (80%) IOUtil (62%) Ignore (100%) IgnoreBranches (0%)

Package A	# Classes	Line Coverage		Branch Coverage		Complexity
All Packages	55	7596	1625/2179	64%	472/73 <mark>8</mark>	2.319
net.sourceforge.cobertura.ant	11	5296	170 <mark>/</mark> 330	43%	<mark>4</mark> 0/94	1.848
net.sourceforge.cobertura.check	3	096	0/150	0%	0/76	2.429
net.sourceforge.cobertura.coveragedata	13	N/A	N/A	N/A	N/A	2.277
net.sourceforge.cobertura.instrument	10	90%	460/510	75%	123/164	1.854
net.sourceforge.cobertura.merge	1	86%	30/35	88%	14/16	5.5
net.sourceforge.cobertura.reporting	3	87%	116/134	80%	43/54	2.882
net.sourceforge.cobertura.reporting.html	4	91%	475/523	77%	156/202	4.444
net.sourceforge.cobertura.reporting.html.files	1	87%	39/45	62%	5/8	4.5
net.sourceforge.cobertura.reporting.xml	1	100%	155/155	95%	21/22	1.524
net.sourceforge.cobertura.util	9	60%	175/291	69%	70/102	2.892
someotherpackage	1	83%	5/6	N/A	N/A	1.2

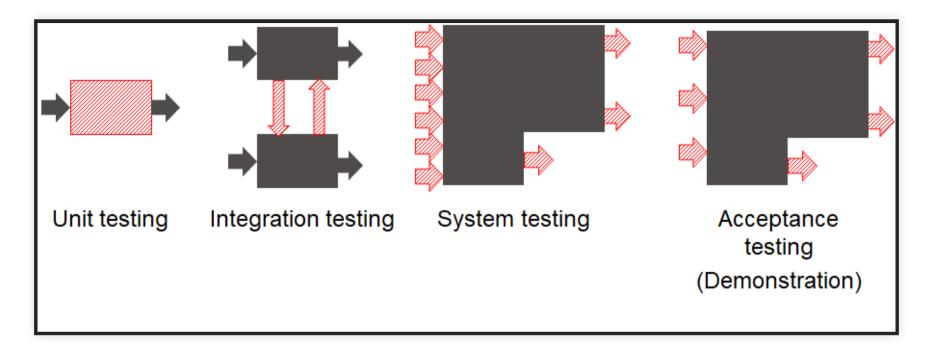
Report generated by Cobertura 1.9 on 6/9/07 12:37 AM.

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TESTABLE CODE

- Think about testing when writing code
- Unit testing encourages you to write testable code
- Separate parts of the code to make them independently testable
- Abstract functionality behind interface, make it replaceable
- Test-Driven Development: A design and development method in which you write tests before you write the code

INTEGRATION AND SYSTEM TESTS



INTEGRATION AND SYSTEM TESTS

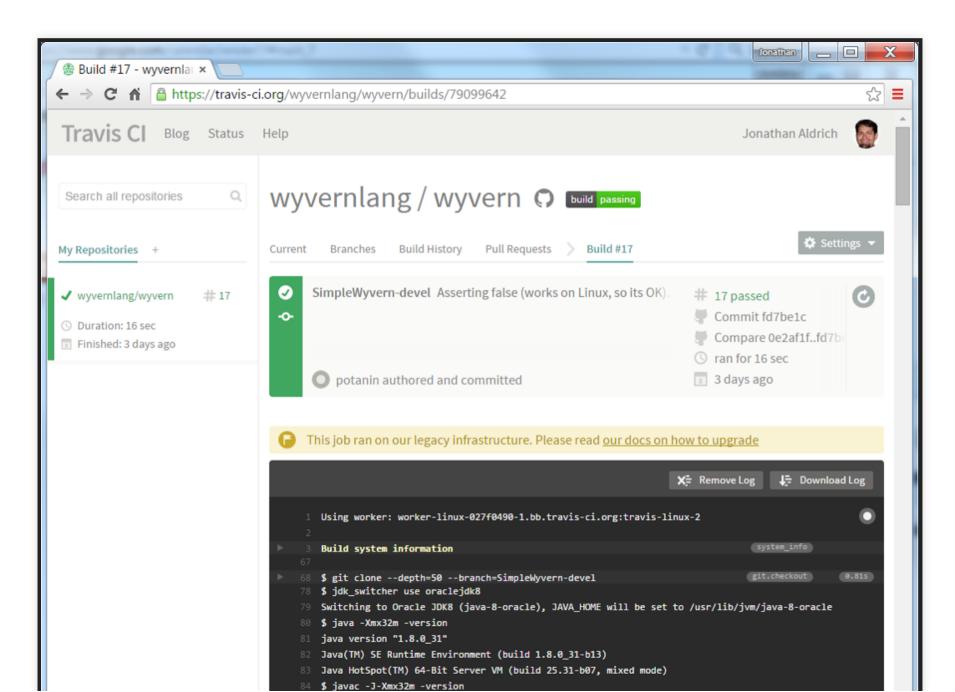
Test larger units of behavior

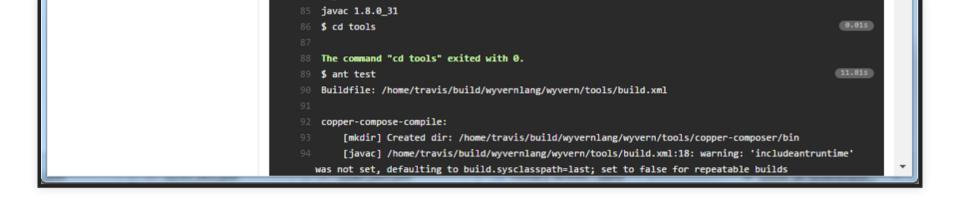
Often based on use cases or user stories -- customer perspective

```
@Test void gameTest() {
    Poker game = new Poker();
    Player p = new Player();
    Player q = new Player();
    game.shuffle(seed)
    game.add(p);
    game.deal();
    p.bet(100);
    q.bet(100);
    p.call();
    q.fold();
    assert(game.winner() == p);
}
```

BUILD SYSTEMS & CONTINUOUS INTEGRATION

- Automate all build, analysis, test, and deployment steps from a command line call
- Ensure all dependencies and configurations are defined
- Ideally reproducible and incremental
- Distribute work for large jobs
- Track results
- Key CI benefit: Tests are regularly executed, part of process





TRACKING BUILD QUALITY

Track quality indicators over time, e.g.,

- Build time
- Test coverage
- Static analysis warnings
- Performance results
- Model quality measures
- Number of TODOs in source code

Jenkins search $\textcircled{\blue}{\blue}$ Stop-tabac dev ENABLE AUTO REFRESH Jenkins Suisse Back to Dashboard **Project Stop-tabac dev** Status CI build Changes 0 Zedit description Disable Project Workspace Test Result Trend Build Now 140 Coverage Report 120 N Delete Project 100 Configure count Workspace 80 Set Next Build Number 60 Duplicate Code 40 Recent Changes Coverage Report 20 0 SLOCCount Latest Test Result (no failures) 06 <u></u>#171 £210 263 #345 1426 #438 52 Git Polling Log (just show failures) enlarge Code Coverage Permalinks 9 Build History (trend) Classes 45% Conditionals 74% Files 45% Lines 28% Packages 88% 100 #977 Aug 27, 2012 4:37:27 PM Last build (#977), 3 min 17 sec ago Last stable build (#977), 3 min 17 sec ago 90 #438 Jun 28, 2012 8:47:42 AM Last successful build (#977), 3 min 17 sec ago 80 #426 Jun 26, 2012 1:39:39 PM 70 Classes 60 #345 Jun 19, 2012 9:02:20 AM Conditionals ≫ 50 Files #263 Jun 6, 2012 9:14:42 PM 40 30 Lines #210 May 31, 2012 8:42:29 AM 20 Packages #171 May 23, 2012 9:58:18 PM 10 0 #90 May 15, 2012 11:49:41 AM £210 #426 #171 £263 F345 #438 £977 RSS for all RSS for failures SLOCCount Trend 8 000 6 0 0 0 Sa 4 000 2 000 0 #210 ±426 #171 £263 #345 #438 #977

📕 objc

Source: https://blog.octo.com/en/jenkins-quality-dashboard-ios-development/

TEST MONITORING

- Inject/simulate faulty behavior
- Mock out notification service used by monitoring
- Assert notification

```
class MyNotificationService extends NotificationService {
    public boolean receivedNotification = false;
    public void sendNotification(String msg) { receivedNotificat
}
@Test void test() {
    Server s = getServer();
    MyNotificationService n = new MyNotificationService();
    Monitor m = new Monitor(s, n);
    s.stop();
    s.request();
    s.request();
    wait();
    assert(n.receivedNotification);
}
```

TEST MONITORING IN PRODUCTION

- Like fire drills (manual tests may be okay!)
- Manual tests in production, repeat regularly
- Actually take down service or trigger wrong signal to monitor

CHAOS TESTING



http://principlesofchaos.org

Speaker notes

Chaos Engineering is the discipline of experimenting on a distributed system in order to build confidence in the system's capability to withstand turbulent conditions in production. Pioneered at Netflix

CHAOS TESTING ARGUMENT

- Distributed systems are simply too complex to comprehensively predict
- -> experiment on our systems to learn how they will behave in the presence of faults
- Base corrective actions on experimental results because they reflect real risks and actual events
- Experimentation != testing -- Observe behavior rather then expect specific results
- Simulate real-world problem in production (e.g., take down server, inject latency)
- *Minimize blast radius:* Contain experiment scope

NETFLIX'S SIMIAN ARMY

Chaos Monkey: randomly disable production instances

Latency Monkey: induces artificial delays in our RESTful client-server communication layer Conformity Monkey: finds instances that don't adhere to best-practices and shuts them down Doctor Monkey: monitors other external signs of health to detect unhealthy instances Janitor Monkey: ensures that our cloud environment is running free of clutter and waste Security Monkey: finds security violations or vulnerabilities, and terminates the offending instances

10–18 Monkey: detects problems in instances serving customers in multiple geographic regions

Chaos Gorilla is similar to Chaos Monkey, but simulates an outage of an entire Amazon availability zone.

CHAOS TOOLKIT

- Infrastructure for chaos experiments
- Driver for various infrastructure and failure cases
- Domain specific language for experiment definitions

```
{
   "version": "1.0.0",
    "title": "What is the impact of an expired certificate on ou
    "description": "If a certificate expires, we should graceful
    "tags": ["tls"],
    "steady-state-hypothesis": {
        "title": "Application responds",
        "probes": [
                "type": "probe",
                "name": "the-astre-service-must-be-running",
                "tolerance": true,
                "provider": {
                    "type": "python",
                    "module": "os.path",
                    "func" "oviete"
```

http://principlesofchaos.org, https://github.com/chaostoolkit, https://github.com/Netflix/SimianArmy

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CHAOS EXPERIMENTS FOR ML INFRASTRUCTURE?



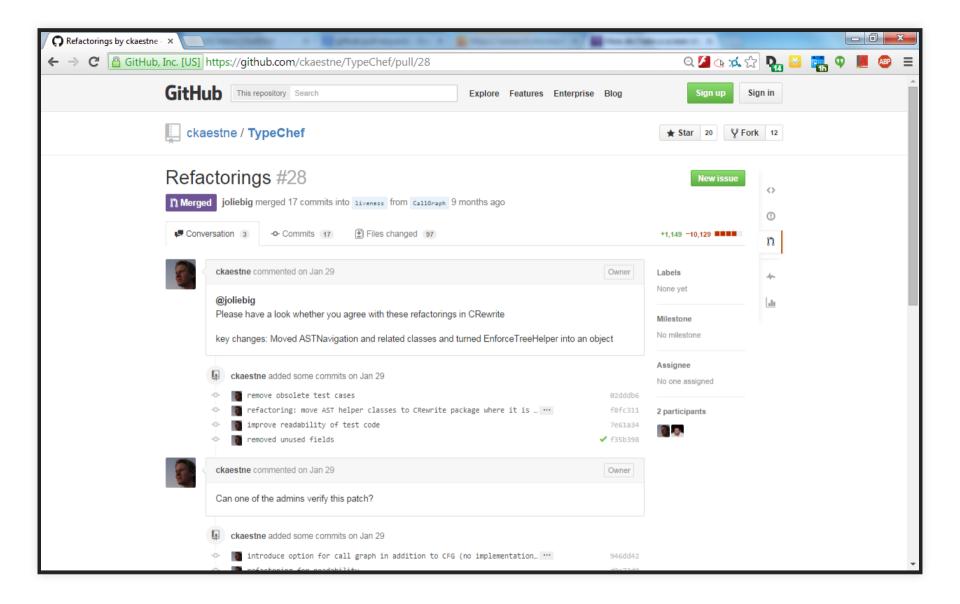
Speaker notes

Fault injection in production for testing in production. Requires monitoring and explicit experiments.

CODE REVIEW AND STATIC ANALYSIS

CODE REVIEW

- Manual inspection of code
 - Looking for problems and possible improvements
 - Possibly following checklists
 - Individually or as group
- Modern code review: Incremental review at checking
 - Review individual changes before merging
 - Pull requests on GitHub
 - Not very effective at finding bugs, but many other benefits: knowledge transfer, code imporvement, shared code ownership, improving testing



STATIC ANALYSIS, CODE LINTING

Automatic detection of problematic patterns based on code structure



PrintWriter log = null; if (anyLogging) log = new PrintWriter(...); if (detailedLogging) log.println("Log started");

PROCESS INTEGRATION: STATIC ANALYSIS WARNINGS DURING CODE REVIEW



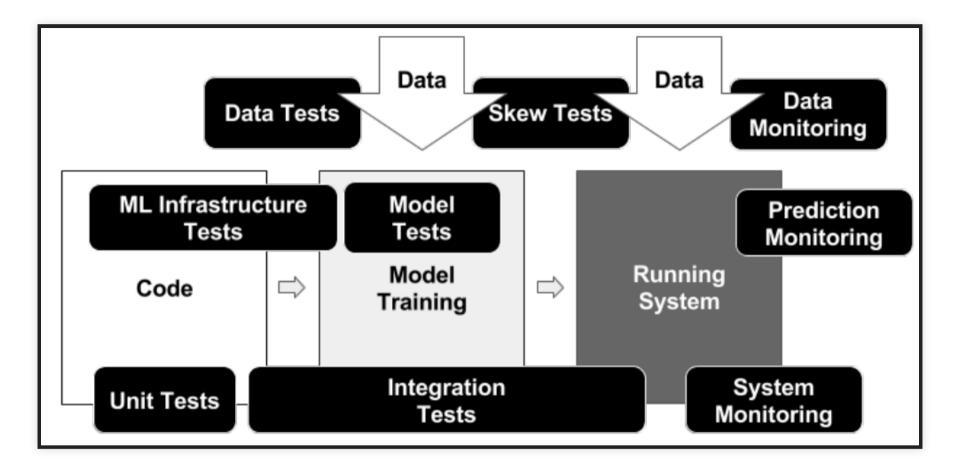
Sadowski, Caitlin, Edward Aftandilian, Alex Eagle, Liam Miller-Cushon, and Ciera Jaspan. "Lessons from building static analysis tools at google." Communications of the ACM 61, no. 4 (2018): 58-66.

5.5

Social engineering to force developers to pay attention. Also possible with integration in pull requests on GitHub.

INFRASTRUCTURE TESTING





CASE STUDY: SMART PHONE COVID-19 DETECTION



(from midterm; assume cloud or hybrid deployment)

DATA TESTS

- 1. Feature expectations are captured in a schema.
- 2. All features are beneficial.
- 3. No feature's cost is too much.
- 4. Features adhere to meta-level requirements.
- 5. The data pipeline has appropriate privacy controls.
- 6. New features can be added quickly.
- 7. All input feature code is tested.

TESTS FOR MODEL DEVELOPMENT

- 1. Model specs are reviewed and submitted.
- 2. Offline and online metrics correlate.
- 3. All hyperparameters have been tuned.
- 4. The impact of model staleness is known.
- 5. A simpler model is not better.
- 6. Model quality is sufficient on important data slices.
- 7. The model is tested for considerations of inclusion.

ML INFRASTRUCTURE TESTS

- 1. Training is reproducible.
- 2. Model specs are unit tested.
- 3. The ML pipeline is Integration tested.
- 4. Model quality is validated before serving.
- 5. The model is debuggable.
- 6. Models are canaried before serving.
- 7. Serving models can be rolled back.

MONITORING TESTS

- 1. Dependency changes result in notification.
- 2. Data invariants hold for inputs.
- 3. Training and serving are not skewed.
- 4. Models are not too stale.
- 5. Models are numerically stable.
- 6. Computing performance has not regressed.
- 7. Prediction quality has not regressed.

BREAKOUT GROUPS

- Discuss in groups:
 - Team 1 picks the data tests
 - Team 2 the model dev. tests
 - Team 3 the infrastructure tests
 - Team 4 the monitoring tests
- For 15 min, discuss each listed point in the context of the Covid-detection scenario: what would you do?
- Report back to the class

DEV VS. OPS



COMMON RELEASE PROBLEMS?



COMMON RELEASE PROBLEMS (EXAMPLES)

- Missing dependencies
- Different compiler versions or library versions
- Different local utilities (e.g. unix grep vs mac grep)
- Database problems
- OS differences
- Too slow in real settings
- Difficult to roll back changes
- Source from many different repositories
- Obscure hardware? Cloud? Enough memory?

DEVELOPERS

Coding

. . .

- Testing, static analysis, reviews
- Continuous integration
- Bug tracking
- Running local tests and scalability experiments

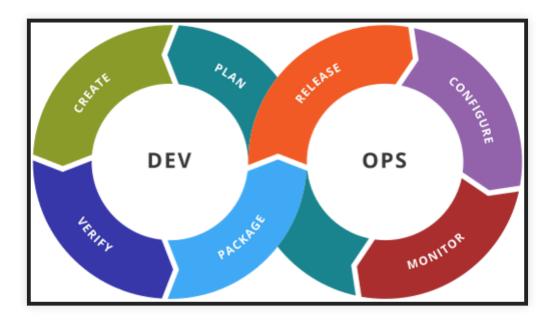
- **OPERATIONS**
- Allocating hardware resources
- Managing OS updates
- Monitoring performance
- Monitoring crashes
- Managing load spikes, ...
- Tuning database performance
- Running distributed at scale
- Rolling back releases
- ...

QA responsibilities in both roles

QUALITY ASSURANCE DOES NOT STOP IN DEV

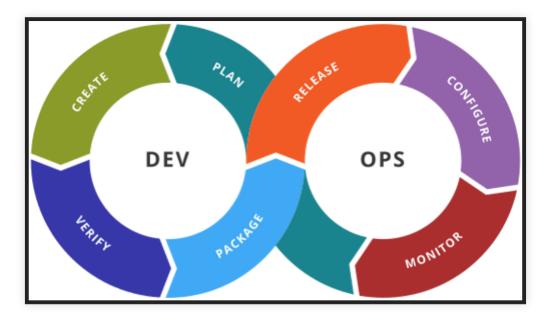
- Ensuring product builds correctly (e.g., reproducible builds)
- Ensuring scalability under real-world loads
- Supporting environment constraints from real systems (hardware, software, OS)
- Efficiency with given infrastructure
- Monitoring (server, database, Dr. Watson, etc)
- Bottlenecks, crash-prone components, ... (possibly thousands of crash reports per day/minute)

DEVOPS



KEY IDEAS AND PRINCIPLES

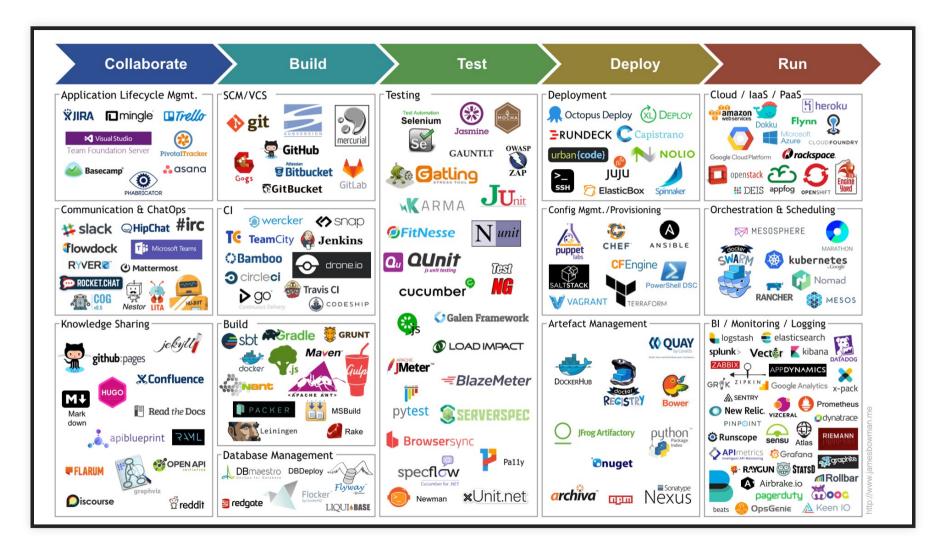
- Better coordinate between developers and operations (collaborative)
- Key goal: Reduce friction bringing changes from development into production
- Considering the *entire tool chain* into production (holistic)
- Documentation and versioning of all dependencies and configurations ("configuration as code")
- Heavy automation, e.g., continuous delivery, monitoring
- Small iterations, incremental and continuous releases
- Buzz word!



COMMON PRACTICES

- All configurations in version control
- Test and deploy in containers
- Automated testing, testing, testing, ...
- Monitoring, orchestration, and automated actions in practice
- Microservice architectures
- Release frequently

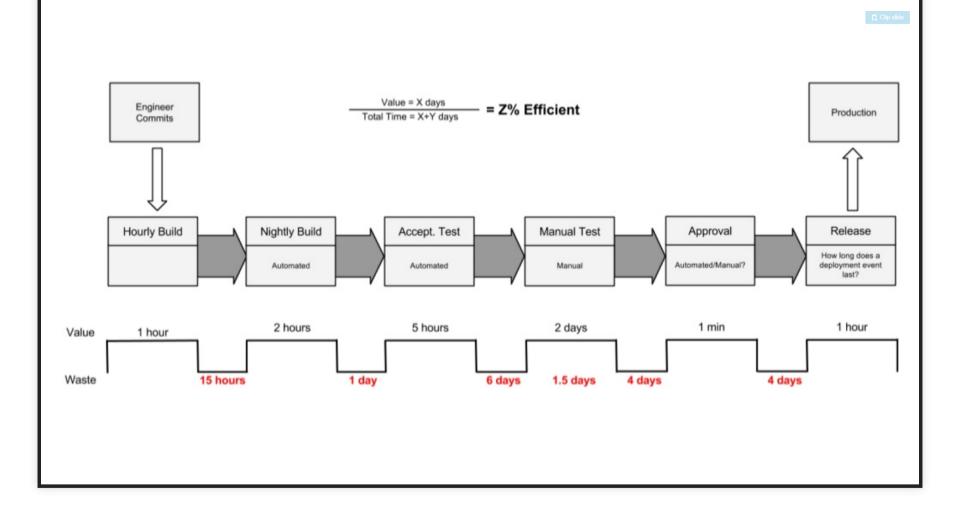
HEAVY TOOLING AND AUTOMATION



HEAVY TOOLING AND AUTOMATION -- EXAMPLES

- Infrastructure as code Ansible, Terraform, Puppet, Chef
- CI/CD Jenkins, TeamCity, GitLab, Shippable, Bamboo, Azure DevOps
- Test automation Selenium, Cucumber, Apache JMeter
- Containerization Docker, Rocket, Unik
- Orchestration Kubernetes, Swarm, Mesos
- Software deployment Elastic Beanstalk, Octopus, Vamp
- Measurement Datadog, DynaTrace, Kibana, NewRelic, ServiceNow

CONTINUOUS DELIVERY



Source: https://www.slideshare.net/jmcgarr/continuous-delivery-at-netflix-andbeyond

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TYPICAL MANUAL STEPS IN DEPLOYMENT?

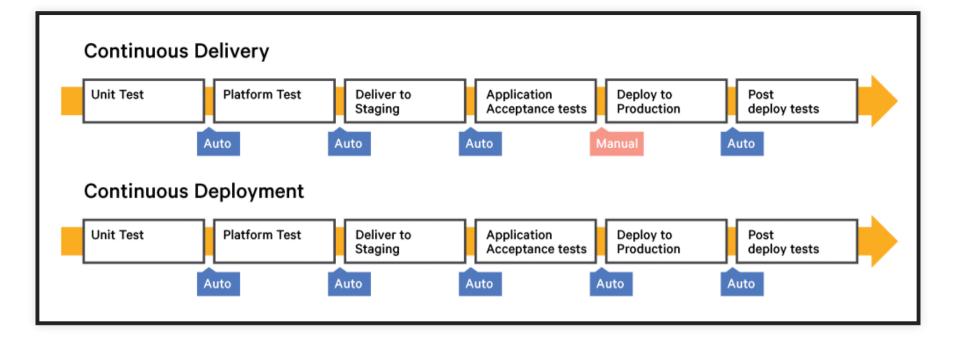


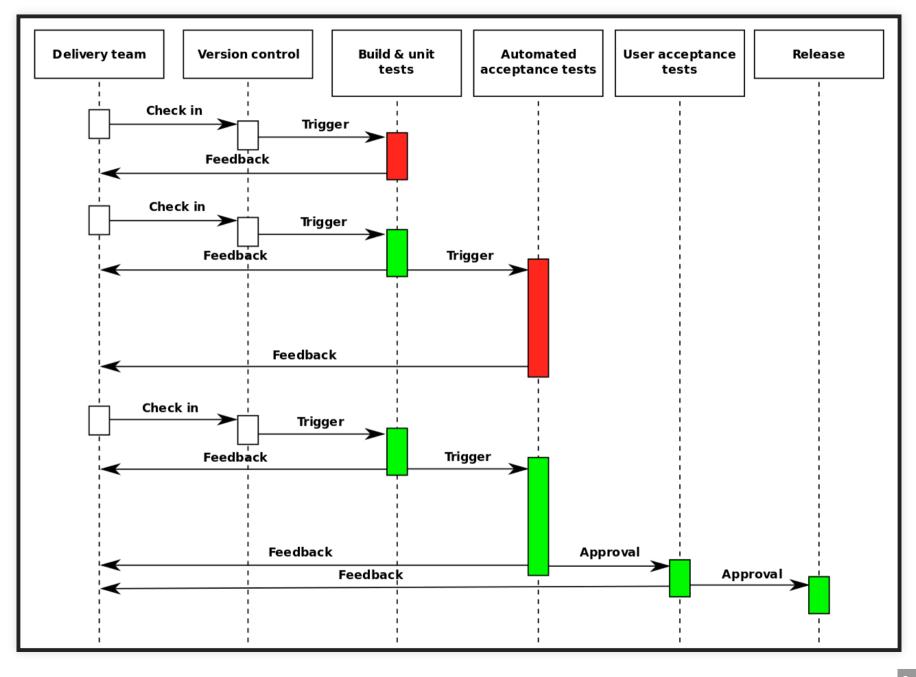
CONTINUOUS DELIVERY

- Full automation from commit to deployable container
- Heavy focus on testing, reproducibility and rapid feedback
- Deployment step itself is manual
- Makes process transparent to all developers and operators

CONTINUOUS DEPLOYMENT

- Full automation from commit to deployment
- Empower developers, quick to production
- Encourage experimentation and fast incremental changes
- Commonly integrated with monitoring and canary releases





FACEBOOK TESTS FOR MOBILE APPS

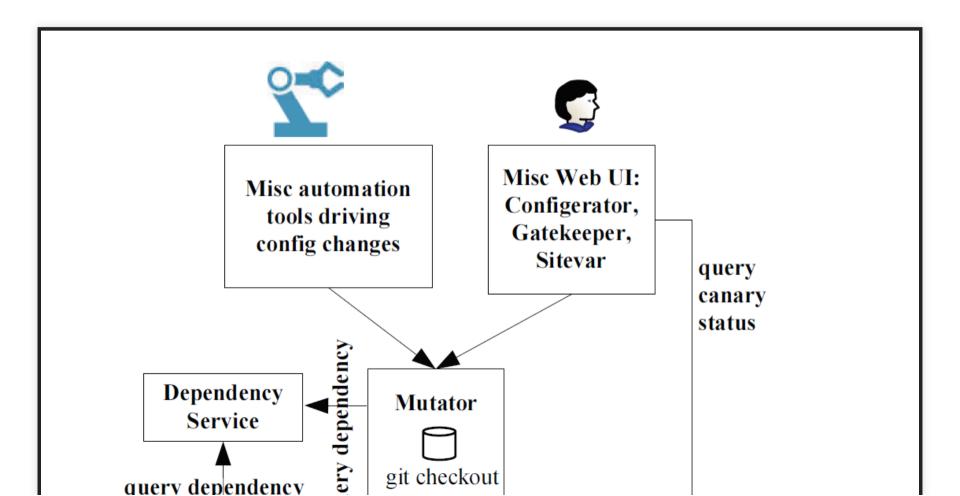
- Unit tests (white box)
- Static analysis (null pointer warnings, memory leaks, ...)
- Build tests (compilation succeeds)
- Snapshot tests (screenshot comparison, pixel by pixel)
- Integration tests (black box, in simulators)
- Performance tests (resource usage)
- Capacity and conformance tests (custom)

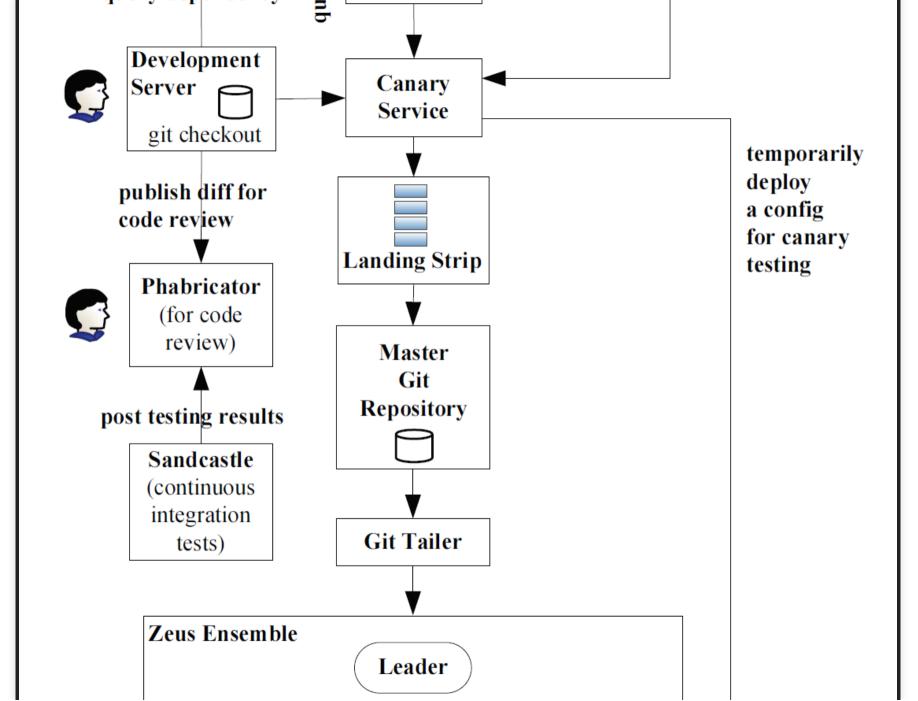
Further readings: Rossi, Chuck, Elisa Shibley, Shi Su, Kent Beck, Tony Savor, and Michael Stumm. Continuous deployment of mobile software at facebook (showcase). In Proceedings of the 2016 24th ACM SIGSOFT International Symposium on Foundations of Software Engineering, pp. 12-23. ACM, 2016.

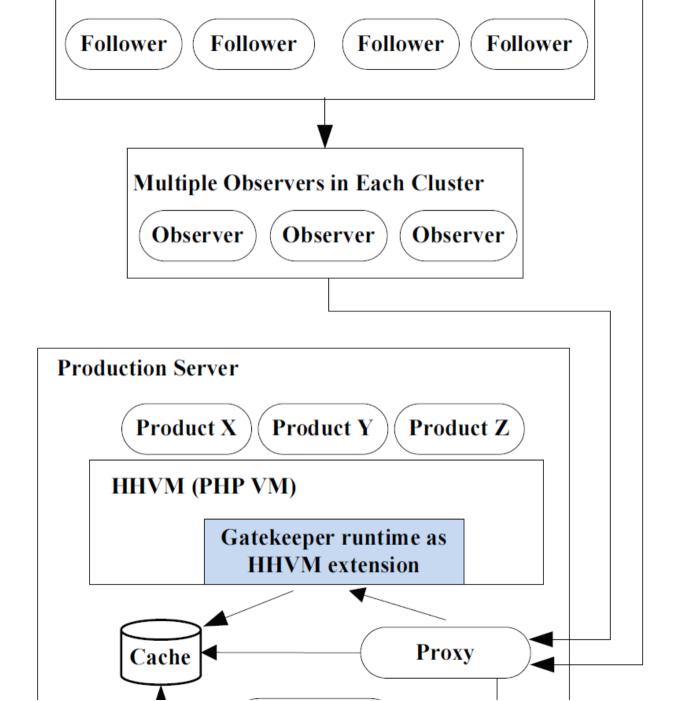
RELEASE CHALLENGES FOR MOBILE APPS

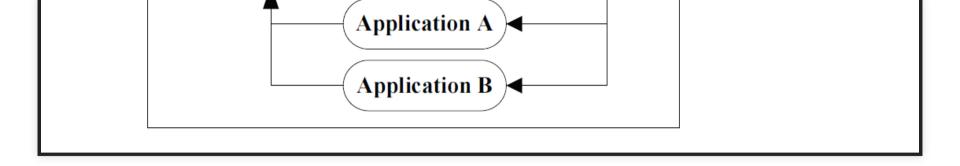
- Large downloads
- Download time at user discretion
- Different versions in production
- Pull support for old releases?
- Server side releases silent and quick, consistent
- -> App as container, most content + layout from server

REAL-WORLD PIPELINES ARE COMPLEX





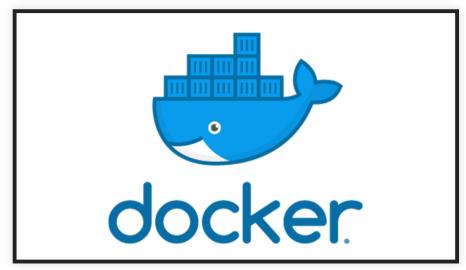




CONTAINERS AND CONFIGURATION MANAGEMENT

CONTAINERS

- Lightweight virtual machine
- Contains entire runnable software, incl. all dependencies and configurations
- Used in development and production
- Sub-second launch time
- Explicit control over shared disks and network connections



DOCKER EXAMPLE

```
FROM ubuntu:latest
MAINTAINER ...
RUN apt-get update -y
RUN apt-get install -y python-pip python-dev build-essential
COPY . /app
WORKDIR /app
RUN pip install -r requirements.txt
ENTRYPOINT ["python"]
CMD ["app.py"]
```

Source: http://containertutorials.com/docker-compose/flask-simple-app.html

COMMON CONFIGURATION MANAGEMENT QUESTIONS

- What runs where?
- How are machines connected?
- What (environment) parameters does software X require?
- How to update dependency X everywhere?
- How to scale service X?

ANSIBLE EXAMPLES

- Software provisioning, configuration management, and applicationdeployment tool
- Apply scripts to many servers

[webservers]

web1.company.org
web2.company.org
web3.company.org

[dbservers]

db1.company.org db2.company.org

[replication_server

This role deploys the mongod processes and - name: create data directory for mongodb file: path={{ mongodb_datadir_prefix }}/mon delegate_to: '{{ item }}' with_items: groups.replication_servers

- name: create log directory for mongodb
file: path=/var/log/mongo state=directory o

- name: Create the mongodb startup file template: src=mongod.j2 dest=/etc/init.d/mo delegate_to: '{{ item }}' with_items: groups.replication_servers

name Create the monordh configuration file

PUPPET EXAMPLE

Declarative specification, can be applied to many machines

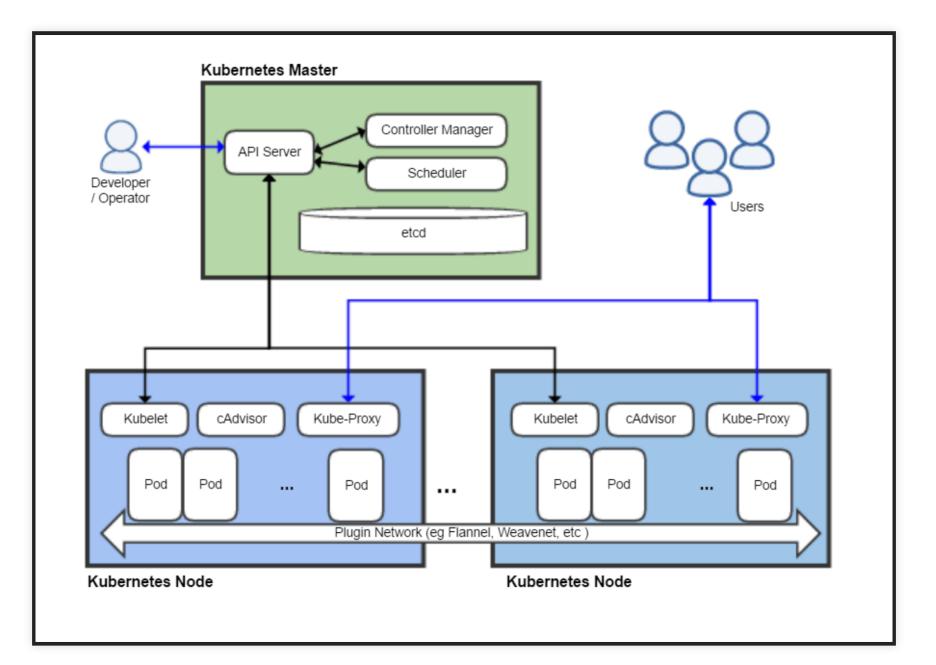
```
$doc_root = "/var/www/example"
exec { 'apt-get update':
 command => '/usr/bin/apt-get update'
package { 'apache2':
 ensure => "installed",
 require => Exec['apt-get update']
file { $doc_root:
 ensure => "directory",
 owner => "www-data",
 group => "www-data",
 mode => 644
```

Speaker notes

source: https://www.digitalocean.com/community/tutorials/configuration-management-101-writing-puppet-manifests

CONTAINER ORCHESTRATION WITH KUBERNETES

- Manages which container to deploy to which machine
- Launches and kills containers depending on load
- Manage updates and routing
- Automated restart, replacement, replication, scaling
- Kubernetis master controls many nodes

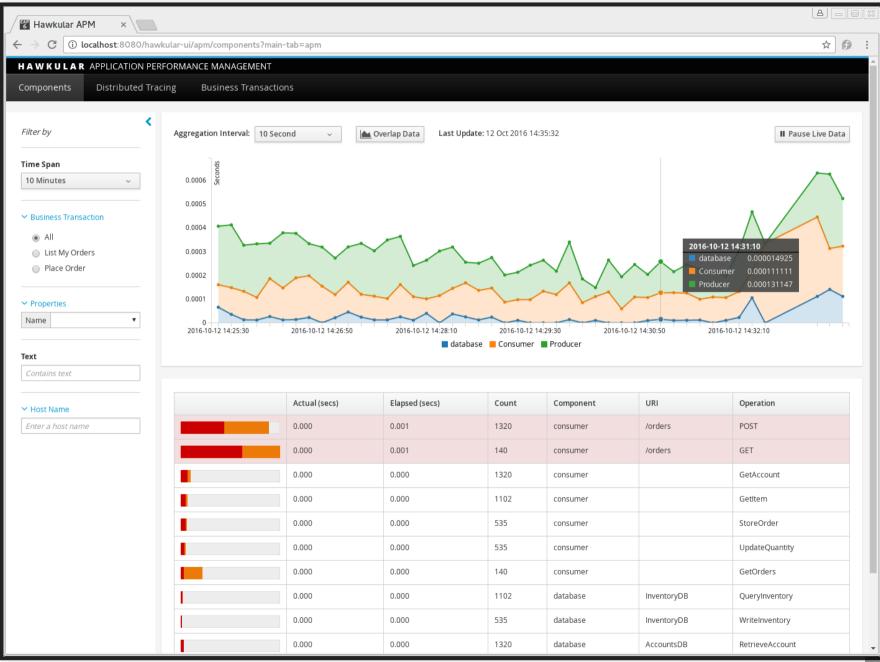


CC BY-SA 4.0 Khtan66

MONITORING

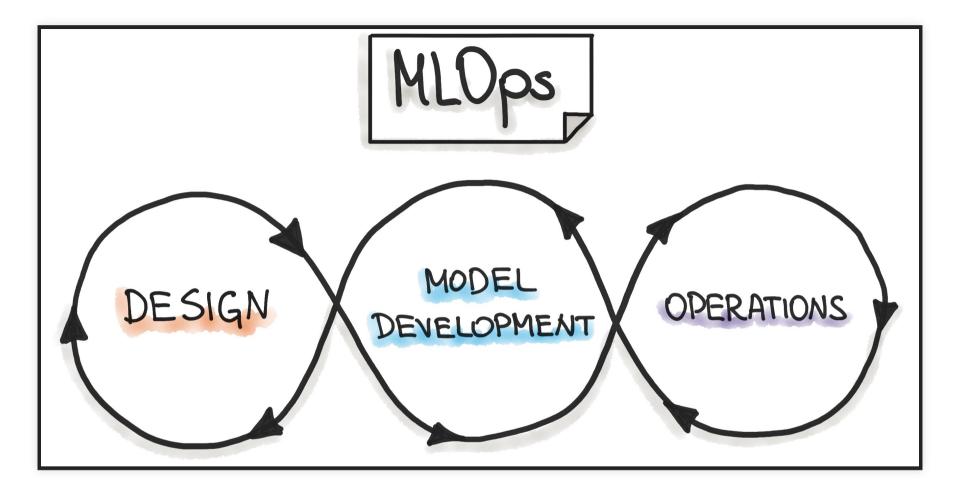
- Monitor server health
- Monitor service health
- Collect and analyze measures or log files
- Dashboards and triggering automated decisions
- Many tools, e.g., Grafana as dashboard, Prometheus for metrics, Loki + ElasticSearch for logs
- Push and pull models

HAWKULAR



HAWKULAR

Hawkular APM ×		8 - 0 ×
		☆ 🕖 :
HAWKULAR APPLICATION PERFORMANCE MANAGEMENT		
Components Distributed T	racing Business Transactions	
Filter by	Initial Endpoint OrderManager: /orders[POST] ~ Show 2 Instance(s) Details	• Reset Zoom
Time Span 1 Hour Business Transaction Image: Image	InventoryManager [UpdateQuantity] 0 0116m OrderManager [0 0.07ams OrderLog [0 0.12ams OrderLog [0 0.12ams OrderLog [Cettern] OrderLog [Cettern] OrderLog (Cettern) OrderLog (Cettern) OrderLog (Cettern) OrderLog Ord	



https://ml-ops.org/

ON TERMINOLOGY

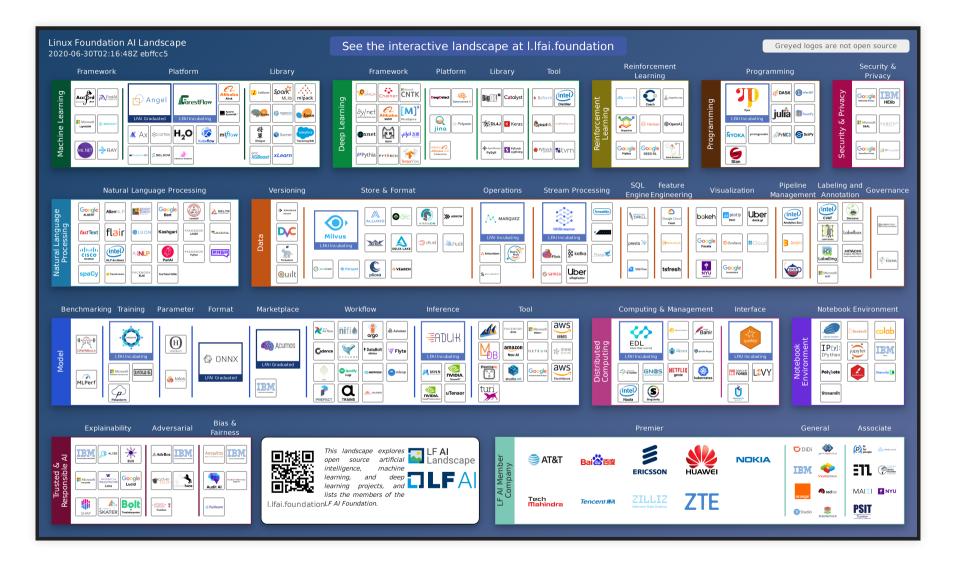
- Many vague buzzwords, often not clearly defined
- *MLOps*: Collaboration and communication between data scientists and operators, e.g.,
 - Automate model deployment
 - Model training and versioning infrastructure
 - Model deployment and monitoring
- *AlOps*: Using Al/ML to make operations decision, e.g. in a data center
- DataOps: Data analytics, often business setting and reporting
 - Infrastructure to collect data (ETL) and support reporting
 - Monitor data analytics pipelines
 - Combines agile, DevOps, Lean Manufacturing ideas

MLOPS OVERVIEW

- Integrate ML artifacts into software release process, unify process
- Automated data and model validation (continuous deployment)
- Data engineering, data programming
- Continuous deployment for ML models
 - From experimenting in notebooks to quick feedback in production
- Versioning of models and datasets
- Monitoring in production

Further reading: MLOps principles

TOOLING LANDSCAPE LF AI



Linux Foundation AI Initiative

11.4

SUMMARY

- Beyond model and data quality: Quality of the infrastructure matters, danger of silent mistakes
- Many SE techniques for test automation, testing robustness, test adequacy, testing in production useful for infrastructure quality
- Lack of modularity: local improvements may not lead to global improvements
- DevOps: Development vs Operations challenges
 - Automated configuration
 - Telemetry and monitoring are key
 - Many, many tools
- MLOps: Automation around ML pipelines, incl. training, evaluation, versioning, and deployment

FURTHER READINGS

- Eric Breck, Shanqing Cai, Eric Nielsen, Michael Salib, D. Sculley. The ML Test Score: A Rubric for ML Production Readiness and Technical Debt Reduction. Proceedings of IEEE Big Data (2017)
- Zinkevich, Martin. Rules of Machine Learning: Best Practices for ML Engineering. Google Blog Post, 2017
- Serban, Alex, Koen van der Blom, Holger Hoos, and Joost Visser. "Adoption and Effects of Software Engineering Best Practices in Machine Learning." In Proc. ACM/IEEE International Symposium on Empirical Software Engineering and Measurement (2020).
- O'Leary, Katie, and Makoto Uchida. "Common problems with Creating Machine Learning Pipelines from Existing Code." Proc. Third Conference on Machine Learning and Systems (MLSys) (2020).
- Earysa Visengeriyeva. Machine Learning Operations A Reading List, InnoQ 2020

