

# The Use of Designed Experiments in Process Development

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IMAPS NE 5/7/2019

# What is a Designed Experiment?

- In an experiment, we deliberately change one or more process variables (or factors) in order to observe the effect the changes have on one or more response variables. The (statistical) design of experiments (*DOE*) is an efficient procedure for planning experiments so that the data obtained can be analyzed to yield valid and objective conclusions.
  - Change Process Variables ➡ **Effect** on Response
  - Statistical Validity, signal-to-noise testing
  - Valid and Objective Conclusions
- Statistical Analysis Software
- Constraints
  - To randomize or not
  - What's manageable
  - How much does it COST???
  - How much material do I have available?

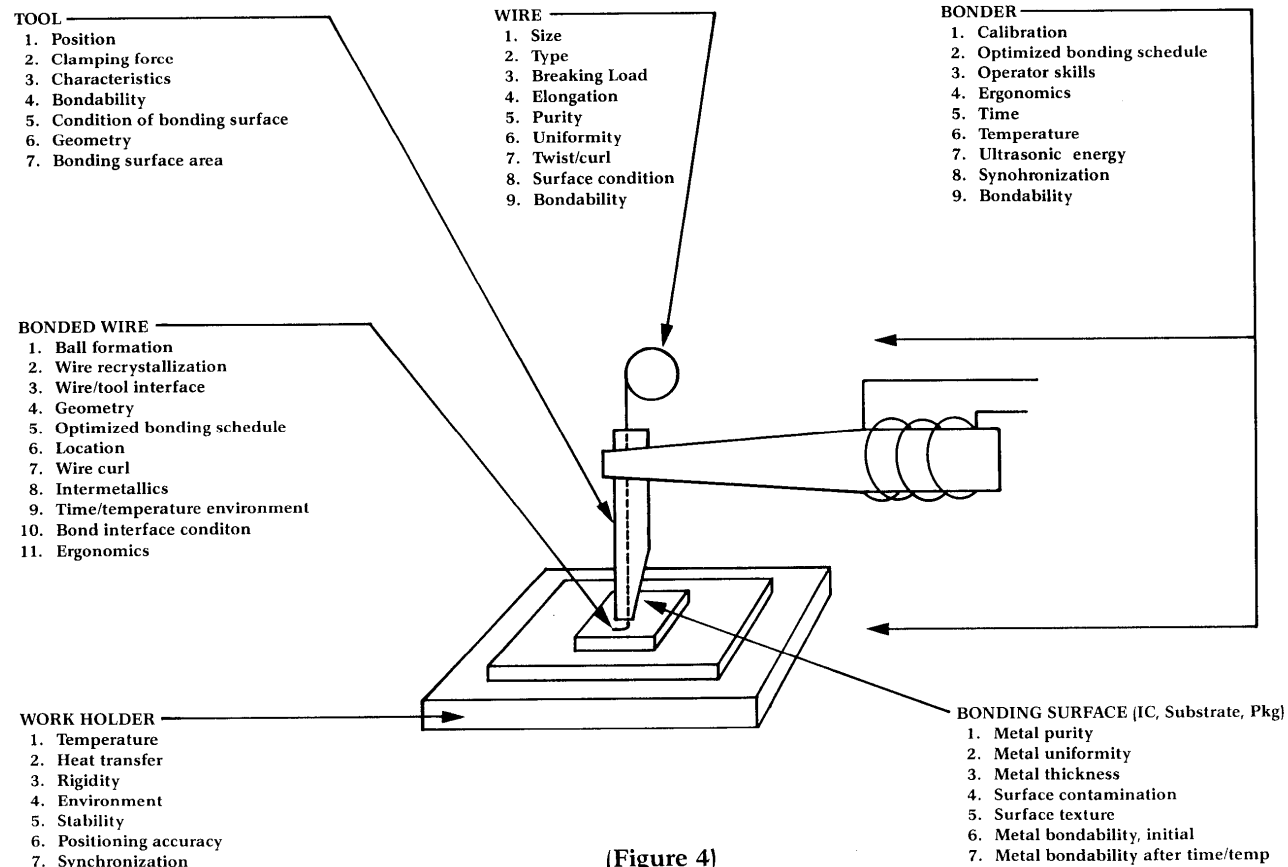
# Pareto Diagrams

- Break the problem up into smaller components (PARETO)
- Analyze the Pareto data
- Find the **MOST SIGNIFICANT DEFECTS**
  - Most numerous
  - Most costly
  - Most dangerous- long-term reliability
- What is breaking and where is it breaking?
- Focus on what is most important
- What is **EASIEST** to fix?

# Design of Experiments

- The best way to study a complicated process while you are waiting for INSPIRATION!!!
- Do designed experiments to understand the Process
- How do the process programmable parameters really effect the results
- Statistical significance and validity
  - Signal to Noise (F-test), More data can resolve smaller differences
  - Noisy data means that you can't resolve small signals
  - Is it real or just noisy data?
- Learn and Use the software that you have

# 50 Bonding Variables



(Figure 4)

## 50 BONDING VARIABLES

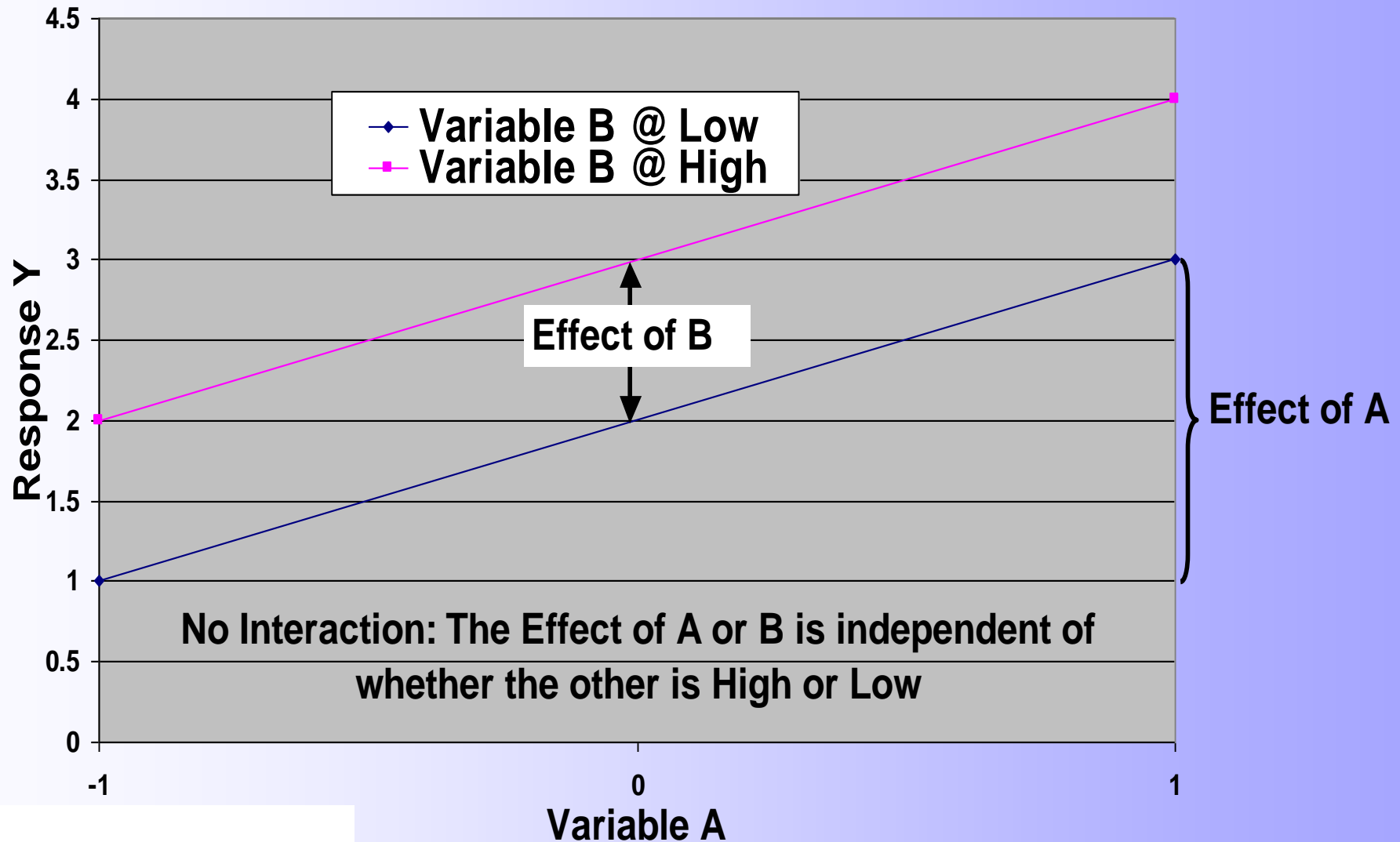
There are a number of variables that occur in either wedge or ball bonding. These are assembled for easy reference. (24)

Is this all of them...

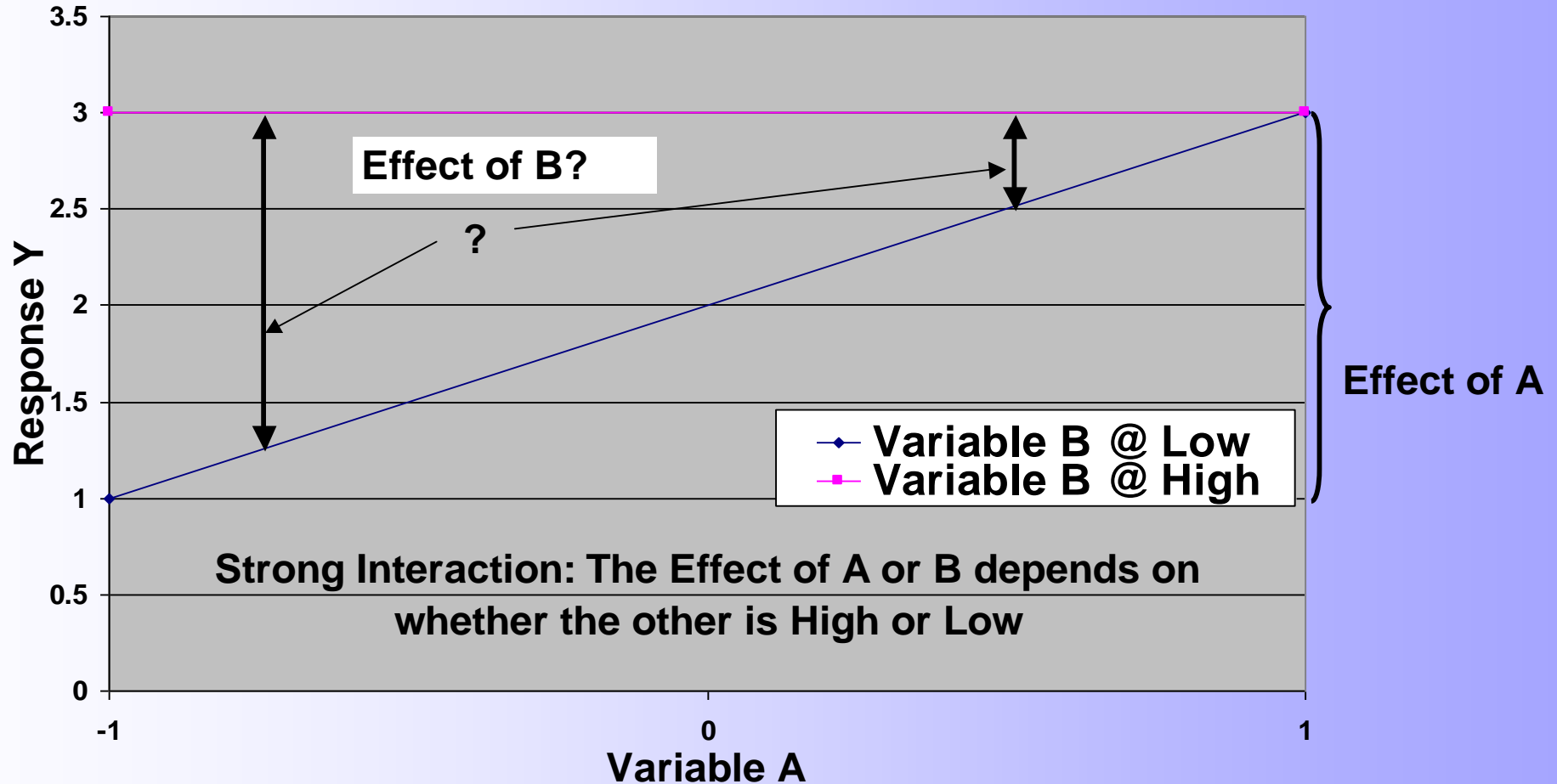
Not even close!!!

- Complex processes often have **HUNDREDS** of possible variables
- What are the major process variables
- What is failing
- Pareto analysis of the defectives
- Focus on the biggest problem or easiest to solve?

# What's an Effect?

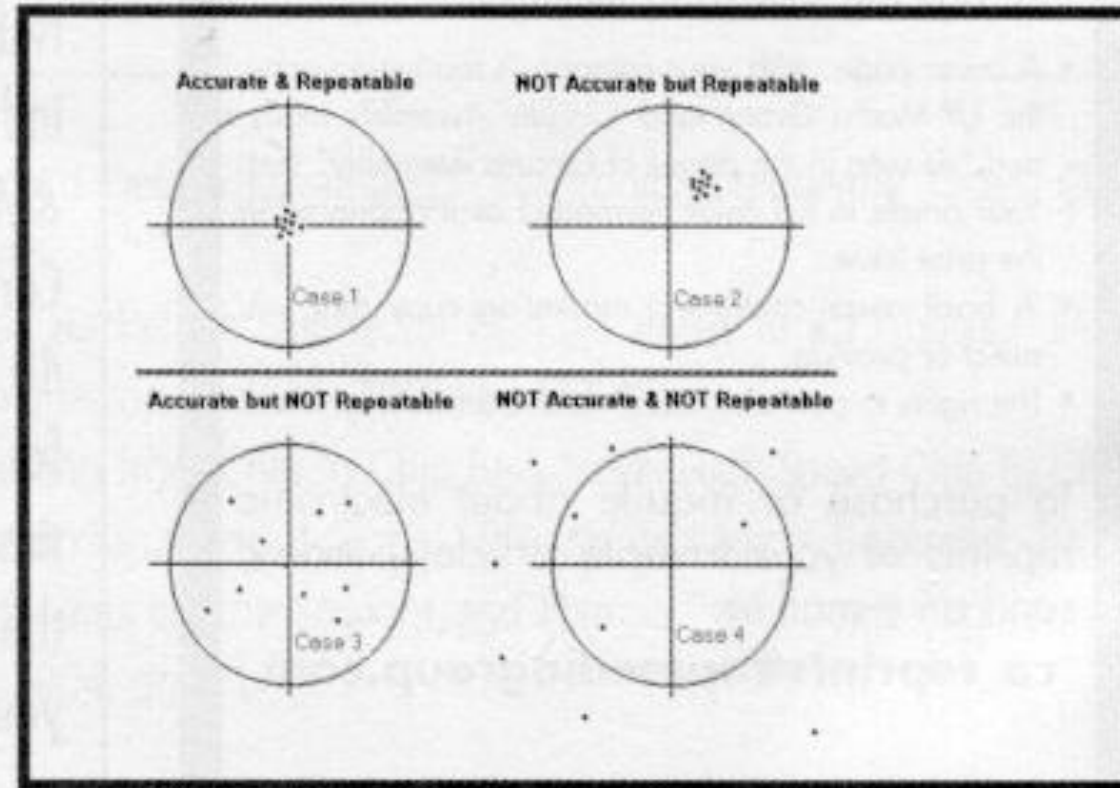


# Whats an Interaction?



*One of the important benefits of DOE is that we not only learn the effects and interactions, but also get a statistical validity? Are they real or just random variations in the data?*

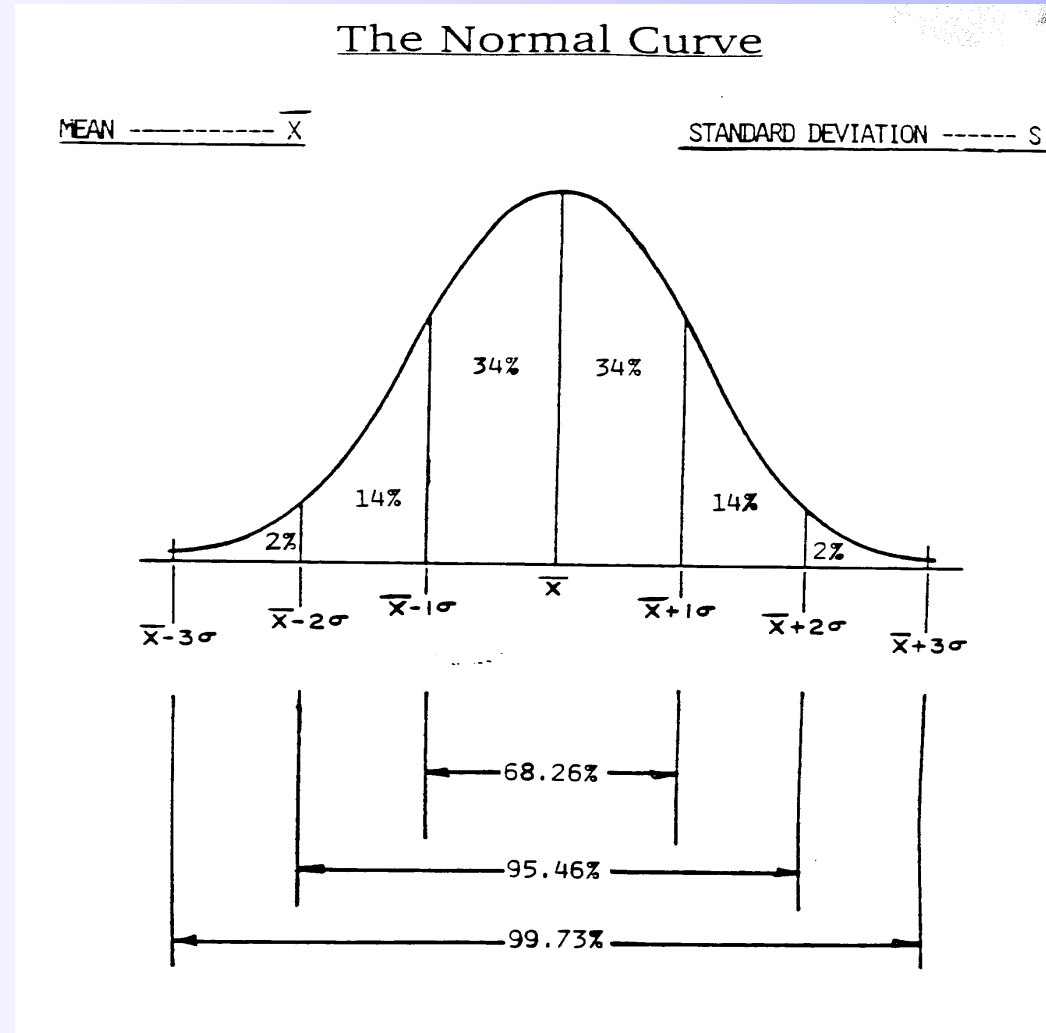
# Accuracy vs Repeatability



**FIGURE 1:** Small standard deviation does not guarantee accuracy. Case 1 shows a repeatable machine, Case 2 a repeatable machine that is not very accurate.



# A Histogram Describes the Variation of Normal Data



$$Cp = \frac{(USL - LSL)}{6\sigma} = \frac{\text{Design Variation}}{\text{Process Variation}}$$

$$Cpk_{Upper} = \frac{(USL - \bar{X})}{3\sigma}$$

$$Cpk_{Lower} = \frac{(\bar{X} - LSL)}{3\sigma}$$

$$Cpk = \text{Smaller of } Cpk_{Upper} \text{ and } Cpk_{Lower}$$

# Screening Vs. Response Surface

- SCREENING:
  - Sift out of many control variables the critical few
  - Less cells, less time, less cost, assumes planar response (linear)
  - No interactions
- RESPONSE SURFACE:
  - Obtain map of process that accurately predicts
  - Includes quadratic and interaction terms of the model

# Benefits of using Response Surface Experimental Design

- Most information in fewest trials
- Pictorial & numerical understanding
- “What if ...?” questions answered (predictions)
- Optimization is easy
- Interactions accounted for

# Recommended Designs

- Taguchi for Screening (L-9 + center)
- LINEAR WITH CENTER POINT (screening)
- QUADRATIC (response surfaces)

# Other Designs

- Factorial
- Fractional Factorial
- Taguchi
- Central Composite
- .....
- There are Pros and Cons to All of Them

# Taguchi L-9 + Center Point

	A	B	C	D
Taguchi	-1	-1	-1	-1
	-1	0	0	0
	-1	1	1	1
	0	-1	0	1
	0	0	1	-1
	0	1	-1	0
	1	-1	1	0
	1	0	-1	1
	1	1	0	-1
CP	0	0	0	0

- Benefits
  - Small # cells
  - 3 Levels
  - Center Point
- Disadvantages
  - Regression
  - No Interactions
  - Highly Saturated

# Fractional Factorial w Center Points

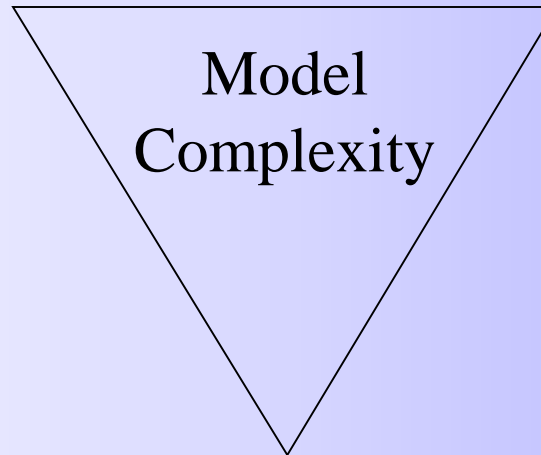
	A	B	C	D	E
<b>Factorial</b>	-1	-1	-1	-1	1
	1	-1	-1	-1	-1
	-1	1	-1	-1	-1
	1	1	-1	-1	1
	-1	-1	1	-1	-1
	1	-1	1	-1	1
	-1	1	1	-1	1
	1	1	1	-1	-1
	-1	-1	-1	1	-1
	1	-1	-1	1	1
	-1	1	-1	1	1
	1	1	-1	1	-1
	-1	-1	1	1	1
	1	-1	1	1	-1
	-1	1	1	1	-1
	1	1	1	1	1
<b>Center</b>	0	0	0	0	0
	0	0	0	0	0
	0	0	0	0	0

- Advantages
  - Interactions
  - Centerpoints
- Disadvantages
  - Not as efficient as ECHIP
  - Regression or special S/W
- Should be randomized



# What size sample is best?

Sample Variation  
(Standard Deviation)



Least Significant  
Difference

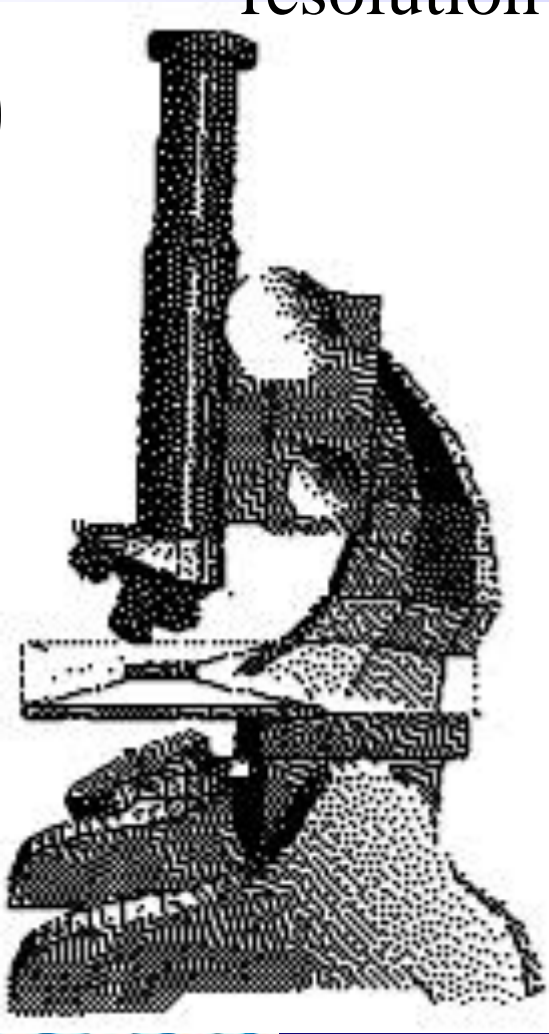
Number of Samples

What is the cost/sample?

How many are available?

# RESOLVING POWER - tradeoff desired resolution and number of trials needed

RESOLVING POWER



30X

100X

300X



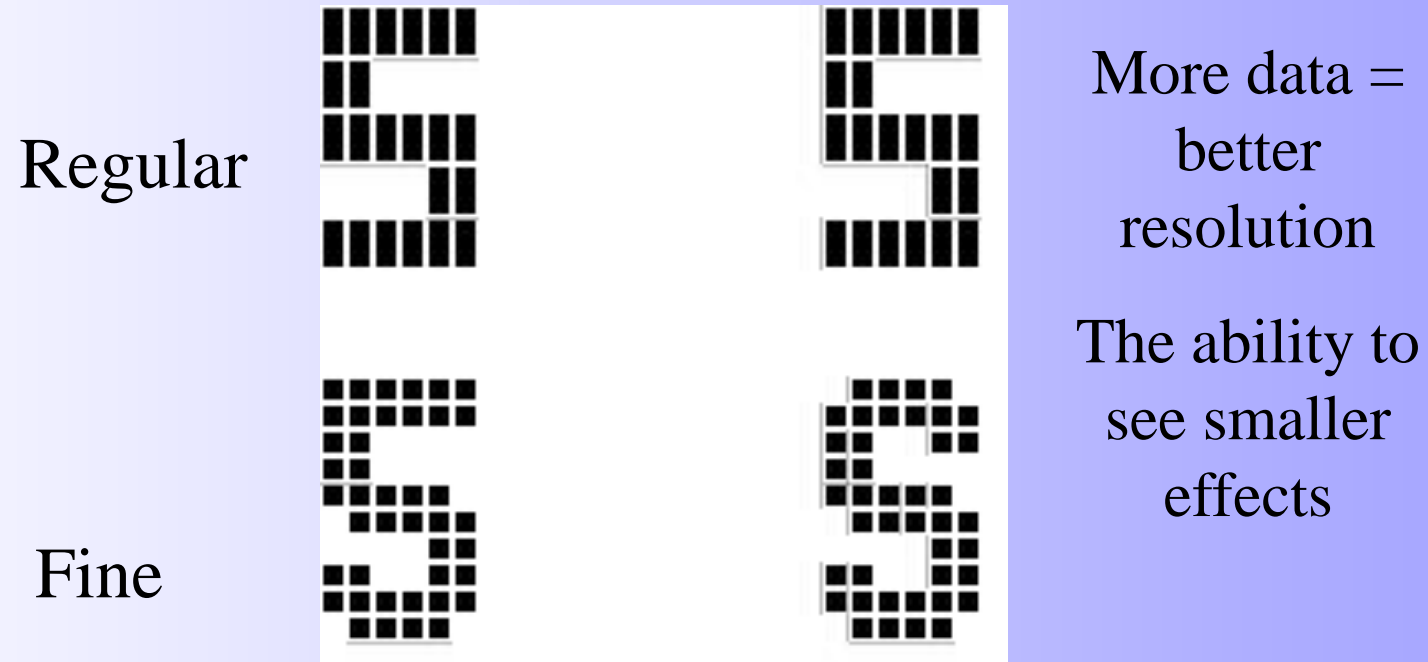
More trials: More magnification

Smaller *effects* can be “seen”

30 observations are not always enough to see what we want!

# RESOLUTION: Smallest detectable effect

e.g. FAX machines transmit in regular and fine modes. Compare the characters “5” & “S”



The resolution equals the dot size.

# What Do You Do If You Have Lack-Of-Fit?

Try a data transformation

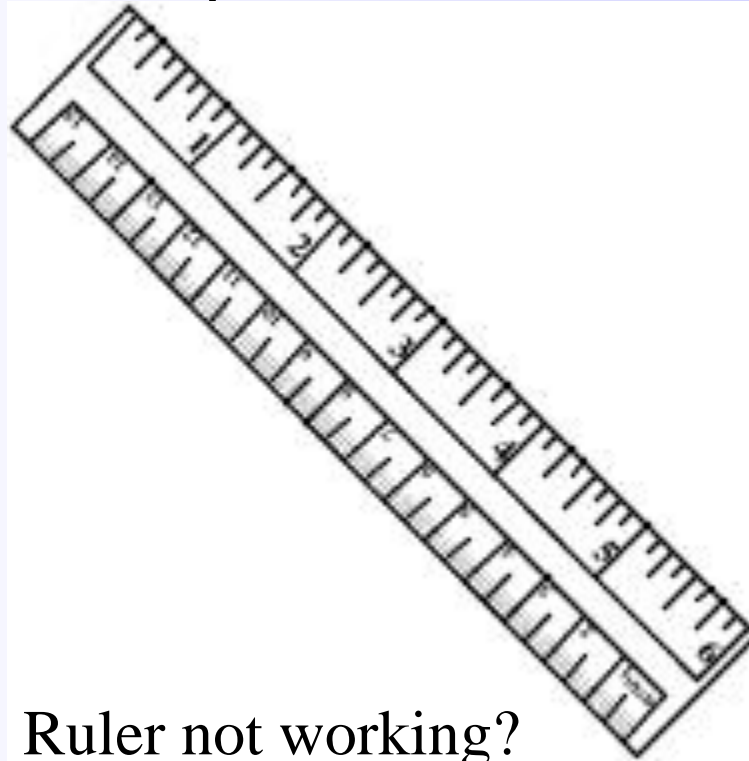
Try a more complex model

These remedies assume your lack-of-fit is not due to one or more “bad data” points in your set of observations.

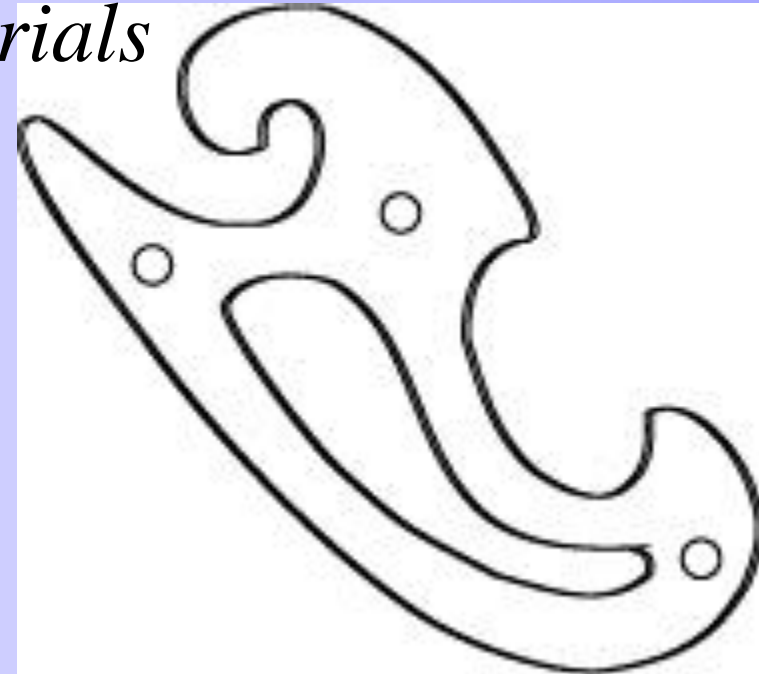
# Remedies for Lack-Of-Fit

More complex model

*Requires additional trials*



Ruler not working?

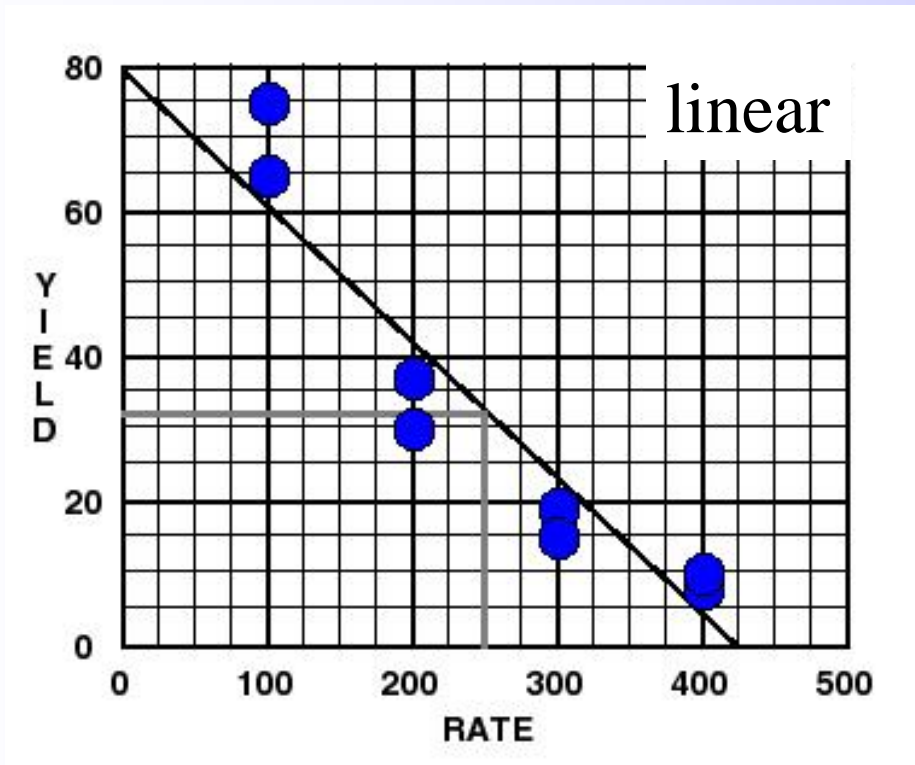


Try a French Curve

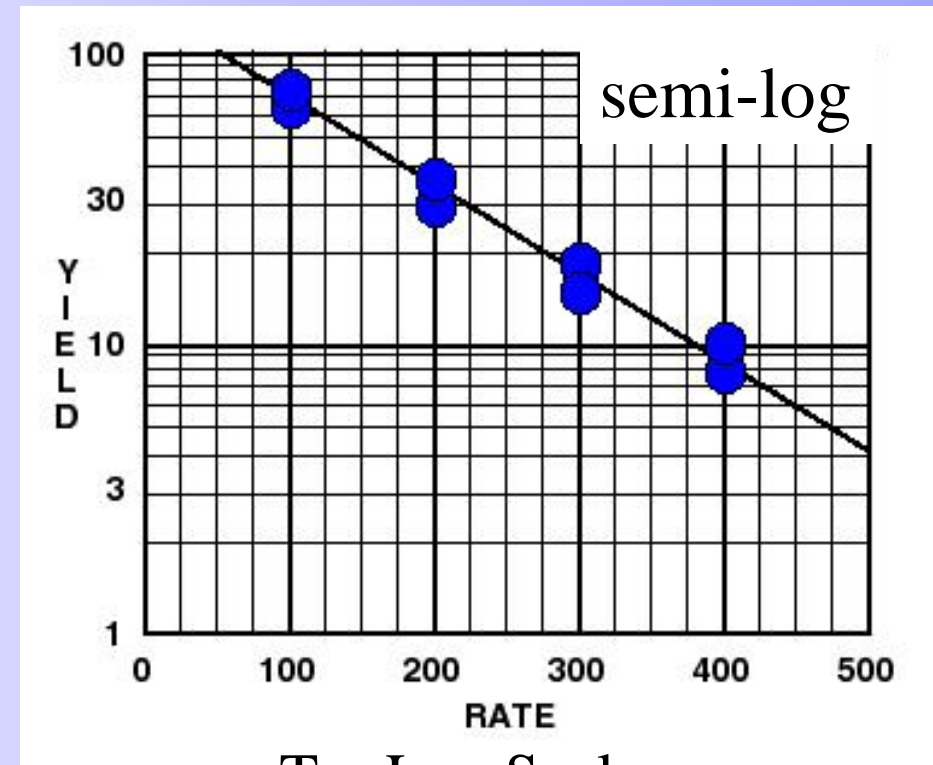
# Remedies for Lack-Of-Fit

Data transformation

*Requires additional expertise*



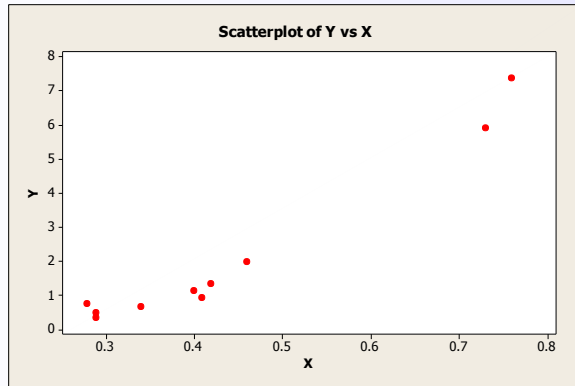
Linear Scale NOT Working



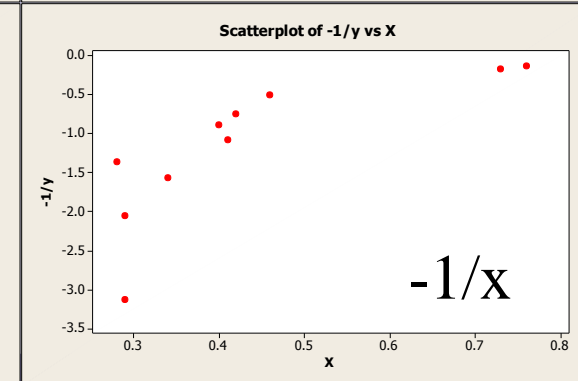
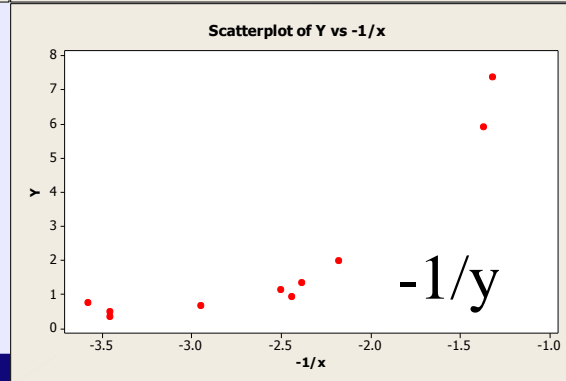
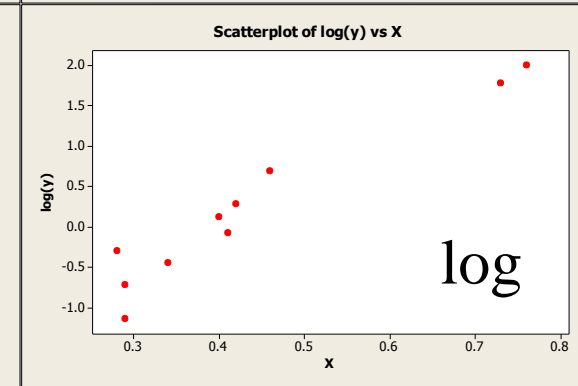
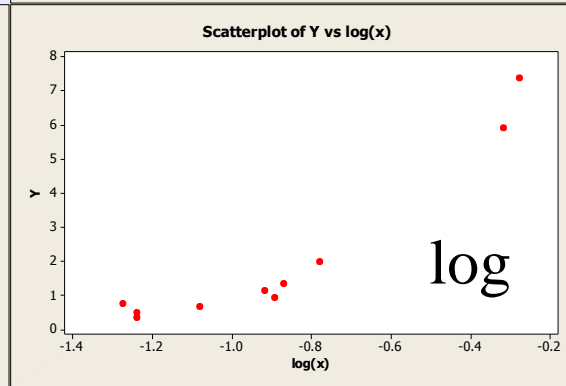
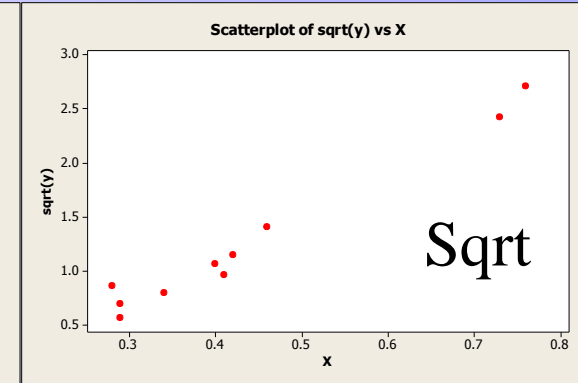
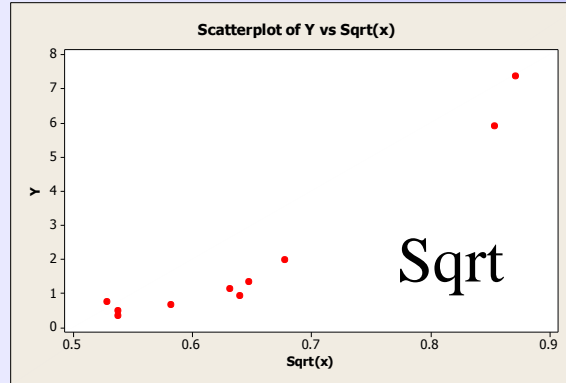
Try Log Scale

# Transformations

Transformations  
just change the  
grid of the graph  
paper



Y vs X  
Goal is to fit a  
straight line AFTER  
transformation



Y Transforms

X Transforms

## What's a process capability study?

- A designed experiment with just one variable- TIME
- It captures all of the natural and unnatural variation in the process
- It includes drift, operators, shifts, machines... ALL of the variation that the process can see.



# Conclusions

- Best method for understanding complex processes
- Best for resolving yield and reliability problems
- Provides statistical validity to know what REALLY matters
- Allows understanding the interactions between process parameters
- Required for High Reliability processes