Application of Simulation to Sample Size Calculations and Teaching Statistical Concepts



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■ JMC Data Experts

Question: Which Door Should I Take?



One Car and Two Goats

- 1. Contestant choses a door.
- 2. Monty opens one door with a goat.
- 3. Monty offers the contestant the remaining door or stay with their door.

1. Switch to the other door ?

2. Stay with your door?

3. It doesn't matter?





The Right Answer Explained (?)

Since you seem to enjoy coming straight to the point, I'll do the same. You blew it! Let me explain. If one door is shown to be a loser, that information changes the probability of either remaining choice, neither of which has any reason to be more likely, to 1/2. As a professional mathematician, I'm very concerned with the general public's lack of mathematical skills. Please help by confessing your error and in the future being more careful. Robert Sachs, Ph.D. George Mason University

You blew it, and you blew it big! Since you seem to have difficulty grasping the basic principle at work here, I'll explain. After the host reveals a goat, you now have a one-in-two chance of being correct. Whether you change your selection or not, the odds are the same. There is enough mathematical illiteracy in this country, and we don't need the world's highest IQ propagating more. Shame! Scott Smith, Ph.D. University of Florida



Could a Quick Simulation Answer the Question Definitively?

GMacro Goat3 Name C1 "Door" Name C2 "Car" Name c3 "Guess" Name c4 "Wins" Do k1 = 1 : 1000Sample 1 'Door' 'Car' Sample 1 'Door' 'Guess' Let c4[k1] = 'Car' = 'Guess'Enddo Name k2 "Stay Wins" Name k3 "Switch Wins" Let $k^2 = Sum(c^4)$ Let k3 = 1000 - k2Print k2 k3 endmacro



https://math.ucsd.edu/~crypto/Monty/montybg.html



Agenda

- The Monty Hall Paradox
- Confidence Intervals for the Mean and Standard Deviation
 - Percent variability for the mean and standard deviation
 - Impact on process capability
 - Effect on MSA measures of capability and measuring repeatability
- Sampling Plans for Expanded Gage R & R
 - Simulation results for sources of variation
 - ✤ The best sampling plan when variation sources cannot be estimated
- Analysis of Variance Study when Factors are Nested
 - Syringe needle strength where is the variability coming from?
 - Study results for the complete study
 - Simulation results on the effect of sample size on study results
- Cleaning Process Factorial with a Non-normal Covariate
 - Factorial design for a cleaning process with a non-normal covariate
 - Power and sample size calculations
 - Simulation for the effect of sample size on averaging the covariate
- Questions & Discussion

Non-disclosure Statement



"YOU CAN TELL ME ANYTHING, MRS., ROJAK. I'D NEVER VIOLATE THE SACRED FISH STORE-CUSTOMER RELATIONSIHIP."

All industrial experiments, results and scenarios are based on the authors' actual experiences. Data units, variable names, etc have been changed for demonstration purposes only.

The Difference between a Statistician and an Engineer



The engineer takes three measurements (5.5, 6.1, 6.4), takes the average and concludes that the thickness of the fabric is $6.0/1000^{\circ}$, give or take.

The statistician concludes that the engineer took three samples from a population that is likely normally distributed, centered at 6.0 /1000" and has a standard deviation of 1 /1000".

The difference is critical. The second interpretation understands that future samples from the population will be different even if nothing has changed.



How do We Teach this Important Concept of Variability

📠 Minitab - Simulation for Mean and Standard Deviation.mpj -								
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÷	C1	C2	C3	C4	C5			
	Population	Sample						
1	7.44434	6.06817						
2	7.27192	5.58695						
3	4.34465	6.19379						
4	7.46776	6.43897						
5	4.78513	6.29044						
6	5.95534	7.77908						
7	5.09990	5.95274						
8	6.92940	5.93298						
9	4.88650							
10	4.03013							
11	7.49095							
12	4.42031							
13	6.22871							
14	6.04138							
15	5.93097							
16	5.28613							
17	6.32382							
18	6.18136							

Classroom Exercise

- 1. Create a population with a given mean and std deviation.
- 2. Take a random sample of 8 measurements.
- 3. Calculate the descriptive statistics.

Statistics

Variable	Ν	Mean	StDev	Minimum	Maximum
Sample	8	5.169	1.653	2.242	7.128

Statistics

Variable	Ν	Mean	StDev	Minimum	Maximum
Sample	8	6.280	0.658	5.587	7.779

Statistics

Variable	Ν	Mean	StDev	Minimum	Maximum
Sample	8	5.923	1.051	4.935	7.542

Statistics

Variable	N	Mean	StDev	Minimum	Maximum
Sample	8	6.382	0.653	5.110	7.013

Repeat sampling 10,000 times. What is the distribution of the estimates of the mean?

Distribution of 10,000 Estimates of the Mean



Conclusions

- 1. 95% of the estimates of the mean fall between 5.3 and 6.7.
- Your estimate from 8 samples will likely be within + / 12% of the right answer.
- 3. The confidence interval for the mean is Lower limit = M - $Z_{.95}\sigma_M$ Upper limit = M + $Z_{.95}\sigma_M$
 - (6-1.96*.3536,6,6+1.96*.3536) (5.31,6,6.69)

What about the standard deviation?

Confidence interval simulation when the standard deviation is not known for various sample sizes is shown by David M. Lane: http://onlinestatbook.com/2/estimation/ci_sim.html

Distribution of 10,000 Estimates of the Standard Deviation



Conclusions

- 1. 95% of the estimates of the standard deviation between .5 and 1.5.
- Your estimate from 8 samples will likely be within + / - 50% of the right answer.

How does this affect our perspective on process capability estimates and sample size? How does this affect our perspective on measurement system capability?



Source of Variation in Process Capability Estimates (Cpk)



To estimate the mean, 5 – 10 samples might be acceptable. Because of the variability in the standard deviation, 30 samples or more are used to estimate Cpk



Estimating Measurement System Capability

$$\sigma_{Total}^{2} = \sigma_{P}^{2} + \left(\left(\sigma_{O}^{2} + \sigma_{PO}^{2} \right) + \sigma_{\varepsilon}^{2} \right) \\ \sigma_{Total}^{2} = \sigma_{P}^{2} + \left(\sigma_{reproducibility}^{2} + \sigma_{repeatability}^{2} \right) \\ \sigma_{Total}^{2} = \sigma_{P}^{2} + \sigma_{measurement}^{2}$$

% Study Variation =
$$\frac{\sigma_{measurement}}{\sigma_{total}} \times 100\%$$
 Rarely Can Trust This
% Tolerance = $\frac{6 \times \sigma_{measurement}}{USL - LSL} \times 100\%$ My First Choice
% Process Variation = $\frac{\sigma_{measurement}}{\sigma_{Historical}} \times 100\%$ My Second Choice

Fxperts

Expanded Gage R & R Overview Overall Process Variation Part-to-Part Variation



Sampling Plan for an Expanded Gage R & R Study



Sampling Plans for Expanded Gage R & R Studies, Lou Johnson and Daniel Griffith <u>https://www.youtube.com/watch?v=KtGhnimE6Qw</u>

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How the Simulation Works



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95% Confidence Interval for Repeatability – 4 vs 2 Repeats



Sampling Plan for Chromatograph Peak Height Expanded Study

4 Chromatographs (Gage)

4 Techs

(Operator)

4 Columns

(Part)

- 4 Columns used to run the 4 Chromatographs and 4 Techs in parallel
- 2 Repeats
- 96 measurements

This turns out to be the best overall sampling plan for an expanded study when the variation for each component can not be estimated.

Sample Size for Analysis of Variance Study

A syringe needle manufacturer measures force – till – failure on an Instron mechanical properties gauge.

Product is made and measured at three manufacturing sites. The third manufacturing site does not always meet specifications whereas the other two do. The goal is to determine which sources of variation (process /site, gage, operator, within lot) are the strongest contributors to the variation and therefore out-of-specification measurements.

Sampling Plan to Determine Sources of Variation with Nesting

10 repeat measurements (needles) of 2 Parts crossed with 12 Operators nested in 6 Gages nested in 3 Locations.

Nesting makes it hard to separate sources of variation. More samples (needles) helps, but how many more?

Results of the Study -- Focus on Location and Repeatability

Variance Components

Source	VarComp	%Contribution (of <u>VarComp</u>)
Total Gage R&R	0.0979	12.82
Repeatability	0.8338	10.91
Reproducibility	0.1459	1.91
Location	0.0134	1.76
Operator(Location)	0.0009	0.12
Instron(Location)	0.0002	0.03
Part-To-Part	0.6661	87.18
Part	0.6661	87.18
Total Variation	0.7641	100.00

Repeatability `is the largest source of variation, but this contains the needle – to – needle variation as well as the measurement variation.

Location is the next largest source of variation, eclipsing Operator and Instron.

So this study was successful, but did we need to use all those samples? 10 per condition?

Sample and Resample the Dataset with Different Sample Sizes

Analysis of Variance Model:

Simulations by Cathy Akritas at Minitab, Inc <u>cakritas@minitab.com</u>

Results of the Simulations

Results of the Simulations

Sample Size - 3 Factor Factorial with a Non-Normal Covariate

Sample Size for a 3 Level Factorial with a Non-Normal Covariate

The power calculation can seem reasonably straight forward. 10 samples per treatment is reasonable.

But the response:

Turbidity Before - Turbidity After

is function of **Turbidity Before**.

1.1 - .7 = .4 and 16.4 - 16.0 = .4have completely different meanings even though they are equal.

Even with random sampling , enough samples are needed to present each condition with the same distribution of incoming contamination.

Sample Size for a 3 Level Factorial with a Non-Normal Covariate

How many samples must be taken from this distribution before the samples will duplicate the parent distribution with reasonable consistency.

Randomly sample 5, 10, 15, 20, etc., twenty times and plot their distribution with the parent.

Sample Size to Consistently Replicate the Distribution

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Questions?

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