



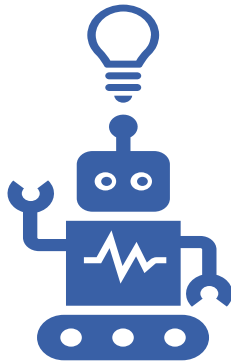
# DevOps for Data Science/DataOps

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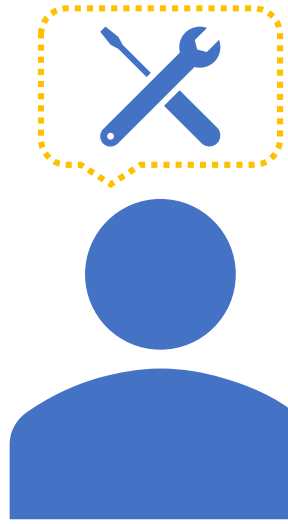
*Applied Machine Learning Scientist*

# Machine Learning Revolution

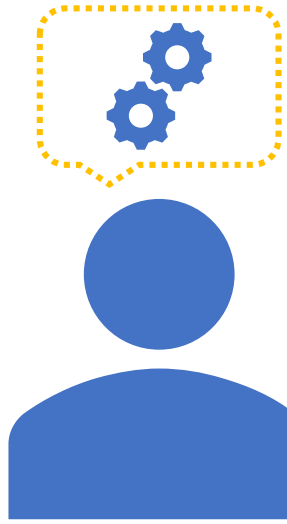
- Data-focused companies
- Data-driven decision and automation
- New type of software components that need to be managed



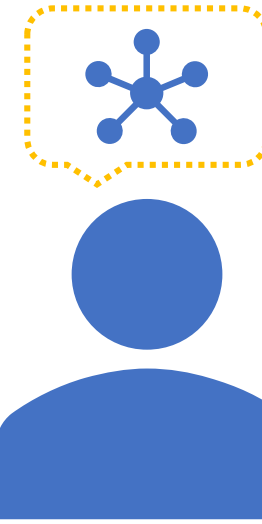
# Teams and new roles



Data Engineer

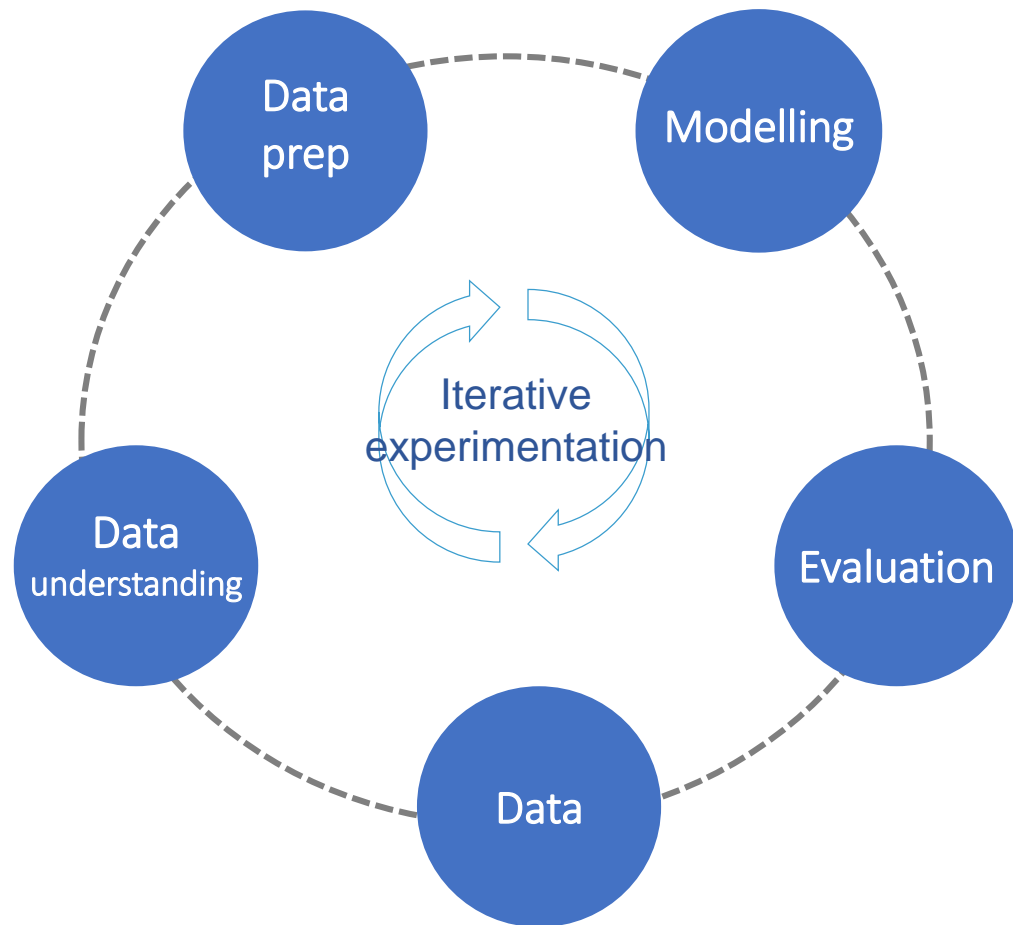


Data Scientist



DataOps Engineer

# Typical Machine Learning Process



# Modelling + Evaluation

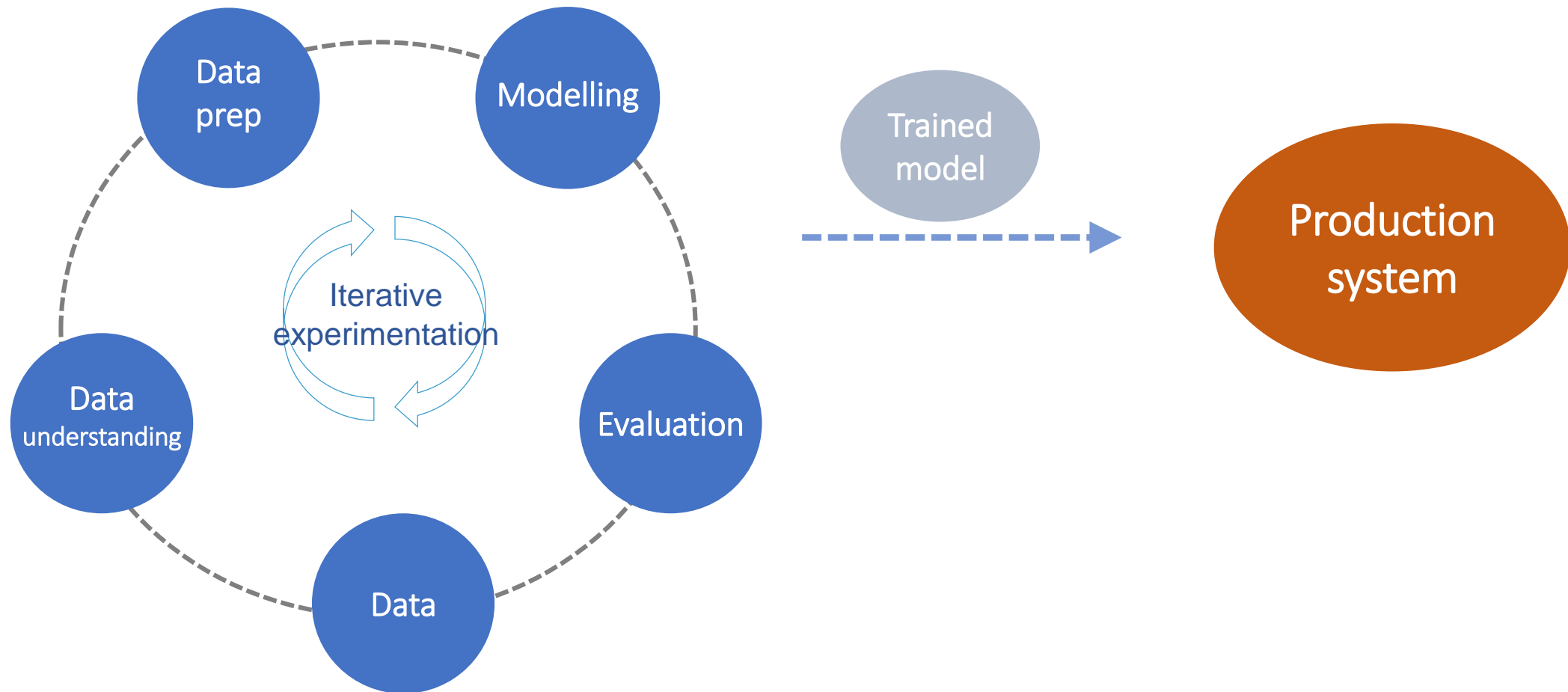
- Most talked about – most exciting
- Experimentation and research happens here
- Some data scientist develop this completely in isolation



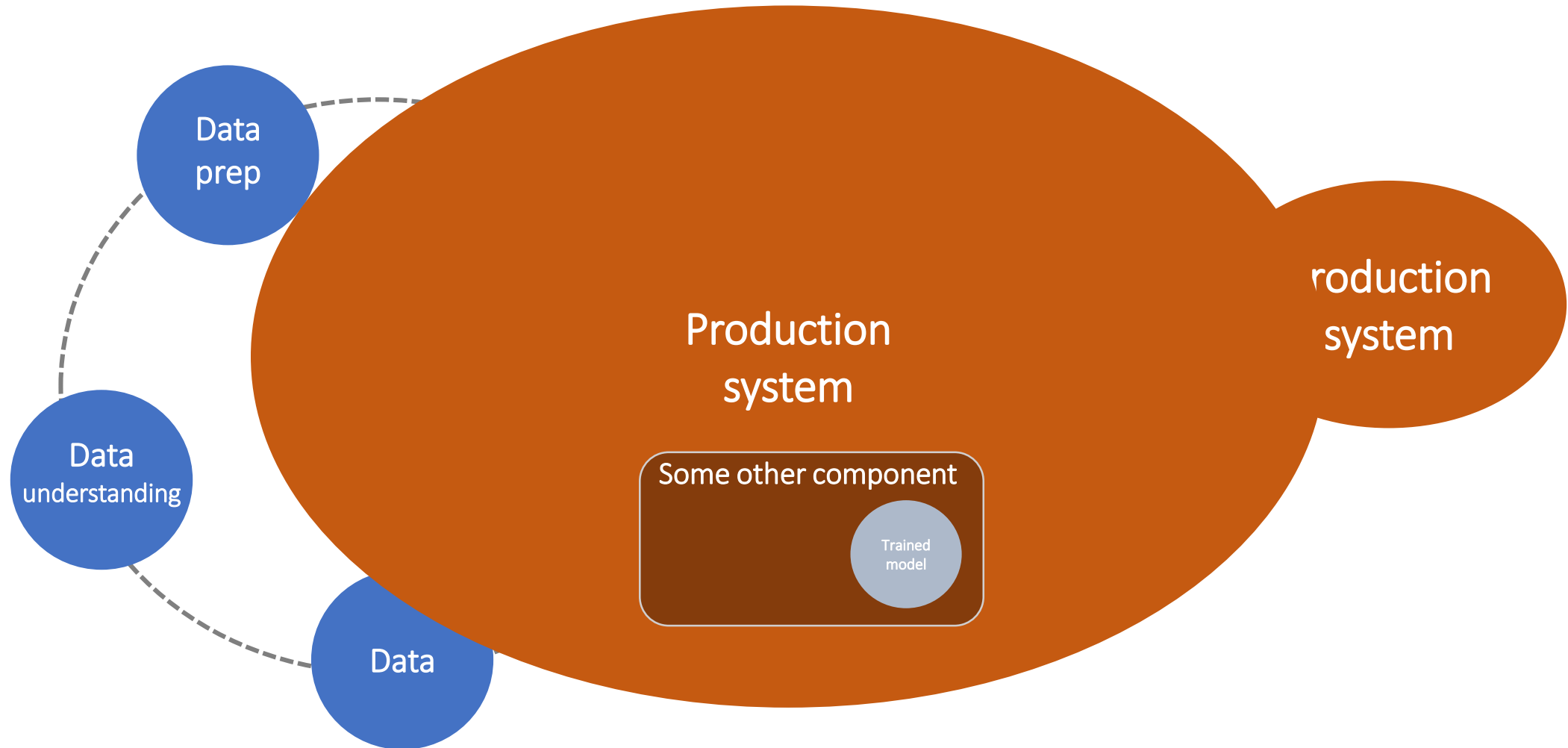
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- Model versioning and source control
  - Need to understand the context of how it is going to be used in order to have the right input/output
  - Where does it fit?



# Typical Machine Learning Process



# Typical Machine Learning Process



# Production steps for ML models

- Define model interface (input/output)
- Store the model and the environment requirements (versioning)
- Wrap the model up as a consumable service (e.g. API – Flask app)
- Deploy this service as part of your system
- Run it along side previous solutions to compare performance



# Where does DevOps fit?

- CI/CD
- Build data pipelines
- Build and test (API level)
- Monitor and learn
- Automate retraining
- Source control for code, notebooks, model

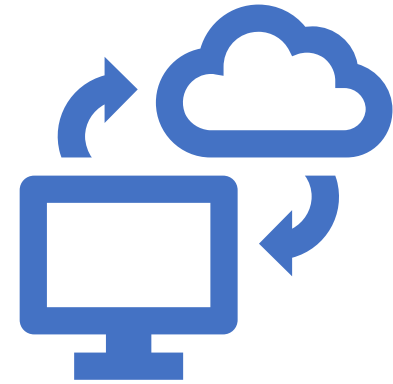
# Once in production ...



Monitor model  
performance



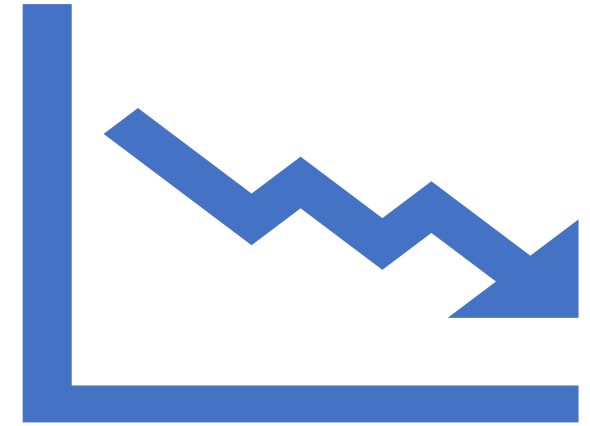
Validate live  
results



Automatic  
retraining

# Data quality monitoring

- ✓ Ensure the data types are consistent
- ✓ Ensure the data is within expected values
- ✓ Ensure no NAs are flowing through the system



**Data Validation at each  
stage of your data pipeline**

# Conclusions

- Culture change
- Where does it fit?
- Data validation is important
- Follow DevOps principles for ML artifacts
- New teams and new roles -> new processes and tools



**Thank you!**