



What will this session be about?

Who am I?

@AnnieTalvasto
CMO at VSHN

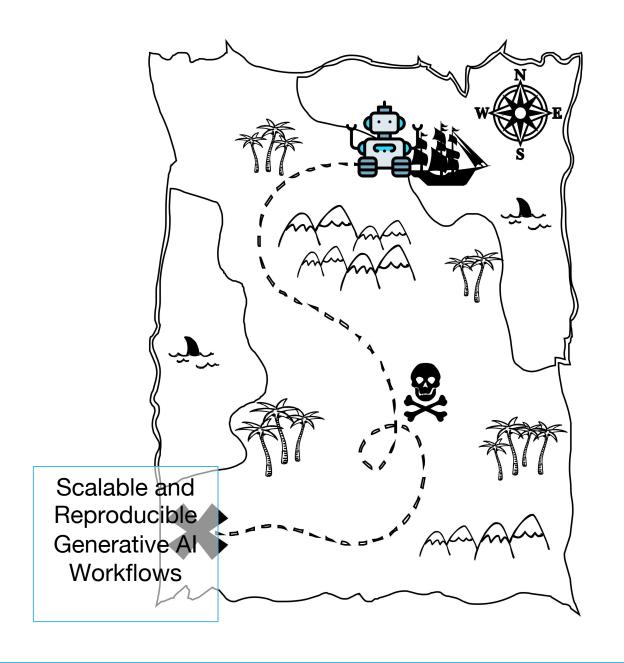
- **OCNCF** Ambassador
- OAzure MVP



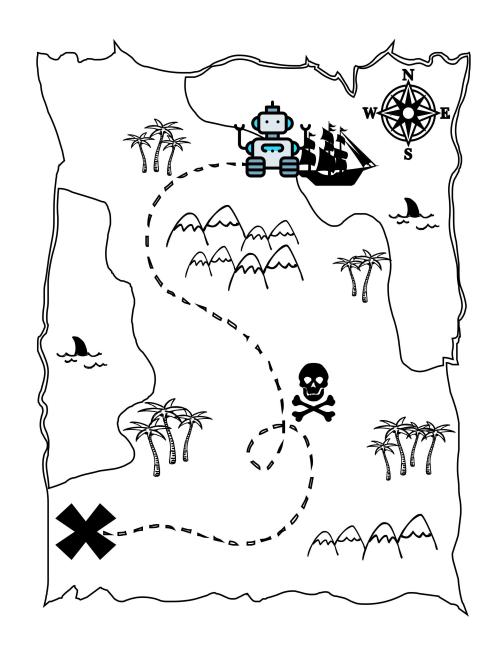
- Kubernetes & CNCF meetup co-organizer
- Startup-coach
- Co-host of Cloudgossip podcast cloudgossip.net

- oIntroduction
- Definitions
- OWhy is AI/ML different
- Kubernetes & MLOps
- OKubeflow
- Best practices
- Considerations
- Wrap up & Resources

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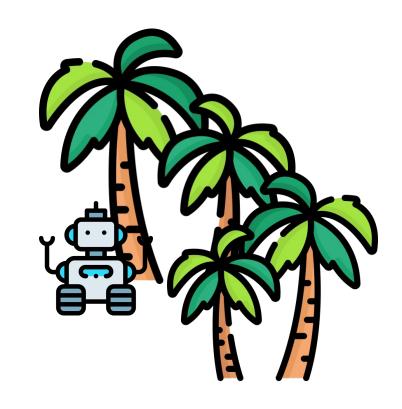


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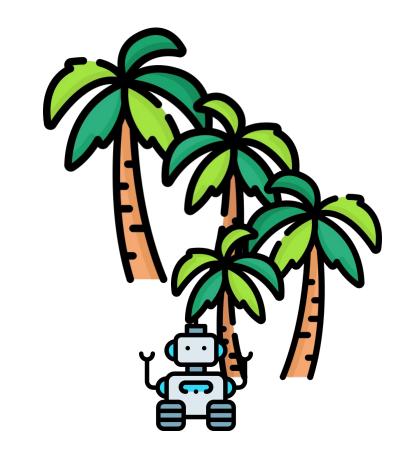
Al

OArtificial Intelligence (AI) refers to the simulation of human intelligence in machines that are programmed to think and learn like humans.



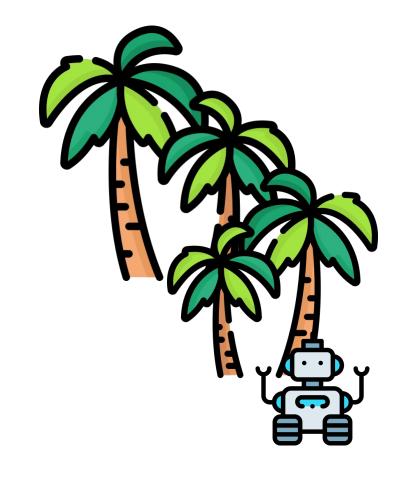
ML

• Machine Learning (ML) is a subfield of artificial intelligence (AI) that focuses on the development of algorithms and statistical models that enable computer systems to improve their performance on a specific task through learning from data, without being explicitly programmed.



Data science

 Data science is a multidisciplinary field that uses various techniques, algorithms, processes, and systems to extract insights and knowledge from data.



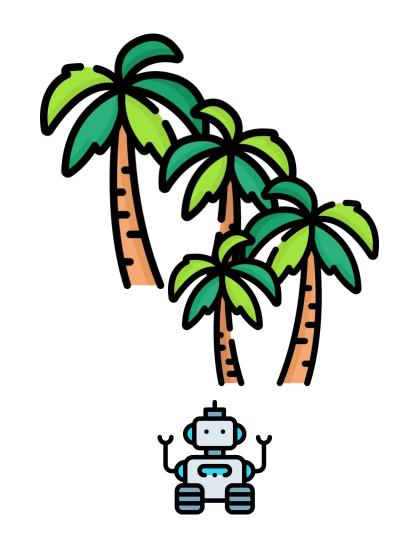
DevOps

Operations," is a set of practices, principles, and cultural philosophies that aim to improve and streamline collaboration between software development (Dev) and IT operations (Ops) teams.

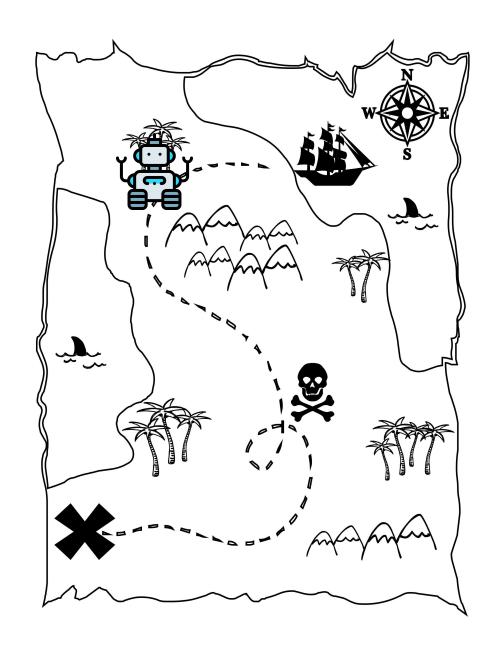


MLOps

 MLOps, short for "Machine Learning Operations," is a set of practices, principles, and tools that combine machine learning (ML) with the practices of DevOps.

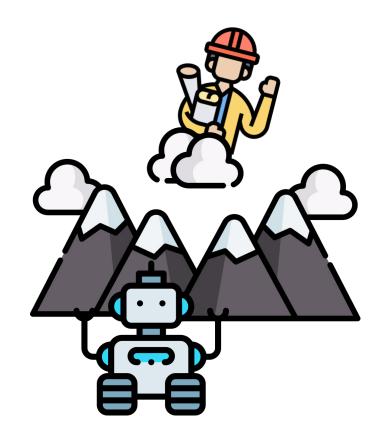


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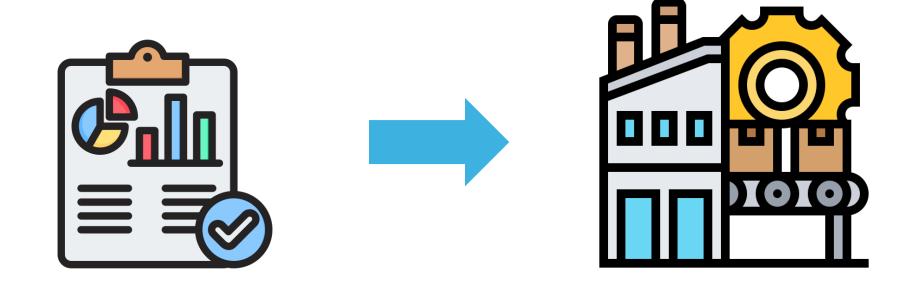


Differences with AI/ML

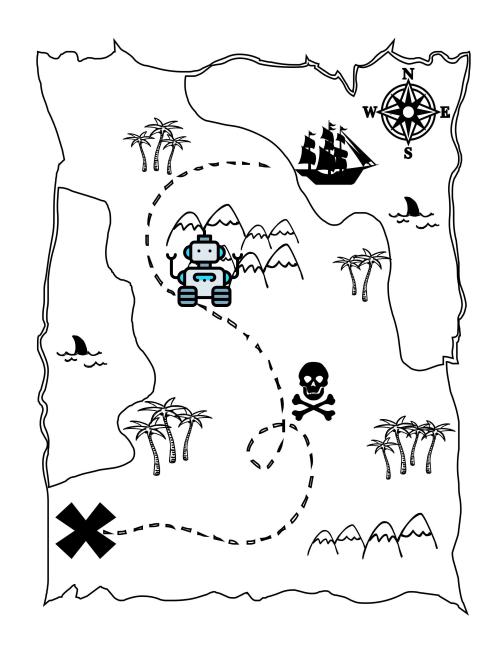
- Data-Centric Nature
- Model Complexity
- Iterative Development
- Scalability
- Monitoring and adaptation
- Testing and validation
- Explainabity and bias
- Rapid advancements
- Collaboration
- Deployment challenges
- Cost



From Research to Production

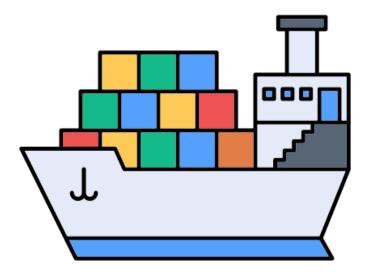


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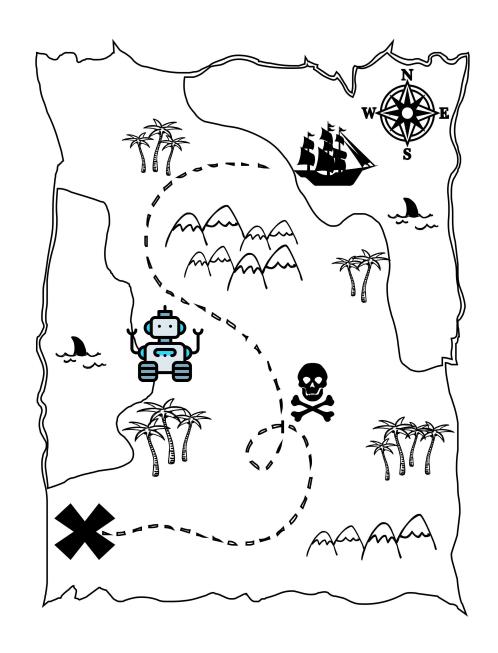


Kubernetes & MLOps

- Portability
- Customizability
- Performance
- Consistency
- Microservices
- Composability



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Kubeflow

The machine learning toolkit for Kubernetes.

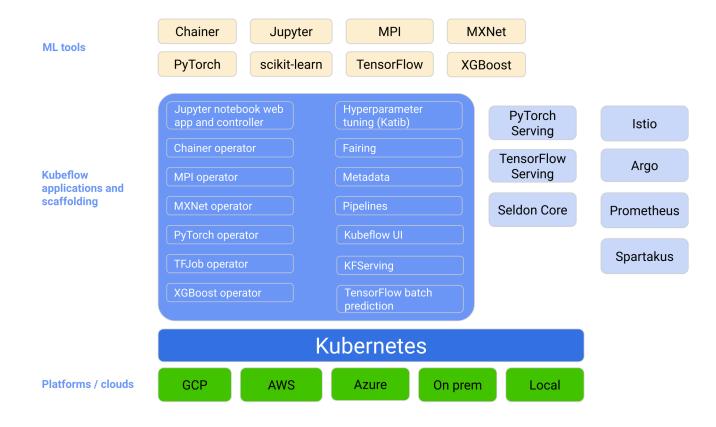


What is Kubeflow?

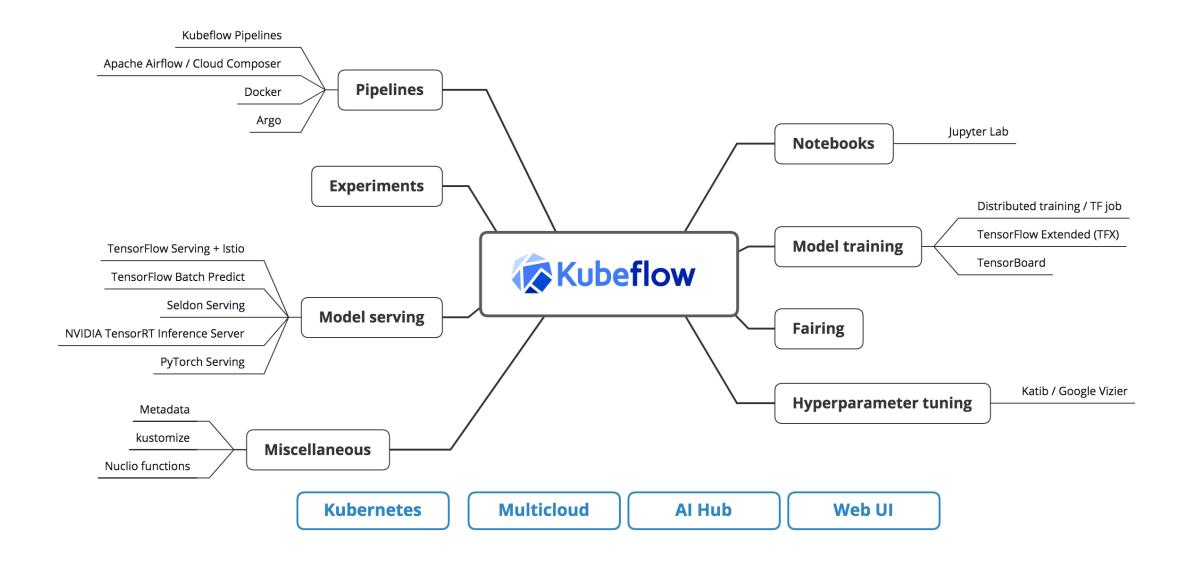
○ End-to-End ML workflow

Scalability and resource management

Reproducibility and collaboration

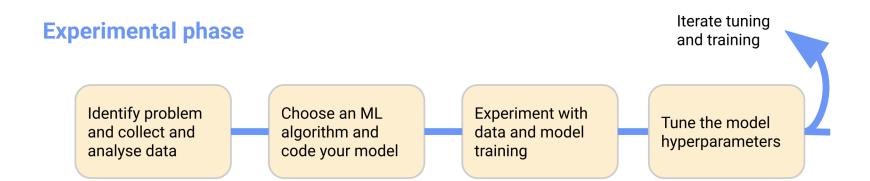


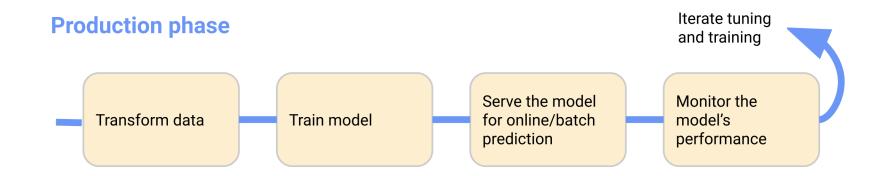
Source: https://www.kubeflow.org/docs/started/architecture/



Source: https://www.analyticsvidhya.com/blog/2023/01/kubeflow-streamlining-mlops-with-efficient-ml-workflow-management//

Before Kubeflow





Source: https://www.kubeflow.org/docs/started/architecture/

With Kubeflow

Choose an ML Identify problem Experiment with Tune the model and collect and algorithm and data and model hyperparameters code your model analyse data training PyTorch **Jupyter Notebook** Katib Fairing scikit-learn TensorFlow **Pipelines** XGBoost **Production phase** Iterate tuning and training Serve the model Monitor the Transform data Train model for online/batch model's prediction performance Chainer **KFServing** Metadata MPI **NVIDIA TensorRT TensorBoard MXNet PyTorch** PyTorch **TFServing** TFJob Seldon **Pipelines**

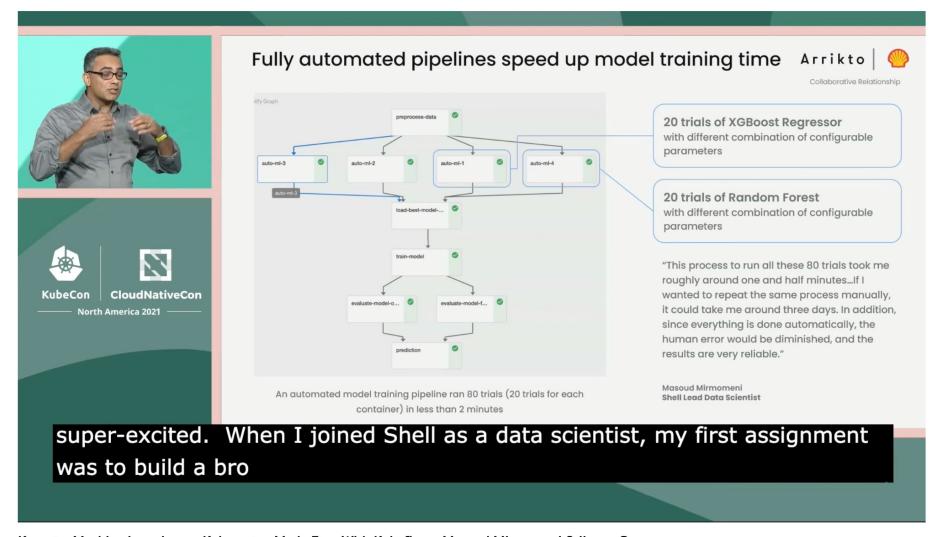
Experimental phase with Kubeflow

Source: https://www.kubeflow.org/docs/started/architecture/

Iterate tuning

and training

Example: Shell



Keynote: Machine Learning on Kubernetes Made Easy With Kubeflow - Masoud Mirmomeni & Jimmy Guerrero



CNCF [Cloud Native Computing Foundation]
105K subscribers





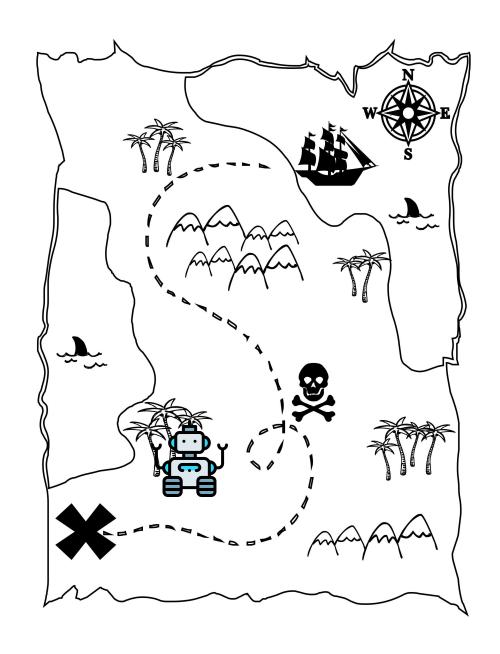






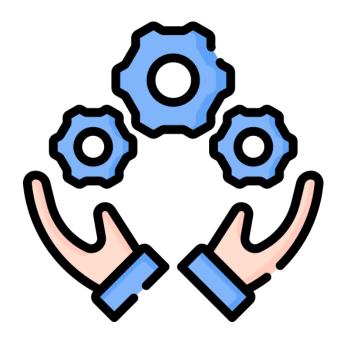


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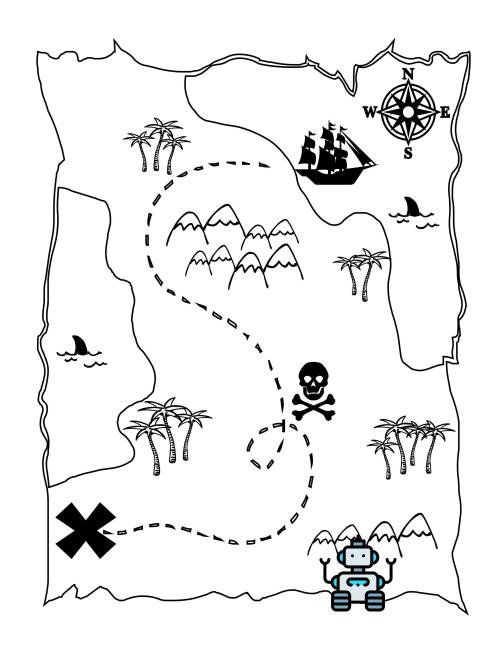


Best Practices

- OIt is always start with data
- Master all the necessary skills
- Know the tools you need and choose them wisely
- Understand the bottlenecks you face
- ○MLOps ♥ DevOps
- Consider serverless
- AutoML, Kaizen



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Ethical considerations

- Fairness & Bias
- Transparency
- Privacy
- Accountability
- Consent and user empowerment
- Security
- Human-Centric Design
- Data Governance
- Long-term consequences
- Regulatory Compliance

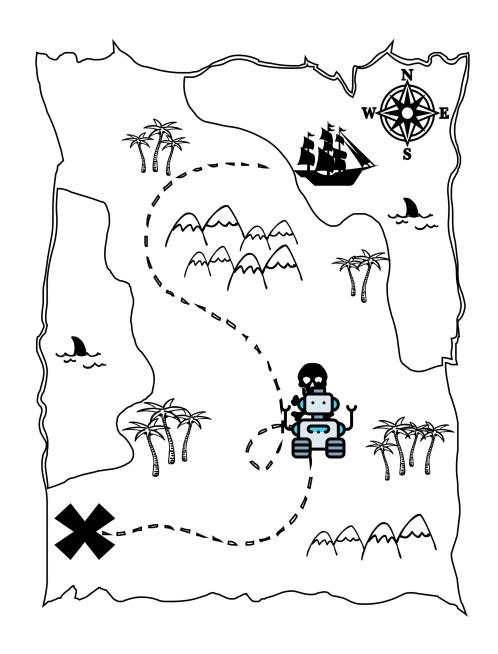


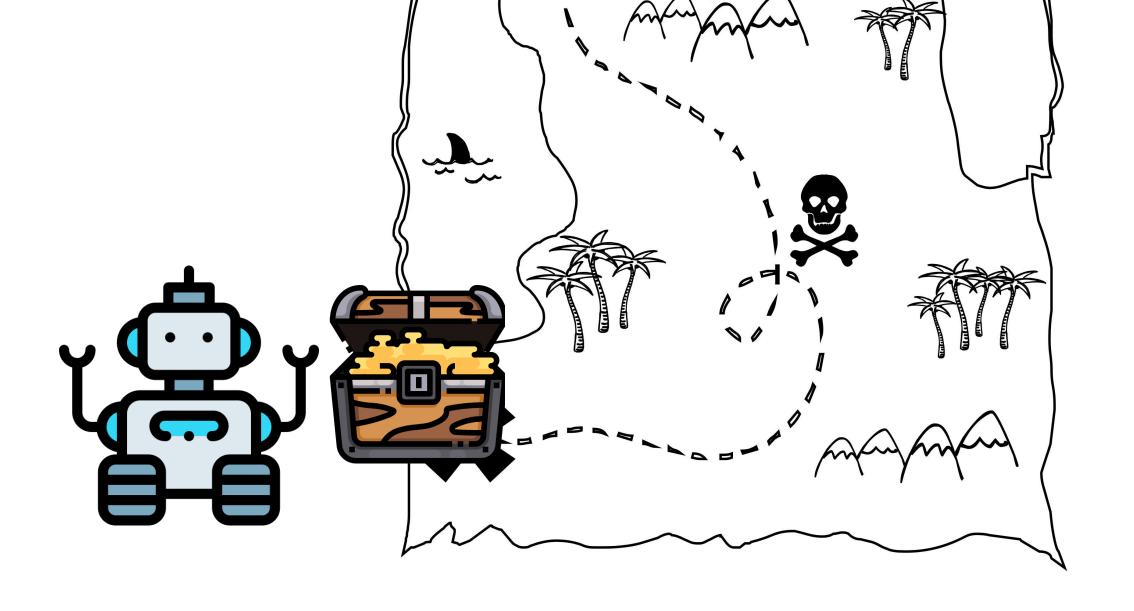
Learn more

- Links and slides: github.com/annietalvasto
- Practical MLOps by Noah Gift and Alfredo Deza (O'Reilly). Copyright 2021 Noah Gift and Alfredo Deza, 978-1-098-10301-9.
- Keynote: Machine Learning on Kubernetes Made Easy With Kubeflow - Masoud Mirmomeni & Jimmy Guerrero
 - https://www.youtube.com/watch?v=ick5hI5YI0k
- Kubeflow for Machine Learning Holden Karau & Adi Polak • GOTO 2022
 - https://www.youtube.com/watch?v=dUZ5TrSReZ c&t

- Webinar: MLOps automation with Git Based CI/CD for ML
 - https://www.youtube.com/watch?v=VCUDo9um KEQ
- All Kubernetes AI day Sessions
- Welcome + Opening Remarks: The State of Production MLOps in the Cloud Native... -Alejandro Saucedo
 - https://www.youtube.com/watch?v=xymbp8RW aCQ&list=PLj6h78yzYM2M9oVaU3amsqL5RXUwc u

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Thank you!

