

Intro to MLOps

Quickly Deploying Data & ML Solutions via the Cloud

Victor Geislinger



What's Hardest about Data Science / Machine Learning?



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- Getting something useful to stakeholders



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Solutions?

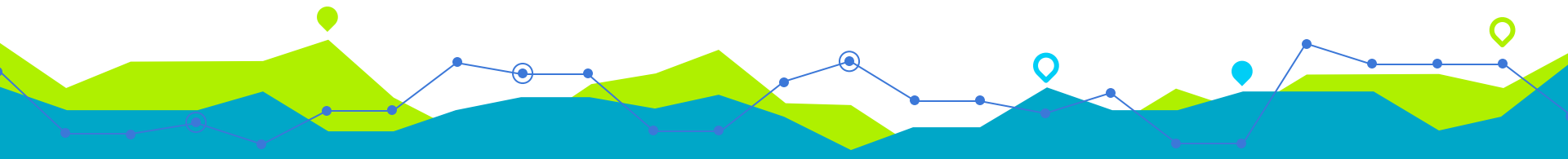


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Solutions?

- More focus on automation
- Iteratively improving
 - Rolling out
 - Working with SEs



What's Hardest about Data Science / Machine Learning?

Warning: Opinion Incoming...



What's Hardest about Data Science / Machine Learning?

- Moving trend in DS/ML



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- Less time in developing algorithms



What's Hardest about Data Science / Machine Learning?

- Moving trend in DS/ML
- AutoML
- Reliance on understanding of fundamentals (what you've been working on!)
- Less time in developing algorithms



A Solution – MLOps: What is it?

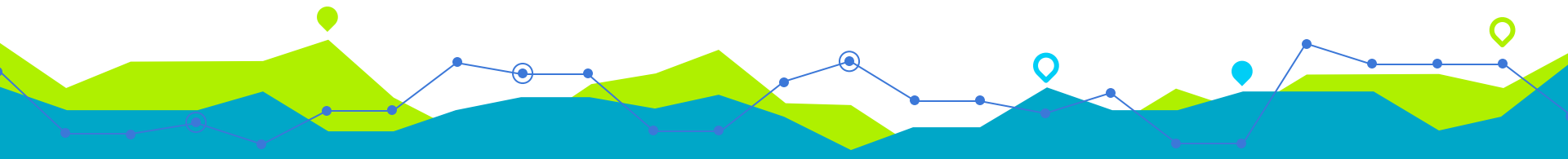


A Solution – MLOps: What is it?

- Machine Learning Operations
 - (similar to DevOps)



Pictured: Performing “ML Operations”...



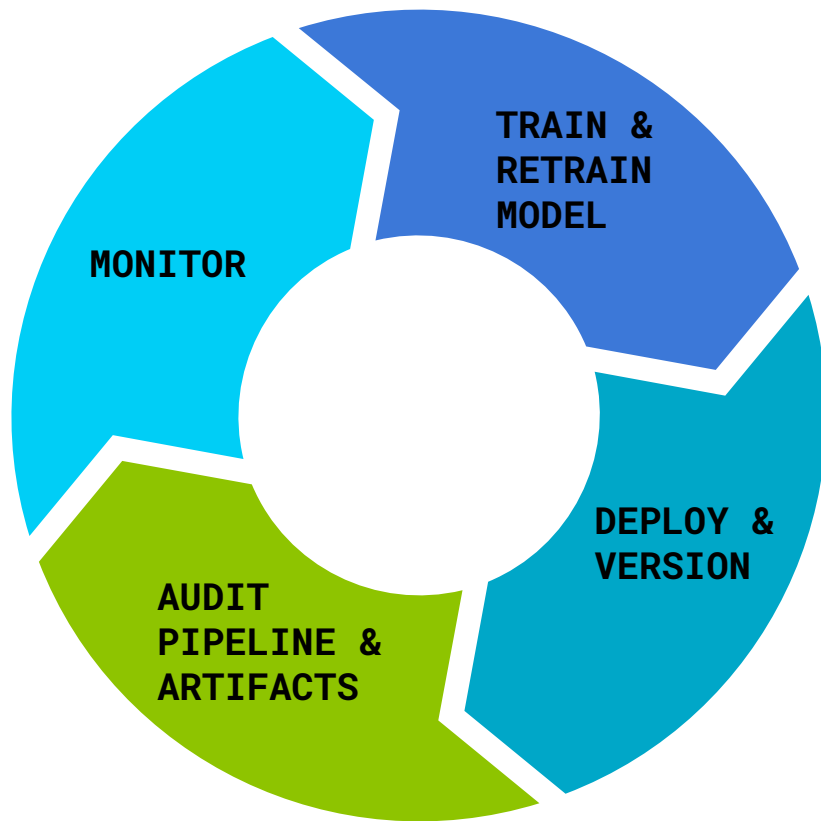
A Solution – MLOps: What is it?

- Machine Learning Operations
 - (similar to DevOps)
- Methods to incorporate the ML lifecycle
 - (move faster into production/usefulness)



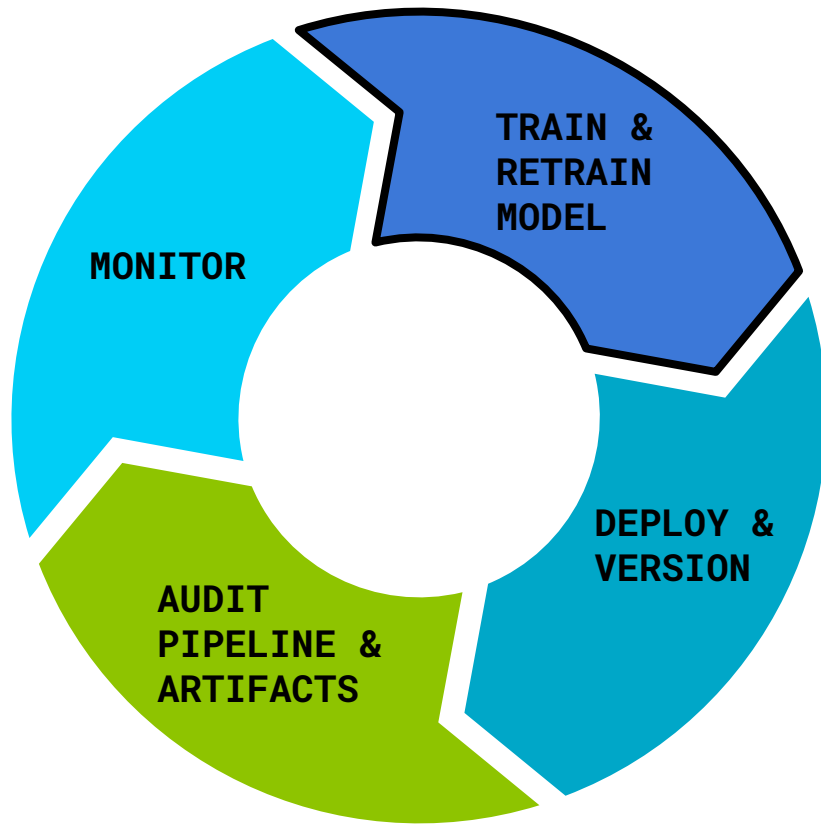
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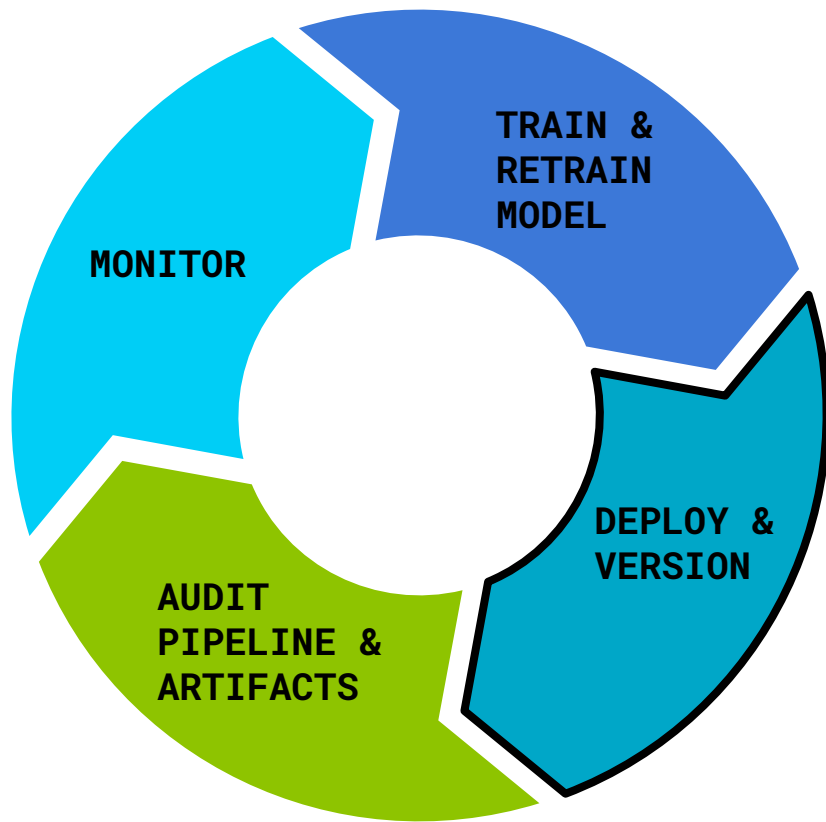
MLOps Feedback Loop

Adopted from *Practical MLOps* by Noah Gift and Alfredo Deza (O'Reilly).
Copyright 2021, 978-1-098-10301-9



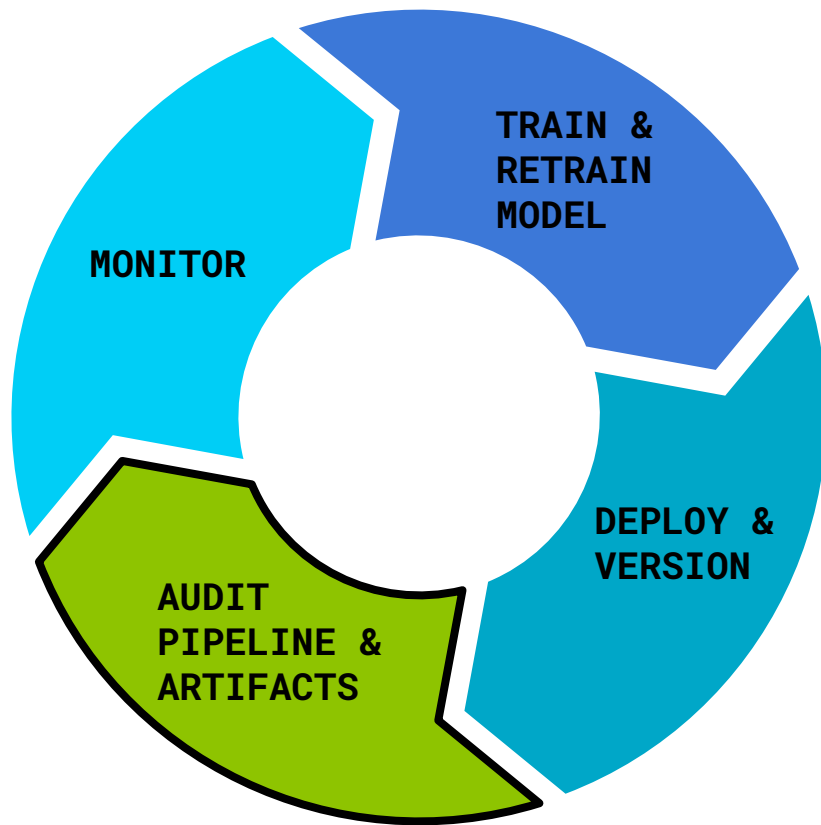
- Data can change
- People can change
- Reusable & automatic pipelines help





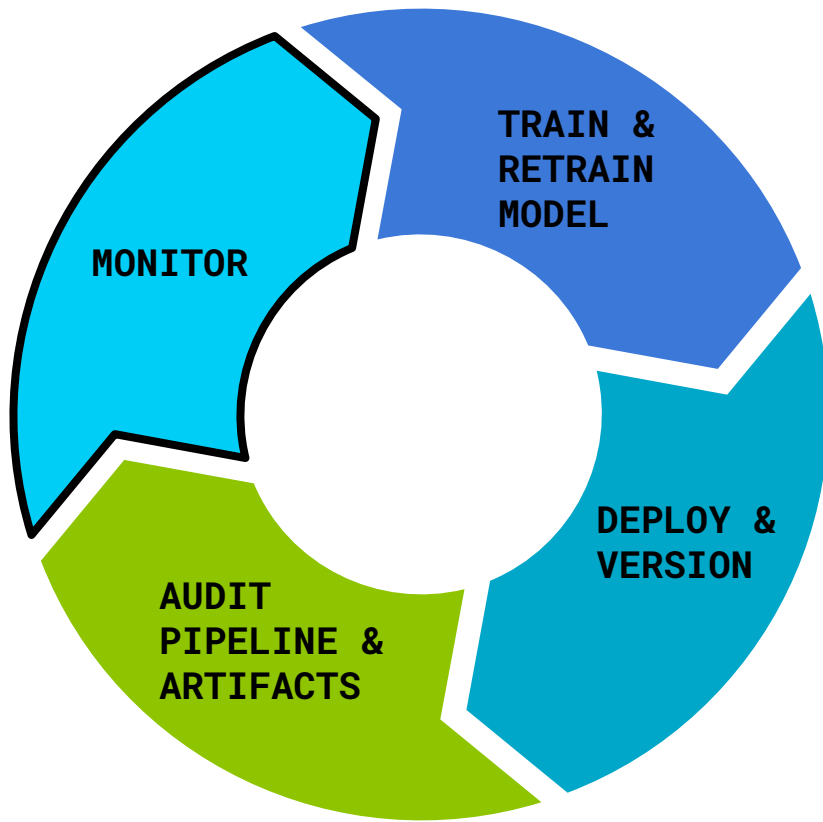
- Continuous Delivery (CD)
- Automated steps (infrastructure too!)
- Minimal friction to deploy new models





- Problems in accuracy, bias, & security arise
- Identify where they occur
- Useful for continuous improvement





- Data drift!
- Prevents accuracy issues



Let's Learn By Doing!



Scenario



Scenario



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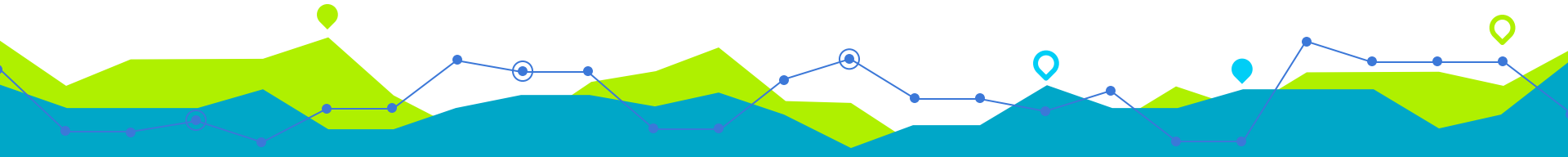


Scenario



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- Use case? What's important?
 - Parse Intent: What *needs* to be done?



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- How do we evaluate?
- How do I get data? Difficult?
- Think of CD/ improvement



Continuous Deployment / Continuous Improvement

WORK HARDER MORE
MAKE IT DO US FASTER THAN
EVER MAKES STRONGER OUR
HOUR NEVER BETTER
AFTER WORK IS OVER



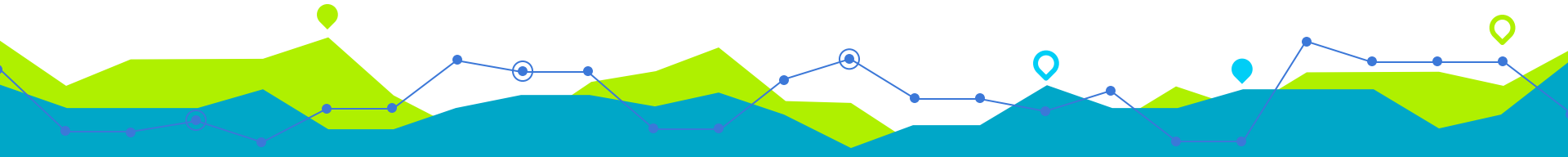
Continuous Deployment / Continuous Improvement

- Proof of concept (MVP)
- What parts can be iterated?



Continuous Deployment / Continuous Improvement

- Proof of concept (MVP)
- What parts can be iterated?
- Work w/SE? Build our own solution?
 - Cost of maintenance?
- Evaluation, Monitoring, etc.



Looking for Data



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- Is this data open/accessible?
 - Solutions already exist? (Pleasssse!!!)



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- Is there “enough”?
 - Other sources + company’s data



Looking for Data

- Is this data open/accessible?
 - Solutions already exist? (Pleasssse!!!)
- Is there “enough”?
 - Other sources + company’s data
- Do we have a “special” use case
 - How flexible is this?
 - How inflexible is it?
 - Issues? Fixes?



Looking for Data

- We'll use this:



[UCI Machine Learning Repository: SMS Spam Collection Data Set](#)



Simple Model

- Why?
- Make a baseline!
- What if a simple solutions *is* the solution?!
- (Story time)
-



Simple Model

- Why?
- Make a baseline!
- What if a simple solutions *is* the solution?!
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-
- Does it have “spammy” words (free, win, etc.)



Checking Out Code!

We'll look at this repo:

<https://github.com/MrGeislinger/mlops-example-spam-detector>



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Let's go to a notebook!



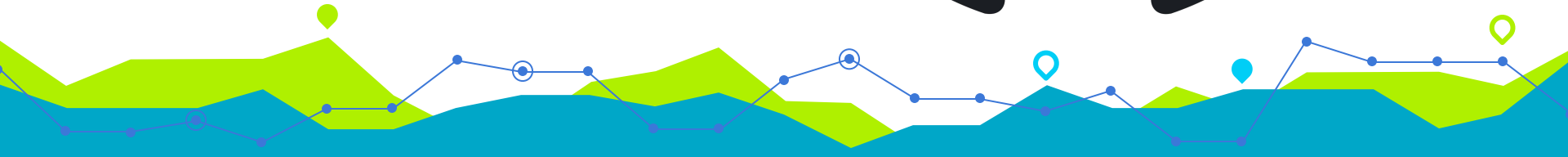
Revisit for Modularization

- Focus on separating out the parts
- Allows us to make *different* UI or allow an interface we don't know (prescribed from other team)



Checking Out Code!

Let's get modularize!
Back to some code!!



Revisit for UI

*I am but a lowly data scientist;
I don't know much about them UI design...*

- Victor Geislinger (everyday)



Revisit for UI

- We'll use Streamlit (Flask-based)

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 - Pretty easy to use with low-learning curve

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Revisit for UI

- We'll use Streamlit (Flask-based)
 - It's "so hot"
 - Pretty easy to use with low-learning curve
- Flask might be better to make a simple API
 - Streamlit is arguably prettier and we're doing a presentation. Sooo...

I am but a lowly data scientist; I don't know much about them UI design...



Revisit for UI

- We'll use Streamlit (Flask-based)
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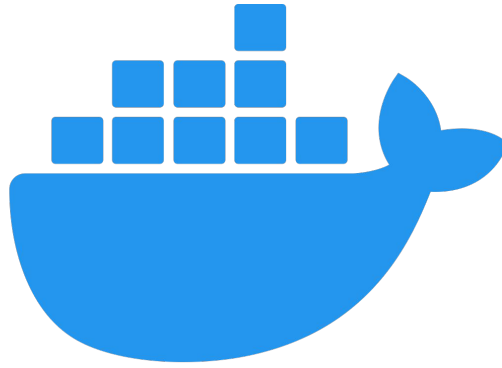
Checking Out Code!

Back to some code!!

Gotta make it pretty for our stakeholders!



Docker-ize It!

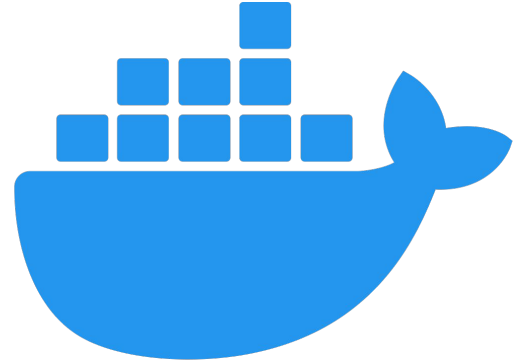


docker[®]



Docker-ize It!

- Allows to “package” up a working environment (called a “container”)
- Makes deploying a lot easier/simpler
- Used a lot by software engineers



docker[®]



Checking Out Code!

Let's make a Dockerfile and
container-ize our code!



Cloud Deployment: Serverless



Cloud Deployment: Serverless

- Deploy to the “cloud”
- Past: Manage a server to run our code
- Future: “Serverless”
 - Run code only when needed
 - Easier to implement/rollout versions



Checking Out Code!

We'll use Google Cloud Platform (GCP)

Specifically using “Cloud Run” that uses Docker images

(But there are countless of different methods!)



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- Hey we need an improvement!
 - Can be from the DS/ML team
 - Can be a “signal”
- Ideally continuously monitor & train, automatically replacing the model



Iterate (if we have time)

- Hey we need an improvement!
 - Can be from the DS/ML team
 - Can be a “signal”
- Ideally continuously monitor & train, automatically replacing the model
- How you update model depends on use case
 - Emergency:
 - Stop everything; rollback to working version
 - Signal new model:
 - auto-rollout
 - Signal drift: investigate



Other Considerations

- Data & Concept Drift
- Monitoring & Logging
- CI / CD
- Tools
 - GitHub Action
 - GCP Vertex AI
 - AWS Sagemaker



My Advice

- Learn CLI
 - Get comfortable in the terminal

```
ROBCO INDUSTRIES UNIFIED OPERATING SYSTEM
COPYRIGHT 2075-2077 ROBCO INDUSTRIES
-Server 1-

=====
>\WELCOME, Overseer Bambi

>\ Fallout 4 Release Date
>\ Tremor Beggar Story
>\ Access Mail
>\ Disengage Vault Lock
```



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My Advice

- Learn CLI
 - Get comfortable in the terminal
- Learn to modularize your code!
 - Python files, Classes, packages, etc.
- Docker (see above points)
- Cloud - especially serverless!

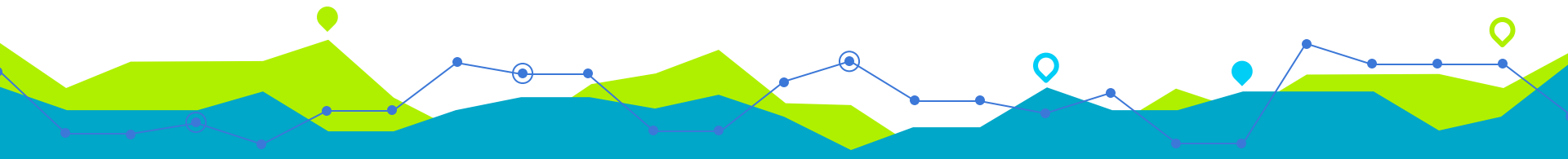


Further Reading

- Practical MLOps
 - Focus on ideas
- Building Machine Learning Pipelines
 - How to automate model lifecycles
- Data Science at the Command Line
 - Useful way to introduce CLI
 - Free online
- Building Machine Learning Powered Applications
 - More about ideas and some use cases

Other books

- Build a Career in Data Science
- Hands-On ML w/ Scikit-Learn ... (2nd Ed).
- Python Data Science Handbook
- Machine Learning Design Patterns



THANKS!

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