

Managing ML Models @ Scale

Intuit's ML Platform

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ML in Action

ML driven experiences





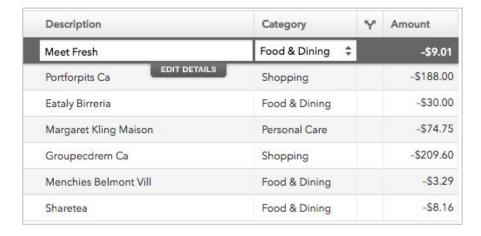
Self-help in TurboTax





Forecasting cash flow in QuickBooks

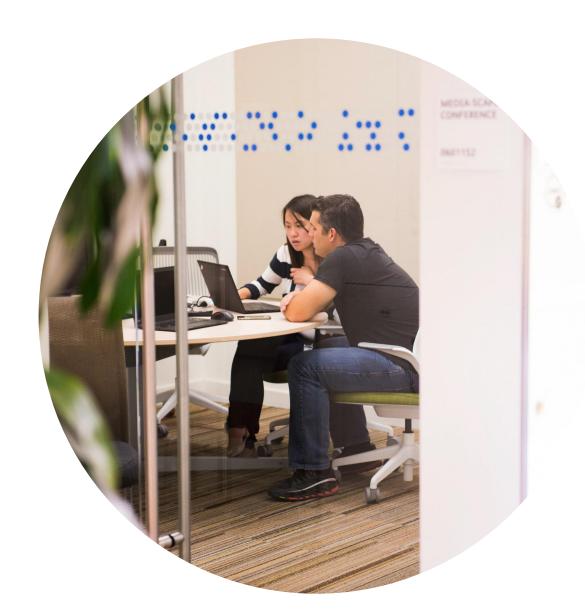




Automatic categorization in Mint



Managing Models in Production is hard

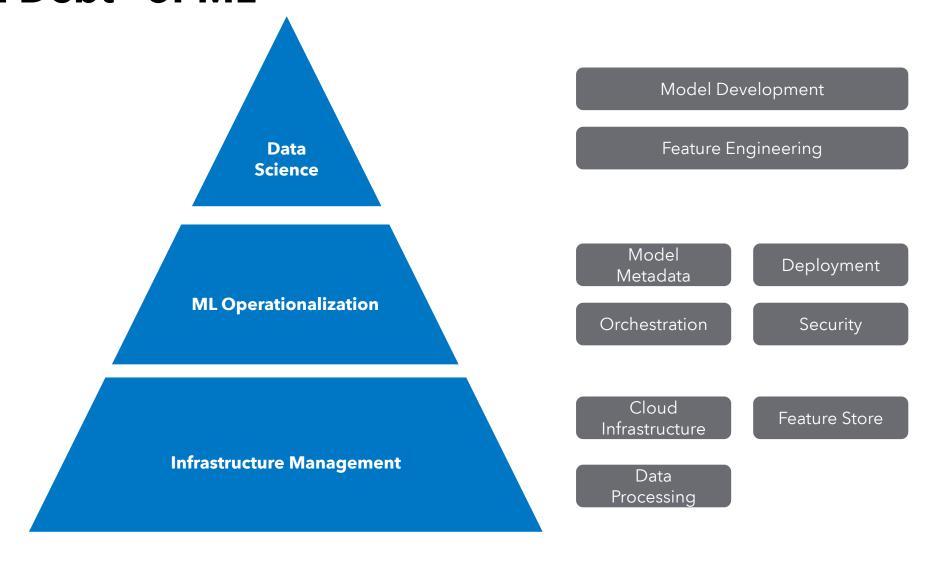




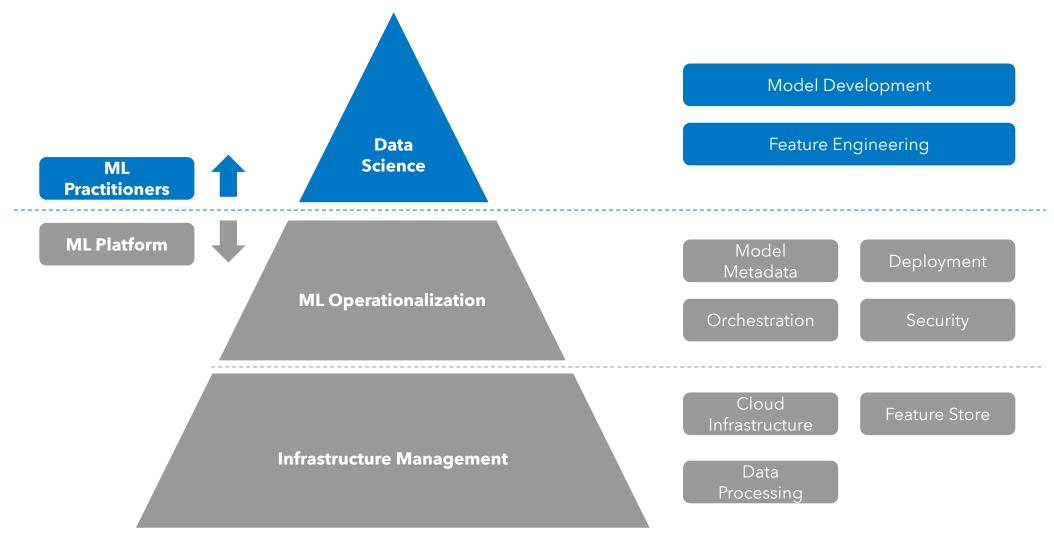
We spend more time bringing the model to production than developing and training it

– Data Scientists, 2018

"Hidden Debt" of ML



How can ML Practitioners focus on their craft?





Principles of Intuit ML Platform



Collaboration



Self-Service



Security

Powering insights and experiences with ML



Data Scientist

"I can quickly train and deploy models that improve the customer experience"



ML Engineer

"I can quickly iterate and make models performant"



Data Analyst

"I can quickly train and deploy models that inform stakeholders how the business is doing"



Product Manager

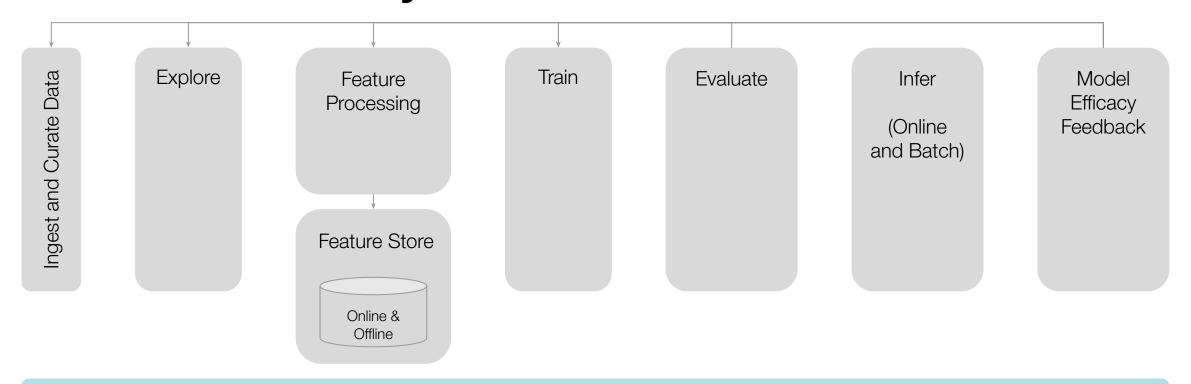
"I can easily run experiments that rely on models to get a customer benefit"

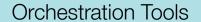


Marketer

"I can easily target particular users that have a certain characteristics"

Generic Model Lifecycle





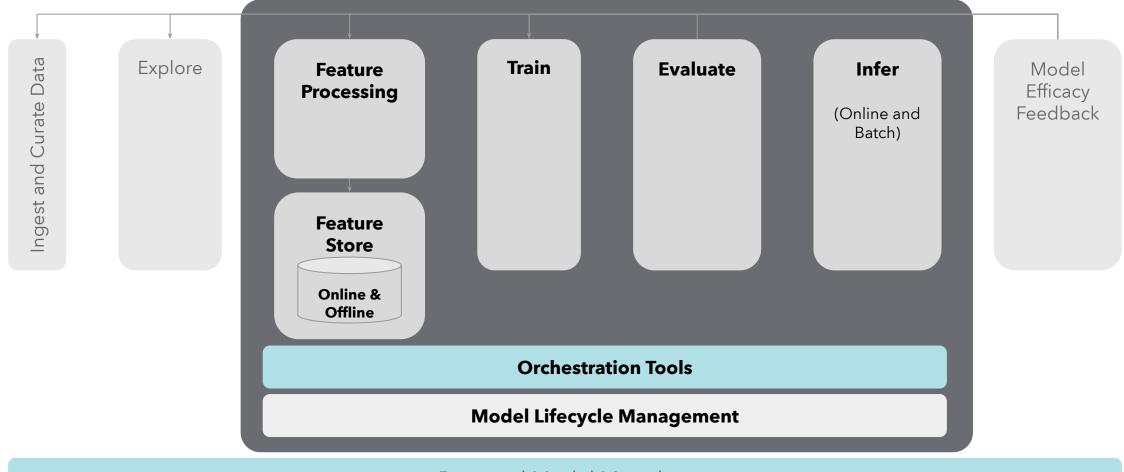
Model Lifecycle Management

Data and Model Metadata

Data Processing Infrastructure



Intuit's ML Platform Today



Data and Model Metadata

Data Processing Infrastructure



Technologies in use









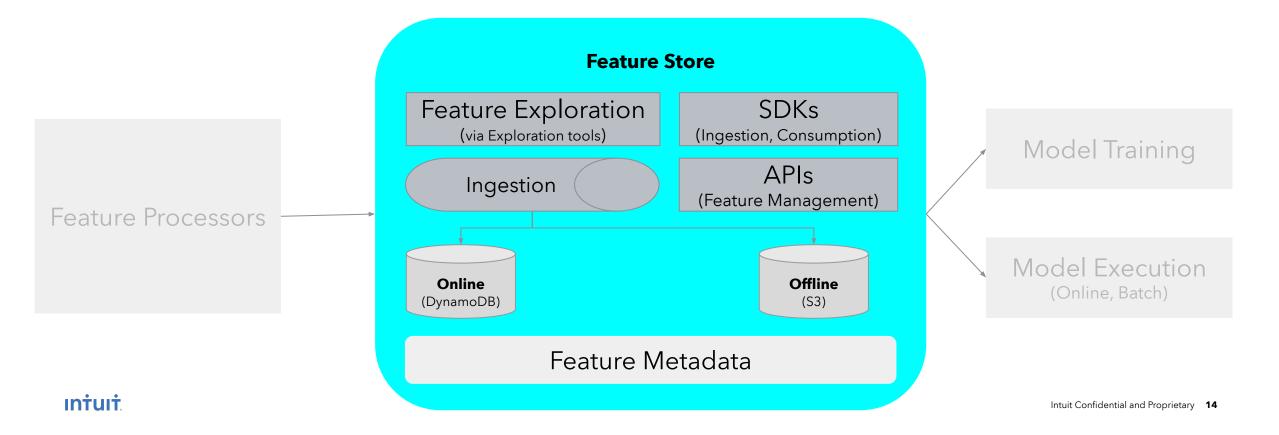




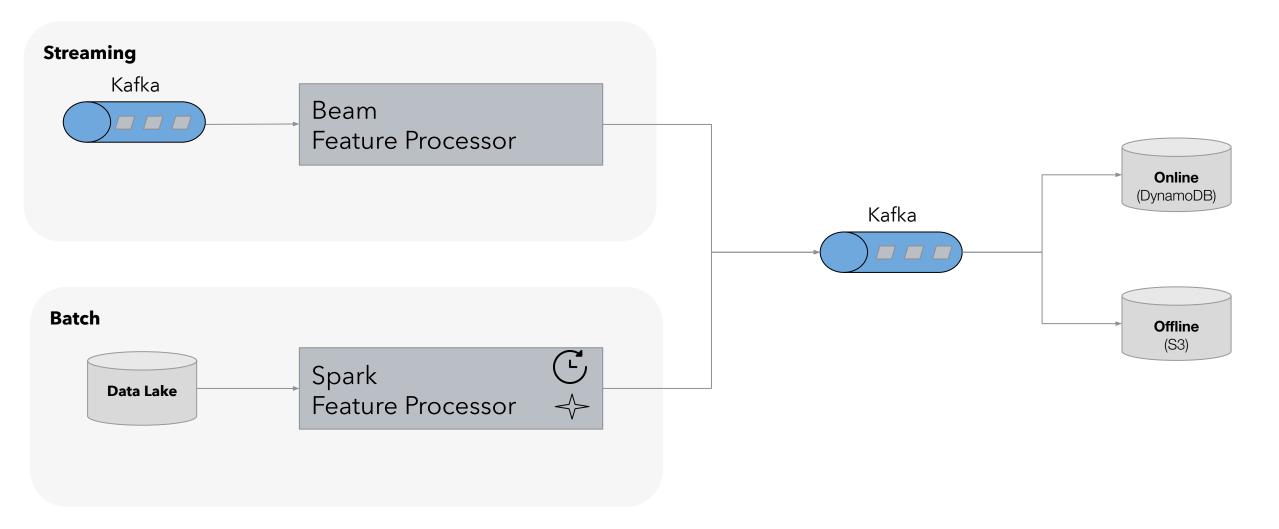


Feature Store

- Store Features in offline (durable) and online (transactional) stores
- Metadata about Features
- Find and re-use Features
- Feature access during model inference and model training

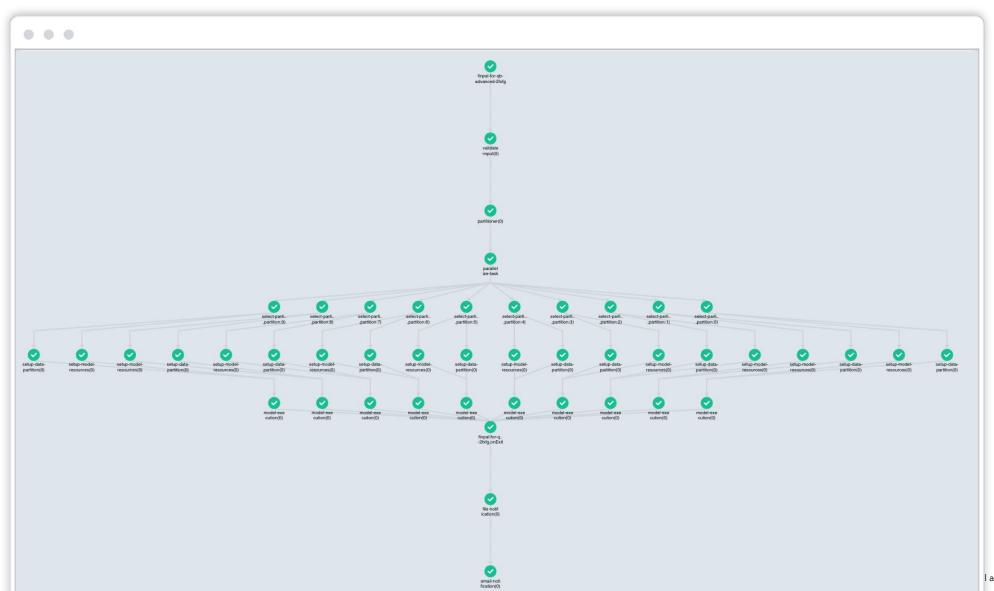


Feature Processing





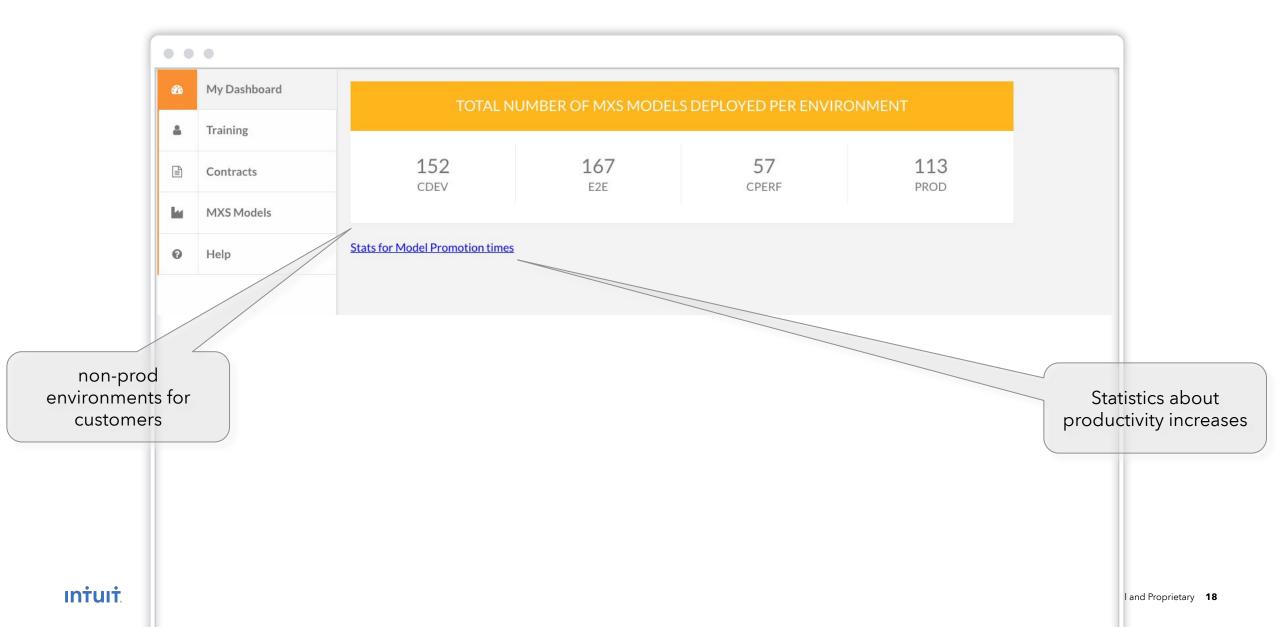
Model Training



Model Training

```
. . .
          pipeline_type: sagemaker-training
          pipeline_version: master
         artifact_url: test.docker.repo/creditcard-approval-model/
         project_name: creditcard-approval-model
          kms_key_arn: arn:aws:kms:us-west-2:1234567890:key/key
         email: firstNameLastName@intuit.com
         training_data_config:
           data_uri: s3://creditcard-approval-input/data/training/
           content type: csv
         validation_data_config:
           data_uri: s3://creditcard-approval-input/data/validation/
           content_type: csv
         additional_data_configs:
           - data_config_name: test
              data_uri: s3://creditcard-approval-input/data/validation/
              content_type: csv
         output_bucket_uri: s3://creditcard-approval-output/
         tags:
           - Key: intuit:billing:fp
             Value: 3po4uih3-9084h49iu3fn3-894n
          resource_config:
           training:
              instance count: 1
              instance_type: ml.m5.xlarge
             max_run_time_sec: 86400
              ebs_volume_gb: 1
```

Self Service Interfaces



Automate the simple things



With great speed, comes cost

Visibility and Self-service

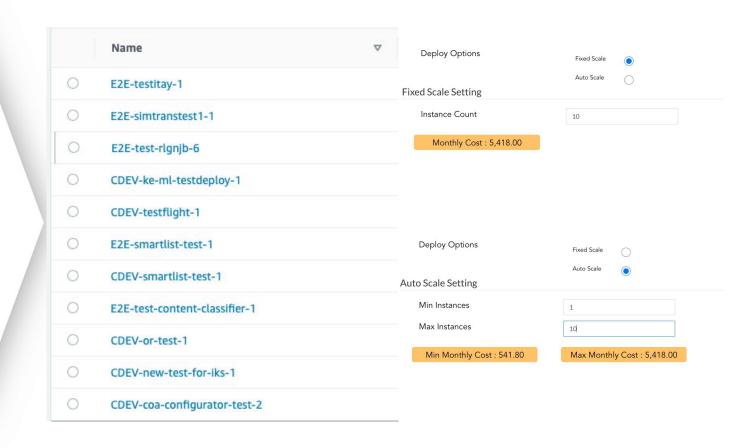
Platform success = Lots of experiments

Cost Transparency

- Upfront pricing
- Cost dashboards

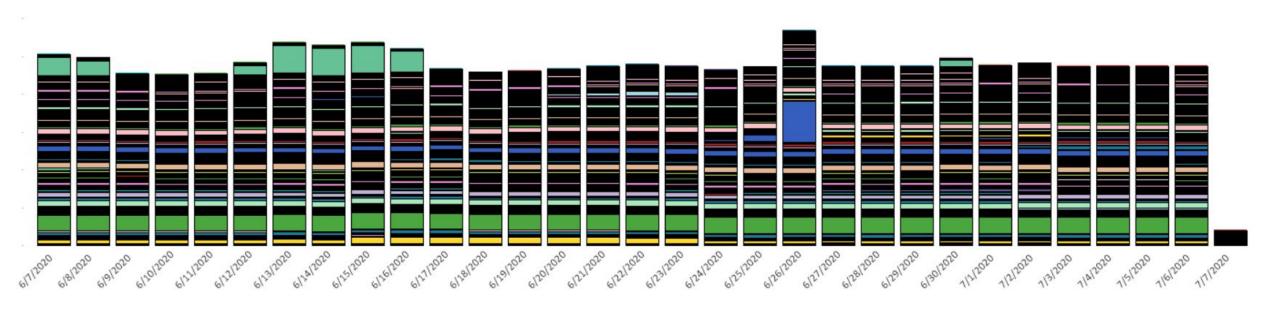
Cost Assignment

- Tagging
- Business Unit assignments
 - Platform ensures correct BU chargebacks



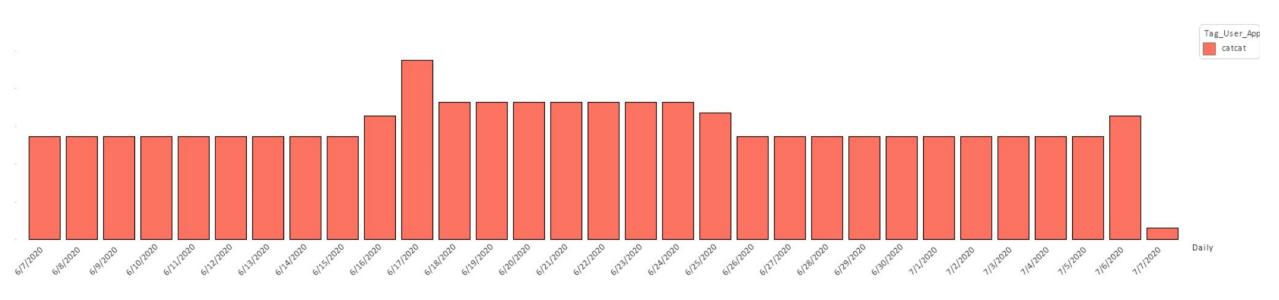


Cost transparency



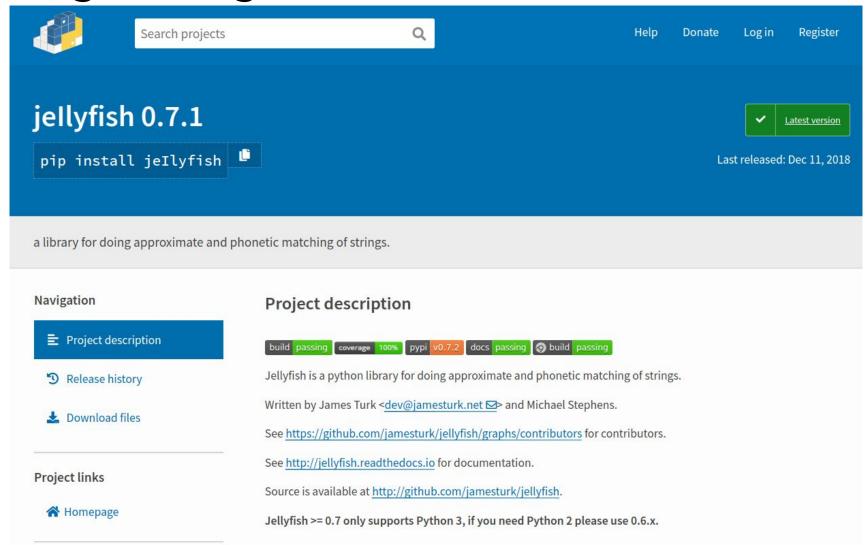


Cost transparency





What can go wrong?





Security

ML Models are as vulnerable as any other piece of software

ML Model Vulnerabilities

- ML models vulnerable for attacks
- Need security gates at every stage of development

Quality Gates

- Source image validations
- Standardized framework for container images
- Image hardening, certification and signing

Security Gates in our AWS setup

- Secure VPC endpoints
- VPC flow logs monitor traffic
- Security groups restrict outbound access
- KMS encryption secure data at rest
- Least privilege IAM roles secure management of the service





Compliance

Complex regulations

- The well known
 - CCPA, GDPR
- More common
 - PCI
- The arcane (very specific to tax compliance)
 - INDOR, NIST

Platform enforces compliance

- Any customer data that is used in the ML model lifecycle is managed centrally
- Achieving compliance for all models reduces complexity in the process for applications using them
- Automating the necessary controlling actions for the platform ensures that future models remain compliant



Learnings



Models as Software

- Accelerated deployment through self service
- Using a GitOps approach allows for declarative development of models
- Production monitoring and alerting
- Security, compliance built-in



Automated Workflows

- Workflows for standard MI operations help with complexity
- Hosting a workflow engine allows for easy extensibility
- Small threshold for customers starting out with the platform through templates with short configuration



Curated Feature Store

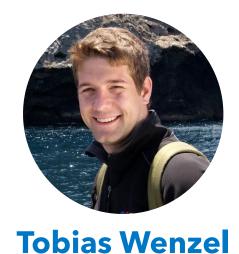
- Model quality **C** Feature quality
- Curated feature store offers library of meaningful features
- Search, explore, share features across models enables acceleration of model development



In Future...

- Custom ML Operator to orchestrate and manage ML Resources (Spark, SageMaker Training/Deploy, Feature Store)
- Declarative Management of MDLC such as re-training of models, monitoring of features etc.
- Managed Notebook Service

Contact Us



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