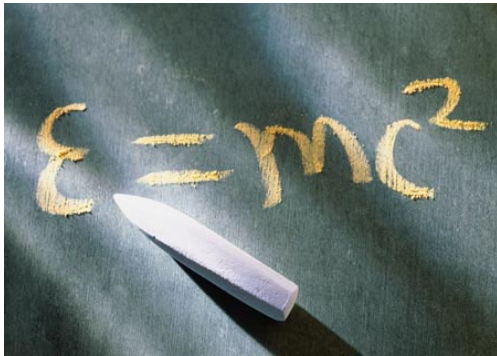


# DFSS RELIABILITY SYSTEM THINKING ASSESSMENT

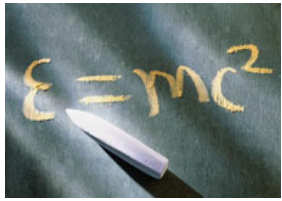
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**Forrest W. Breyfogle III**  
**CEO, Founder**  
**Smarter Solutions, Inc.**

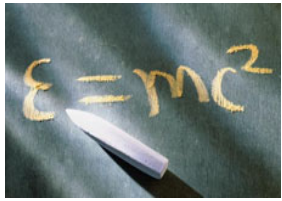
**Example 43.3 *Implementing Six Sigma*, 2<sup>nd</sup> edition**





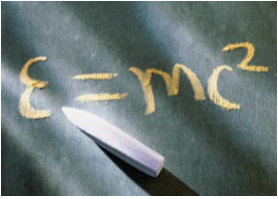
## Objective

- ◆ Compare traditional reliability testing to a Smarter Six Sigma Solutions (S<sup>4</sup>) alternative
- ◆ In this S<sup>4</sup> alternative, a unique Design of Experiments (DOE) analysis technique is described, which can provide more insight to potential future product customer issues than traditional reliability testing
- ◆ These benefits are achieved with much less test effort and could be used as a component of an organization's DFSS development process



## Test Definition

- ◆ A power supply has sophisticated design requirements
  - In addition, the specification indicates an aggressive MTBF (mean time between failures) criterion of  $10 \times 10^6$  hr
- ◆ A test organization is to evaluate the reliability and function of the non-repairable power supply
  
- ◆ Tests of this type seem to be directed toward emphasizing testing the failure rate criterion by exercising enough units long enough to verify the criterion
  - Considering that the failure rate of the units is constant with age, as the criterion implies

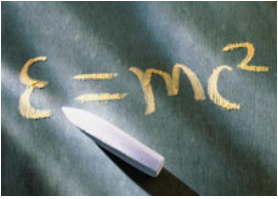


# Reliability Failure Criteria Testing

- ◆ Table K in *Implementing Six Sigma*, 2nd edition can be used to determine a factor for determining the total number of test hours needed
- ◆ If we desire 90% confidence with a test design that allows no failures, then the factor would be 2.303, which would yield a total test time of

$$T = 2.303 (10 \times 10^6) = 23.03 \times 10^6 \text{ hr}$$

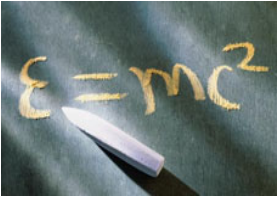
**If had a hypothesis test where tried to protect both  $\alpha$  and  $\beta$  the test hours could be over 600,000,000 hrs**



# Reliability Failure Criteria Testing

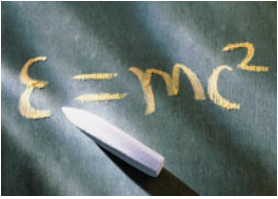
- ◆ Consider that the expected annual number of power-on hours for the power supply is 5000 hr, and each unit is tested to this expected usage
  - This test would require 4605 (i.e.,  $23.03 \times 10^6 / 5000 = 4605$ ) units,
  - While a 5-year test would require 921 units [i.e.,  $23.03 \times 10^6 / (5000)(5) = 921$ ]

Accelerated test alternatives can be helpful to reduce test duration; however, the same basic problems still exist with less magnitude



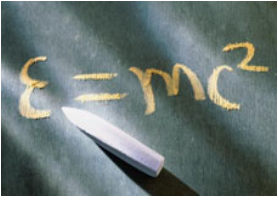
# Reliability Failure Criteria Testing

- ◆ For most scenarios involving complex assemblies, neither of these two test alternatives are reasonable
  - since the unit costs would be prohibitive, the test facilities would be very large,
  - the test would be too long, and
  - information obtained late in the test would probably be "too late" for any "value added"



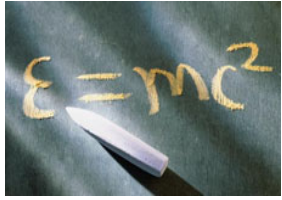
# Reliability Failure Criteria Testing

- ◆ Even if the time and resources are spent to do this test and no failure occurs, customer reliability problems can still exist
- ◆ Two basic assumptions are often overlooked with the preceding test strategy
  - The first assumption is that the sample is a random sample of the population
    - If this test were performed early in the manufacturing process, the "sample" may be the first units built, which is not a random sample of future builds that will go to the customer
    - Problems can occur later in the manufacturing process causing field problems, which this test will not detect since test samples were from an earlier production vintage



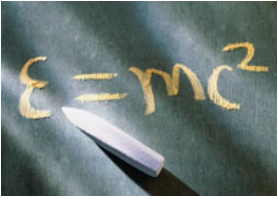
# Reliability Failure Criteria Testing

- The second assumption is that the test replicates customer usage
  - If a test does not closely replicate customer situations, real problems may not be detected
  - For example, if the customer turns off a system unit that contains the power supply each evening and our test units are just exercised continuous operation, the test may miss some thermal cycling component fatigue failures



# Reliability Failure Criteria Testing and DOE Reliability Alternative

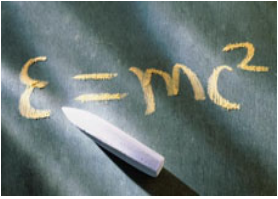
- ◆ Or, another example could be that the customer puts more electrical load on the system than was done on the test system
  - The test may again miss a failure mode caused by this additional loading
- ◆ Perhaps a fractional Design of Experiments (DOE) could better define how the power supplies should be loaded and exercised when trying to identify customer reliability problems



# DOE Reliability Assessment: Design

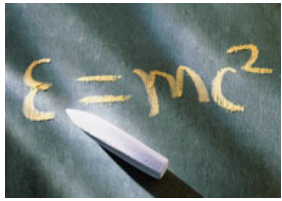
- ◆ For the preceding test, one pre-production unit could be functionally tested at the extremes of its operating environment using a fractional factorial test strategy
- ◆ The following is such a strategy where input factors are evaluated for their effect on the various important output characteristic requirements of the power supply

Inputs			
		Levels	
Factors		-	+
A:	Ambient Temperature	47 deg C	25 deg C
B:	Input ac voltage range	110V	220 V
C:	Mode of programmable output	3.4 V	5.1 V
D:	ac line voltage (within range in B)	Min	Max
E:	Frequency at ac input	Min	Max
F:	Load on -12 V output	Min	Max
G:	Load on -5 V Output	Min	Max
H:	Load on 12 V output	Min	Max
I:	Load on 5.1 V output	Min	Max
J:	Load on 3.4 V output	Min	Max
K:	Load on programmable output	Min	Max
Outputs			
Output voltage on each output (-12V, -5V, 12V, 5.1V, 3.4V, programmable volt output)			
	Ripple/noise		
	Noise		
	Input (power factor)		
	Efficiency		
	Line current		
	Line power		



# DOE Reliability Assessment: Design

- ◆ From Table M4 (*Implementing Six Sigma*, 2<sup>nd</sup> edition), a 32-trial resolution IV design was chosen
  - With this design, the main effects would not be confounded with two-factor interactions
    - however, there would be confounding of two-factor interactions with each other
- ◆ The 11 contrast columns from Table M4 were assigned alphabetical factor designations from left to right (A - K)

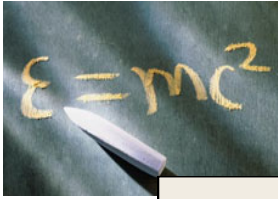


# DOE Reliability Assessment: Results

- ◆ Example data from these test trials, along with two of the experimental trial outputs (-12 and 3.4 voltage level outputs) are:

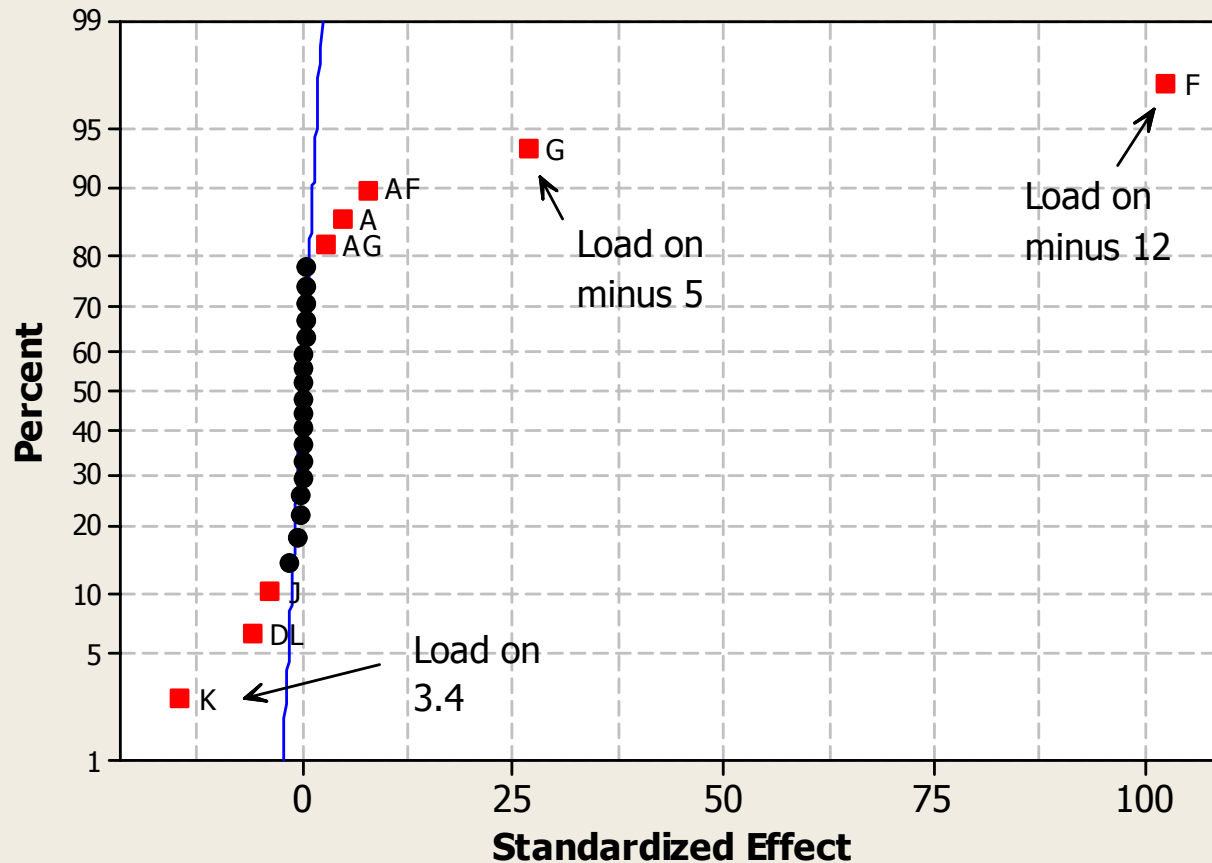
Trial	A	B	C	D	E	F	G	H	I	J	K	-12V	3.4V
1	+	-	-	-	+	+	+	+	-	+	-	-11.755	3.1465
2	+	+	-	-	-	+	+	-	+	-	-	-11.702	3.3965
3	+	+	+	-	-	-	+	+	-	+	+	-12.202	3.147
4	+	+	+	+	-	+	-	+	+	-	-	-11.813	3.4038
5	+	+	+	+	+	+	+	+	+	+	+	-11.761	3.1537
6	-	+	+	+	+	-	+	-	+	+	-	-12.2	3.1861
7	-	-	+	+	+	-	-	+	-	+	+	-12.325	3.1902
8	+	-	-	+	+	+	-	+	+	-	+	-12.292	3.398
9	+	+	-	-	+	+	-	-	+	+	+	-11.872	3.1498

# DOE Reliability Assessment: -12V Analysis



## Normal Probability Plot of the Standardized Effects

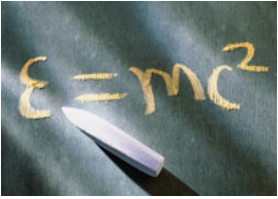
(response is -12V, Alpha = .10)



Effect Type	
●	Not Significant
■	Significant

Factor	Name
A	A
B	B
C	C
D	D
E	E
F	F
G	G
H	H
J	I
K	J
L	K



# DOE Reliability Assessment: -12 V Analysis

Fractional Factorial Fit

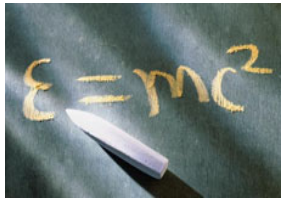
Estimated Effects and Coefficients for -12V (coded units)

Term	Effect	Coef	StDev Coef	T	P
Constant		-12.02	0.001804	-7E-03	0.000
A	0.02	0.01	0.001804	5.54	0.000
B	0.00	0.00	0.001804	0.42	0.683
C	0.00	0.00	0.001804	0.42	0.683
D	0.00	0.00	0.001804	0.17	0.864
E	0.00	0.00	0.001804	0.69	0.497
F	0.43	0.22	0.001804	120.03	0.000
G	0.11	0.06	0.001804	31.35	0.000
H	-0.01	-0.00	0.001804	-1.80	0.088
I	-0.02	-0.01	0.001804	-4.54	0.000
J	-0.06	-0.03	0.001804	-17.08	0.000
K	0.00	0.00	0.001804	0.66	0.519
A*F	0.03	0.02	0.001804	9.01	0.000
D*K	-0.03	-0.01	0.001804	-6.96	0.000

Smaller magnitudes

Load on -12V

Load on -5V



# DOE Reliability Assessment: -12 V Analysis

## Fractional Factorial Fit

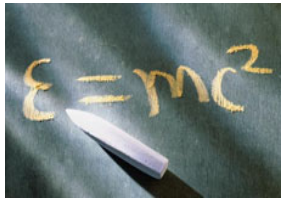
### Estimated Effects and Coefficients for -12V (coded units)

Term	Effect	Coef	StDev Coef	T	P
Constant		-12.02	0.001804	-7E-03	0.000
A	0.02	0.01	0.001804	5.54	0.000
B	0.00	0.00	0.001804	0.42	0.683
C	0.00	0.00			
D	0.00	0.00			
E	0.00	0.00			
F	0.43	0.22			
G	0.11	0.06			
H	-0.01	-0.00			
I	-0.02	-0.01			
J	-0.06	-0.03			
K	0.00	0.00			
A*F	0.03	0.02			
D*K	-0.03	-0.01	0.001804	-6.96	0.000

▲ The effect of the -12-V loading (factor F), for example, on the -12-V output level is simply the difference in average output response for the trials at the high load to those at low load, which is:

Average effect on -12-V output by -12-V load (F effect)

$$\begin{aligned} &= \frac{(-11.755 - 11.702, \dots)}{16} - \frac{(12.202 - 12.200, \dots)}{16} \\ &= -0.43 \text{ V} \end{aligned}$$



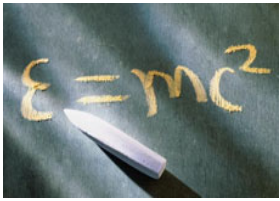
# DOE Reliability Assessment: -12 V Analysis

Analysis of Variance for -12V (coded units)

Source	DF	Seq SS	Adj SS	Adj M	F	P
Main Effects	11	1.63935	1.63935	0.149032	1E+03	0.000
2-Way Interactions	2	0.01350	0.01350	0.006750	64.80	0.000
Residual Error	18	0.00187	0.00187	0.000104		
Total	31	1.65473				

- ◆ Consider where the error estimate coming from
  - Response: Other two-factor and higher interaction term effects

# DOE Reliability Assessment: -12 V Analysis



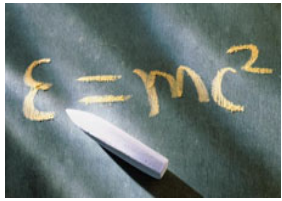
Alias Structure (up to order 2)

I  
A  
B  
C  
D  
E  
F  
G  
H  
I  
J  
K

- ◆ Consider what the resolution is for this design
  - Response: Two-factor interactions are confounded with each other but not with the main effects

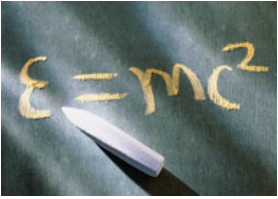
A\*F + B\*I + C\*D + E\*J  
D\*K + E\*H + F\*G

- ◆ A subset model of only the significant factors would next be appropriate; this analysis would show
  - Because of the design balance the effects do not change
  - The changes in probability value would not affect our basic decision
  - The residual plots would not change much



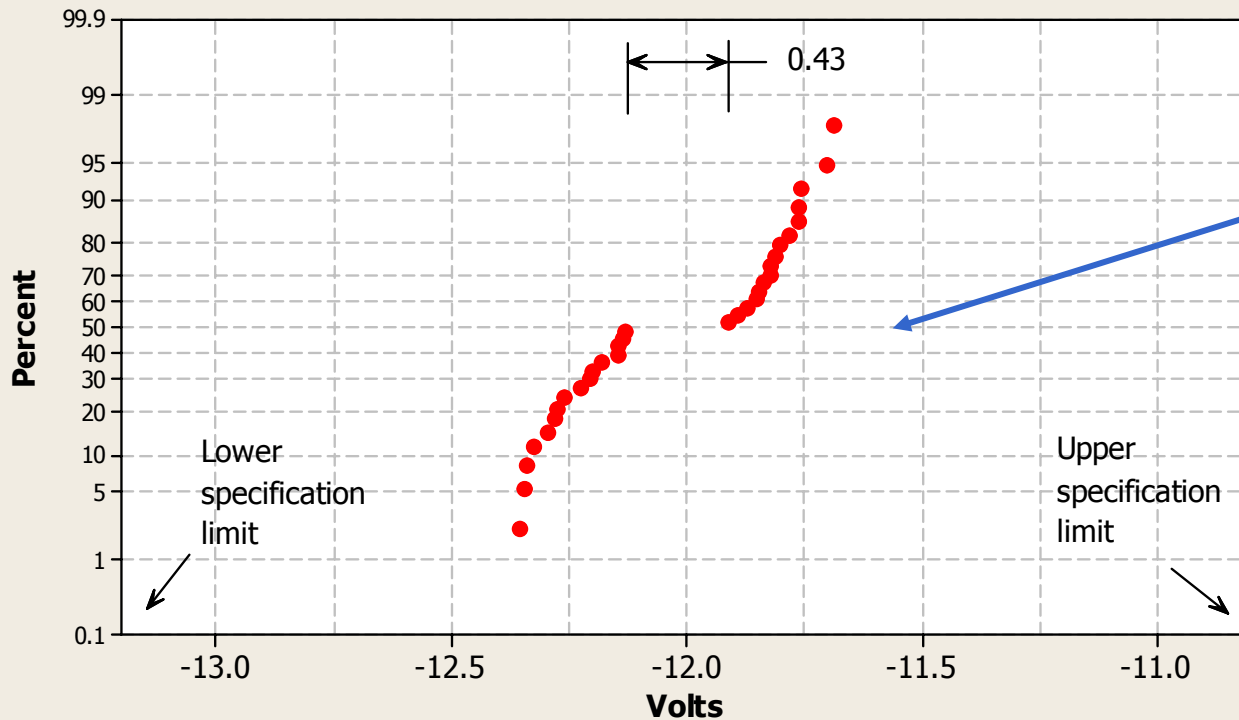
# DOE Reliability Assessment: -12 V Analysis

- ◆ The results of statistical analyses are commonly presented as significance statements
  - However, a practitioner may be interested in the overall effects relative to specification limits
- ◆ One approach to size this consideration is to make a probability plot of the outputs of the 32 trials and include the specification limits on the plot
  - however, note that this plot is not a true random sample plot of a population
- ◆ This approach is call DOE Collective Response Capability Assessment (DCRCA)



# DOE Reliability Assessment: -12 V Analysis

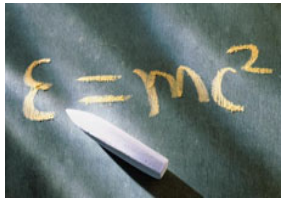
Probability Plot of -12V  
Normal



This plot reflects the variability of one machine, given various worst-case loading scenarios

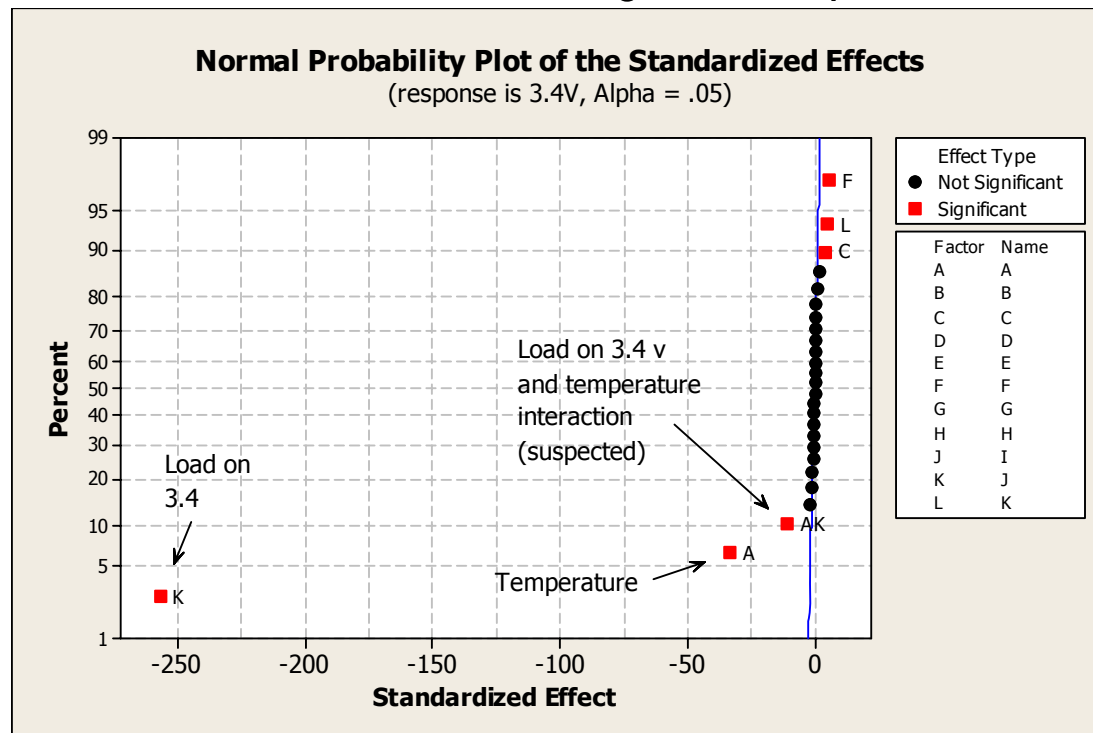
The gap between points is approximately the same magnitude as the effect from circuit loading

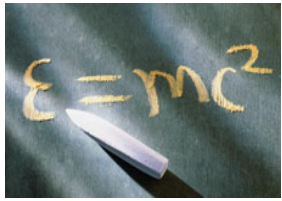
More on next slide



# DOE Reliability Assessment: 3.4 V Analysis

- ◆ The 3.4-V loading effect appears most significant, followed by the temperature effect
  - The third most significant effect is noted to be a suspicion of an interaction between this loading and temperature



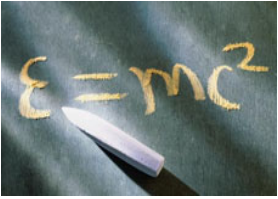


# DOE Reliability Assessment: 3.4 V Analysis

Fractional Factorial Fit

Estimated Effects and Coefficients for 3.4V (coded units)

Term	Effect	Coef	StDev Coef	T	P	
Constant		3.2891	0.000399	8238.12	0.000	
A	-0.0317	-0.0159	0.000399	-39.70	0.000	← Temperature
B	-0.0017	-0.0009	0.000399	-2.19	0.041	
C	0.0045	-0.0022	0.000399	5.59	0.000	← Smaller Effects
D	-0.0007	-0.0004	0.000399	-0.94	0.359	
E	-0.0004	-0.0002	0.000399	-0.47	0.644	
F	0.0055	0.0028	0.000399	6.93	0.000	← Temp*Load 3.4 V suspected
G	-0.0002	-0.0001	0.000399	-0.28	0.781	
H	0.0000	0.0000	0.000399	0.05	0.963	
I	0.0002	0.0001	0.000399	0.30	0.769	
J	-0.2447	-0.1224	0.000399	-306.50	0.000	← Load on 3.4 V output
K	0.0049	0.0025	0.000399	6.15	0.000	
A*J	-0.0100	-0.0050	0.000399	-12.55	0.000	

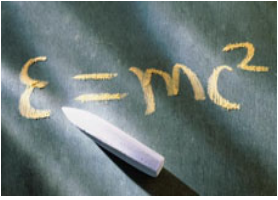


# DOE Reliability Assessment: 3.4 V Analysis

Fractional Factorial Fit	Estimated Effects and
Term	Effect
Constant	
A	-0.0317
B	-0.0017
C	0.0045
D	-0.0007
E	-0.0004
F	0.0055
G	-0.0002
H	0.0000
I	0.0002
J	-0.2447
K	0.0049
A*J	-0.0100

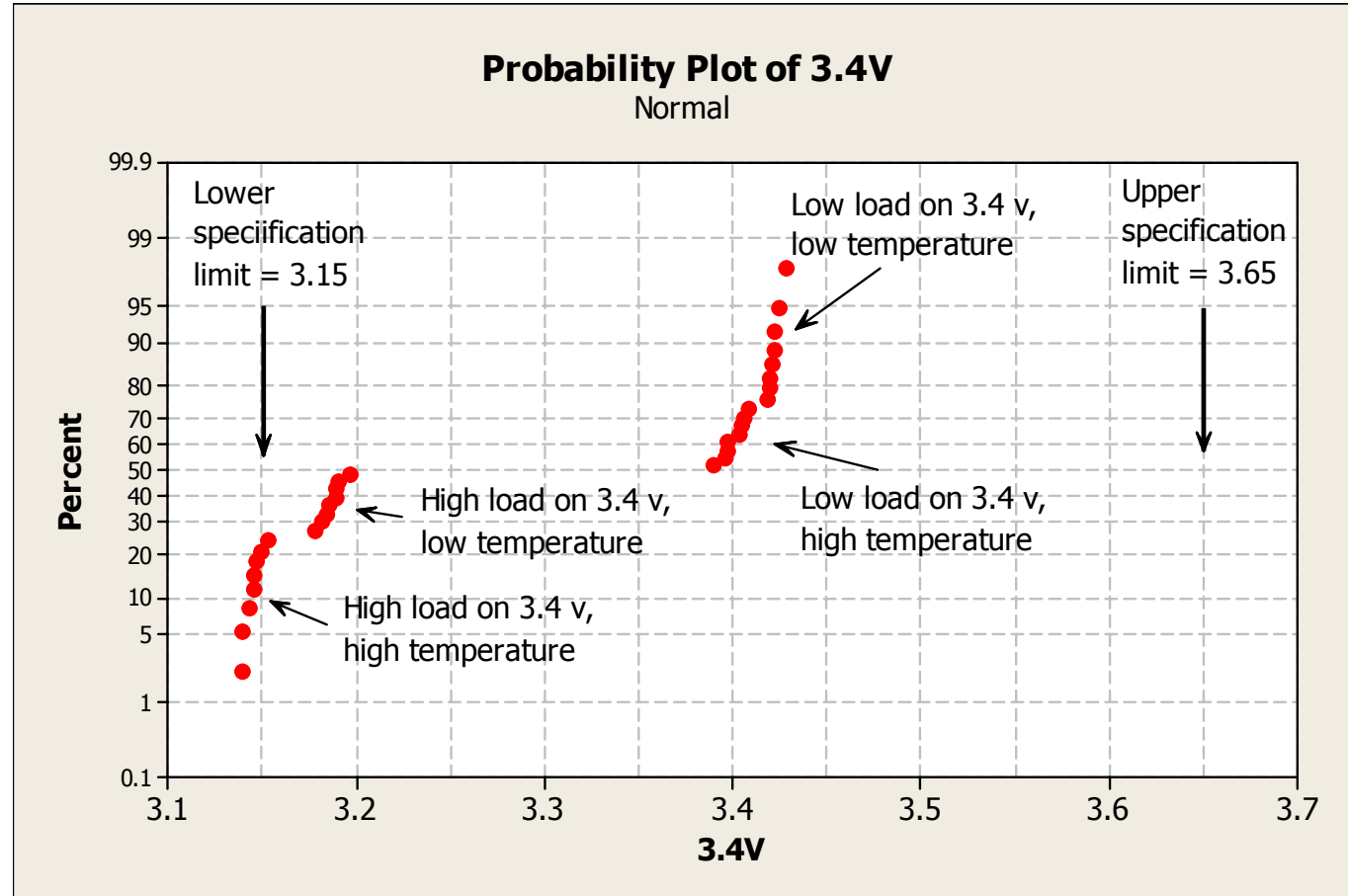
- Two-factor interactions are confounded within this resolution IV experiment design
  - This factor is in reality contrast column 15 (see the design matrix in Table M4)
- Table N2 indicates that contrast 15, in general, has many interactions that are confounded
  - However, the number of interaction possibilities is reduced, since this experiment had only 11 factors (designated A through K)
    - because of this the number of possibilities is reduced to EF, GH, AJ, and BK
- Of these possibilities we believe that the AJ interaction is the significant one

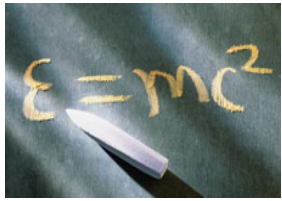
-0.0050      0.000599      -12.55      0.000



# DOE Reliability Assessment: 3.4 V Analysis

- ◆ Similarly to the -12-V analyses, let's next create a probability plot of the 32 trial outputs
  - remembering that this plot is not a true random sample plot of the population

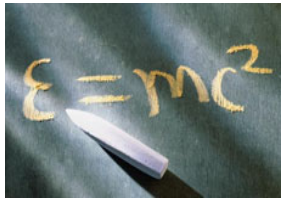




## DOE Reliability Assessment: 3.4 V Analysis

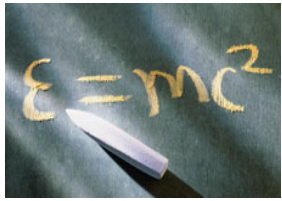
- ◆ However, some of the data points fall below the lower specification limit
  - It appears that the supplier is adjusting the power supply to a 3.4-V output under a low-load condition
    - However, with additional load, the output decreases to a value that is close to the specification limit
- ◆ It is apparent from the out-of-specification condition that the voltage adjustment procedure at the supplier should be changed to "center" the high/low loading conditions within the specification limits

Note, this power supply could cause an intermittent problem (i.e., failure at high load when it heats up)



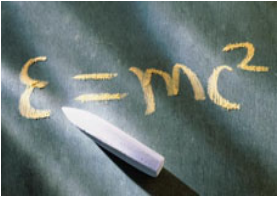
## DOE Reliability Assessment: 3.4 V Analysis

- ◆ The next question of concern is whether there are any other parameters that should be considered
  - One possible addition to the variability conditions is circuitry drift with component age
    - Another is variability between power supply assembly units
- ◆ To address the between-assemblies condition, multiple units could have been used for these analyses
  - However, the test duration and resource requirements could be much larger
- ◆ An alternative is to consider an addition derived from historical information; however, this information is not often available
  - Another alternative is to evaluate a sample of parts, held at constant conditions, to assess the magnitude of this variability
    - This effort can also serve as a confirmation experiment to assess whether the conclusions that were drawn from the fractional factorial experiment are valid



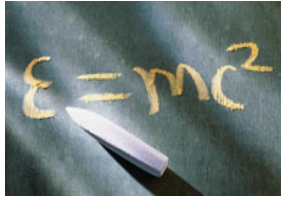
## DOE Reliability Assessment: 3.4 V Analysis

- ◆ A test of 10 power supply assemblies is plotted
- ◆ The test data consists of two points for each assembly, taken at low load/low temperature and high load/high temperature on the 3.4-V output and
  - The plot indicated that approximately 99.8% of the population variability is within a 0.1-V range
- ◆ Allowance could be made for this 0.1-V variation adder in the 3.4-V analyses



# DOE Reliability Assessment: Benefits

- ◆ After the power supply qualification test process is completed satisfactorily,
  - information from this test can be used to determine which parameters need to be monitored within the manufacturing process
- ◆ If this problem escaped development tests, manufacturing would probably be reporting this issue as a “no trouble found” when failing units were returned by the customer
  - “No trouble found” issues can become very expensive problems for an organization



## Summary

- ◆ The described methodology can help organizations improve their enterprise design process efficiency and effectiveness
- ◆ In a computer-industry DFSS Design-Measure-Analyze-Design-Verify (DMADV) process, a verify-phase application of this procedure can help reduce the number of future post-ship non-trouble finds

