

WINDPOWER2012

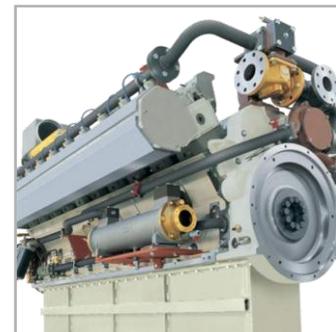
# Analytic Approaches for Improving Wind Farm Operations & Maintenance

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Atlanta, GA

Information provided in this presentation is based on data that has been scaled and made anonymous to protect proprietary information. This does not impact the overall methodology or capabilities of the tools and algorithms used.



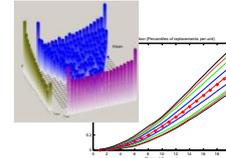
# Analytics for Physical Asset Management

## Technology / Capability

## Business Impact

Decision Optimization

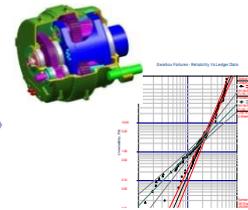
- Long term service agreements
- Optimize Spares & Logistics
- Optimize Maintenance
- Risk-Based Capital Budgeting



- Optimal capital outlays
- Minimal O&M costs
- Maximize financial returns
- Enterprise Risk Mgmt.

Reliability & Life Mgmt.

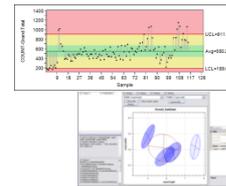
- Weibull Analysis
- Prognostics Health Mgmt.
- Life Odometers
- Performance-Life Tradeoff



- Identify "high risk" subpopulations in fleet
- Prioritize maintenance
- Safety .. Low OPEX

Anomaly Detection

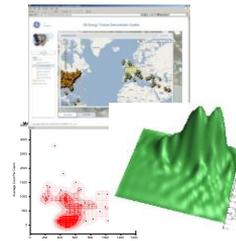
- Statistical Process Control
- Fault Detection
- Fault Root Cause Diagnosis
- Sensor & Data Fusion



- Reduce unplanned maintenance costs
- Improve availability
- Improve performance

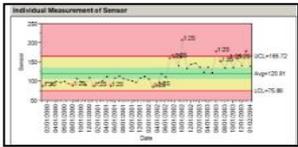
Data Mining & Reporting

- Statistical Summaries Of Performance & Reliability
- Unit Vs. Fleet Comparisons
- Mission & Usage Analysis

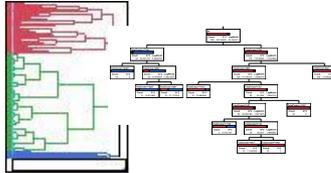
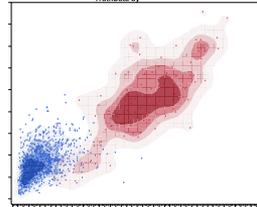
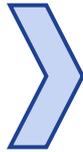


- Quantify costs, resources
- Enable new O&M deals
- Define & manage programs

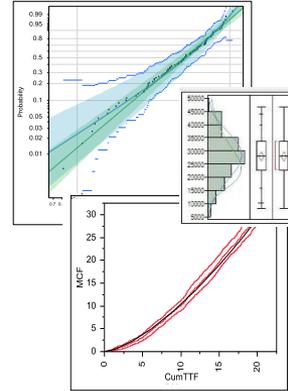
# O&M Analytics Approach



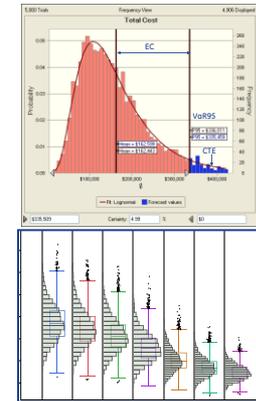
Text & Data Mining  
on Services/repair  
data



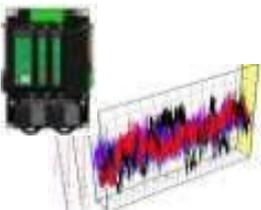
Fleet Segmentation,  
clustering &  
classification



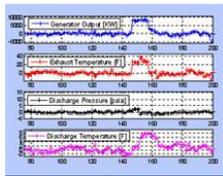
Survival Analysis  
Recurrent Event Models  
Partial Renewal Models



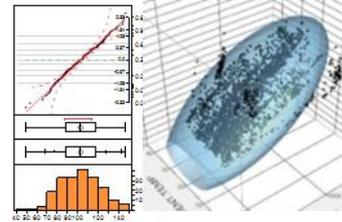
Scenario and  
Simulation-  
based  
optimization of  
O&M costs,  
spares, crews,  
RCM and CBM  
system design



Sensor Data



Multiple anomaly  
Detection



Prognosis



# Case Study Overview

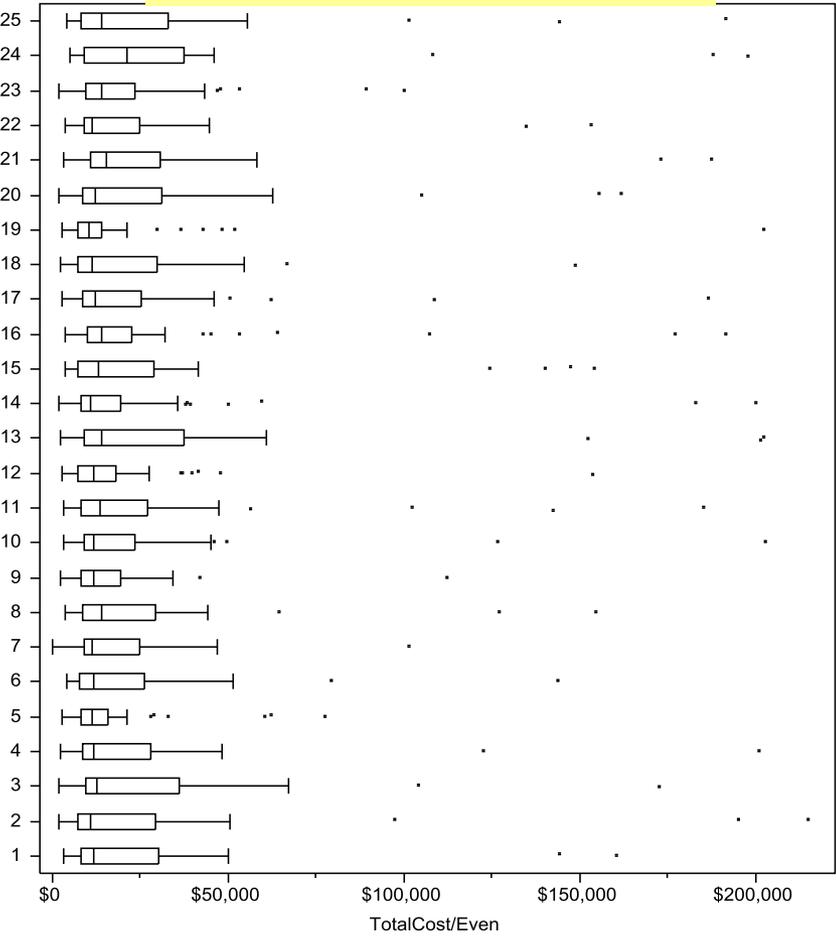
- A fleet of 25 units tracked for ~ 20 years from commissioning in early 1990 until end of observation in 2010.
- Operations and maintenance records are available and events manually classified as Major Repair, Minor Repair or Replacement (establish “truth data”)
- Services data aggregated to include event date, Turbine ID, Total parts costs (\$/event) , total labor hours and costs (\$/event) and total event costs
- Intent was to demonstrate how data-mining methods can be used to extract information about the underlying repair and renewal processes .. and build a predictive model to statistically estimate future costs

	TotalCost/Event	X (Parts)	Y (Labor)	
UnitID	Sum	Sum	Sum	N
1	\$891,256.00	\$732,622.00	\$158,634.00	37
2	\$1,359,585.00	\$1,113,758.00	\$245,827.00	54
3	\$899,574.00	\$729,138.00	\$170,436.00	35
4	\$1,386,120.00	\$1,143,546.00	\$242,574.00	61
5	\$647,767.00	\$535,182.00	\$112,585.00	40
6	\$994,097.00	\$807,202.00	\$186,895.00	48
7	\$642,511.00	\$519,321.00	\$123,190.00	33
8	\$954,903.00	\$776,798.00	\$178,105.00	38
9	\$872,841.00	\$692,877.00	\$179,964.00	54
10	\$1,232,365.00	\$989,857.00	\$242,508.00	57
11	\$916,032.00	\$751,655.00	\$164,377.00	32
12	\$1,083,290.00	\$908,070.00	\$175,220.00	51
13	\$1,561,625.00	\$1,269,248.00	\$292,377.00	51
14	\$1,127,644.00	\$938,414.00	\$189,230.00	50
15	\$1,380,994.00	\$1,113,258.00	\$267,736.00	56
16	\$1,318,246.00	\$1,115,271.00	\$202,975.00	52
17	\$998,526.00	\$810,504.00	\$188,022.00	44
18	\$973,761.00	\$781,638.00	\$192,123.00	46
19	\$684,169.00	\$571,752.00	\$112,417.00	35
20	\$1,320,797.00	\$1,074,194.00	\$246,603.00	52
21	\$1,128,581.00	\$900,972.00	\$227,609.00	41
22	\$776,298.00	\$630,680.00	\$145,618.00	33
23	\$1,285,913.00	\$1,022,990.00	\$262,923.00	64
24	\$1,435,259.00	\$1,132,172.00	\$303,087.00	47
25	\$1,099,025.00	\$870,330.00	\$228,695.00	40

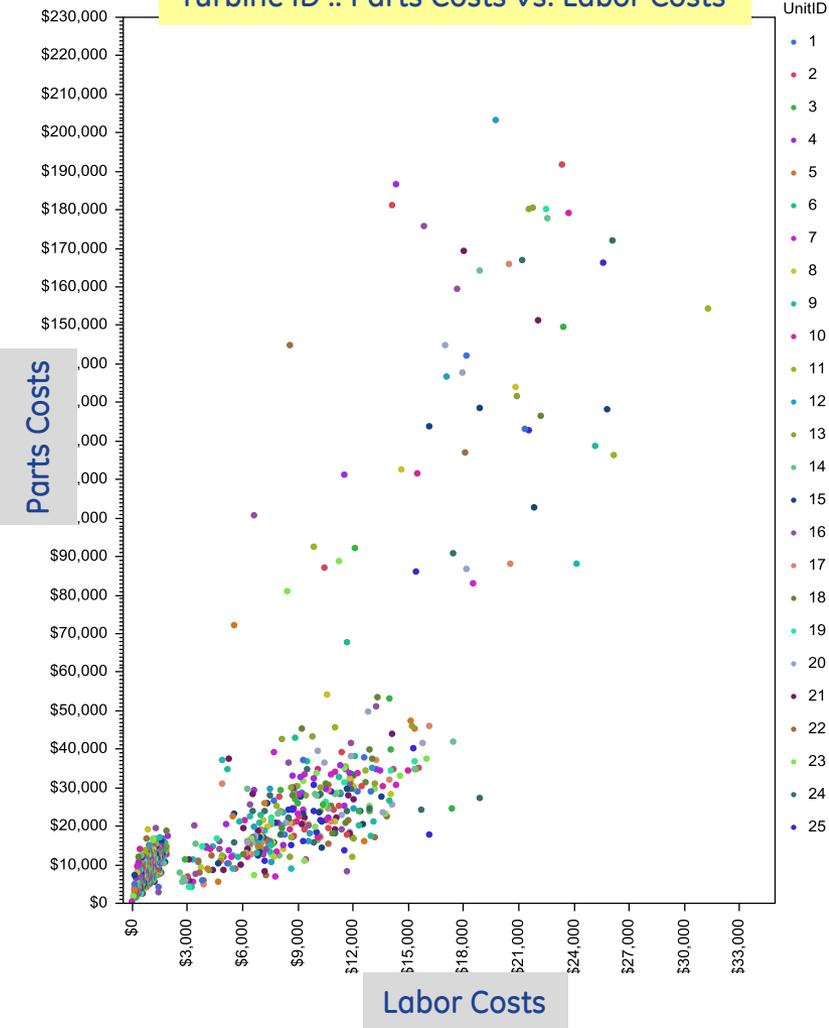
**Note : All data has been scaled and made anonymous**

# Exploratory Data Analysis (1)

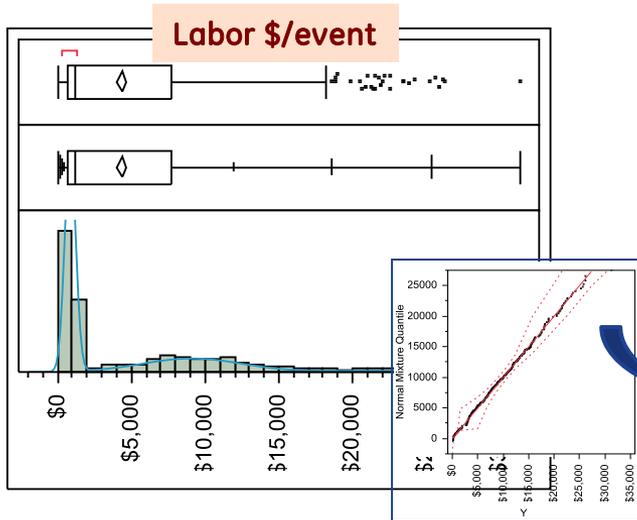
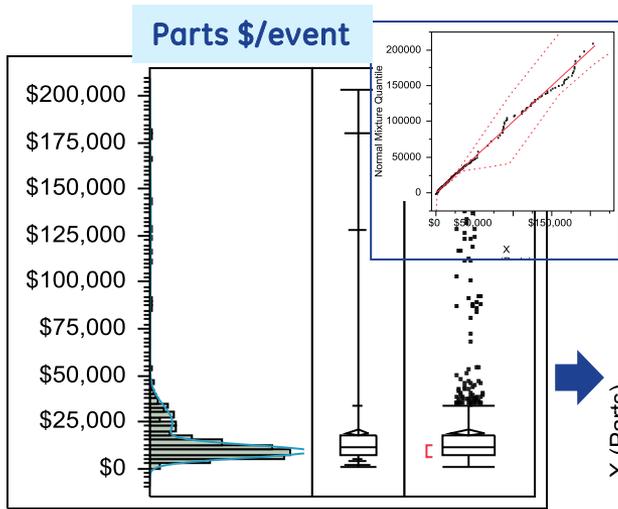
Turbine ID .. Total Cost/Event Box Plot



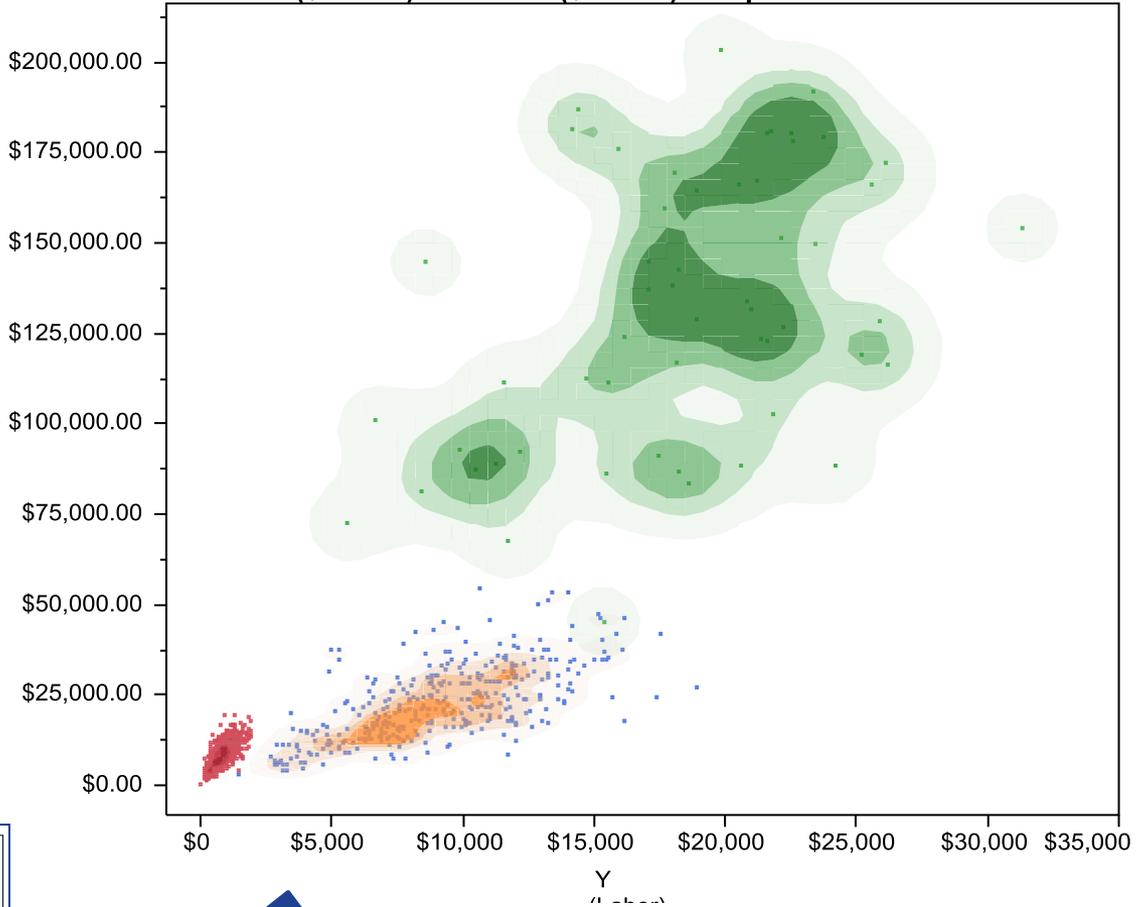
Turbine ID .. Parts Costs Vs. Labor Costs



# Exploratory Data Analysis (2)



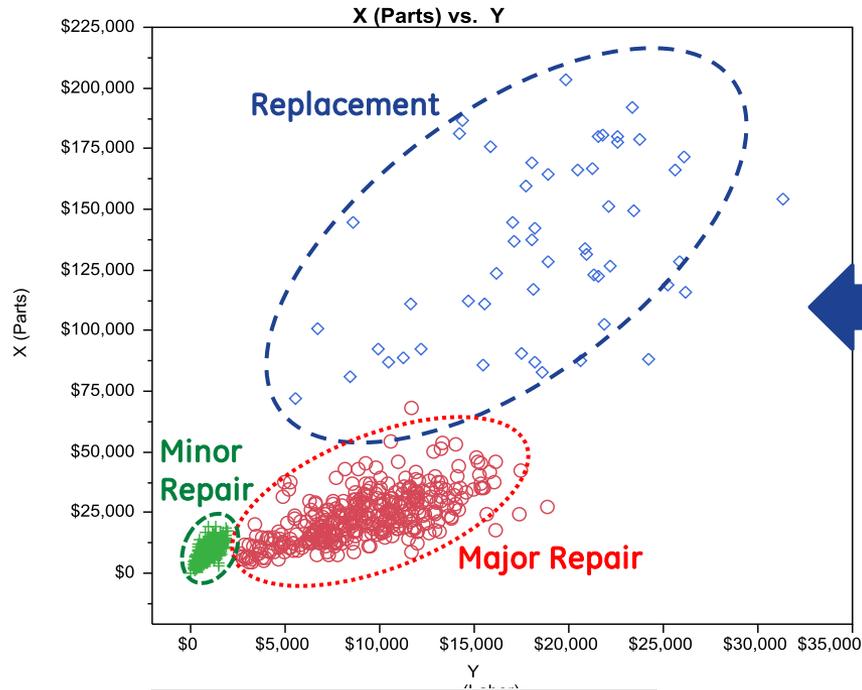
### Parts (\$/event) Vs Labor (\$/event) - Unplanned



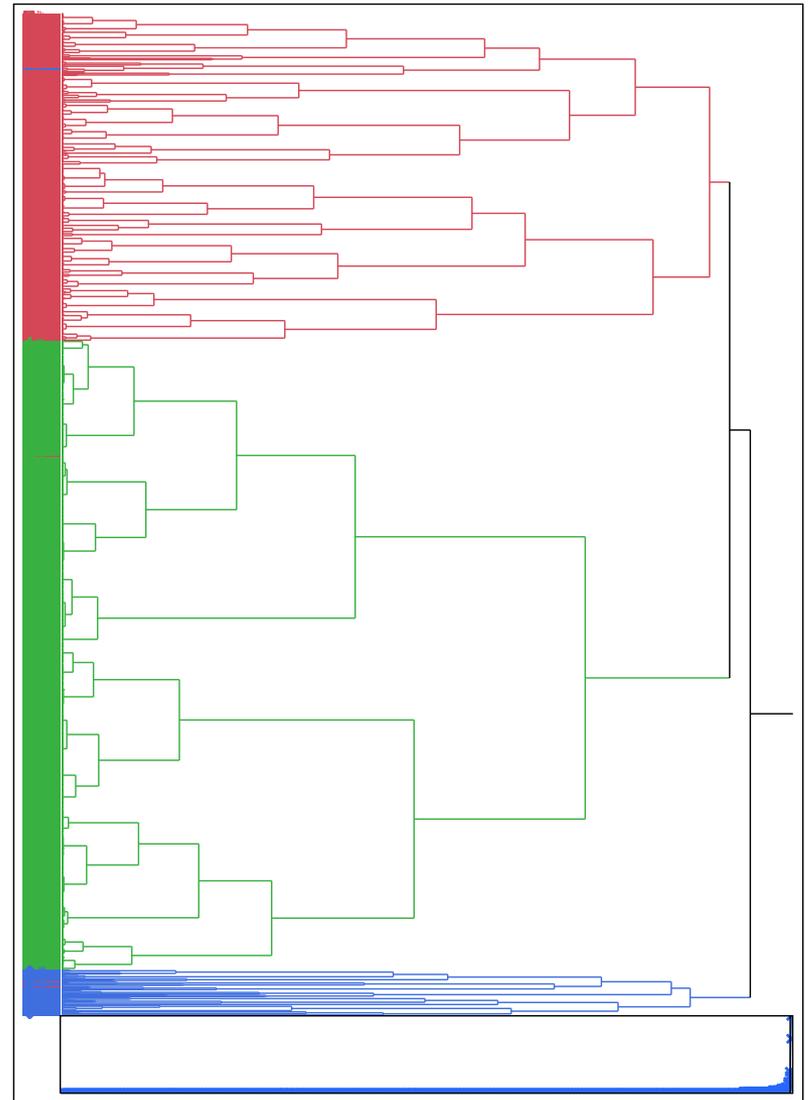
Mixture distributions fitted to the parts & labor cost data indicate potential sub-groups within the dataset ..

# Segmenting the Data (Hierarchic Clustering)

Hierarchic clustering (unsupervised learning) used parts & labor costs to classify the events into three categories with 99.73% accuracy ..



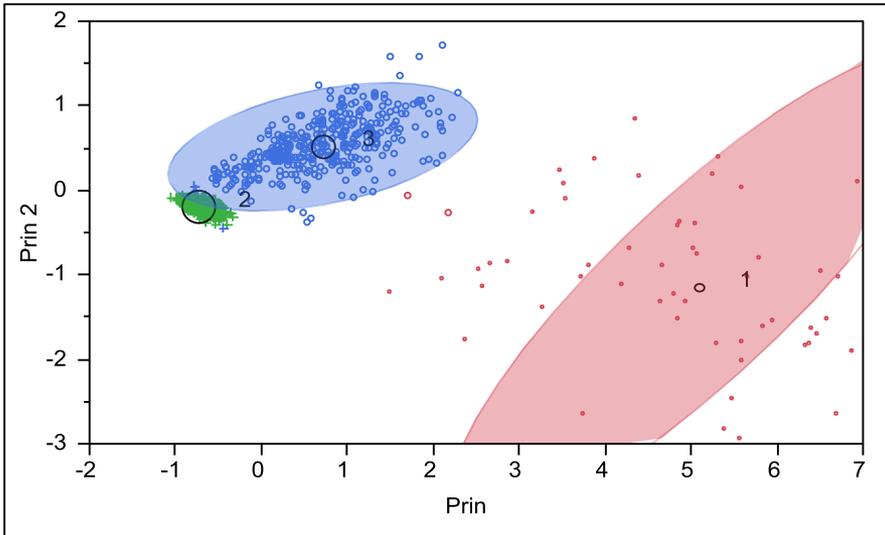
Repair Category	ClusterID	Heirarchic	3
Major_Repair	374	1	0
Minor_Repair	0	721	0
Replacement	2	0	53



# Other Machine Learning Methods ..

Limited pre-classified data from other wind farms can be used to train models like k-means clustering & decision trees to classify new data into pre-defined repair/replacement categories ..

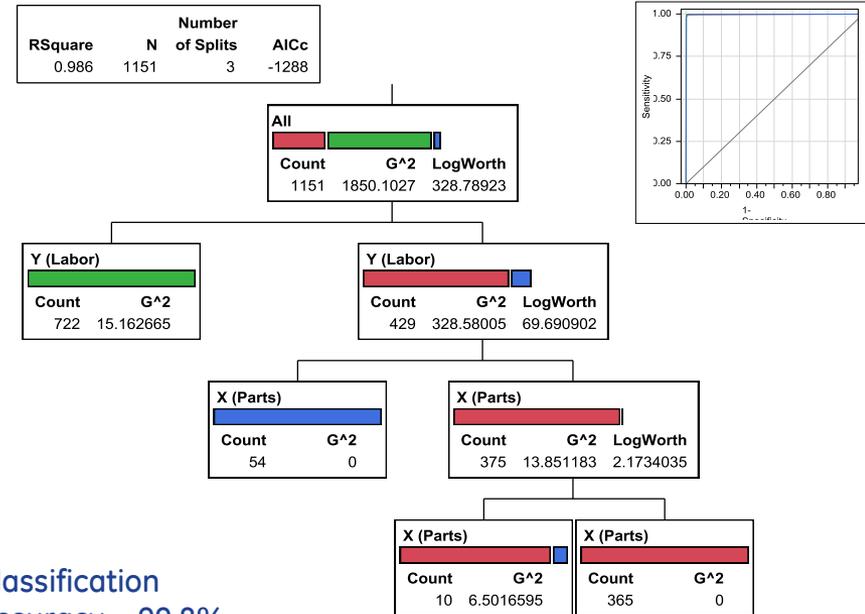
## K-Means Clustering Algorithm Using Robust Normal Mixtures



	ClusterID_KmeansRobNor		
Repair Category	1	2	3
Major_Repair	1	0	374
Minor_Repair	0	719	2
Replacement	54	0	1

Classification Accuracy ~ 99.6%

## Decision Tree / Partition Classifier

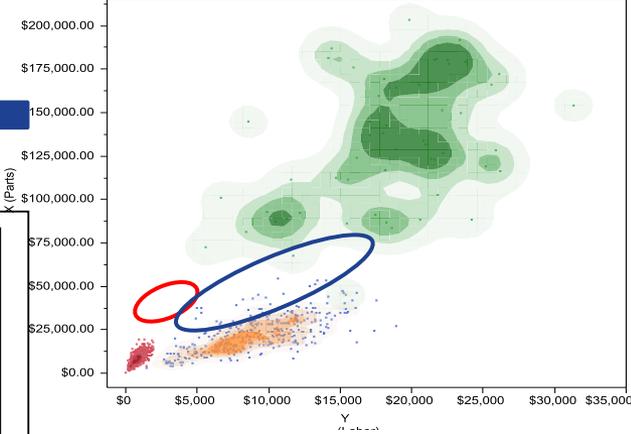


Classification Accuracy ~ 99.8%

Actua	Predicted		
Training	Major_Repair	Minor_Repair	Replacement
Major_Repair	374	1	0
Minor_Repair	0	721	0
Replacement	1	0	54

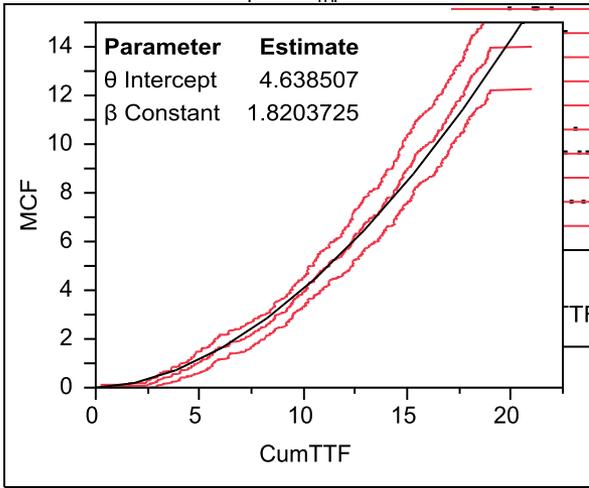
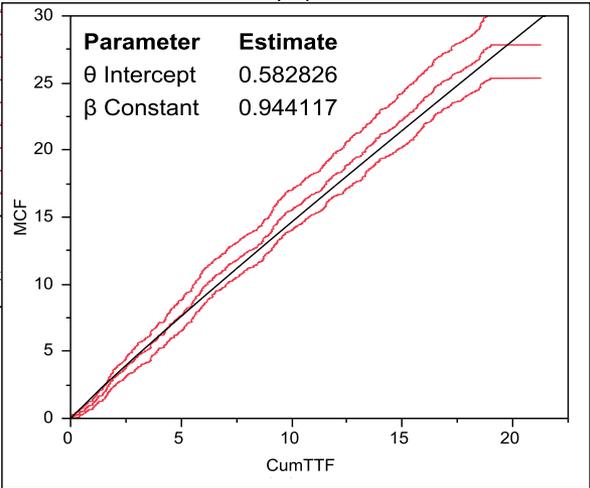
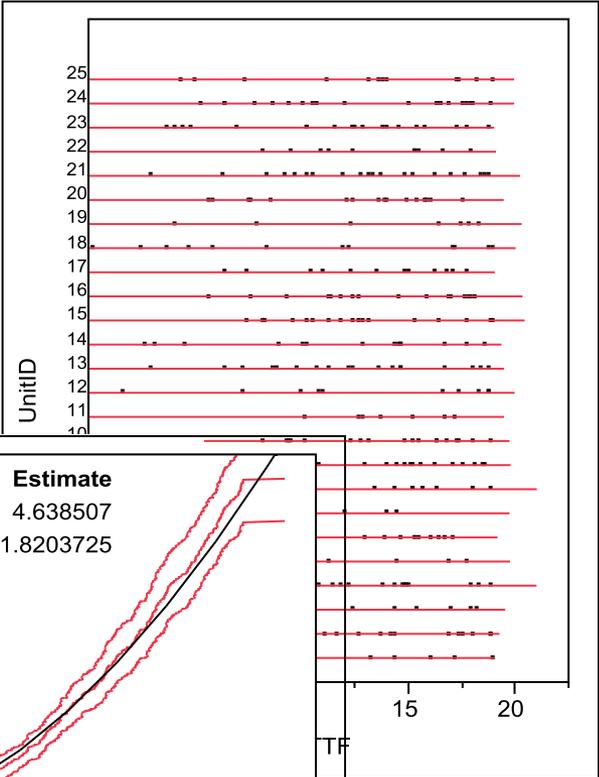
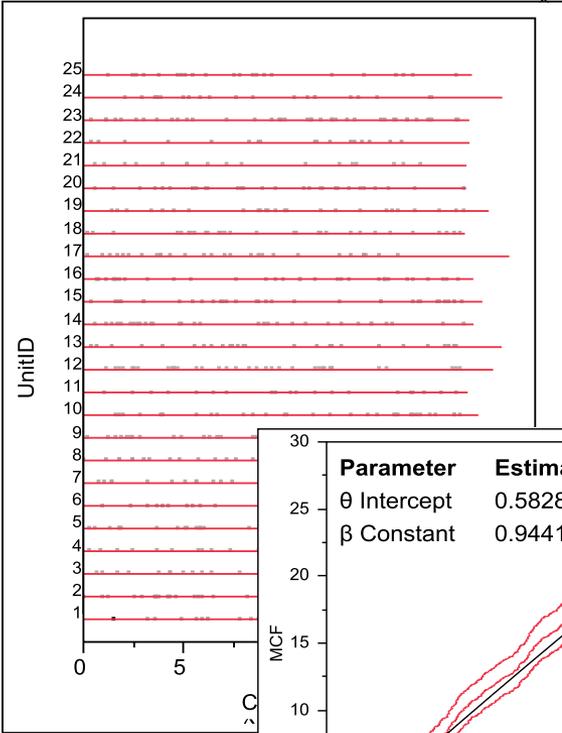
# Predictive Models for Repairs (MCF-NHPP)

Parts (\$/event) Vs Labor (\$/event) - Unplanned



Minor Repairs

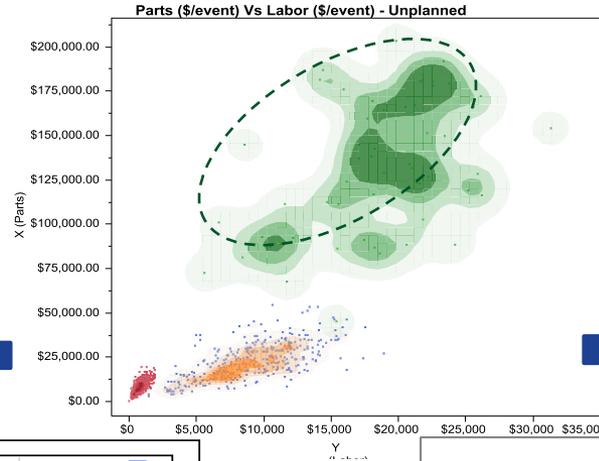
Major Repairs



# Predictive Models for Replacements (E.g. Weibull)

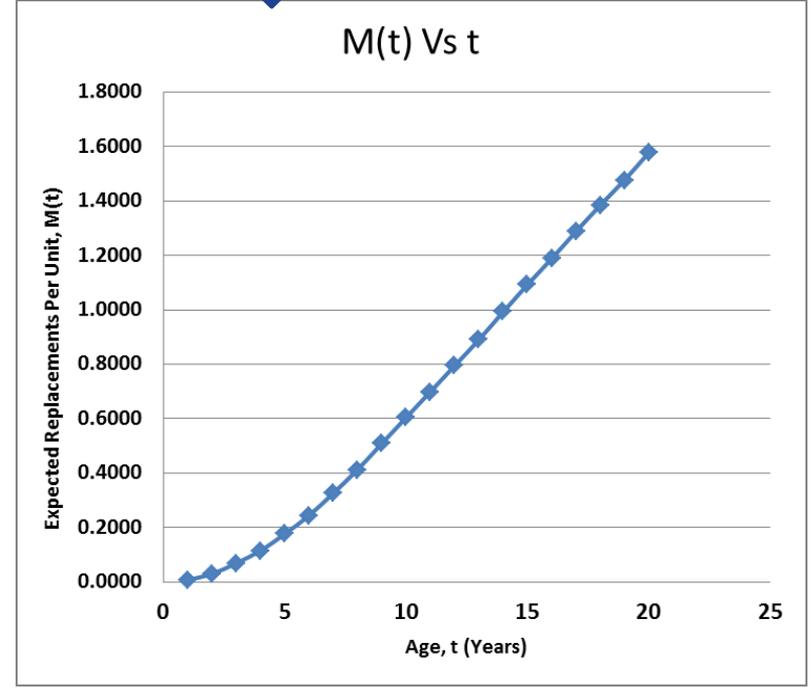
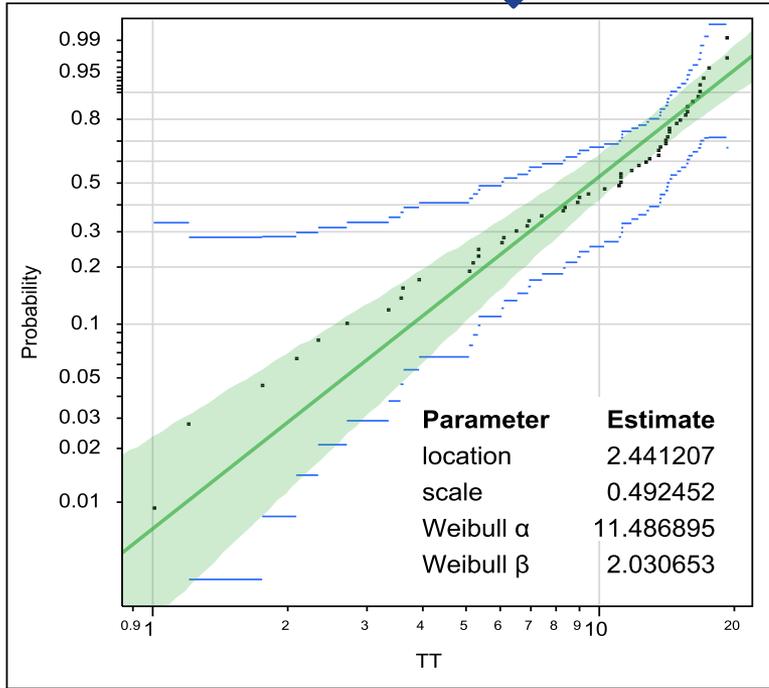
Survival Model for Replacements,  $F(t)$

$$F(t) = 1 - \exp\left[-\left(\frac{t}{\alpha}\right)^\beta\right]$$



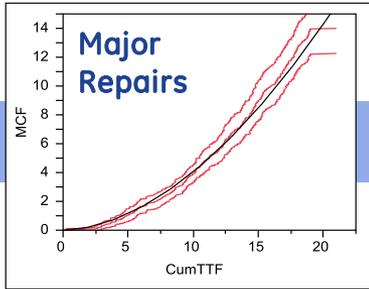
Renewal analysis from survival model,  $M(t)$

$$M(t) = F(t) + \int_{u=0}^{u=t} F(t-u)dF(u)$$

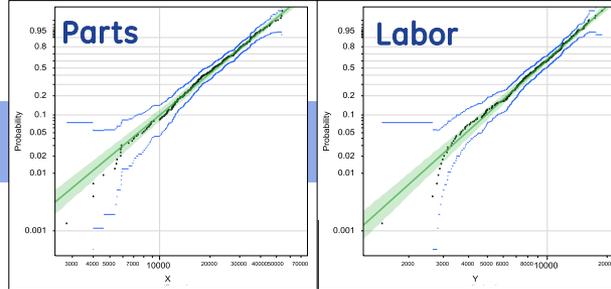


# Predicting Costs via Simulation

## Event Models



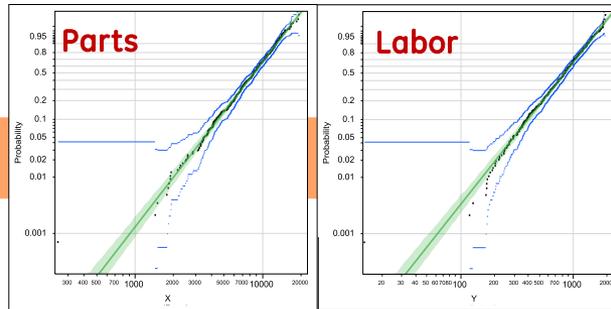
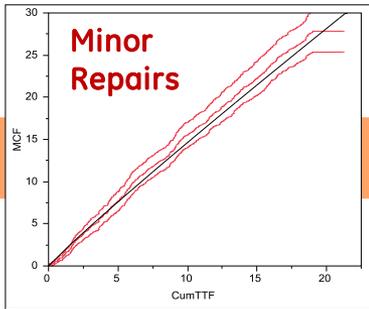
## Part & Labor Cost Distributions



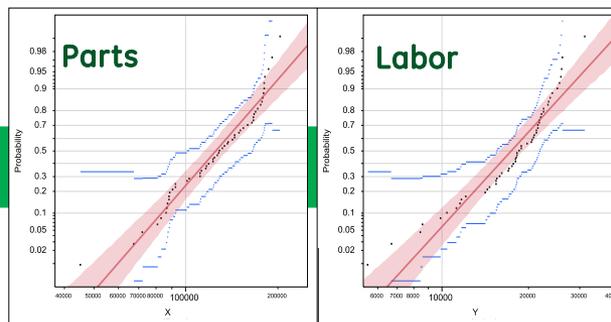
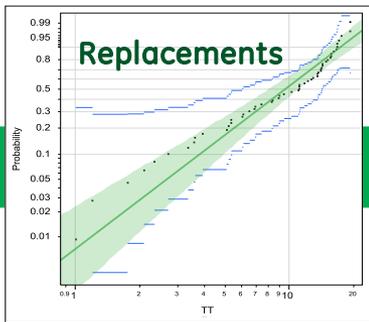
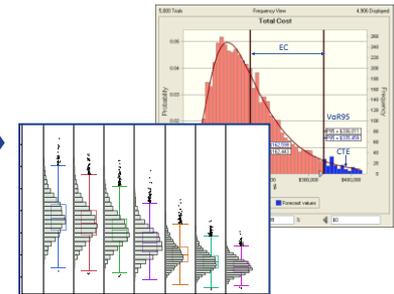
## O&M Capabilities

- Scheduled Maintenance
- Inspection Capability
- Repair Effectiveness
- Crew availability
- Crane availability
- Spare Pools
- Condition Monitoring

Monte-Carlo Simulation Engine



## Results



- Risk & Confidence Bounds
- Cost Distributions
- Availability Distributions
- Optimal CBM/PHM design,
- Optimal schedules for logistics, inspection & maintenance, spares etc.

# Conclusions

**Analytics for Physical Asset Management** can be applied to existing maintenance, cost & condition monitoring data to build predictive models that optimize wind farm O&M costs

This goes beyond “classical reliability” (E.g. Weibull’s) and includes a range of data-mining and actuarial analysis techniques

This methodology is a low-risk, low-cost approach to systematically exploit your data and drive productivity into your wind farm operations.

Please contact us for more information if interested

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## Thank You !