

DAMA-CHICAGO, JUNE 15,2016

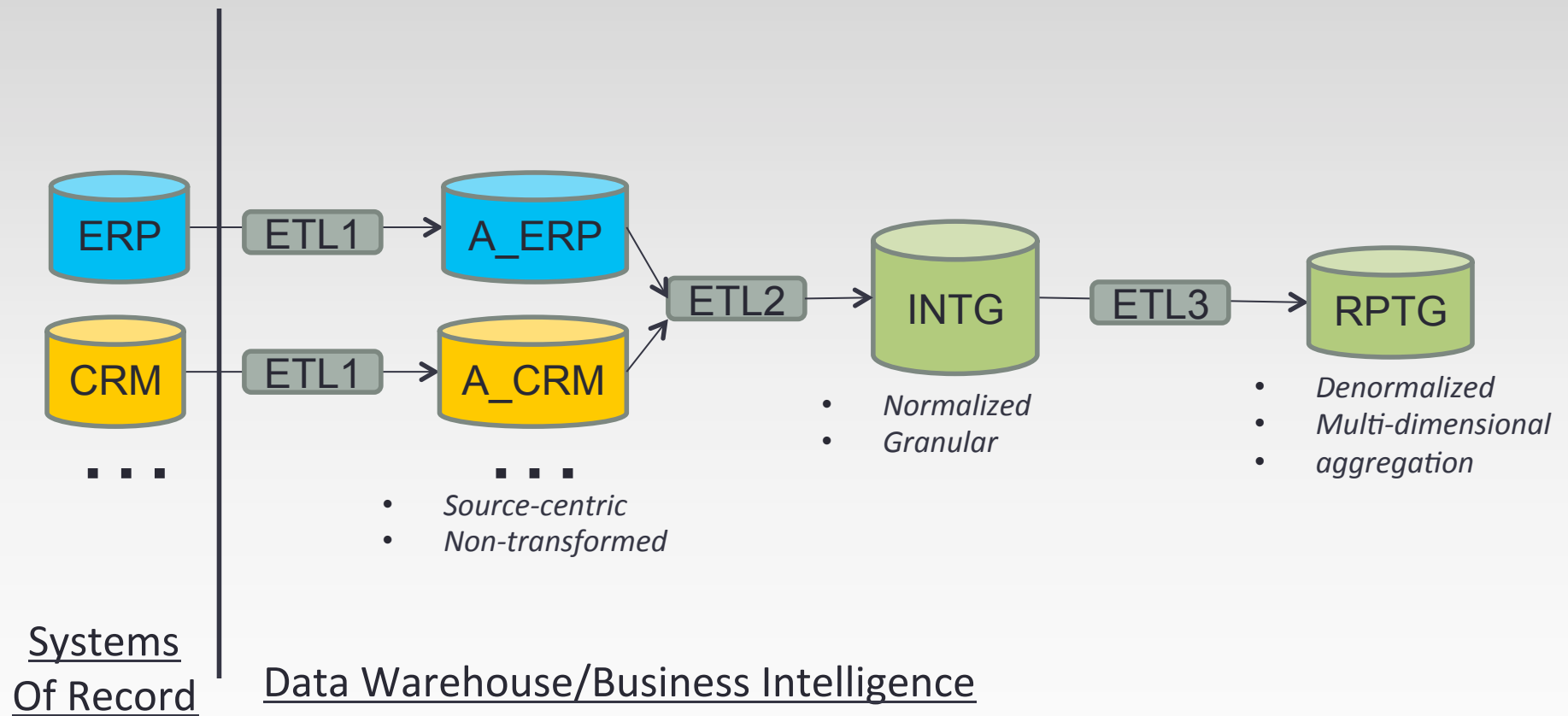
*Predictive Analytics:
A Statistical Primer for Data Modelers*

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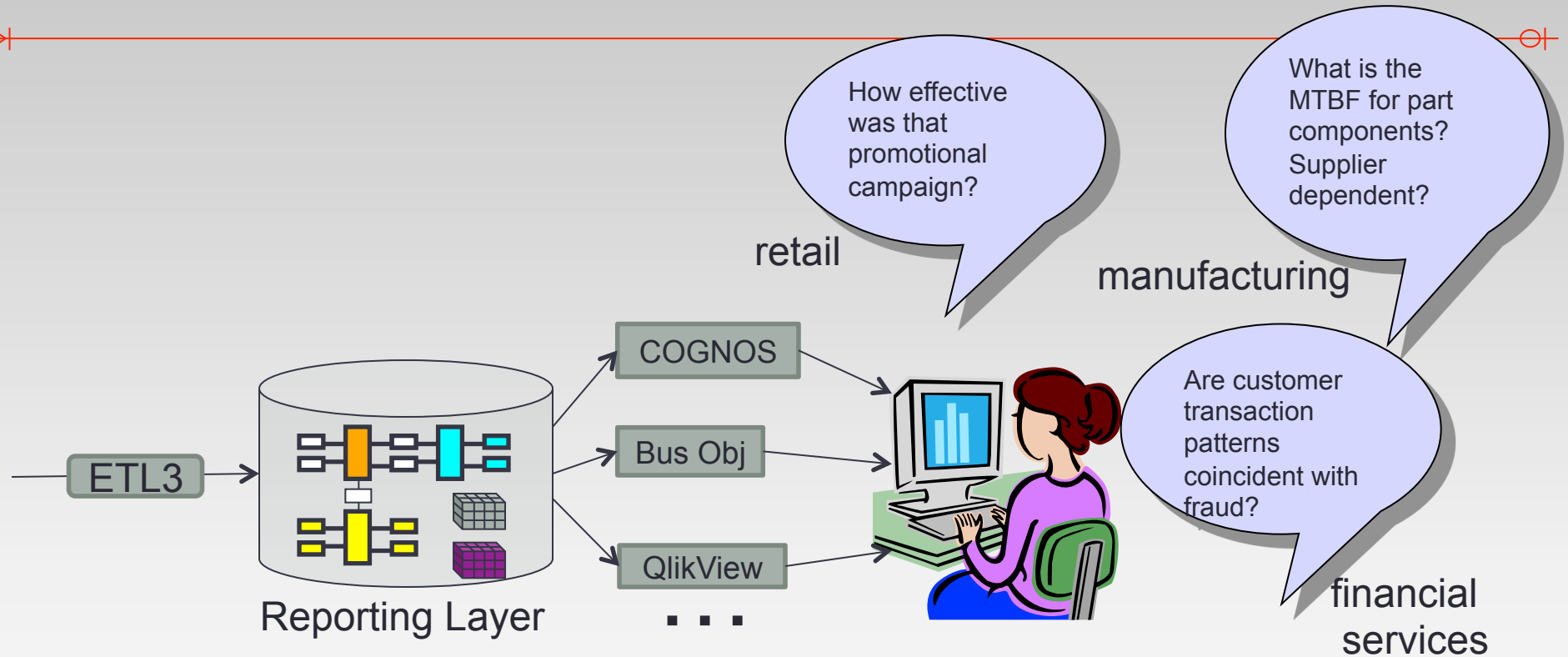
Predictive Analytics: Agenda

- **What** – Contrast to Descriptive Analytics
- **Why** – Value Proposition for Predictive Analytics
- **How** – Statistical basis of Predictive Analytics
- **Getting Started** with Predictive Analytics
- **Q&A**

'Traditional' DW/BI Architecture



Descriptive Analytics (OLAP)

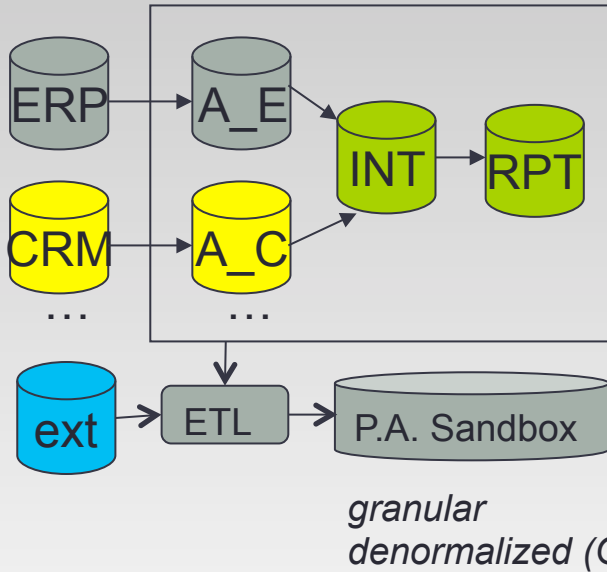


Descriptive Analytics – What has happened?

Trends, patterns, exceptions in historic data

Monitor/Control, business process improvement

Predictive Analytics



Forecast future sales?
→ Inventory level
→ Labor needs?
retail

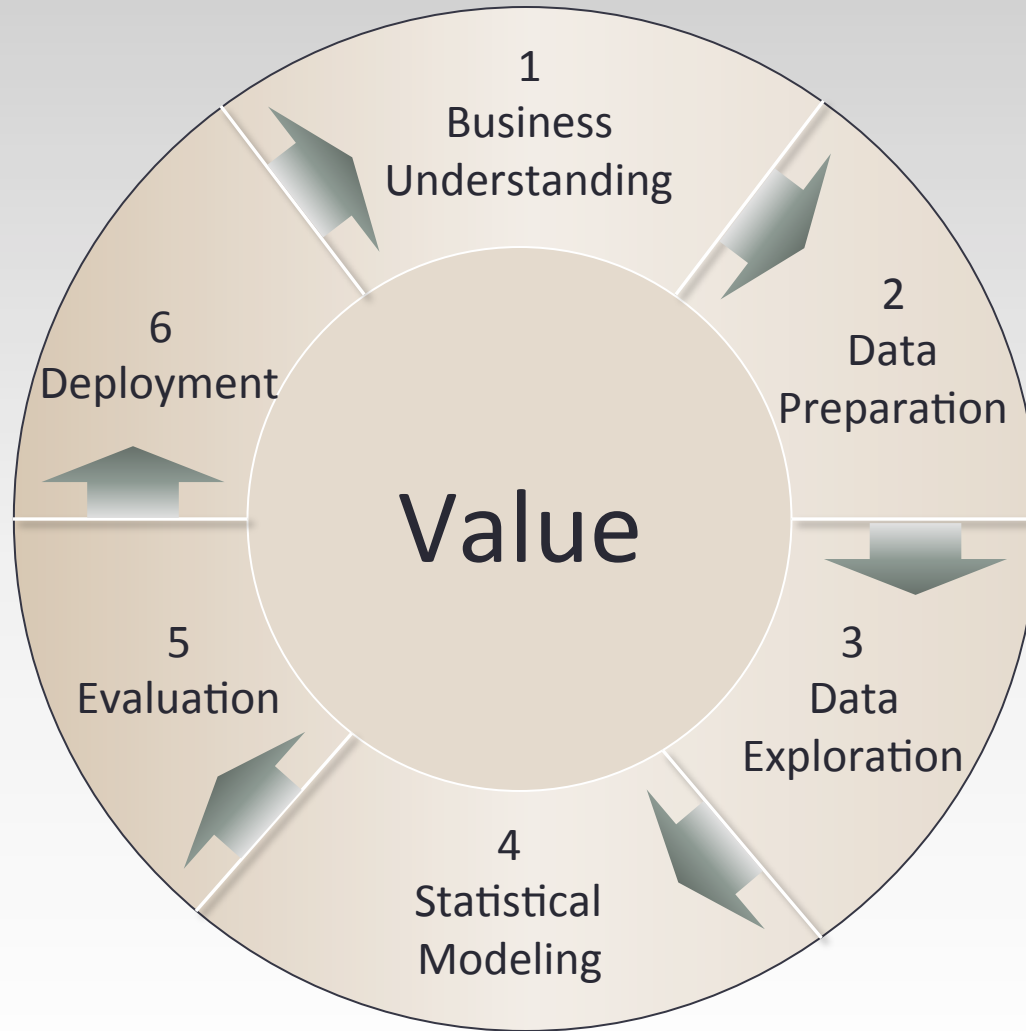
Optimal equipment maintenance schedule?
equip operation

Future jet fuel prices for hedge contracts?
airlines

Predictive Analytics – What will happen?

Advanced statistics → mathematical models
Forecast future state/behavior

Predictive Analytics Lifecycle

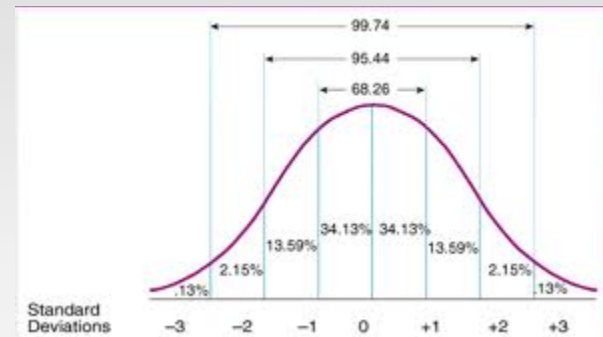




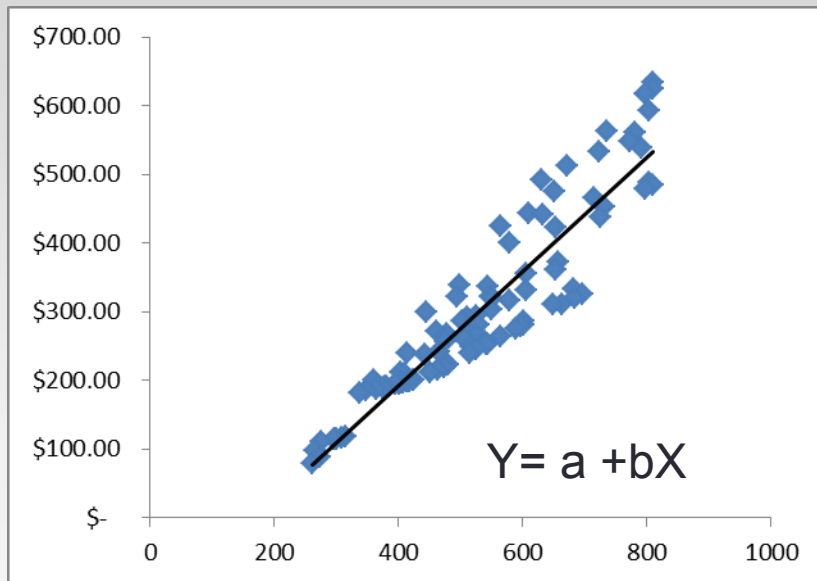
Mean (average), $\underline{X} = \sum (X_i) / N$

Variance, $S_x^2 = \sum (X_i - \underline{X})^2 / (N-1) = s_x^2$

Standard Deviation, $s_x = \sqrt{s_x^2}$



Simple Linear Regression



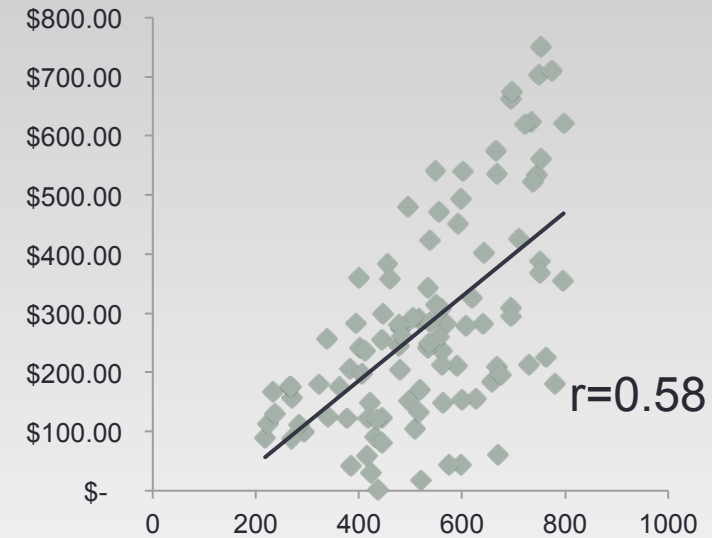
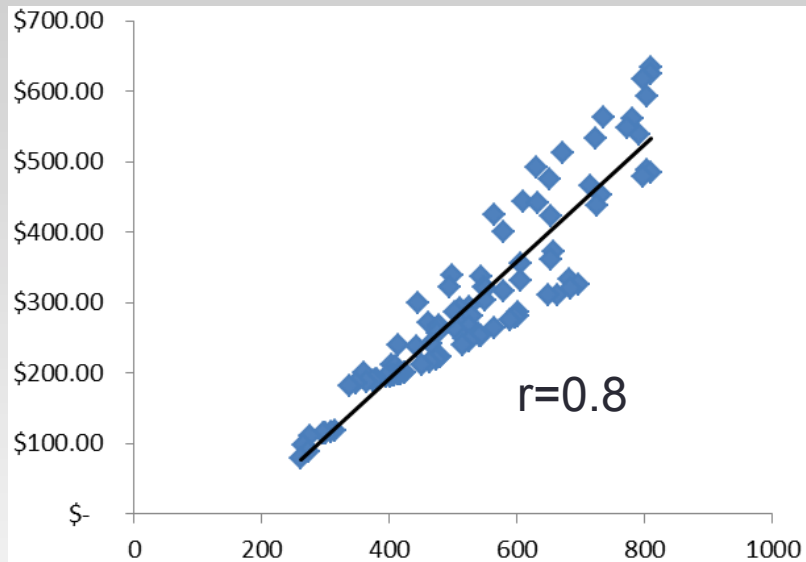
$$\text{Covariance, } S_{xy}^2 = \frac{\sum(X_i - \underline{X})(Y_i - \underline{Y})}{(N-1)} = s_{xy}^2$$

$$\text{Slope, } b = s_{xy} / s_x$$

$$\text{Intercept, } a = \underline{Y} - b\underline{X}$$

Best Fit: minimizes variance between predicted values and observed values

Simple Correlation



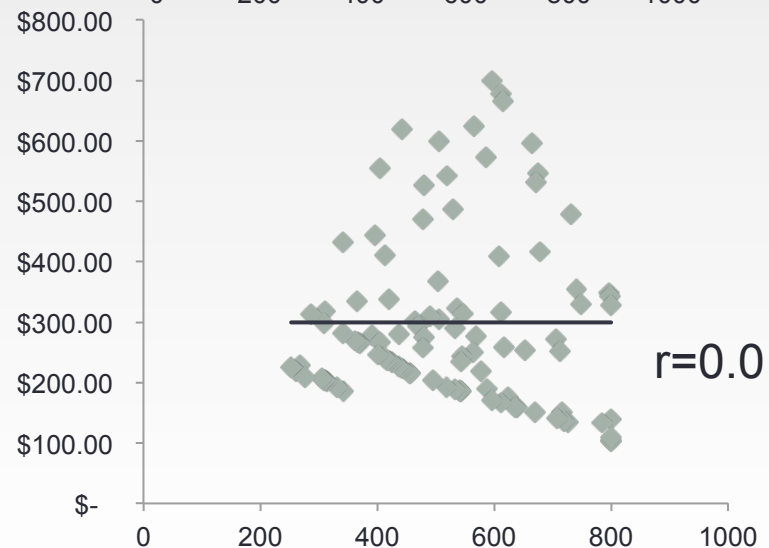
Correlation Coefficient, $r = s_{xy} / s_x s_y$

$$-1 \leq r \leq +1$$

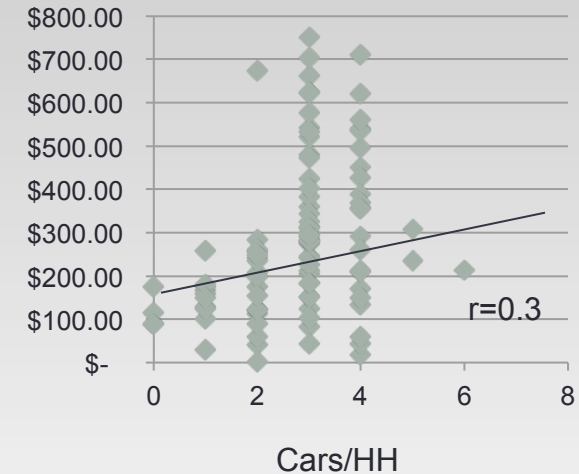
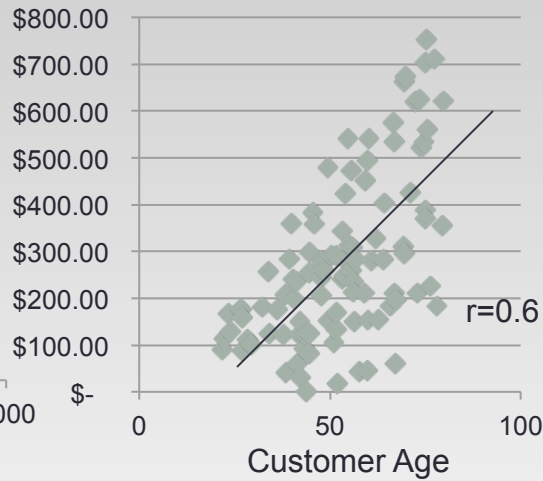
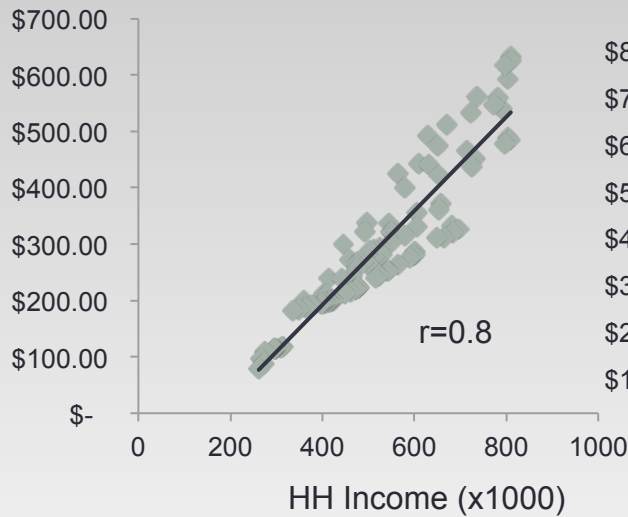
Correlation \neq Cause and Effect

$$0 \leq r^2 \leq +1$$

$r^2 \Rightarrow$ fraction of Y variance attributable to X



Multiple Correlation & Regression



$$\text{Mercedes Sales} = a + b(\text{Income}) + c(\text{Age}) + d(\text{HH Cars})$$

$$R^2 = 0.89$$

<u>Variable</u>	<u>Sales</u>	<u>Income</u>	<u>Age</u>	<u>Cars</u>
<u>Sales</u>	1.0	0.8	0.6	0.3
<u>Income</u>		1.0	0.7	0.4
<u>Age</u>			1.0	0.3
<u>Cars</u>				1.0

Partial Correlation

Original Correlation Matrix

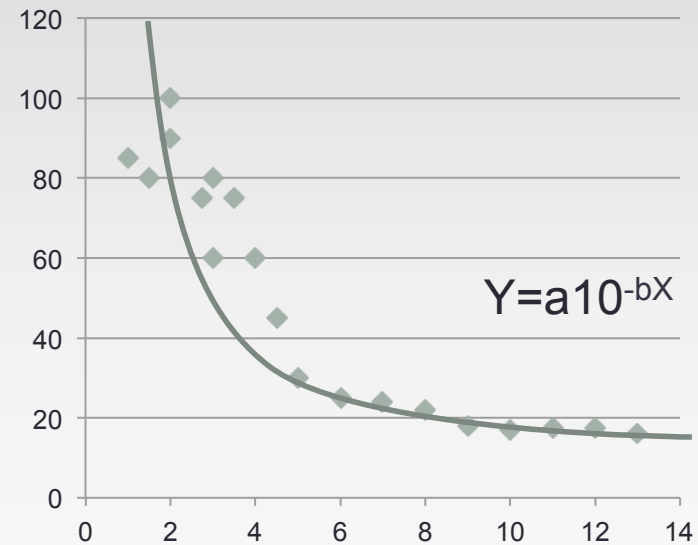
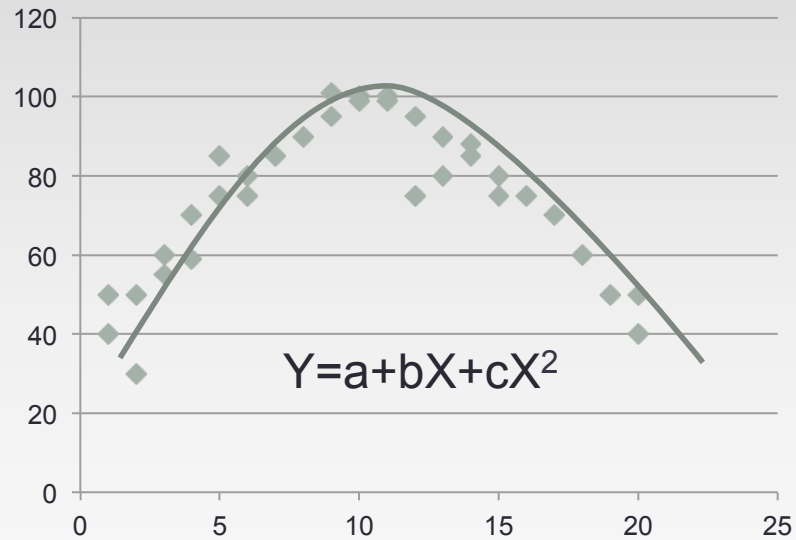
<u>Variable</u>	<u>Sales</u>	<u>Income</u>	<u>Age</u>	<u>Cars</u>
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<u>Income</u>		1.0	0.7	0.4
<u>Age</u>			1.0	0.3
<u>Cars</u>				1.0

Stepwise (Partial) Correlation Matrix

<u>Variable</u>	<u>Sales</u>	<u>Income</u>	<u>Age*</u>	<u>Cars*</u>
<u>Sales</u>	1.0	0.79	-0.48	0.09
<u>Income</u>		1.0		
<u>Age</u>			1.0	
<u>Cars</u>				1.0

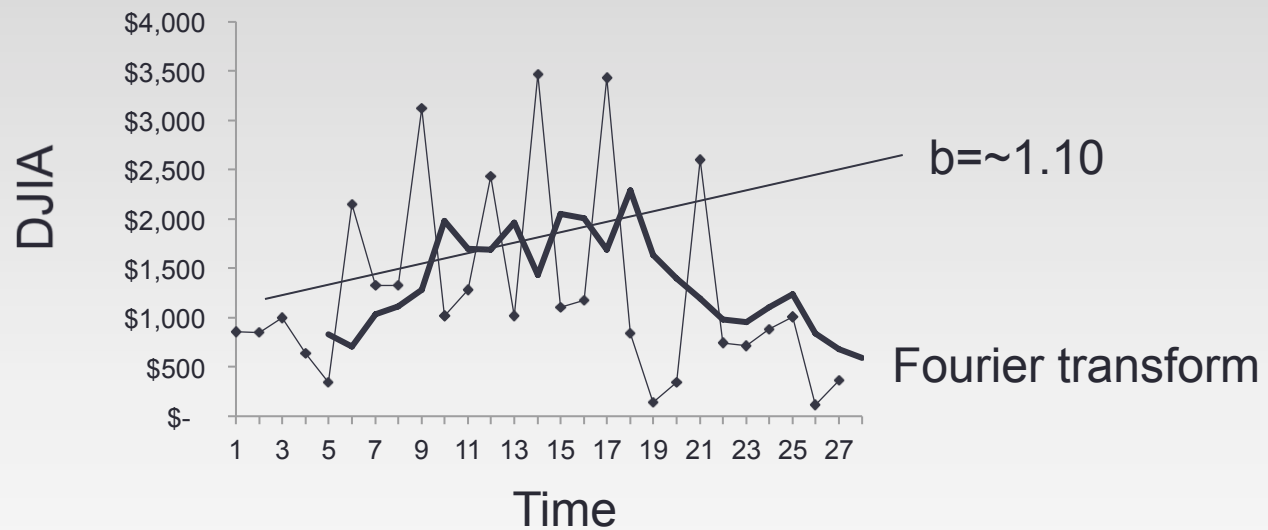
$$\text{Mercedes Sales} = a' + b'(\text{HH Income}) - c'(\text{Age})$$

Nonlinear Regression



Does the mathematical model make business sense?

Periodic (Cyclic) Models



$$DJIA = a + bT + c\text{SIN}(T_c - T) + d\text{SIN}(T_d - T) + \dots$$

Getting Started Suggestions

1. Bone up on statistics/predictive modeling
 - www.coursera.org – free, online classes
 - Books – “Predictive Analytics” by Conrad Carlberg
 - Leverage in-house methods/tools expertise
2. Engage a user-partner/business problem
 - Evaluate current forecast process and impact
 - Explore the data – metrics, sensitivity variables
 - Prototype a model – test it against actual data
3. Estimate business impact (\$) – what if scenarios, precasting
4. Get started with simple tools (Excel with advance stat plug-ins)
5. Parallel trials with current process and proposed predictive model
 - Continuous monitoring and refinement

Questions

