



QuantUniversity, LLC

[www.quantuniversity.com](http://www.quantuniversity.com)



# Scalable Data Science Pipelines with QuSandbox & MATLAB Online Server

**Presented By:**

Sri Krishnamurthy, CFA, CAP

[sri@quantuniversity.com](mailto:sri@quantuniversity.com)

[www.quantuniversity.com](http://www.quantuniversity.com)

**MathWorks Computational  
Finance Conference 2021**

**Online**

# Speaker bio



Sri Krishnamurthy  
Founder and CEO  
QuantUniversity



- Advisory and Consultancy for Financial Analytics
- Prior Experience at MathWorks, Citigroup and Endeca and 25+ financial services and energy customers.
- Columnist for the [Wilmott Magazine](#)
- Author of forthcoming book [“Pragmatic AI and ML in Finance”](#)
- Teaches AI/ML and Fintech Related topics in the MS and MBA programs at [Northeastern University, Boston](#)
- **Reviewer:** Journal of Asset Management

# QuantUniversity

- Boston-based Data Science, Quant Finance and Machine Learning training and consulting advisory
- Trained more than 1000 students in Quantitative methods, Data Science and Big Data Technologies using MATLAB, Python and R
- Building  a platform for AI and Machine Learning Experimentation



Get the app  English  Log

QuantUniversity Meetup

[Home](#) [Members](#) [Sponsors](#) [Photos](#) [Discussions](#) [More](#)

[Join us!](#)



# Agenda

1. Data Science Pipelines & QuSandbox
2. Case study:
  - NLP Pipeline for Sentiment Analysis of EDGAR filings

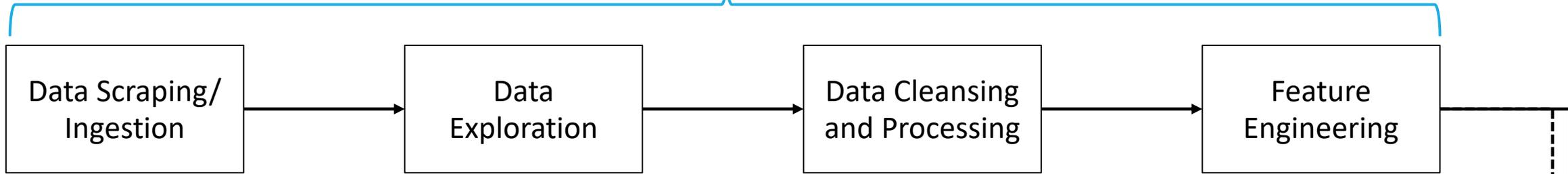


# Pipelines and the QuSandbox



# Machine Learning Workflow

Data Engineer, Dev Ops Engineer



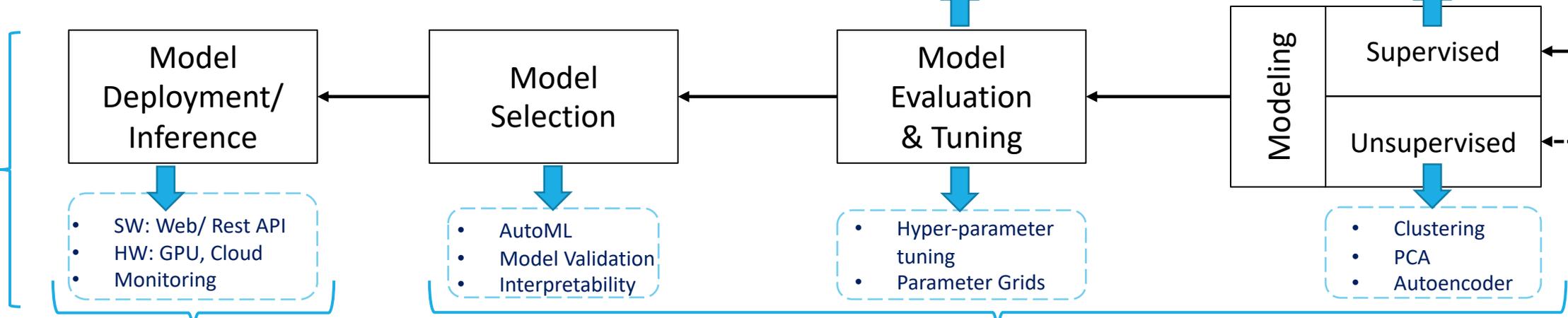
Robotic Process Automation (RPA) (Microservices, Pipelines)

 Risk Management/ Compliance(All stages)

- RMS
- MAPS
- MAE
- Confusion Matrix
- Precision/Recall
- ROC

- Regression
- KNN
- Decision Trees
- Naive Bayes
- Neural Networks
- Ensembles

Analysts & Decision Makers



Software/Web Engineer

Data Scientist/Quants

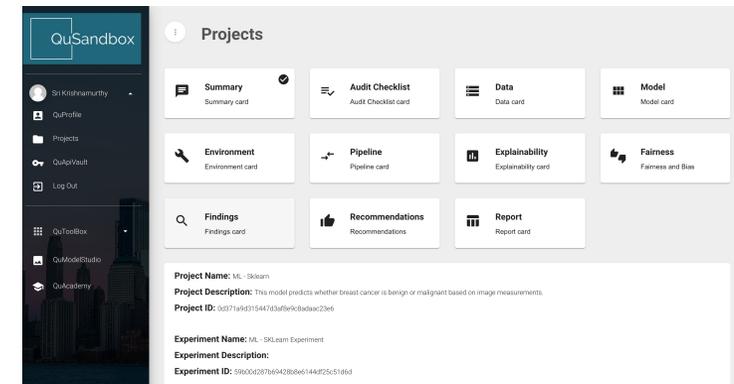
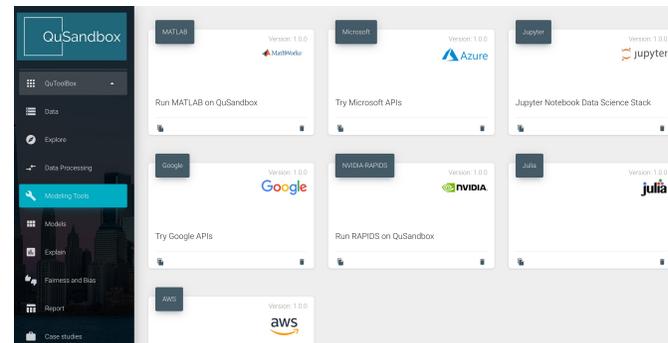
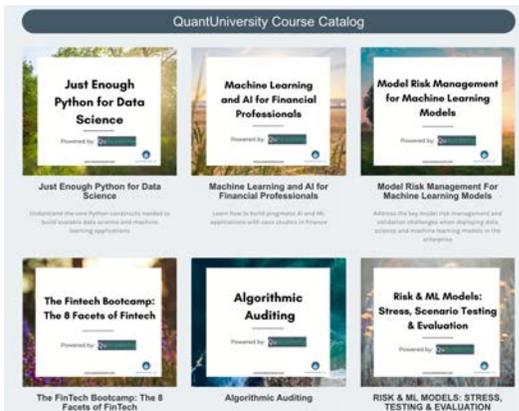


QuantUniversity, LLC  
www.quantuniversity.com

# The QU approach



Request DEMO at  
[info@qusandbox.com](mailto:info@qusandbox.com)





Filter By Courses

Risk and ML Models

Filter By Topics

All

Search Project

**RISK & ML MODELS**  
**STRESS, SCENARIO TESTING & EVALUATION**  
 1.0 Introduction to Machine Learning, AI and Risk

Presented By: Sri Krishnamurthy, CFA, CAP  
 QuantUniversity

Powered by: QuSandbox

2021 Copyright QuantUniversity LLC.

## 1. Introduction to ML, AI and Risk

Duration: 1 hr    QUCredits: 1

Introduction to Machine Learning, AI and Risk

[MORE DETAILS](#)

**RISK & ML MODELS**  
**STRESS, SCENARIO TESTING & EVALUATION**  
 2.0 Stress testing and Scenario generation

Presented By: Sri Krishnamurthy, CFA, CAP  
 QuantUniversity

Powered by: QuSandbox

2021 Copyright QuantUniversity LLC.

## 2. Stress testing and Scenario generation

Duration: 1 hr    QUCredits: 1

Introduction to Machine Learning, AI and Risk

[MORE DETAILS](#)

**RISK & ML MODELS**  
**STRESS, SCENARIO TESTING & EVALUATION**  
 3.0 Metrics and Evaluation for risk in ML models

Presented By: Sri Krishnamurthy, CFA, CAP  
 QuantUniversity

Powered by: QuSandbox

2021 Copyright QuantUniversity LLC.

## 3. Metrics and Evaluation for risk in models

Duration: 1 hr    QUCredits: 1

Introduction to Machine Learning, AI and Risk

[MORE DETAILS](#)

**RISK & ML MODELS**  
**STRESS, SCENARIO TESTING & EVALUATION**  
 5.0 Model Validation of ML models

Presented By: Sri Krishnamurthy, CFA, CAP  
 QuantUniversity

Powered by: QuSandbox

2021 Copyright QuantUniversity LLC.

## 4. Anomalies and Outliers

**RISK & ML MODELS**  
**STRESS, SCENARIO TESTING & EVALUATION**  
 6.0 Frontier topics and Wrap up

Presented By: Sri Krishnamurthy, CFA, CAP  
 QuantUniversity

Powered by: QuSandbox

2021 Copyright QuantUniversity LLC.

## 5. Model Validation of ML

**RISK & ML MODELS**  
**STRESS, SCENARIO TESTING & EVALUATION**  
 5.0 Model Validation of ML models

Presented By: Sri Krishnamurthy, CFA, CAP  
 QuantUniversity

Powered by: QuSandbox

2021 Copyright QuantUniversity LLC.

## 6. Frontier topics and Wrap



# QuSandbox

- QuToolBox
- Data
- Explore
- Data Processing
- Modeling Tools**
- Models
- Explain
- Fairness and Bias
- Report
- Case studies

**MATLAB** Version: 1.0.0  
  
 Run MATLAB on QuSandbox

**Microsoft** Version: 1.0.0  
  
 Try Microsoft APIs

**Jupyter** Version: 1.0.0  
  
 Jupyter Notebook Data Science Stack

**Google** Version: 1.0.0  
  
 Try Google APIs

**NVIDIA-RAPIDS** Version: 1.0.0  
  
 Run RAPIDS on QuSandbox

**Julia** Version: 1.0.0  
  
 Run Julia on QuSandbox

**AWS** Version: 1.0.0  
  
 Run AWS on QuSandbox

-  Sri Krishnamurthy ▲
-  QuProfile
-  Projects
-  QuApiVault
-  Log Out
-  QuToolBox ▼
-  QuModelStudio
-  QuAcademy

⋮ **Projects**

 <b>Summary</b> Summary card 	 <b>Audit Checklist</b> Audit Checklist card	 <b>Data</b> Data card	 <b>Model</b> Model card
 <b>Environment</b> Environment card	 <b>Pipeline</b> Pipeline card	 <b>Explainability</b> Explainability card	 <b>Fairness</b> Fairness and Bias
 <b>Findings</b> Findings card	 <b>Recommendations</b> Recommendations	 <b>Report</b> Report card	

**Project Name:** ML - Sklearn

**Project Description:** This model predicts whether breast cancer is benign or malignant based on image measurements.

**Project ID:** 0d371a9d315447d3af8e9c8adaac23e6

**Experiment Name:** ML - SKLearn Experiment

**Experiment Description:**

**Experiment ID:** 59b00d287b69428b8e6144df25c51d6d

QuSandbox

Sri Krishnamurthy

QuProjects

QuToolBox

Data

Explore

Data Processing

Modeling Tools

Models

Explain

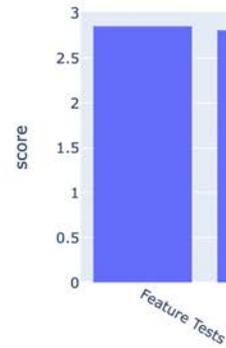
Fairness and Bias

**QUReport** Score Report credit risk

Id: 8cf00cf70ed9499a9b483362ad58eb4c  
Version: 1.1  
Date: 2021-05-14

**Information**  
 Experiment: 8cf00cf70ed9499a9b483362ad58eb4c  
 Owner: Sri Krishnamurthy  
 Contact: info@qusandbox.com  
 References: The ML Test Score: A Rubric for ML Production Readiness and Technical Debt Reduction

Final Score



QuSandbox

Sri Krishnamurthy

QuProjects

QuToolBox

QuAcademy

Findings Board

Recommendations Board

- TESTPLAN
- REPORTS**
- NOTES
- ISSUES

Reports

Name \*

Version \*

Test Plan

- Document the test plan for the report
- How is the model tested (Full, sampled, regression tested)
- Comment on the testing infrastructure (CI/CD, test suites)

# Case study



# Goal

- Understanding sentiments in Earnings call transcripts



## CORPORATE PARTICIPANTS

**Dana Quattrochi** athenahealth, Inc. - IR  
**Jonathan Bush** athenahealth, Inc - Chairman and CEO  
**Tim Adams** athenahealth, Inc - CFO  
**Andy Hurd** Epocrates - President and CEO  
**Rob Cosinuke** athenahealth, Inc. - Chief Marketing Officer

## CONFERENCE CALL PARTICIPANTS

**Sean Wieland** Piper Jaffray & Co. - Analyst  
**Jamie Stockton** Wells Fargo Securities, LLC - Analyst  
**George Hill** Citigroup - Analyst  
**Greg Bolan** Sterne, Agee & Leach - Analyst  
**Ryan Daniels** William Blair & Company - Analyst  
**Rich Close** Avondale Partners - Analyst  
**Sandy Draper** Raymond James - Analyst  
**David Bayer** Northland Securities - Analyst  
**Dave Windley** Jefferies & Co. - Analyst  
**Charles Rhyee** Cowen and Company - Analyst  
**Bret Jones** Oppenheimer & Co. - Analyst  
**Michael Cherny** ISI Group - Analyst  
**Tony Bartsch** Park West Asset Management - Analyst

## PRESENTATION

### Operator

Welcome to the athenahealth conference call. I would now like to turn the call over to Ms. Dana Quattrochi. You may now begin.

**Dana Quattrochi** - *athenahealth, Inc. - IR*

Good morning and thank you for joining us. With me on the call today is Jonathan Bush, our Chairman and CEO; Tim Adams, our Chief Financial Officer; Rob Cosinuke, our Chief Marketing Officer; and Andy Hurd, President and CEO of Epocrates.



# Challenges

- Interpreting emotions
- Labeling data



# NLP pipeline

Stage 1

Data Ingestion  
from Edgar

Stage 2

Pre-Processing

Stage 3

Invoking APIs to  
label data

Stage 4

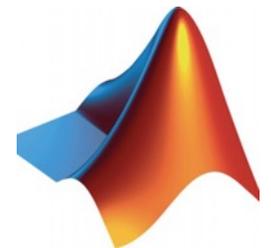
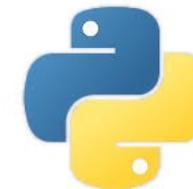
Compare APIs

Stage 5

Build a new  
model for  
sentiment  
Analysis



- Amazon Comprehend API
- Google API
- Watson API
- Azure API





# THE NEW DECALOGUE

ML GOVERNANCE IN THE AGE OF DATA SCIENCE AND AI  
WWW.QUANTUNIVERSITY.COM

## OUR SERVICES

- Model Governance and Algorithmic Audits
- Model Life Cycle Management
- AI/ML Onboarding
- Third-party Model Validation
- Training and Education
- QuSandbox

## ABOUT US

QuantUniversity is a quantitative analytics and model risk advisory based in Boston, MA. We provide Quant Finance, Data Science and Machine Learning based solutions focused on model-risk made accessible through QuSandbox. QuantUniversity has worked with analysts and executives from Bloomberg, Fidelity, Ford, Goldman Sachs, IBM, J.P. Morgan Chase, Nataxis Global Advisors, Pan Agora, T.D. Securities, and other institutions, providing quantitative advisory services, analytics training, and model risk solutions. Contact us at [www.quantuniversity.com](http://www.quantuniversity.com)

## 10 THINGS YOU NEED TO KNOW ABOUT MODEL GOVERNANCE FOR AI/ML MODELS

### 1. DEFINING MODELS

Models are not just restricted to code and associated parameters. You have to factor data, the programming environment and packages, parameters and hyperparameters along with the model code.

### 2. GOVERNING MACHINE LEARNING MODELS

You could have hundreds of machine learning models working alongside traditional models. A comprehensive framework is needed to factor the nuances of machine learning models in your governance process.

### 3. MODEL VERIFICATION AND VALIDATION OF MACHINE LEARNING MODELS

It's not just sufficient to verify if machine learning models work with historical test/validation datasets from a technical perspective. You have to validate if the models can be used for business decision making.

### 4. PERFORMANCE METRICS AND EVALUATION CRITERIA

The choice of performance metrics and evaluation criteria depends on how the models would be used and for what purpose. Evaluate the choices carefully.

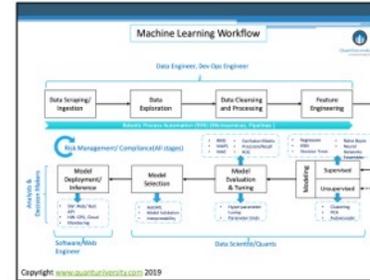
### 5. MODEL INVENTORY AND TRACKING

Avoid model "clutter" by having a formal model inventory and tracking system. You need to track models, data snapshots, parameters, hyperparameters, programming environments etc. In addition, the entire pipeline needs to be tracked. Provenance tracking is important for reproducibility.



# THE NEW DECALOGUE

ML GOVERNANCE IN THE AGE OF DATA SCIENCE AND AI  
WWW.QUANTUNIVERSITY.COM



## 10 THINGS YOU NEED TO KNOW ABOUT MODEL GOVERNANCE FOR AI/ML MODELS (CONT'D)

### 6. DATA GOVERNANCE AND MODEL GOVERNANCE

Machine Learning models are by design data driven. Integrating Data governance and model governance aspects is essential.

### 7. DEVELOPMENT MODELS VS PRODUCTION MODELS

As you design models for inference, scalability, performance considerations need to be factored. Models may have to be redesigned/compiled to factor production requirements. It is important to test models to ensure production models behave as they were designed.

### 8. FAIRNESS, REPRODUCIBILITY, AUDITABILITY, EXPLAINABILITY, INTERPRETABILITY, BIAS

Depending on the application, models should be evaluated to ensure Fairness, Reproducibility, Auditability, Explainability, Interpretability & Bias considerations are met.

### 9. MACHINE LEARNING CHOICES

As the field of machine learning matures, you have multiple options, Automatic Machine Learning, ML as a service, Pre-trained models and models developed from scratch etc. bringing different model governance considerations.

### 10. ROLES AND RESPONSIBILITIES

With AI and ML making strides, you have many new roles in your model building workflow. (Data engineers, scientists, model evaluators, cloud engineers, DevOps, MLOps etc.). Factor the new roles and define clear responsibilities for all the key stakeholders in the model lifecycle.



INFO@QUSANDBOX.COM



WWW.QUSANDBOX.COM



617-283-7904



INFO@QUSANDBOX.COM



WWW.QUSANDBOX.COM



617-283-7904

-  Sri Krishnamurthy ▲
-  QuProfile
-  Projects
-  QuApiVault
-  Log Out
-  QuToolBox ▼
-  QuModelStudio
-  QuAcademy

⋮ **Projects**

 <b>Summary</b> Summary card 	 <b>Audit Checklist</b> Audit Checklist card	 <b>Data</b> Data card	 <b>Model</b> Model card
 <b>Environment</b> Environment card	 <b>Pipeline</b> Pipeline card	 <b>Explainability</b> Explainability card	 <b>Fairness</b> Fairness and Bias
 <b>Findings</b> Findings card	 <b>Recommendations</b> Recommendations	 <b>Report</b> Report card	

**Project Name:** ML - Sklearn

**Project Description:** This model predicts whether breast cancer is benign or malignant based on image measurements.

**Project ID:** 0d371a9d315447d3af8e9c8adaac23e6

**Experiment Name:** ML - SKLearn Experiment

**Experiment Description:**

**Experiment ID:** 59b00d287b69428b8e6144df25c51d6d

**Request DEMO at**  
**[info@qusandbox.com](mailto:info@qusandbox.com)**



QuantUniversity, LLC

[www.quantuniversity.com](http://www.quantuniversity.com)

# Thank you!

## Contact

Sri Krishnamurthy, CFA, CAP  
Founder and CEO  
QuantUniversity LLC.

LinkedIn [srikrishnamurthy](#)

[www.QuantUniversity.com](http://www.QuantUniversity.com)

