STEAMS and DMAIC Curriculum for Data Scientists Using JMP 16©

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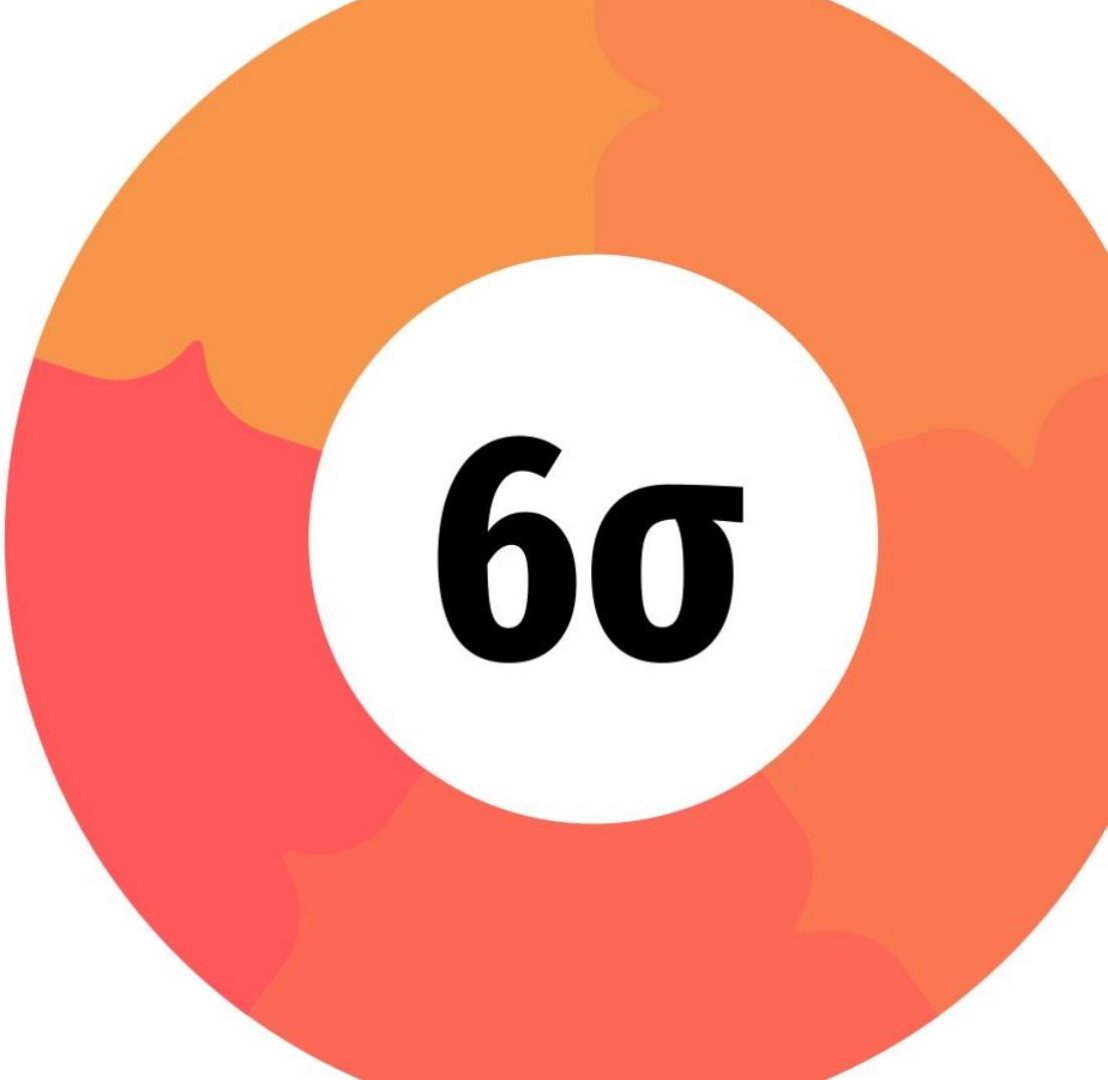
Stanford Online High School STEAMS Club



Overview

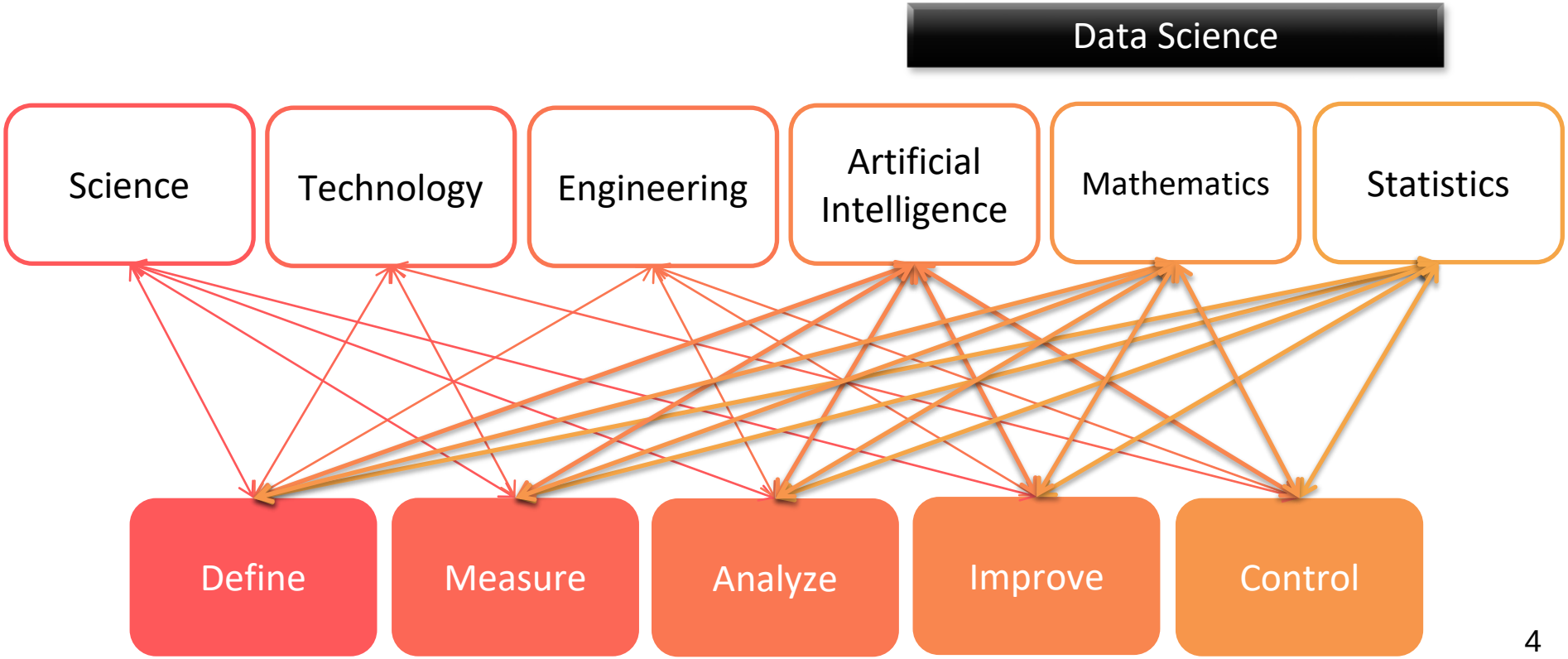
- **Opportunity statement** – the traditional Six Sigma DMAIC process combined with the interdisciplinary STEAMS methodology can help data scientists make greater contributions in the field of Big Data
- **Project objective** – develop a Six Sigma data science training curriculum for high schoolers to industry professionals by mapping JMP 16© platforms onto DMAIC phases

Case Study: Mason Chen's Learning Experience

- 2015 **Big Data Statistics** Summer Camp (10 years old)
- 2016 May IBM **SPSS** Statistics Certified (10 years old)
- 2016 August IASSC **Minitab** DMAIC **Black Belt** Certified (11 years old)
- 2016 August ASQ/ASA/JMP Joint Annual **STEAMS** Speaker (11 years old)
- 2016 October IBM **Modeler Data Mining** Certified (11 years old)
- 2017 April IEOM Rabat **DMAIC EV3** Robotics Best Paper Award (11 years old)
- 2017 April IEOM Rabat **Java** Best Paper Award (11 years old)
- 2017 August Found **STEAMS Organization** (12 years old)
- 2018 October **JMP** USA DS Best Contributed Paper Award (13 years old)
- 2019 Youngest **IEEE** Presenter, **JMP** Principal Component & Clustering (13 years old)
- 2020 **JMP STIPS** Certification- **Data Mining** (14 years old)
- 2020 Learning **JMP** DOE Cert Exam (15 years old)
- 2021 March **JMP** Europe Discovery Summit Best Student Poster Award (15 years old)
- 2021 Learning **JMP** 16 Text Mining and Time Series Forecast (15 years old)
- 2021 June Stanford Summer Course: **Linear Algebra** (16 years old)
- 2021 August Stanford OHS **Data Science R** Course (16 years old)
- **2021 August JMP 16_Based Six Sigma Data Science Program (16 years old)**
- 2021 September **R-Based** Six Sigma DMAIC Statistics Curriculum (16 years old)

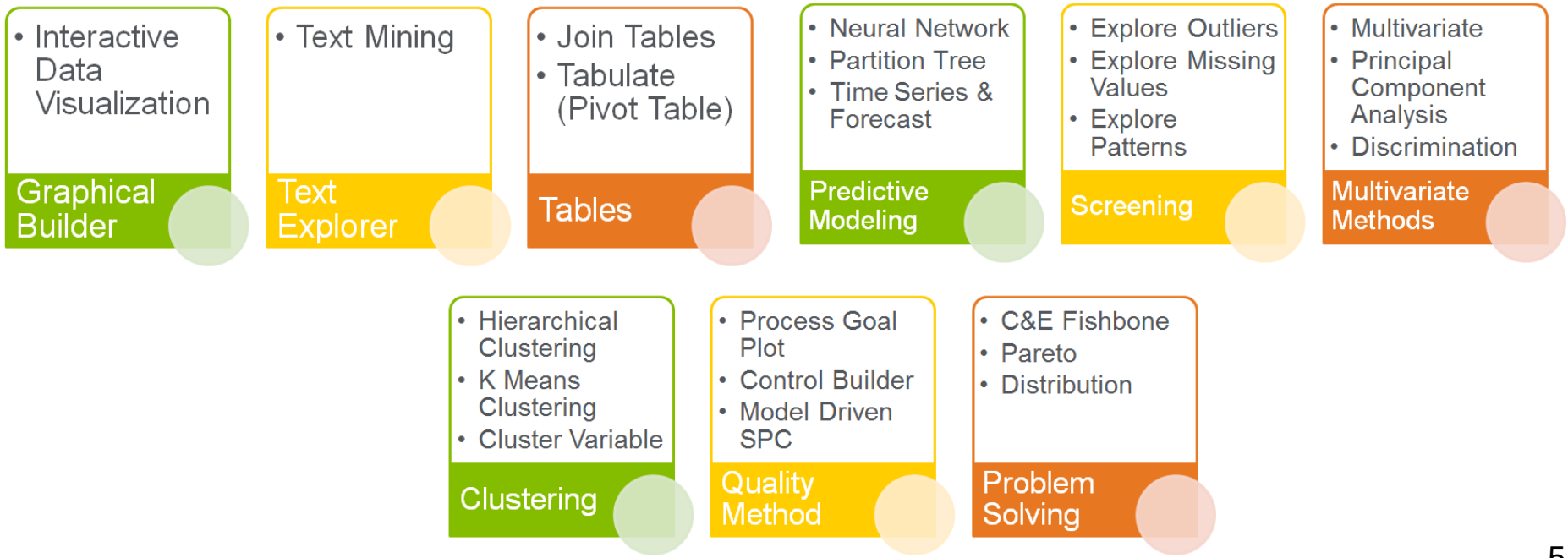
2019 ASA JSM Denver Conference at Rocky Mountain Summit (12,005 ft)

Connecting STEAMS and DMAIC



Data Science JMP 16 Platforms

- Map JMP Platforms to Data Science Certification Program
- Based on Big Data 3Vs: Volume, Variety, and Velocity.



JMP 16 Data Science Statistics

- **DMAIC quality and reliability** – measurement systems analysis (MSA), process capability, statistical process control (SPC), lot acceptance sampling
- **DFSS design modeling** – analysis of variance (ANOVA), regression, design of experiment (DOE), Monte Carlo simulation, robust tolerance
- **Linear algebra** – eigen analysis, principal component analysis (PCA), factor analysis, singular value decomposition (SVD)
- **Data mining** – classification, neural network, partition trees, random forest
- **Time series and forecasting** – time series decomposition, autoregressive integrated moving average (ARIMA) models, forecasting
- **Text mining** – stemming, recoding, tokenization, phrases
- **Survey and consumer research** – sampling plans, choice model, MaxDiff model, marketing segmentation

Six Sigma DMAIC Data Science Curriculum

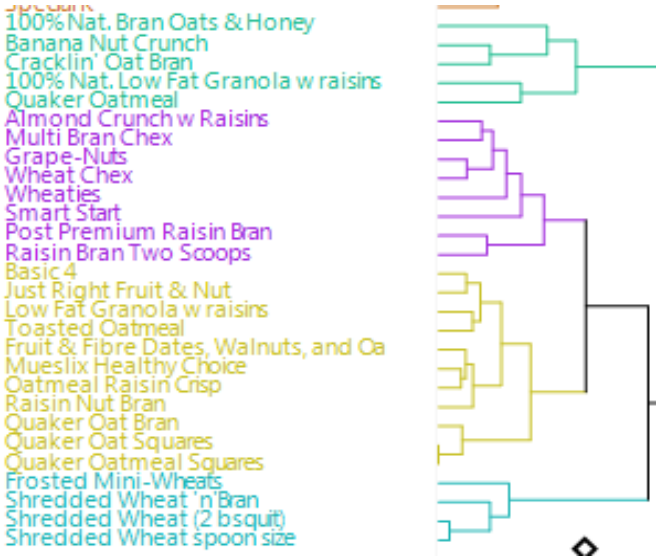
JMP 16 Platforms	A. Regular DMAIC BB	B. Data Mining	C. Text Mining and Categorial
LSS BB 03 Statistics	Basic Statistics, Distributions, Prospective Sample Size and Power, Sample Size Explorer	Basic Data Science Statistics	
LSS BB 04 JMP Introduction	Discovering JMP Book	Using JMP Book	
LSS BB 09 Measure M2 MSA	MSA Design, Variability and Attribute Gauge Charts,		
LSS BB 10 Measure M3 PCA	Process Capability	Quality Utility, Process History Explorer	
LSS BB 13 Analyze A2	Essesstial Graphing Book Part I, Pareto Chart, Cause and Effect Diagram	Essential Graphing Book Part II	Essential Graphing Book Part II, FMEA Plus
LSS BB 14 Analyze A3	Bivariate, Oneway,Contingency, Tabulate, Modeling Utilities, Fit Model (LS, Stepwise), Matched Pairs	Tabulate Plus, Modeling Utilities Plus, Multivariate Correlations, Principal Componets, Discriminat, Partial Least Square, Factor Analysis, Multi Dimensional Scaling, Item Analysis, Hierarchical Clustering, K Means Clustering, Normal Mixture Clustering, Cluster Variables	Tabulate Plus, Text Explorer, Modeling Utilities Plus, Multiple Correspondence, Two-Way Hierarchical Clustering, Latent Class Clustering, Categorical Response
LSS BB 16 Improve I1	Prediction Profiler	Custom Profiler, Excel Profiler, Multiple Factor Analysis	Choice Design, Choice Model, MaxDiff Design, MaxDiff Model,
LSS BB 18 Improve I3	Custom DOE, DSD, Simulator	Neural Network, Partition,Time Series Analysis and Forecast, Response Screening, Process Screening, Predictor Screening	Neural Network, Partition
LSS BB 20 Control A1 SPC	Control Chart Builder	Multivariate and Model Driven Control Chart	

Three training programs – standard, data mining, and text mining

Define Phase

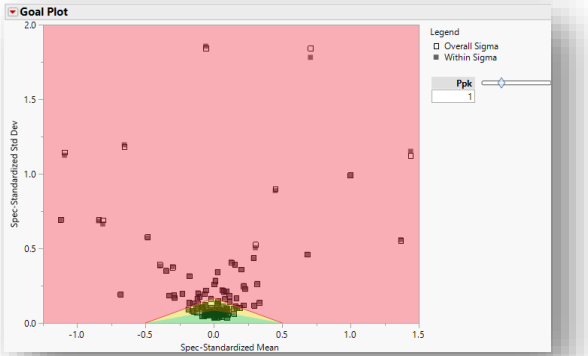
- **Overview**
  - Problem statement (voice of the customer, voice of business)
  - Project goal and objective (critical to quality)
  - Success criteria (specification limits)
  - Team building (forming, storming, norming, performing)

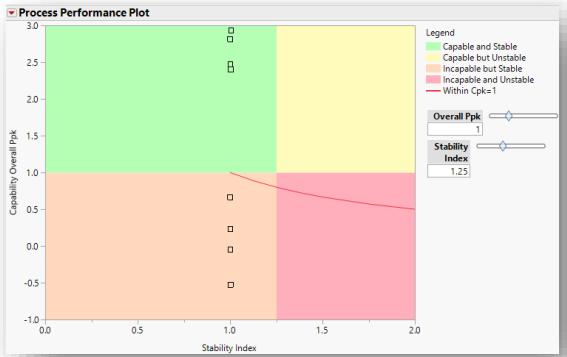
- **JMP 16© platforms**
  - Build JMP database – query builder
  - Data visualization – graph builder, Pareto plot, bubble plot, variability plot
  - Data mining – clustering, multivariate, and partition methods
  - Marketing Research: Consumer Research

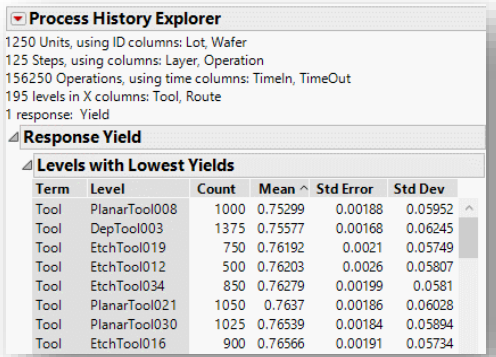


Measure Phase

- Several JMP Platforms can help visualize or/and summarize **Process Capability** and **Process Stability** Index of larger scale MFG production







Goal Plot

How well variables are conforming to specification limits

Process Performance Plot

Divided based on process capability and stability

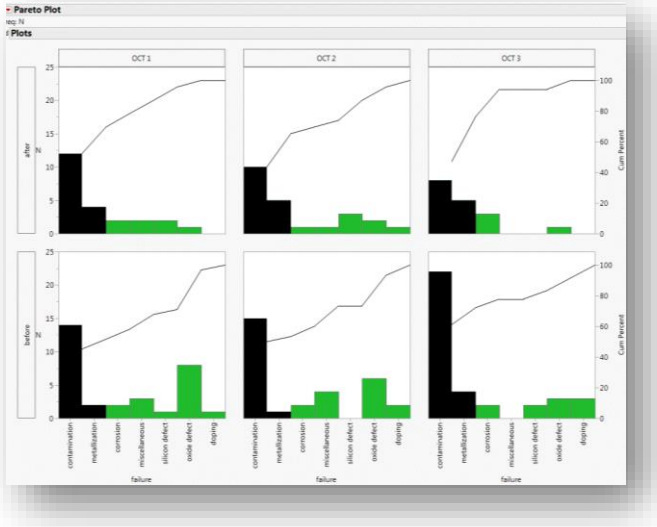
Process History Explorer

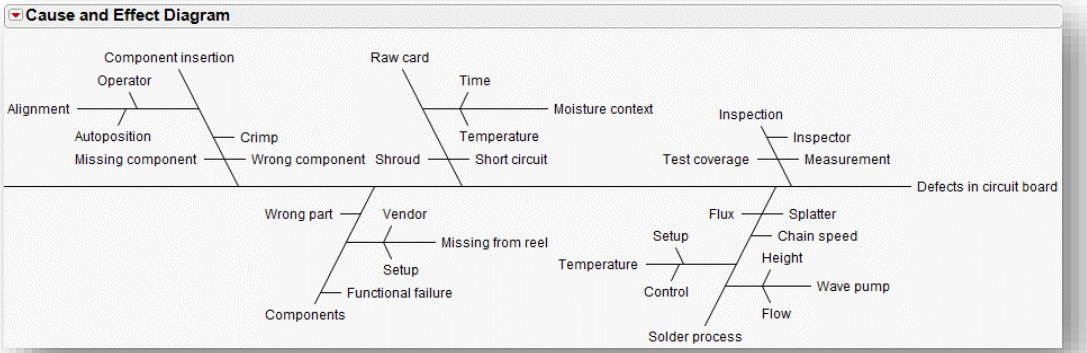
Helps identify factors associated with poor yield

Analyze Phase

- **Overview**
  - Root cause analysis
  - Summarize complex datasets
  - Visualize and discover patterns and insights
  - Isolate and screen for important factors
- **JMP 16© platforms**
  - Root cause analysis – fishbone diagram
  - Table summary – tabulate
  - Text mining – text explorer
  - Multivariate methods – multivariate correlation, factor analysis
  - Clustering – hierarchical clustering, k-means clustering, cluster variables
  - Survey and consumer research – categorical response analysis

Analyze: Identifying the Root Cause





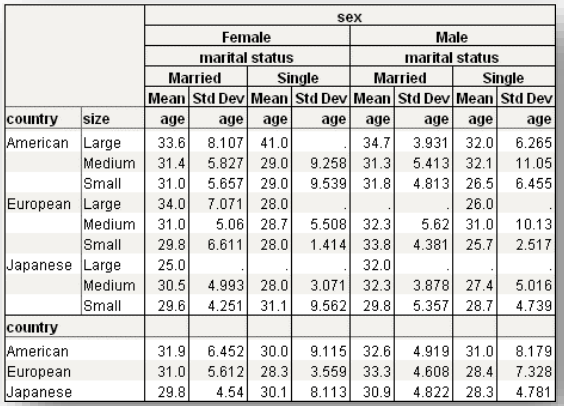
Pareto plot

Highlights the severity of different problems

Fishbone diagram

Brainstorm and organize sources of the problem

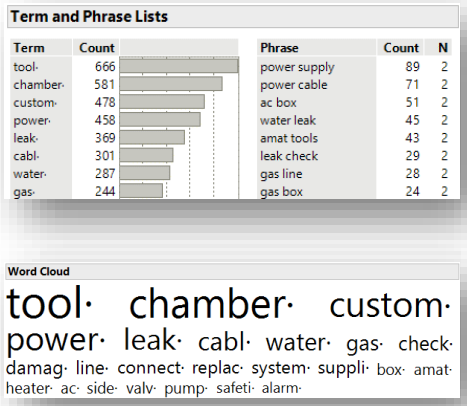
Analyze: Data Summarization



		sex							
		Female				Male			
		marital status				marital status			
		Married		Single		Married		Single	
		Mean	Std Dev	Mean	Std Dev	Mean	Std Dev	Mean	Std Dev
country	size	age	age	age	age	age	age	age	age
American	Large	33.6	8.107	41.0	.	34.7	3.931	32.0	6.265
	Medium	31.4	5.827	29.0	9.258	31.3	5.413	32.1	11.05
	Small	31.0	5.657	29.0	9.539	31.8	4.813	26.5	6.455
European	Large	34.0	7.071	28.0	.	.	.	26.0	.
	Medium	31.0	5.06	28.7	5.508	32.3	5.62	31.0	10.13
	Small	29.8	6.611	28.0	1.414	33.8	4.381	25.7	2.517
Japanese	Large	25.0	.	.	.	32.0	.	.	.
	Medium	30.5	4.993	28.0	3.071	32.3	3.878	27.4	5.016
	Small	29.6	4.251	31.1	9.562	29.8	5.357	28.7	4.739
country									
American		31.9	6.452	30.0	9.115	32.6	4.919	31.0	8.179
European		31.0	5.612	28.3	3.559	33.3	4.608	28.4	7.328
Japanese		29.8	4.54	30.1	8.113	30.9	4.822	28.3	4.781

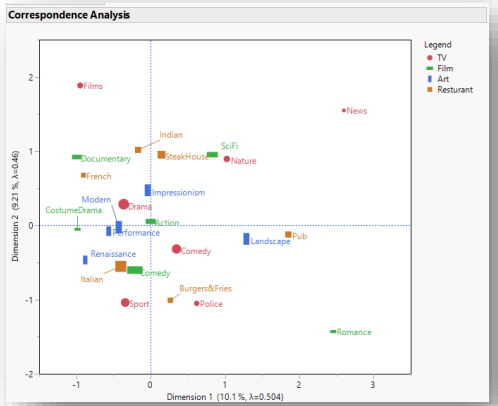
Tabulate

Descriptive statistics and pivot tables



Text Explorer

Analyzes patterns between unstructured text



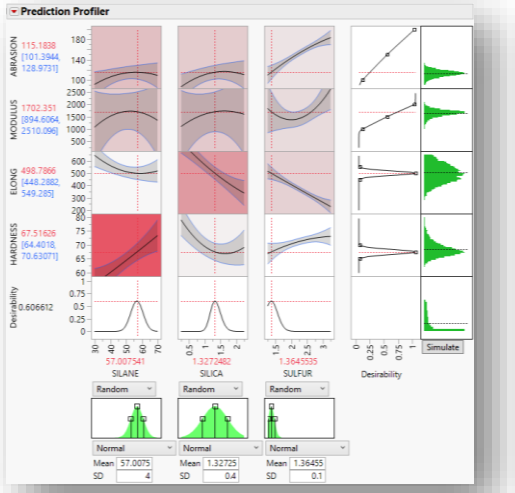
Multiple Correspondence Analysis

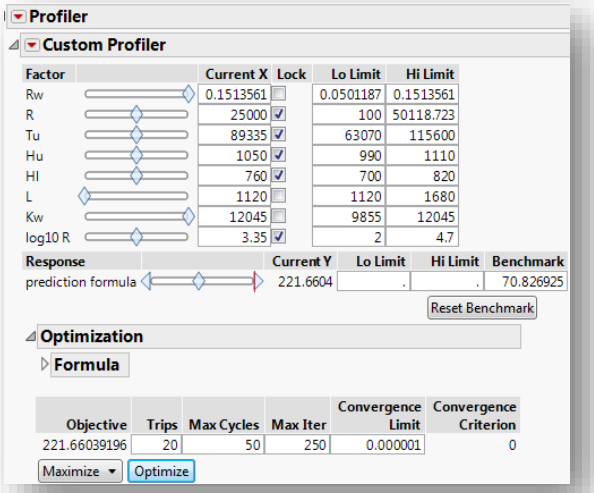
Associations between categorical levels

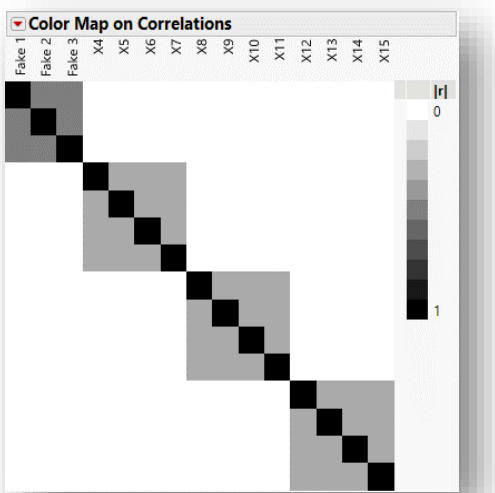
Improve Phase

- **Overview**
  - Build predictive models
  - Design new experiments
  - Improve production quality
- **JMP 16© platforms**
  - Predictive modeling – prediction profiler, custom profiler
  - Design of experiment (DOE) – custom DOE, mixture DOE, group orthogonal supersaturated designs, augmentation
  - Specialized models – neural network, partition model, response screening, process screening, predictor screening
  - Survey and consumer research – choice model, MaxDiff design

Improve: Design Optimization







Prediction Profiler

Studies response distribution and factor sensitivity

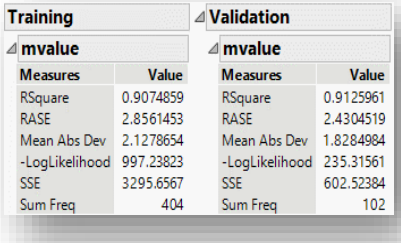
Custom Profiler

Finds optimal factor settings without graphs

Group Orthogonal Supersaturated

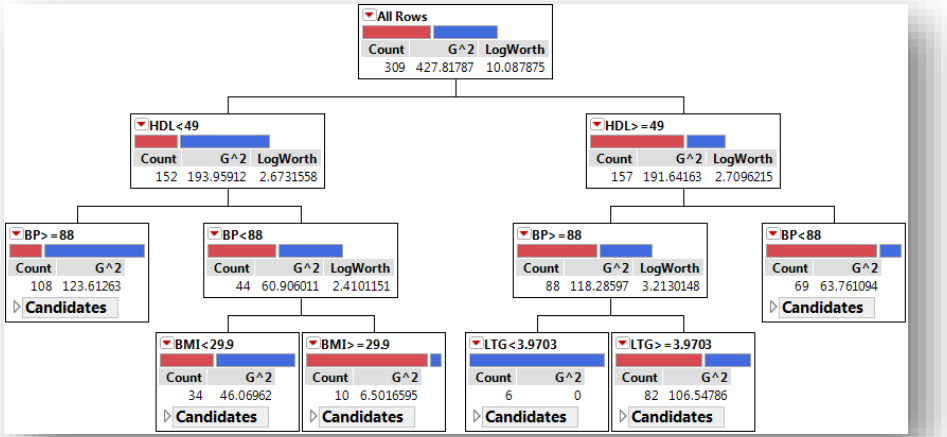
For designs with a greater number of factors than runs

Improve: Predictive Models



Neural network

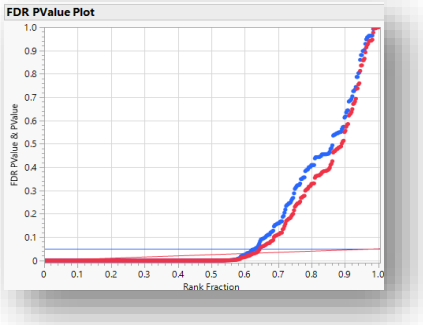
Uses a transfer function to predict response variables

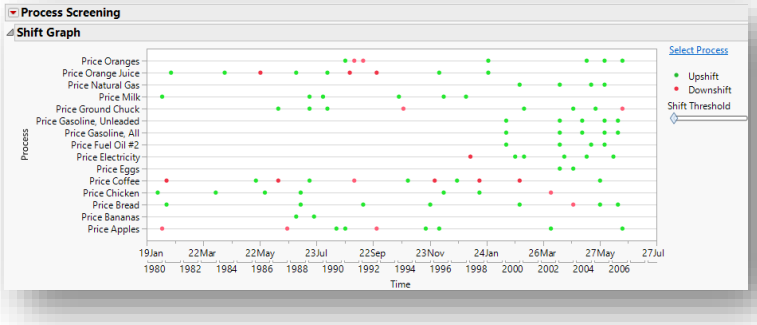


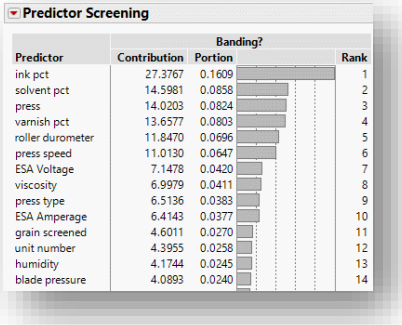
Partition

Creates a decision tree by recursively partitioning data

Improve: Design Optimization







Response Screening

Features that aid the analysis of large datasets

Process Screening

Process capability and stability for many responses

Predictor Screening

Ranks predictors using bootstrap forest partitioning

Improve: Consumer Research

	Choice Set	Choice ID	Disk Size	Speed	Battery Life	Price
1	1	1	40 GB	2.0 GHz	6 Hrs	$1000
2	1	2	80 GB	1.5 GHz	4 Hrs	$1500
3	2	1	80 GB	1.5 GHz	4 Hrs	$1200
4	2	2	40 GB	1.5 GHz	6 Hrs	$1500
5	3	1	40 GB	1.5 GHz	4 Hrs	$1200
6	3	2	80 GB	2.0 GHz	6 Hrs	$1500
7	4	1	40 GB	2.0 GHz	4 Hrs	$1000
8	4	2	80 GB	1.5 GHz	6 Hrs	$1500
9	5	1	80 GB	2.0 GHz	6 Hrs	$1200
10	5	2	40 GB	1.5 GHz	6 Hrs	$1000
11	6	1	40 GB	1.5 GHz	4 Hrs	$1500
12	6	2	80 GB	1.5 GHz	6 Hrs	$1000

	Subject	Choice Set	Candy	Choice
1	1	1	Reese's Cups	•
2	1	1	Hershey Bar	•
3	1	1	Snickers	•
4	1	1	Butterfinger	•
5	1	2	Butterfinger	•
6	1	2	Heath Bars	•
7	1	2	Plain M&Ms	•
8	1	2	Snickers	•
9	1	3	Plain M&Ms	•
10	1	3	Snickers	•
11	1	3	Peanut M&Ms	•
12	1	3	Hershey Bar	•

MaxDiff Results

Marginal Utility	Marginal Probability	Candy
0.9027	0.2664	Plain M&Ms
0.6765	0.2125	Reese's Cups
0.6599	0.2090	Peanut M&Ms
0.2466	0.1382	Heath Bars
0.0302	0.1113	Hershey Bar
-1.061	0.0374	Butterfinger
-1.455	0.0252	Snickers

Parameter Estimates

Likelihood Ratio Tests

Source	L-R ChiSquare	DF	Prob>ChiSq
Candy	39.769	6	<.0001*

Choice Design

Used to find the best combination of features

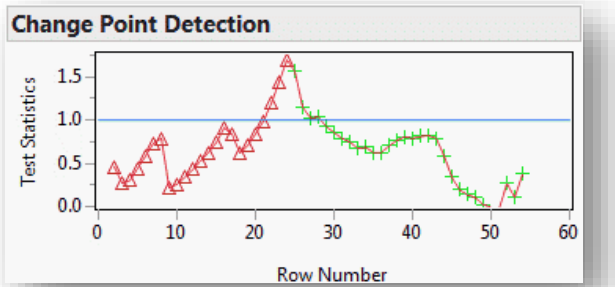
MaxDiff Design

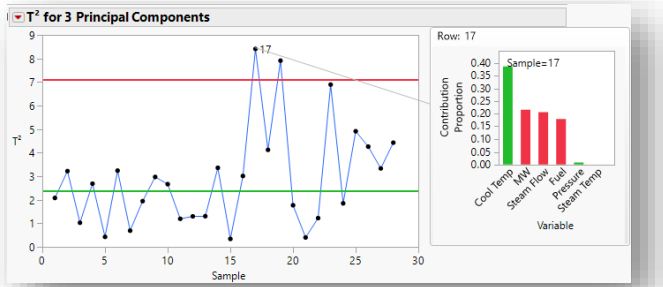
Only considers most and least preferred items

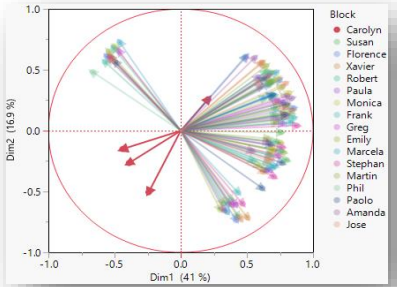
Control Phase

- **Overview**
  - Scale-Up Process Control
  - Sustain Improvement over long period
  - Upstream-Downstream Multivariate Process Control
- **JMP 16© platforms**
  - **Classical Control Charts**: Control Chart Builder
  - **Time Sensitive Control Charts**: CUSUM, EWMA Control Charts
  - **Multivariate Control Charts**: T2 Control Chart, Model Driven Multivariate Control Chart
  - **Consumer Research**: Multiple Factor Analysis
  - **Time Series Ana**lysis: Time Series Decomposition and Smoothing, ARIMA, Forecasting

Control: Multivariate Tools







Change Point Detection plot

Detects a shift in the mean by dividing the data

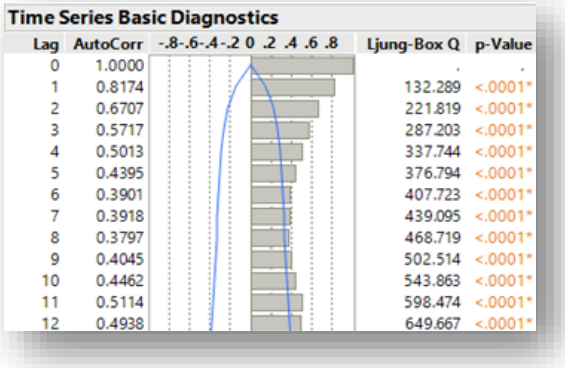
T Square chart

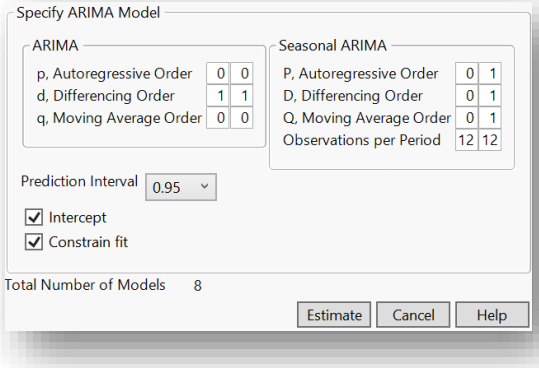
Uses principal components for process stability

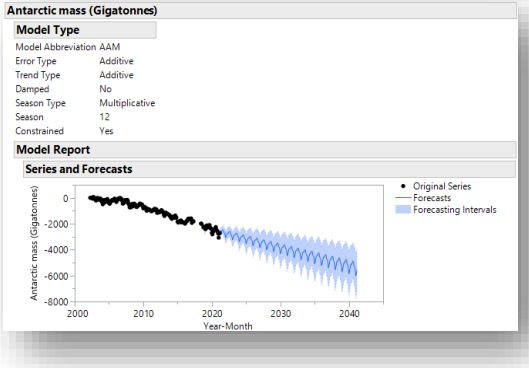
Multiple factor analysis

Uses eigenvalue decomposition to compare items

Control: Time Series Techniques







Model diagnostics

Identify trend, seasonal, and cyclic components

ARIMA models

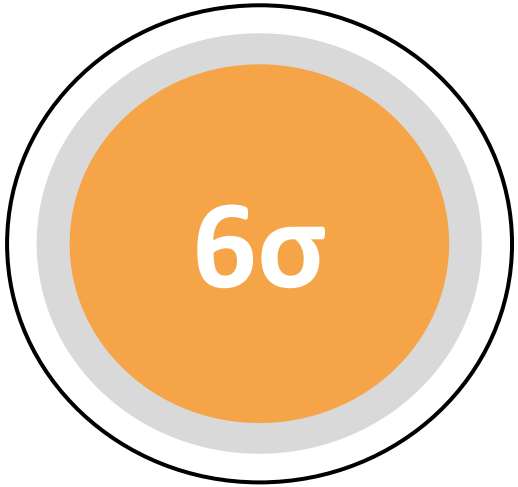
Fits data using seasonal or non-seasonal methods

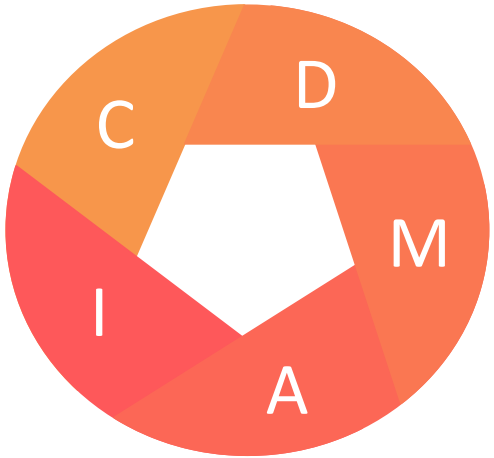
Forecasting

Finds the optimal model to predict future points

Takeaways

- Traditional **Six Sigma DMAIC** and Interdisciplinary **STEAMS** methods can help develop Data Scientist on leadership and team building
- **Modern JMP 16** platforms are mapped to DMAIC Phases to help deploy Six Sigma Projects in **Data Science** fields
- **Database Management, Applied Engineering Statistics, Data Mining and Text Mining** are all critical to today's Data Scientific Analytics





Thanks!