



Doing More With Less:

From Sample to Population

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Table of Contents



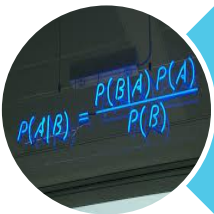
Calculating Sample Size



Bootstrapping

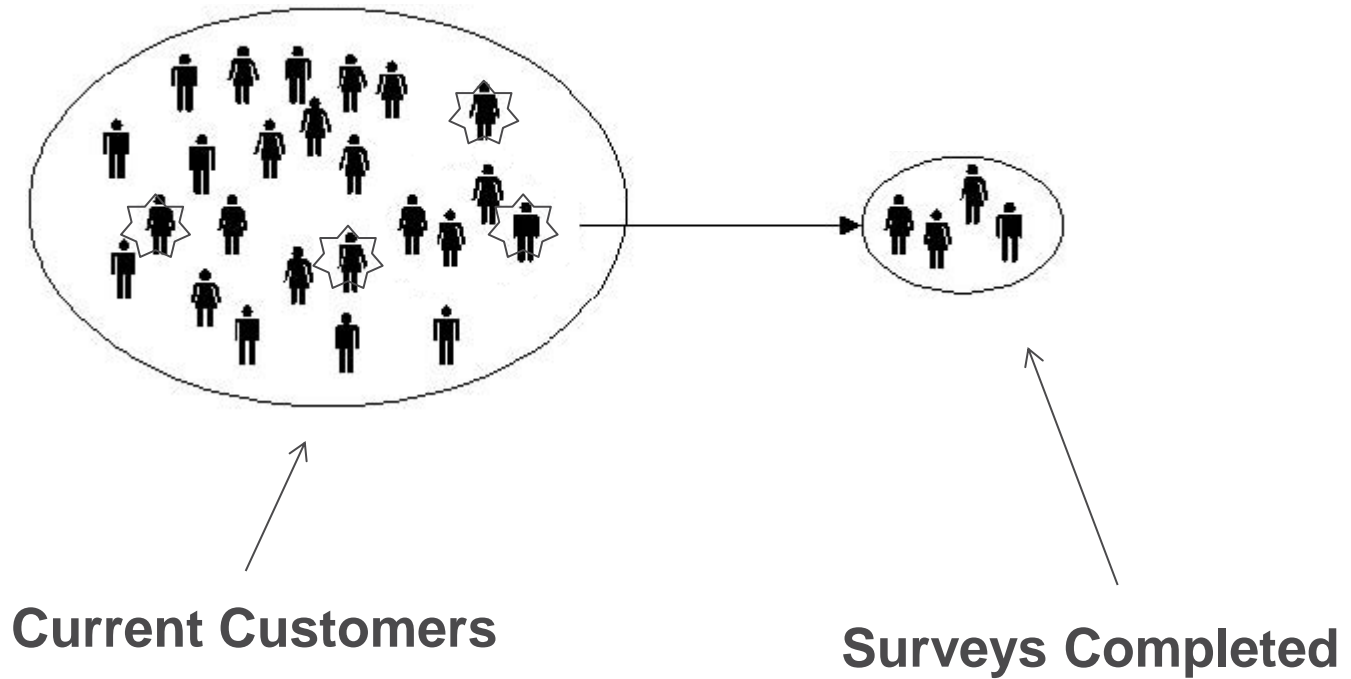


Downsampling



Bayesian Analysis

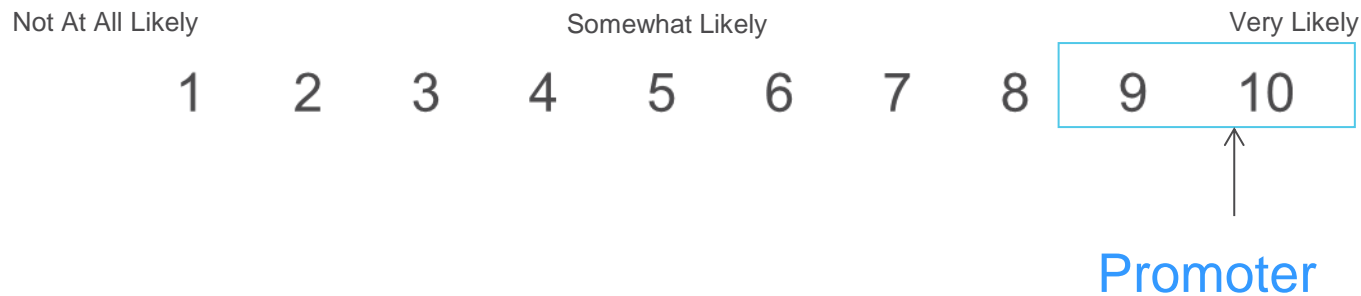
Population and Sample



Measure Proportion of Promoter



What is the likelihood that you would recommend insurance from XXX company to a friend or colleague?



We want to know the proportion of Promoters p .

Hypothesis Testing



Hypothesis testing is a formal process to determine whether to reject a null hypothesis or not, based on the sample data.

- H_0 is null hypothesis
- H_a is alternative hypothesis

In the statistical framework, we either reject a null hypothesis or fail to reject a null hypothesis.

Type I and Type II Error



Decision

Reject H_0

Fail to reject H_0

Truth

H_0

Type I Error

Correct Decision

H_a

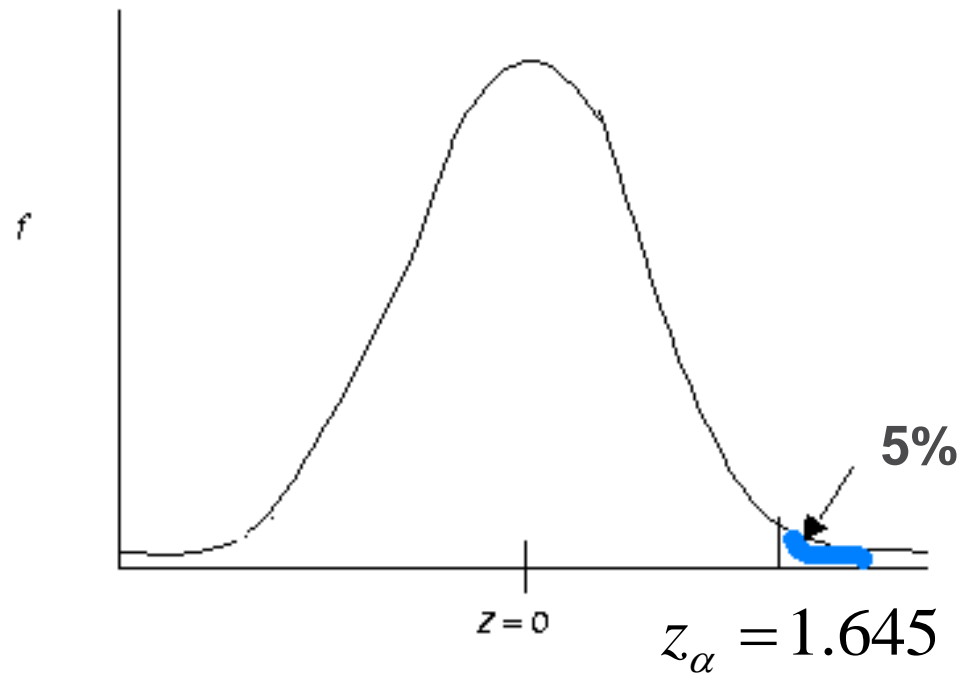
Correct Decision

Type II Error

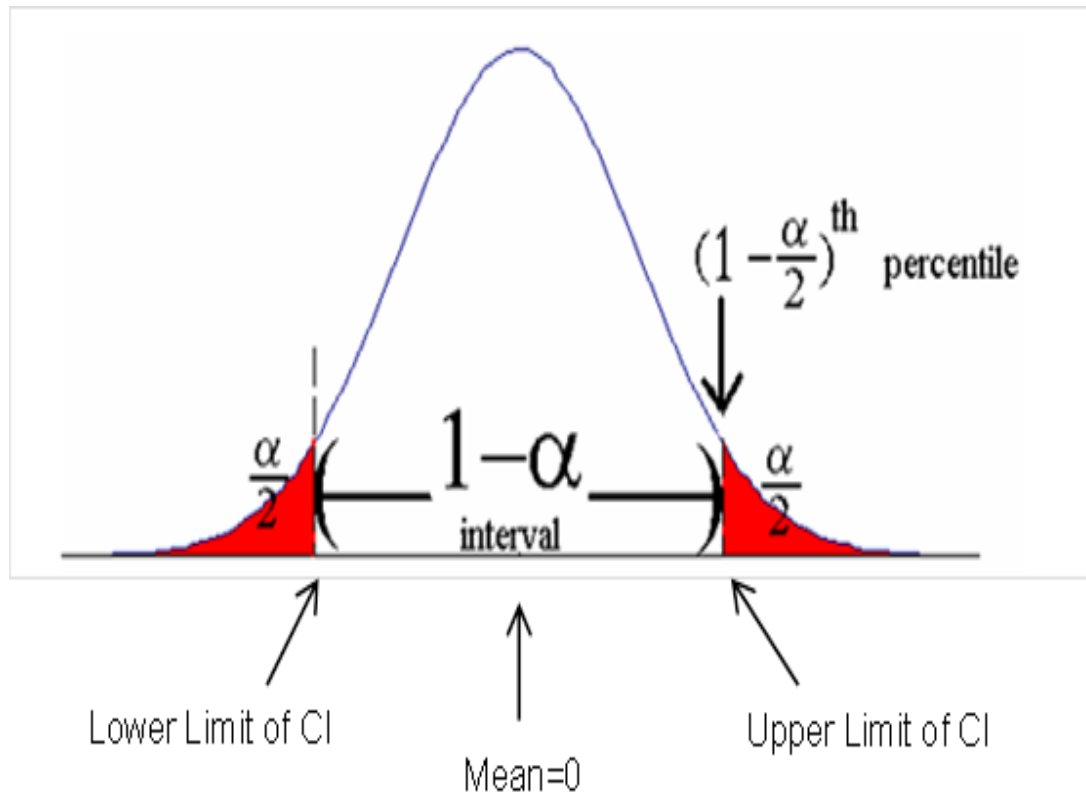
What is z_α



A z-score indicates how many standard deviations an element is from the mean.



Confidence Interval



Traditional Sample Size



The point estimate of p is $\hat{p} = \frac{y}{n}$ and an approximate $1 - \alpha$ confidence interval for p is

$$\hat{p} \pm z_{\alpha/2} \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}}$$

Here $y \sim \text{binomial}(n, p)$.

α is the Type I error.

$\varepsilon = z_{\alpha/2} \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}}$ is the maximum error of the point estimate $\hat{p} = \frac{y}{n}$.



Traditional Sample Size

$$\varepsilon = z_{\alpha/2} \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}}$$

Assume p is about to equal to p^* , then

$$\text{Sample Size}(SS) = n = \frac{z^2_{\alpha/2} p^*(1 - p^*)}{\varepsilon^2}$$

$$SS = \frac{1.645^2 * 0.82(1 - 0.82)}{0.10^2} \approx \mathbf{39.94}$$

From 40 sample surveys, we are 90% confident that the true p belongs to the interval $\hat{p} \pm 0.10$.

Confidence Level	Z score
90%	1.645



Test for Equality

How many surveys per agent are credible in order to reflect a statistically significant change on the Promoter Score?

Steps



1. Set up hypotheses: null and alternative
2. Identify the distributions and key variables
3. Select desirable confidence level and statistical power
4. Calculate sample size needed

Hypothesis on Promoter Score

$$H_0 : p_1 - p_2 = 0 \text{ vs } H_a : p_1 - p_2 \neq 0$$

p_1 is the current proportion of promoters among all customers for an agent.

p_2 is the proportion of promoters for an agent in the next month.

$$\hat{p} = \frac{y}{n} \text{ and } y \sim \text{binomial}(n, p)$$

Derivation



We reject the null hypothesis if

$$\left| \frac{\hat{p}_1 - \hat{p}_2}{\sqrt{(\hat{p}_1(1 - \hat{p}_1) + \hat{p}_2(1 - \hat{p}_2))/n}} \right| > z_{\alpha/2}$$

$$\frac{|p_1 - p_2|}{\sqrt{(\hat{p}_1(1 - \hat{p}_1) + \hat{p}_2(1 - \hat{p}_2))/n}} - z_{\alpha/2} = z_{\beta}$$

$$n_1 = n_2 = \frac{(z_{\alpha/2} + z_{\beta})^2 [p_1(1 - p_1) + p_2(1 - p_2)]}{(p_1 - p_2)^2}$$



Sample Size Needed

$$n_1 = n_2 = \frac{(z_{\alpha/2} + z_{\beta})^2 [p_1(1-p_1) + p_2(1-p_2)]}{(p_1 - p_2)^2}$$

$z_{\alpha/2}$ is z statistic given Type I error α

z_{β} is z statistic given Type II error β

Illustrative Example



$$H_0 : p_1 - p_2 = 0 \text{ vs } H_a : p_1 - p_2 \neq 0$$

$$n_1 = n_2 = \frac{(z_{\alpha/2} + z_{\beta})^2 [p_1(1-p_1) + p_2(1-p_2)]}{(p_1 - p_2)^2}$$

$$n_1 = n_2 = \frac{(1.645 + 0.84)^2 [0.80(1-0.80) + 0.86(1-0.86)]}{(0.80 - 0.86)^2} \approx 480.98$$

To conclude a significant change, more samples are needed.

Aggregate the Results



We want to aggregate the results to a regional level.

We can use weighted average by number of customers to scale it back in order to minimize the sample size needed.

	Promoter Proportion	Agency Satisfaction
Big Sample	59.3%	70.2%
Weighted Avg w/ 40 surveys per agent	59.0%	70.5%

Sample Size Calculation



Survey Analysis
Hypothesis Testing



Collect right
amount of sample

Bootstrapping



Random sampling with replacement.



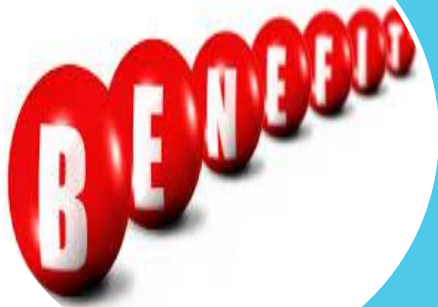


Illustrative Example Cont.

- **Customer Experience Survey**



When Sample is not
Sufficient
Establish Error Bounds



Statistical Inference
with a Small Sample

Downsampling



Down Sampling zero-claims can reduce the time of model convergence during the model development phase.

The standard deviation for parameter estimator will likely increase with the small sample. It won't necessarily reverse modeling decisions.

Modeler can periodically run the model with the entire training data set to verify the model structure.

Once the mode structure is finalized, the entire data set should be used for the final model.

Downsampling



During Model
Development

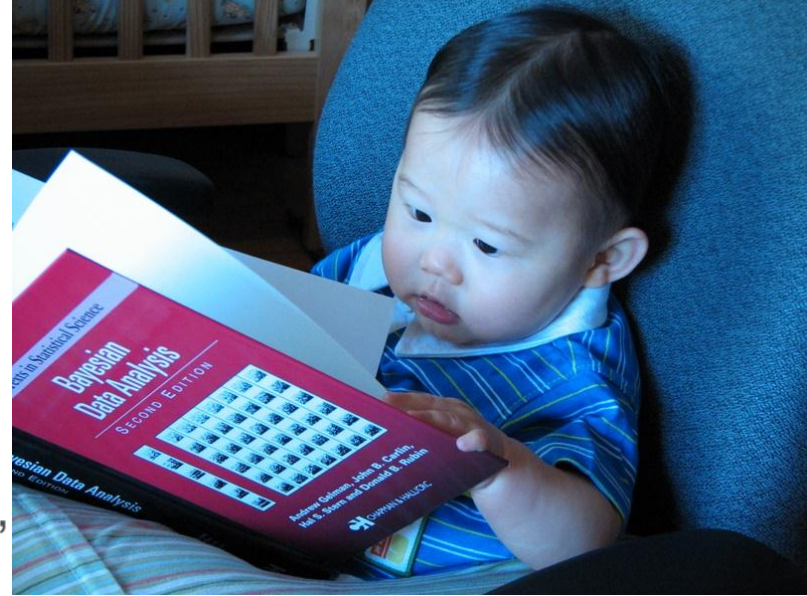


Save Time

Bayesian Analysis

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

Posterior is proportional to “likelihood*prior”



We use known knowledge as input for the prior distribution, and use data collected to calculate the likelihood. The posterior will be updated as more and more data is collected.

As a result, the sample size needed will be reduced.

Bayesian Analysis



Need more flexibility



Less Sample Needed
Allow distributions for
parameters

Summary



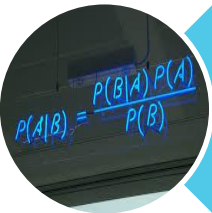
Calculate Sample Size



Bootstrapping



Downsampling



Bayesian Analysis

Questions

