# DevOps Workshop

PRESENTED BY -Aayush Agrawal



### **Aayush Agrawal**

#### **Education -**

- MS in Business Analytics, Carlson School of Management, University of Minnesota, 2017
- B.Tech in Electrical Engineering, Malaviya National Institute of Technology, India, 2013



### Experience -

- >5 years in Data science, Currently working as a Data scientist at Land O' Lakes, Inc. LAND O'LAKES, INC.
- Moderator and rank 3<sup>rd</sup> at <a href="https://www.analyticsvidhya.com/">https://www.analyticsvidhya.com/</a>
- Kaggle Expert <a href="https://www.kaggle.com/aayushmnit">https://www.kaggle.com/aayushmnit</a>

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## Agenda

- Model Building using cookbooks
- Version control by GIT

Deploying models as API using Flask

- Packaging applications using Docker
- Deploying application in cloud

### DevOps at a glance

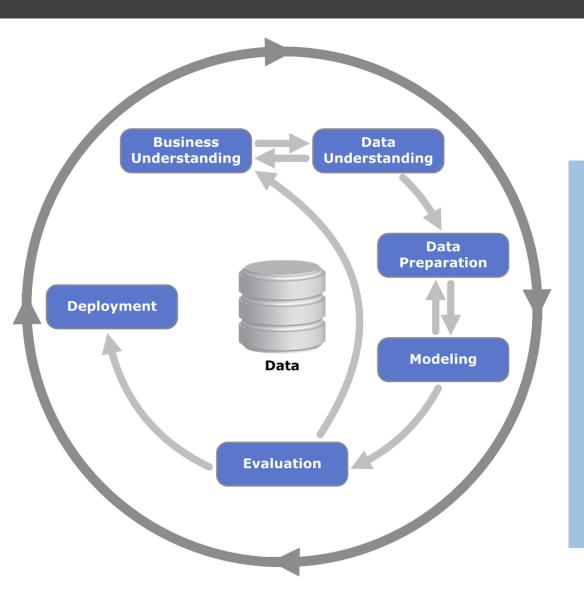
### **DevOps-**

- Continuous improvement(Test and learn)
- Improve the software delivery process (stakeholder management)
  - Reduce rework
  - Reduce overhead by automation

### DevOps in data science -

- Neat, consistent, reliable & structured code
- Deploy model with API
- Use containers
- Use orchestration tools
- Monitor all level of stacks

### CRISP DM – Best methodology for Data science



### **Business Understanding**

Translation of project objectives and requirements from a business perspective to data Science problem

### **Data Understanding**

Data collection, quality checks, exploration, hypothesis testing

#### **Data preparation**

Process of converting raw data to model ready data

### Modeling

Building advance analytics/ machine learning model

#### **Evaluation**

Model quality assessment and results interpretation

### **Deployment**

Automation of data and modeling pipelines

### Use of Cookbooks in Data science

- Many of the data science task are quite repeatable and highly generalizable
- 60% of the time writing code is wasted on functions/program which a data scientist have already written

### Reusability

- Less syntax to remember
- DRY (Don't repeat yourself)

#### Modularity

- Cleaner and well written code
- Reliable performance by multiple re-use in different scenario

#### Faster execution

- Less code to write (10Xer)
- Consistent output to analyze

#### Customizable

- Appending/Removing functionality
- Customizable documentation

### My cookbooks overview

Link to git repo - <a href="https://github.com/aayushmnit/cookbook">https://github.com/aayushmnit/cookbook</a>

Generic Preprocessing - Helper functions to do EDA, missing value analysis & treatment and generic preprocessing like (scaling, encoding etc.)

<u>Machine learning - Classification</u> - Helper functions to solve classification type problems in machine learning. It contains codes for holdout/cross validation, model explanation codes (LIME and variable importance plot), and codes for general classification algorithms (Xgboost, LightGBM, Extra trees, random forest, logistic regression, decision trees, K-nearest neighbors and SVM)

<u>Machine learning - Regression</u> - Helper functions to solve regression type problems in machine learning. It contains codes for holdout/cross validation, model explanation codes (LIME and variable importance plot), and codes for general regression algorithms (Xgboost, LightGBM, Extra trees, random forest, linear regression, regression trees, K-nearest neighbors and SVM)

<u>Recommender systems</u> - Helper functions to build recommender systems using Matrix factorization using LightFM package.

<u>Natural language processing</u> - Helper functions for NLP text processing and analysis like N-grams, word cloud, tokenization, lowercasing, punctuation/stopwords removal, stemmer/lemmatizer, TF-IDF & count vectorizer

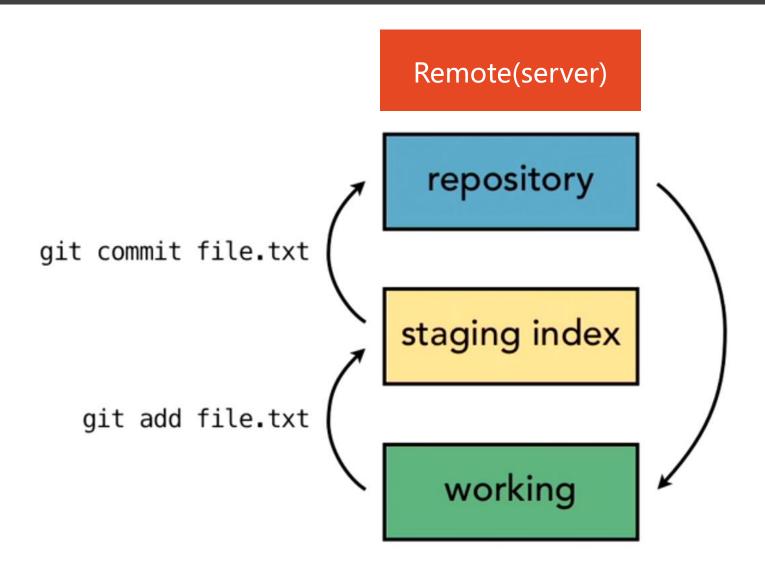
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### Who should use Git?

- Anyone wanting to track edits
  - Review a history log of changes made
  - View differences between versions
  - Retrieve old versions
- Anyone needing to share changes with collaborators
- Anyone not afraid of command line tools

### Git follows a three tree architecture



https://www.linkedin.com/learning/git-essential-training/exploring-the-three-trees-architecture

### **Useful Links**

- Git essentials training <a href="https://www.linkedin.com/learning/git-essential-training">https://www.linkedin.com/learning/git-essential-training</a>
- Git cheat sheet

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## Flask is a micro service web platform framework

Many machine learning models are used real time at the heart of any data product.

Options to implement Machine Learning models –

### Option 1: Rewriting the whole code in the same language as product

Very hard and time consuming process

### Option 2: API-first approach

- Web APIs have made it easy for cross-language applications to work well
- Software product can just call the API with bunch of parameters to get result regardless of backend language







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### Docker is a containerization platform

Docker is a platform for developers and sysadmins to develop, deploy, and run applications with containers.

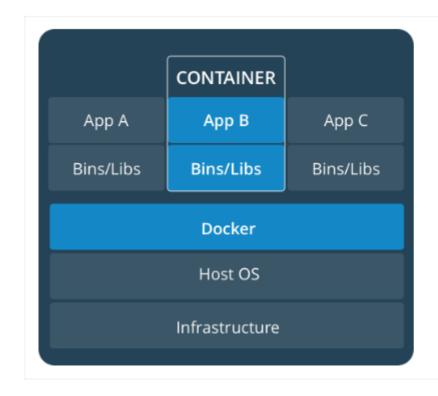
Containerization is increasingly popular because containers are:

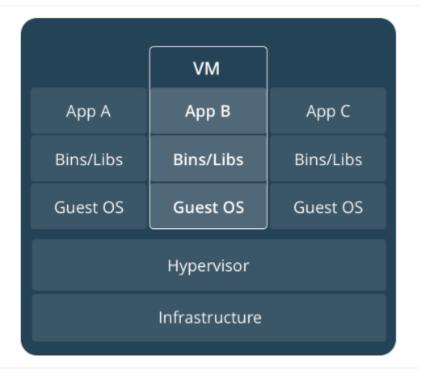
- Flexible: Even the most complex applications can be containerized.
- Lightweight: Containers leverage and share the host kernel.
- Interchangeable: You can deploy updates and upgrades on-the-fly.
- **Portable:** You can build locally, deploy to the cloud, and run anywhere.
- Scalable: You can increase and automatically distribute container replicas.
- Stackable: You can stack services vertically and on-the-fly.

**Image** - An image is an executable package that includes everything needed to run an application-the code, a runtime, libraries, environment variables, and configuration files

**Container** - A container is a runtime instance of an image--what the image becomes in memory when executed

### Container is light weight because it doesn't require an OS





### Creating your own docker file

https://docs.docker.com/get-started/part2/#dockerfile

```
# Use an official Python runtime as a parent image
FROM python: 2.7-slim
# Set the working directory to /app
WORKDIR /app
# Copy the current directory contents into the container at /app
ADD . /app
# Install any needed packages specified in requirements.txt
RUN pip install --trusted-host pypi.python.org -r requirements.txt
# Make port 80 available to the world outside this container
EXPOSE 80
# Define environment variable
ENV NAME World
# Run app.py when the container launches
CMD ["python", "app.py"]
```

### Basic Docker commands

#### Basic commands -

- List docker images: "docker images"
- To run a docker image: "docker run --<option> <image\_name>
- Delete a docker image: "docker rmi <image ID>"
- List all the containers: "docker container Is -la"
- Stop container from running: "docker container stop <container ID>"
- Stop container from running: "docker rm <container ID>"

#### To build a docker image –

- cd to the folder DOCKERFILE
- Run the command "docker build -t <image\_name>"

### To push an image to DockerHub -

- Create a repo on docker hub
- Login using DockerHub credentials: "docker login --username=<user\_id>"
- Run command "docker tag <image\_id> <user\_name>/<repo\_name>:<tag>"
- Run command "docker push <user\_name>/<repo\_name>"

### Scaling with Kubernetes

• <u>Kubernetes</u> is an open-source system for automating deployment, scaling, and management of containerized applications.

#### Kubernetes can perform simple task –

- Starting and stopping containers
- Determine where to run containers
- Checking health of containers
- Restarting or replacing unhealthy containers

#### Master node runs -

- Kube-apiserver
- Etcd(Key-value) for cluster data
- Kube-scheduler for deciding where to run a container
- Kube-controller-manager controls replication processes and endpoints
- Cloud controller manager which interacts with cloud services

#### Minion node runs –

- Kublet to make sure containers run inside a pod
- Kube-proxy to manage network operations
- Container-runtime manages execution of containers







### Some good resources on topic discussed

### Git essential training- Lynda

#### Docker -

- Docker official tutorial Link
- How Docker Can Help You Become A More Effective Data Scientist <u>Hamel Husain</u>
- A Step Towards Reproducible Data Science: Docker for Data Science Workflows <u>Analytics Vidhya</u>

### API using FLASK -

- Tutorial to deploy Machine Learning models in Production as APIs (using Flask) <u>Analytics Vidhya</u>
- How to Deploy Keras Models to Production <u>Siraj Raval</u>
- Flask Tutorial <u>Link</u>

### **Deployment -**

Deployment using Flask, docker and Kubernetes – <u>Medium</u>
Deploy Your First Deep Learning Model On Kubernetes With Python, Keras, Flask, and Docker - <u>Medium</u>
Deployable machine learning for data science – <u>Lynda</u>

# Thank You!