



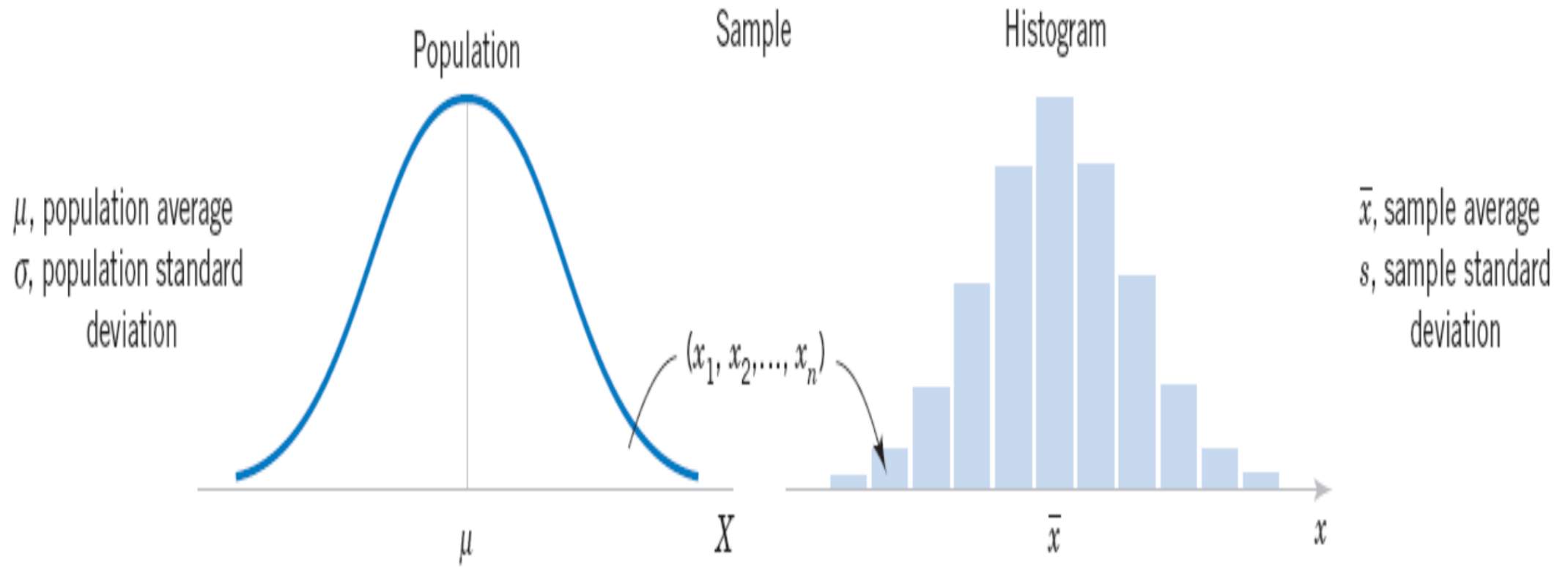
The goal is to turn data into information, and
information into insight.

—Carly Fiorina

Statistical Inference

- The field of statistical inference consists of those methods used to make decisions or draw conclusions about a population.
- These methods utilize the information contained in a sample from the population in drawing conclusions.

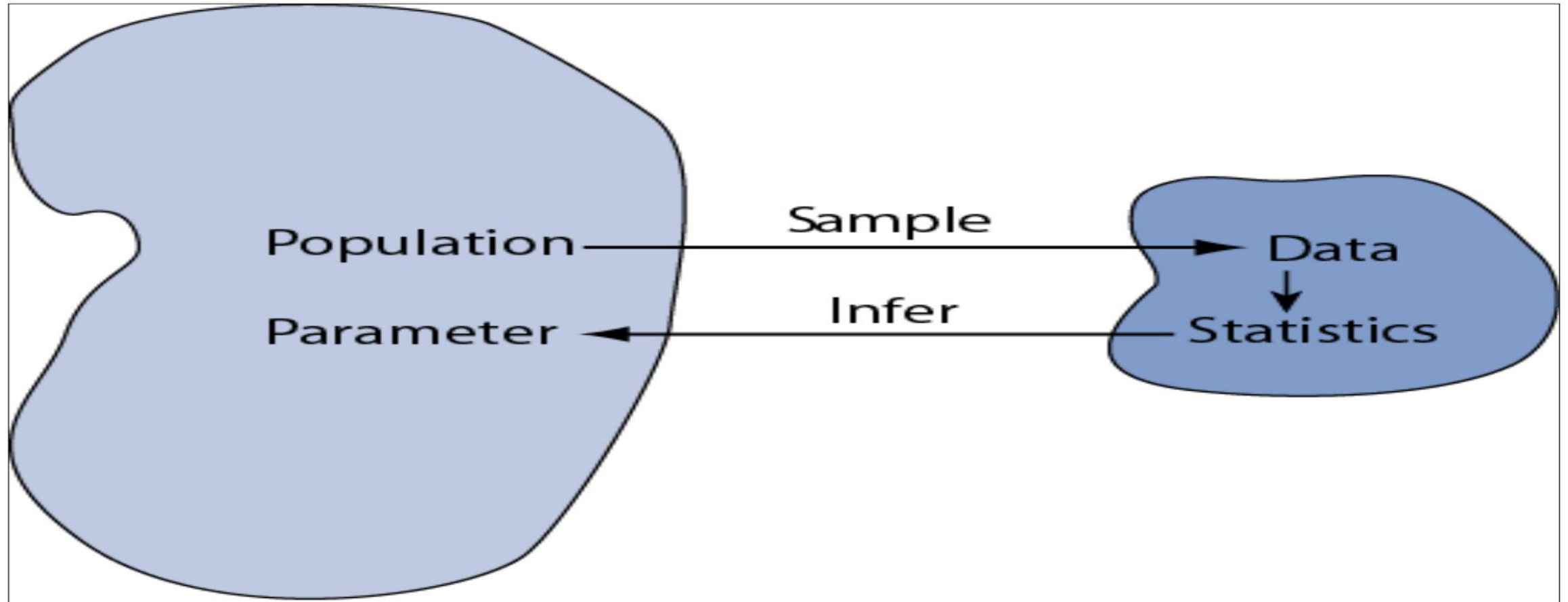
Relationship between a population and sample



Some Important terms

- **Population** \equiv all possible values
- **Sample** \equiv a portion of the population
- **Statistical inference** \equiv generalizing from a sample to a population with calculated degree of certainty
- Two forms of statistical inference
 - **Hypothesis testing**
 - **Estimation**
- **Parameter** \equiv a characteristic of population, e.g., population mean μ
- **Statistic** \equiv calculated from data in the sample, e.g., sample mean ()

Population and sample



Null and Alternative Hypotheses

Convert the research question to null and alternative hypotheses :

- The **null hypothesis (H_0)** is a claim of “no difference in the population”
- The **alternative hypothesis (H_a)** claims “ H_0 is false”
- Collect data and seek evidence against H_0 as a way of bolstering H_a (deduction)

General Procedure of Hypothesis Testing

1. **Parameter of interest:** From the problem context, identify the parameter of interest.
2. **Null hypothesis, H_0 :** State the null hypothesis, H_0 .
3. **Alternative hypothesis, H_1 :** Specify an appropriate alternative hypothesis, H_1 .
4. **Test statistic:** State an appropriate test statistic.
5. **Reject H_0 if:** Define the criteria that will lead to rejection of H_0 .
6. **Computations:** Compute any necessary sample quantities, substitute these into the equation for the test statistic, and compute that value.
7. **Conclusions:** Decide whether or not H_0 should be rejected and report that in the problem context. This could involve computing a P -value or comparing the test statistic to a set of critical values.

Sampling Distributions of a Mean

The **sampling distributions of a mean (SDM)** describes the behavior of a sampling mean

$$\bar{x} \sim N(\mu, SE_{\bar{x}})$$

$$\text{where } SE_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

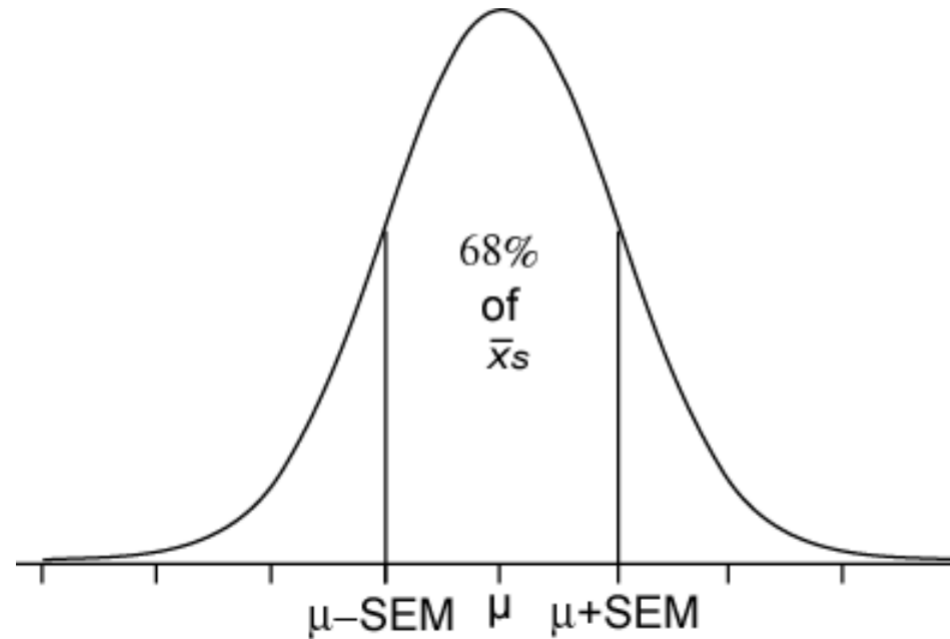


Fig. sdm&se68%.ai

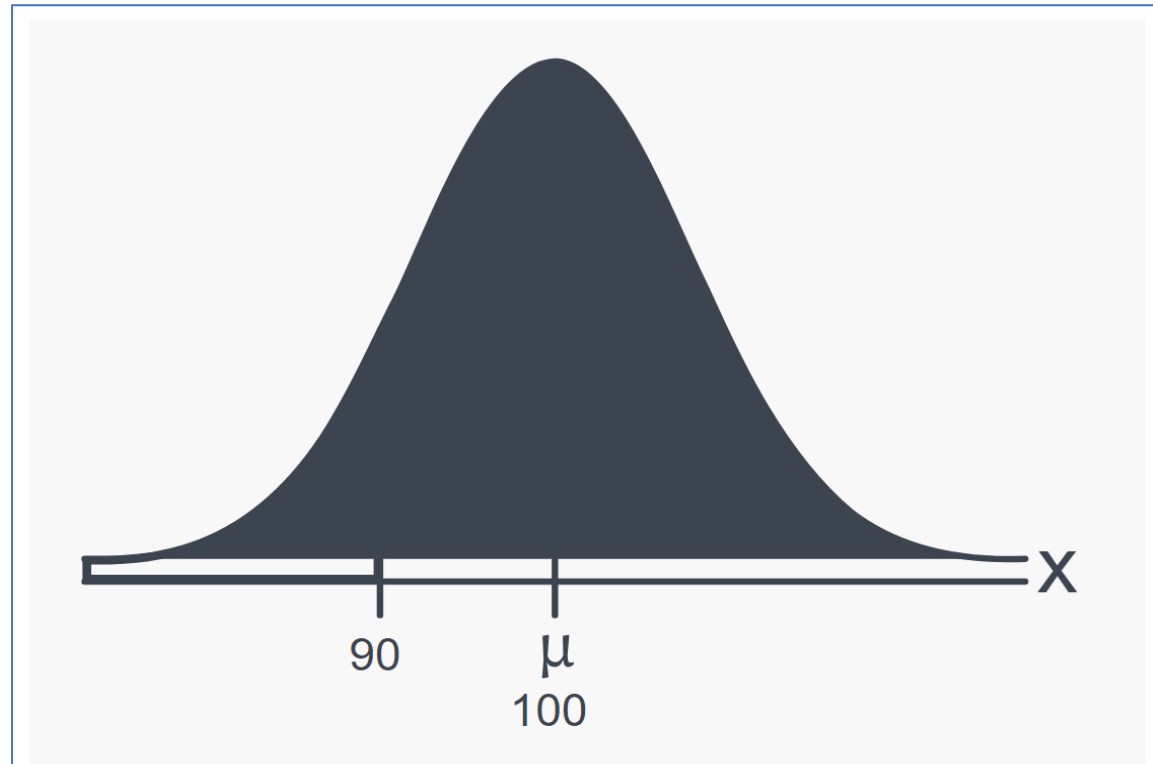
Normal Distribution

- The shape of any normal curve depends on its mean and standard deviation .

- Let X equal the IQ of a randomly selected Indian. Assume mean is 100 and standard deviation is 16. What is the probability that a randomly selected American has an IQ below 90?

If $X \sim N(\mu, \sigma^2)$, then:

$$Z = \frac{X - \mu}{\sigma}$$



Standard Normal Variate

$$P(X) < 90, z = \frac{100-90}{16} = 0.69$$

$$P(Z < 0.69)$$

$$0.75490$$

$$1 - 0.75490 = 24\%$$

Hence, there is 24% probability that the IQ is below 90.

Hypothesis Testing

- We like to think of statistical hypothesis testing as the data analysis stage of a **comparative experiment**, in which the Market Expert is interested, for example, in comparing the mean of a population to a specified value (e.g. mean pull strength).
- Suppose we are interested in knowing that price of a stock is more than 1500.

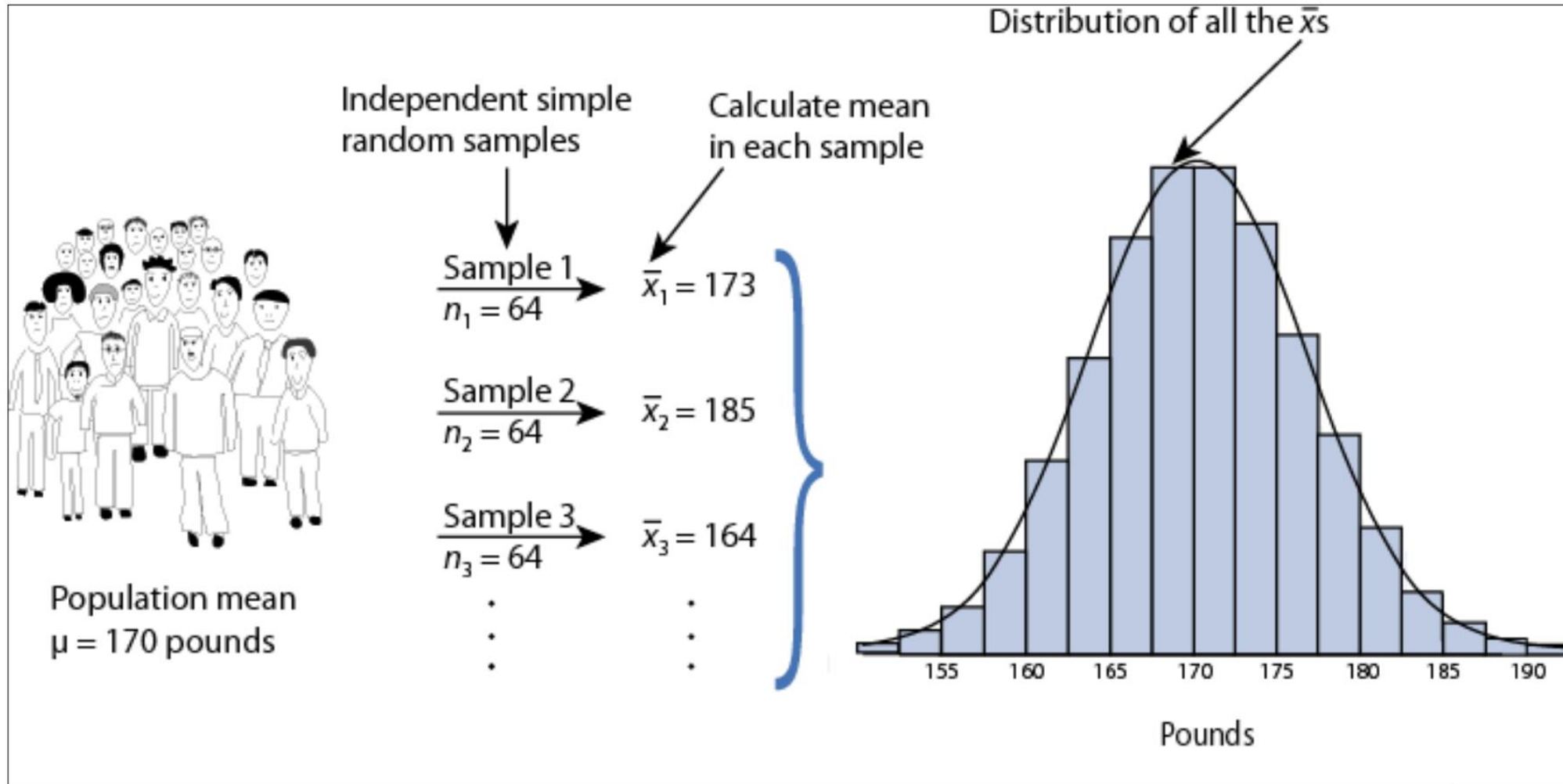
Hypothesis Testing

Two-sided Alternative Hypothesis

$$H_0: \mu = 1500$$

$$H_0: \mu \neq 1500$$

Reasoning Behind $\mu_{z_{stat}}$



Sampling distribution of \bar{x} under $H_0: \mu = 170$ for $n = 64 \Rightarrow \bar{x} \sim N(170, 5)$

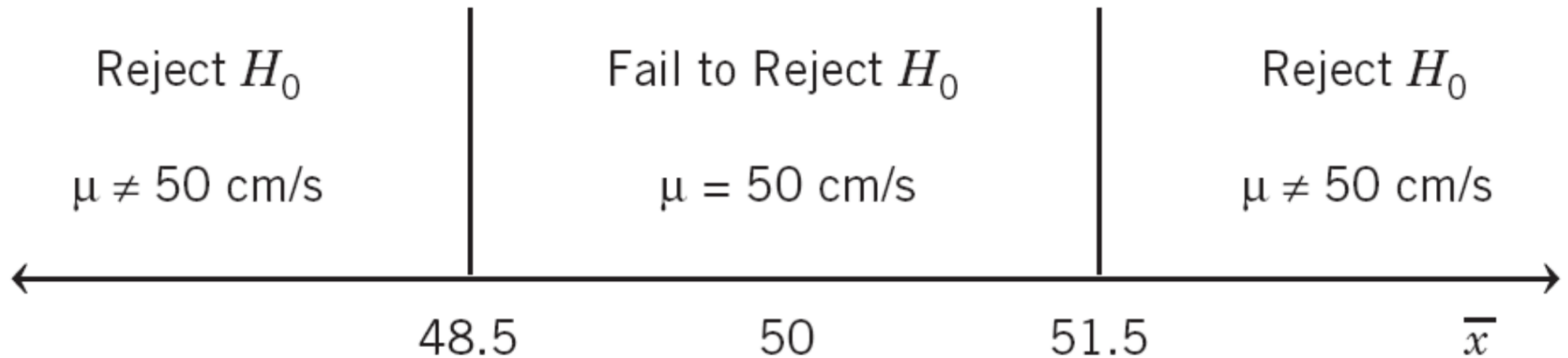
Test of a Hypothesis

- A procedure leading to a decision about a particular hypothesis
- Hypothesis-testing procedures rely on using the information in a **random sample from the population of interest**.
- If this information is *consistent* with the hypothesis, then we will conclude that the hypothesis is **true**; if this information is *inconsistent* with the hypothesis, we will conclude that the hypothesis is **false**.

Testing Statistical Hypotheses

$$H_0: \mu = 50 \text{ cm/s}$$

$$H_1: \mu \neq 50 \text{ cm/s}$$



Testing Statistical Hypotheses

Rejecting the null hypothesis H_0 when it is true is defined as a **type I error**.

Failing to reject the null hypothesis when it is false is defined as a **type II error**.

Testing Statistical Hypotheses

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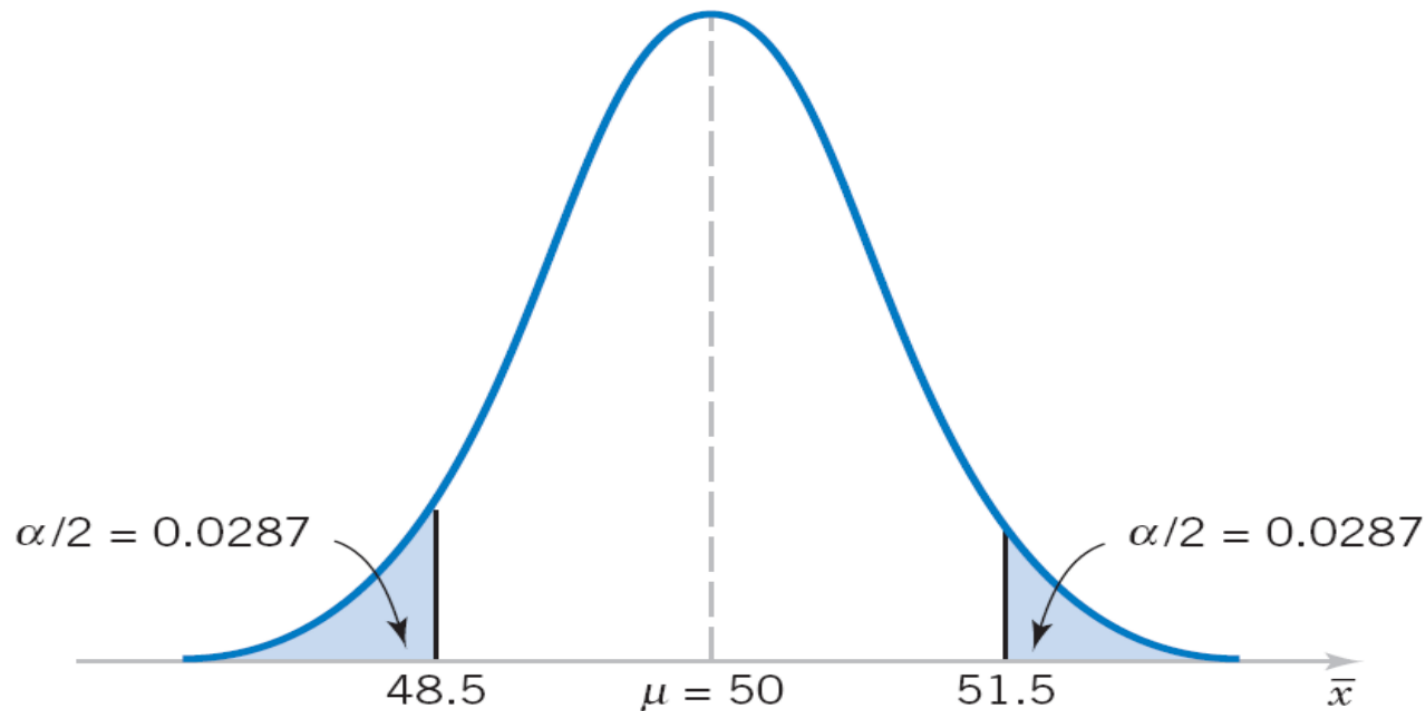
Decision	H_0 Is True	H_0 Is False
Fail to reject H_0	No error	Type II error
Reject H_0	Type I error	No error

$$\alpha = P(\text{type I error}) = P(\text{reject } H_0 \text{ when } H_0 \text{ is true})$$

Sometimes the type I error probability is called the **significance level**, or the **α -error**, or the **size** of the test

Testing Statistical Hypotheses

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Testing Statistical Hypotheses

The **power** of a statistical test is the probability of rejecting the null hypothesis H_0 when the alternative hypothesis is true.

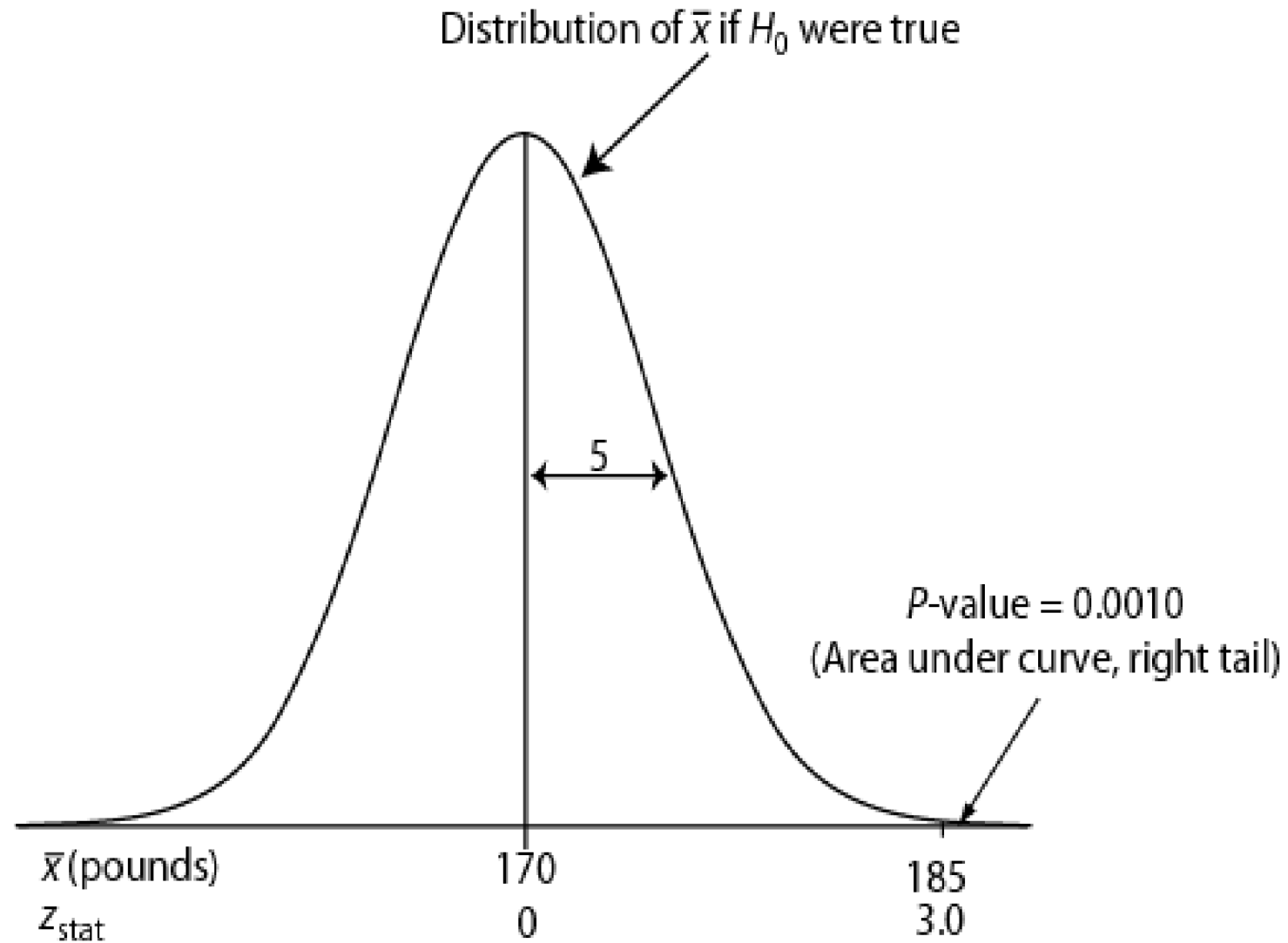
- The power is computed as $1 - \beta$, and power can be interpreted as *the probability of correctly rejecting a false null hypothesis*. We often compare statistical tests by comparing their **power** properties.
- For example, consider the propellant burning rate problem when we are testing $H_0 : \mu = 50$ centimeters per second against $H_1 : \mu$ not equal 50 centimeters per second. Suppose that the true value of the mean is $\mu = 52$. When $n = 10$, we found that $\beta = 0.2643$, so the power of this test is $1 - \beta = 1 - 0.2643 = 0.7357$ when $\mu = 52$.

P -value

What is the probability of the observed test statistic ... **when H_0 is true?**

This corresponds to the AUC in the tail of the Standard Normal distribution beyond the z_{stat}

P-value



P-Value

- Thus, smaller and smaller P -values provide stronger and stronger evidence against H_0
- Small P -value \Rightarrow strong evidence

α -Level

- Let $\alpha \equiv$ probability of erroneously rejecting H_0
- Set α threshold (e.g., let $\alpha = .10, .05, \text{ or } \textit{whatever}$)
- Reject H_0 when $P \leq \alpha$
- Retain H_0 when $P > \alpha$
- Example: Set $\alpha = .10$. Find $P = 0.27 \Rightarrow$ retain H_0
- Example: Set $\alpha = .01$. Find $P = .001 \Rightarrow$ reject H_0

One-Sample z Test

A. Hypothesis statements

$$H_0: \mu = \mu_0 \text{ vs.}$$

$$H_a: \mu \neq \mu_0 \text{ (two-sided) or}$$

$$H_a: \mu < \mu_0 \text{ (left-sided) or}$$

$$H_a: \mu > \mu_0 \text{ (right-sided)}$$

B. Test statistic

$$z_{\text{stat}} = \frac{\bar{x} - \mu_0}{SE_{\bar{x}}} \text{ where } SE_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

C. P-value: convert z_{stat} to P value

D. Significance statement (usually not necessary)

Conditions for z test

- σ known (not from data)
- Population approximately Normal or large sample (central limit theorem)

Statistical Inference

- Let X represent Trader Intelligence scores for stock price forecasting.
- Typically, $X \sim N(100, 15)$
- Take SRS of $n = 9$ from trader population
- Data $\Rightarrow \{116, 128, 125, 119, 89, 99, 105, 116, 118\}$
- Calculate: $\bar{x} = 112.8$
- Does sample mean provide strong evidence that population mean $\mu > 100$?

A. Hypotheses:

$H_0: \mu = 100$ versus

$H_a: \mu > 100$ (one-sided)

$H_a: \mu \neq 100$ (two-sided)

B. Test statistic:

A. Test statistic:

$$SE_{\bar{x}} = \frac{\sigma}{\sqrt{n}} = \frac{15}{\sqrt{9}} = 5$$

$$Z_{\text{stat}} = \frac{\bar{x} - \mu_0}{SE_{\bar{x}}} = \frac{112.8 - 100}{5} = 2.56$$

A. Test statistic:

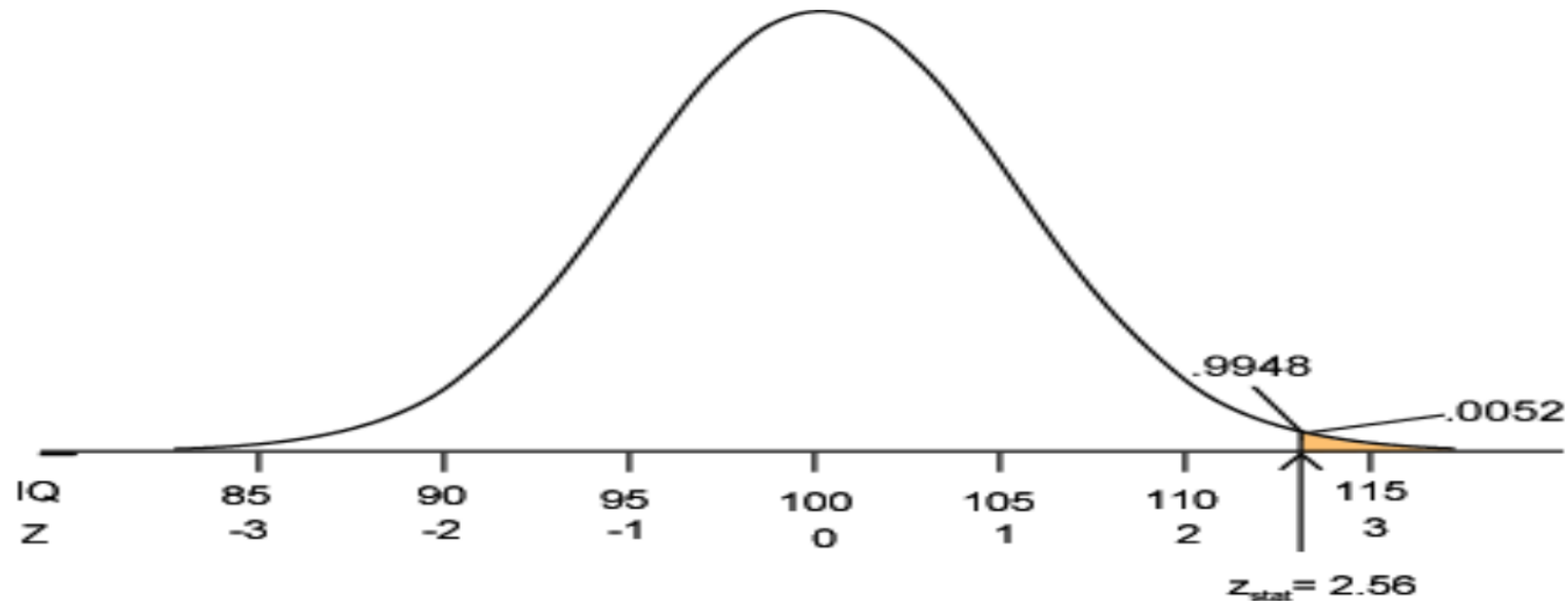
$$SE_{\bar{x}} = \frac{\sigma}{\sqrt{n}} = \frac{15}{\sqrt{9}} = 5$$

$$Z_{\text{stat}} = \frac{\bar{x} - \mu_0}{SE_{\bar{x}}} = \frac{112.8 - 100}{5} = 2.56$$

Calculated Value = 2.56 , tabulated value = 1.96, reject null hypothesis

Statistical Inference

P-value: $P = \Pr(Z \geq 2.56) = 0.0052$



$P = .0052 \Rightarrow$ it is unlikely the sample came from this null distribution \Rightarrow strong evidence against H_0

Two-sided P value

$$H_a: \mu \neq 100$$

Considers random deviations “up” and “down” from

$\mu_0 \Rightarrow$ tails above and below $\pm z_{\text{stat}}$

Thus, two-sided P

$$= 2 \times 0.0052$$

$$= 0.0104$$

Question

Suppose you start up a company that has developed a drug that is supposed to increase IQ. You know that the standard deviation of IQ in the general population is 15. You test your drug on 36 patients and obtain a mean IQ of 97.65. Using an alpha value of 0.05, is this IQ significantly different than the population mean of 100?

Question

$$z = \frac{97.65 - 100}{2.5} \\ = -0.94$$

Level of Significance = 0.05, two tailed, $0.05/2 = 0.025$, Z value = -1., Since calculated value is less than tab null accepted.

