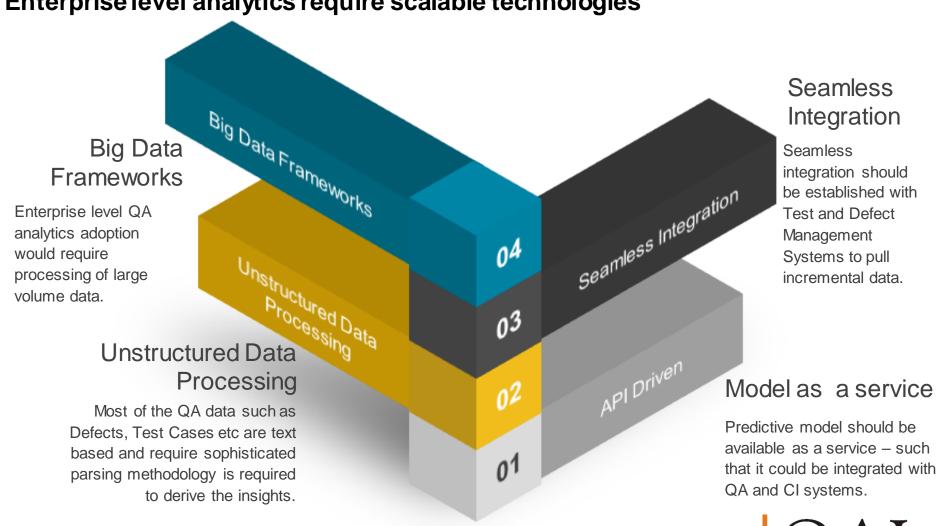


### A Deep Dive Into Best Practices





#### Enterprise level analytics require scalable technologies





## Collaboration between SMEs and Data scientist is the keyto create successful model

Study data dynamics and variations on defect reporting /test execution pattern over time

Build utilities that measure model evaluation metrics on live predictions and its impact on QA efficiency.

Vision

Business

Data Dynamics

Model for current QA release

**Prod Eval Metrics** 

Understand business vision coupled with ROI factor such as Improve Test Efficiency/Optimize Testing. Build model considering current QA release, rather than focusing analysis on historic releases.

- Data scientists should involve Test
   Manager and Leads on the milestone discussions and results evaluation.
- Consult Test Manager about historic data relevancy with current release.





#### Data Analytics - Garbage In, Garbage Out

Data must be **right**: it must be correct, labelled, processed etc; And you must have a **right** data.





**Relevant Data** 

Less relevant, nonrepresentative historic data of current QA release is not useful.



## System Migration Challenges

Migration of QA systems from one product to other impacts data quality. For ex: HP ALM to Jira migration, QTP to Selenium migration.



#### Extensive Feature Engineering

Extensive, complex grooming of features/attributes might result in a over-fitted model.



#### Data is first

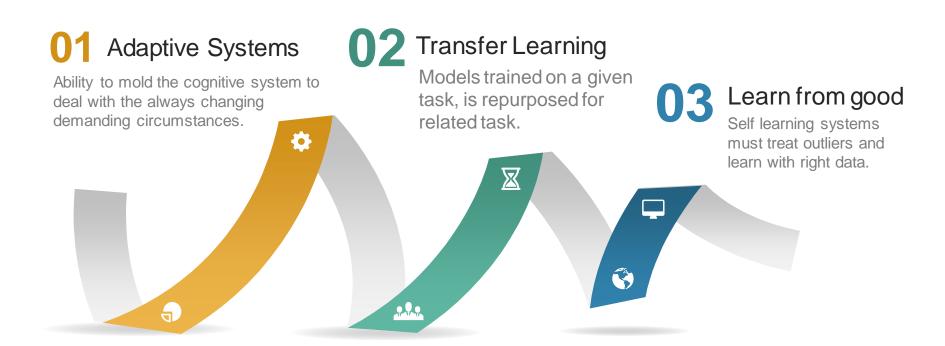
Sophisticated ML algorithms and processing techniques may not yield good results if data quality is poor.





#### Self Learning Systems – Continual Learning is key

Continual Learning is about adaptive learning and about autonomous incremental model development and deployment.







#### Operationalizing Analytics at scale

Successful operationalization strategy of Analytics is key to realizing the benefits of Machine Learning solutions.







Version control the models with hyperparameter and parameter information.



#### **Roll back Strategy**

Strategize rollback plans of analytics models in case of performance degradation.



#### **Disaster Recovery**

Periodically backup models and persist key information as part of disaster management.



#### Benchmarking & Scoring

Benchmark model performance and continuously monitor model scoring in production.



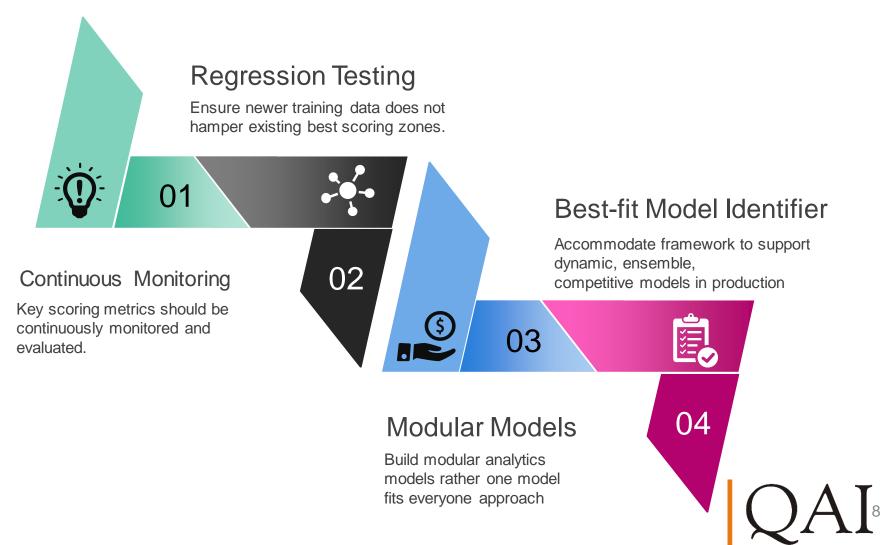
#### Accuracy is not a ONLY qualifying factor of analytics/prediction model.



QAI



# Sustaining Model Performance over time requires discipline and focused effort.





### **Automated Defect Triaging for a Telecom**

#### **Provider**

Leading US Telecomprovider was spending considerable human effort with regard to triaging activity of defects. It was looking forward to automate the process across enterprise leveraging the intelligence from historic defects.



Enterprise wide implementation using big data frameworks, processing 500K defects.



Continuous learning prediction models on a daily basis.

250 K
Defect

95+ %

Environment Uptime

Prediction response time of <2 seconds.



Competing Models Any day 1200+ Defects/Day



Seamless integration with Enterprise Defect Management System.



Model versioning, governance with sophisticated model promotion/rollback mechanism.





#### **Author Biography**



Vasanthkumar Velayudham Manager – Digital Assurance CoE

Engineer and a solution provider, passionate about solving customer problems with the aid of emerging technologies. Evolved from a manual test engineer into test automation architect, Web technologist, security testing engineer and into a Data scientist.



Dattaprasad Kulkarni Director – Digital Assurance CoE

Engineering lead for AI ML (Artificial Intelligence and Machine Learning) based bot products for digital assurance group. Possesses strong data analytics skills and specialized in executing large greenfield AD programs. Passionate about building Al-ML solutions.



Rajeswari Kumaran – Product Lead

Product lead with analytics wing of digital assurance group. Have handled several legacy modernization development projects. Have worked on QA consulting assignments for multiple customers. Interested in data science and leveraging its multiple avenues to solve business issues for customers



# **Q & A**







### Thank You!!!



