

---

# Statistical Hypothesis Testing

---

---

# Statistical Hypothesis Testing

---

- definitions
    - what is it?
  - theories
    - Why are we using it?
  - Examples
    - How are we using it?
  - Applications
    - How is it related to us?
-

# What is it?

---

- A statistical test provides a mechanism for making quantitative decisions about a process or processes. The intent is to determine whether there is enough evidence to "reject" a conjecture or hypothesis about the process.

— NIST/SEMATECH e-Handbook of Statistical Methods, <http://www.itl.nist.gov/div898/handbook/>

---

# Statistical Hypothesis Testing

---

- Statistical
  - Hypothesis
  - Testing
-

# Statistical Hypothesis Testing

---

- statistical model & statistical inference
    - random variables  $X$
    - distribution  $P_{\theta}$  (at least partly unknown)
    - a set of observations/measurement
-

# Statistical Hypothesis Testing

---

- statistical model & statistical inference
  - “Statistical inference is concerned with methods of using this observational material to obtain information concerning the distribution of  $X$  or the parameter  $\Theta$  with which it is labeled”
  - “A statistical inference is a procedure that produces a probabilistic statement about ... a statistic model”

Lehmann, E.L. Testing Statistical Hypotheses  
DeGroot M.H., Schervish M.J. Probability and Statistics

---

# Statistical Hypothesis Testing

---

- Sample vs. Population
- Example (“Lake Wobegon”)
  - Someone claims kids at Lake Wobegon have above average intelligence
  - random samples of 9 kids there with test result of {116, 128, 125, 119, 89, 99, 105, 116, 118}
  - Wechsler scores (the test they take) are scaled to be normally distributed with a mean of 100 and standard deviation of 15.

mean of sample: 112.8  
<http://www.stat.cmu.edu/jeffrey/teaching/StatPrimer/hyp-test.pdf>

---

# Statistical Hypothesis Testing

---

- Null Hypothesis vs. Alternative Hypothesis
- Null hypothesis being “attacked”
- Usually more emphasis on Alternative Hypothesis
  - “Formulate the null hypothesis  $H_0$  (commonly, that the observations are the result of pure chance) and the alternative hypothesis  $H_a$  (commonly, that the observations show a real effect combined with a component of chance variation).”



# Statistical Hypothesis Testing

---

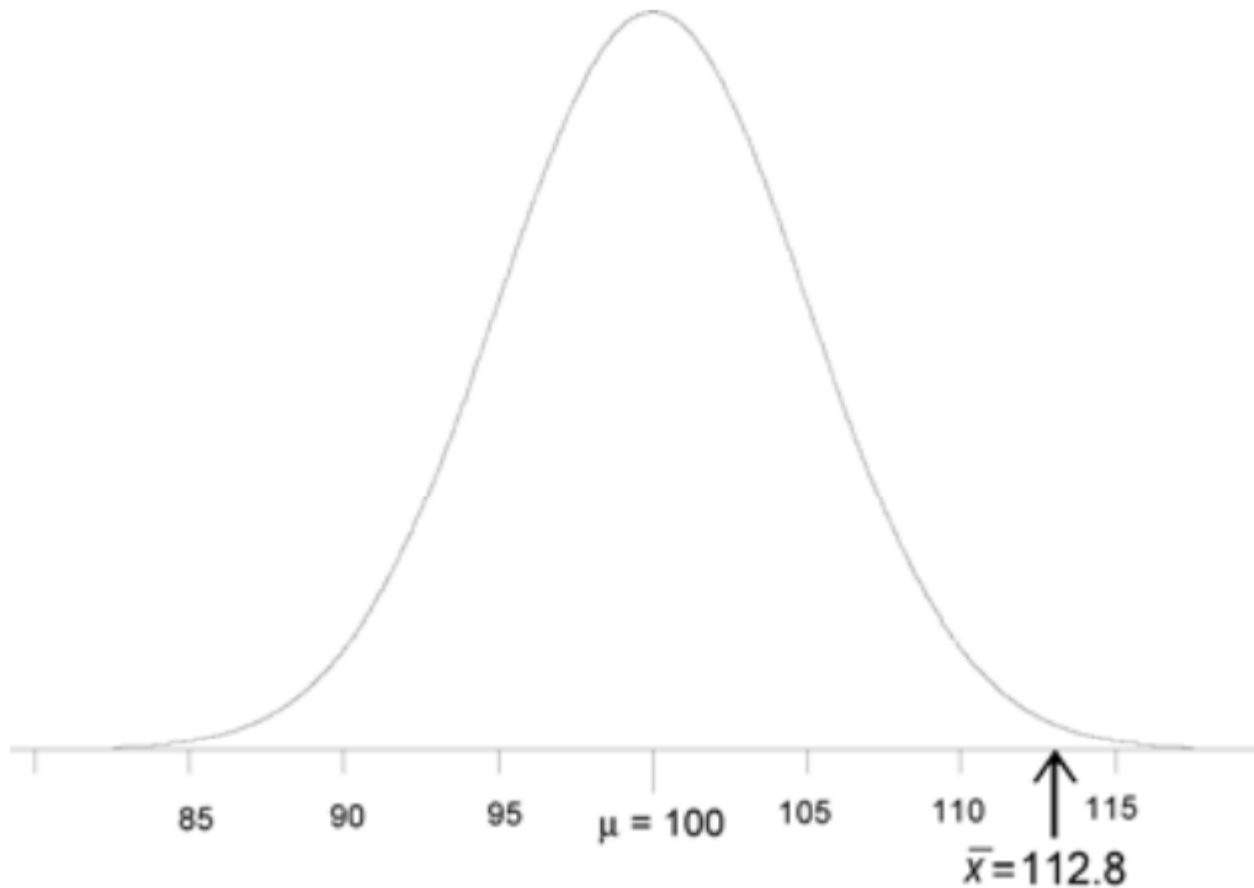
- The idea is to refute  $H_0$  if the sample is statistically far from the population/distribution, which is modeled only under the assumption that  $H_0$  is true.
  - we want to know the confidence level of refuting  $H_0$
  - Failed to reject the null hypothesis does not mean the null hypothesis is true
-

# Statistical Hypothesis Testing

---

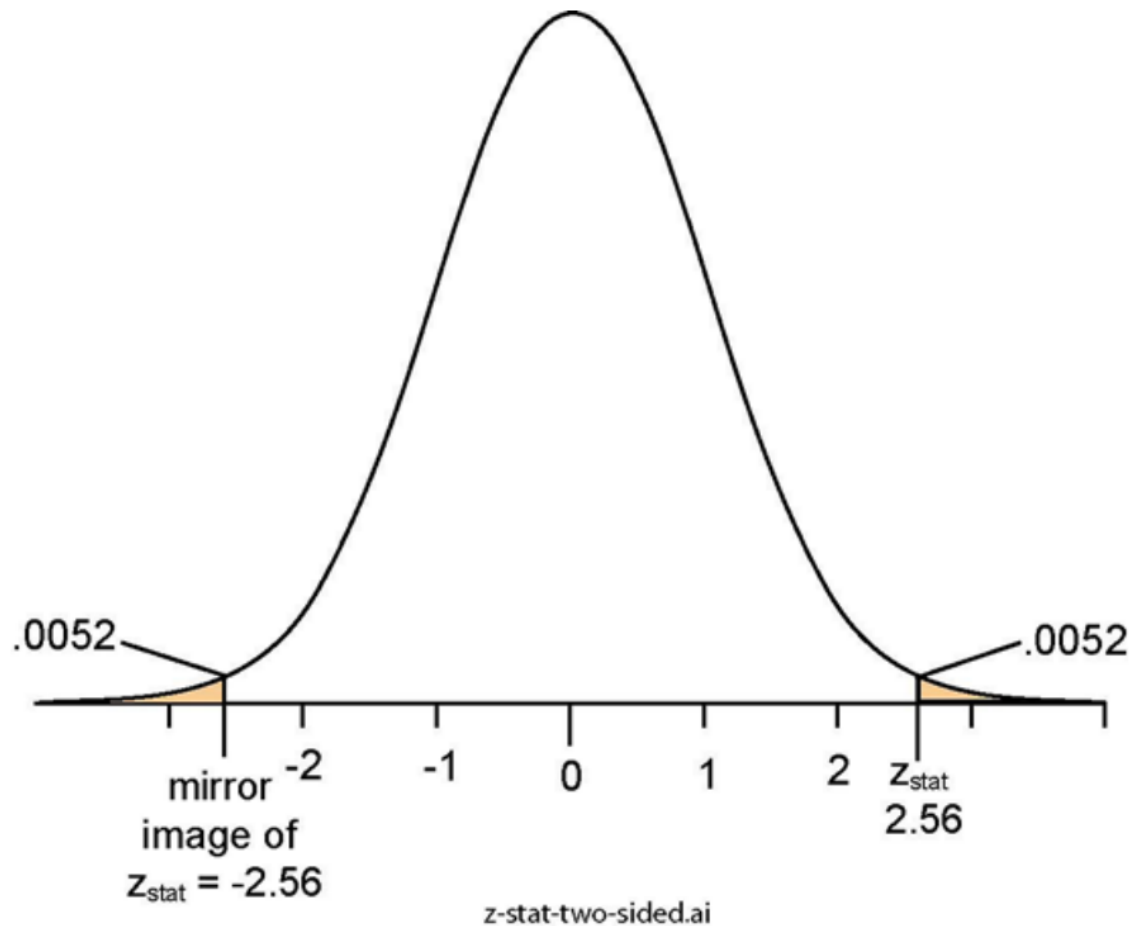
- one-sided vs. two-sided alternatives
- one-sided:
  - $H_0: u \leq 100$
  - $H_a: u > 100$
- two-sided:
  - $H_0: u = 100$
  - $H_a: u \neq 100$

# Statistical Hypothesis Testing



<http://www.sjsu.edu/faculty/gerstman/StatPrimer/hyp-test.pdf>

# Statistical Hypothesis Testing



<http://www.sjsu.edu/faculty/gerstman/StatPrimer/hyp-test.pdf>

# Statistical Hypothesis Testing

---

- Test Statistic:
    - The test statistic is a statistical method based on the specific hypothesis test.
    - $T = r(X)$  where  $X$  is the random sample from the distribution.
    - $H_0$  will be rejected if  $T \in R$
  - Critical Region:
    - the set  $S_1 = \{x: r(x) \in R\}$
-

# Statistical Hypothesis Testing

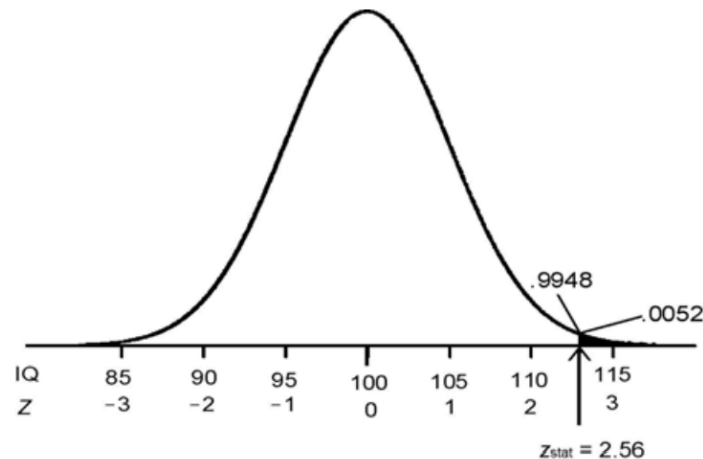
---

- Significance Level:  $\alpha$ 
  - “A value of  $\alpha = 0.05$  means that we inadvertently reject the null hypothesis 5% of the time when it is in fact true. ”
- Power:  $1 - \beta$ 
  - “the probability of accepting the null hypothesis when the alternative hypothesis is, in fact, true, is called  $\beta$ .”
- often referred as Type 1 and Type 2 errors, respectively

# Statistical Hypothesis Testing

---

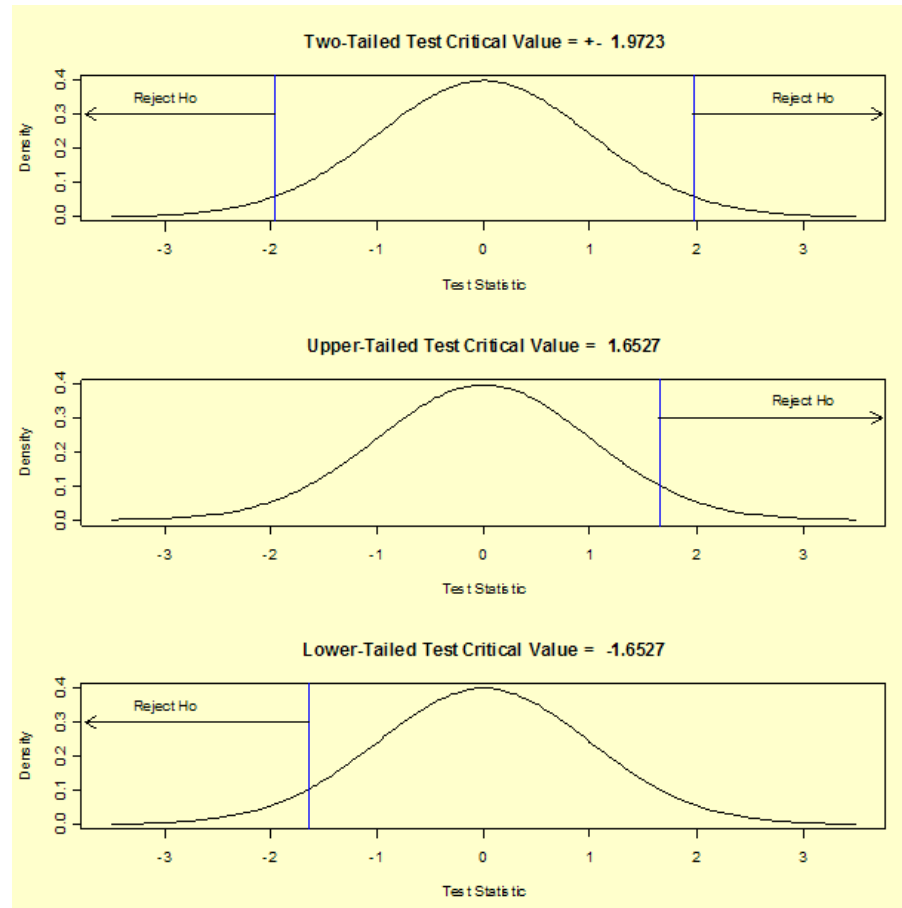
- Critical Region: encompasses those values of the test statistic that lead to a rejection of the null hypothesis.
- P-value: the probability that a test statistic at least as significant as the one observed



z-stat-one-sided.ai

# Statistical Hypothesis Testing

---





# Why are we using it? What's an alternative?

---

- Quantitative understanding of data
- Exploratory Data Analysis (EDA)
  - The primary goal of EDA is to maximize the analyst's insight into a data set and into the underlying structure of a data set
- Quantitative (Classic) Techniques
  - Hypothesis tests
  - Interval estimation
    - An interval estimate quantifies ... uncertainty in the sample estimate by computing lower and upper values of an interval

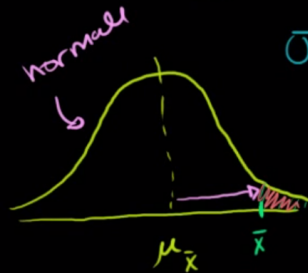
# How are we using it?

## Z-statistics vs. T-statistics

Z-statistic

$$\frac{\bar{X} - \mu_{\bar{X}}}{\sigma_{\bar{X}}}$$

$$\sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$$



$$Z = \frac{\bar{X} - \mu_{\bar{X}}}{\frac{\sigma}{\sqrt{n}}}$$

normally dist-

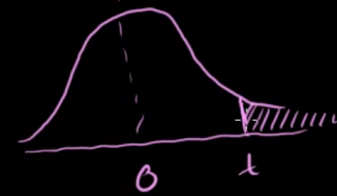
$$Z \approx \frac{\bar{X} - \mu_{\bar{X}}}{\frac{s}{\sqrt{n}}}$$

"ok" if  $n > 30$

t-statistic

$$t\text{-distribution} = \frac{\bar{X} - \mu_{\bar{X}}}{\frac{s}{\sqrt{n}}}$$

small



6:23 / 6:39

# Example

---

- Someone claims kids at Lake Wobegon have above average intelligence
- random samples of 9 kids there with test result of {116, 128, 125, 119, 89, 99, 105, 116, 118}
- Wechsler scores (the test they take) are scaled to be normally distributed with a mean of 100 and standard deviation of 15.
- mean of sample: 112.8

# Example

---

- $\text{mean}(X) = 112.8$
- $u = 100$
- test statistic: z-statistic
- $H_0: u = 100$
- $H_a: u > 100$

- $$z_{stat} = \frac{\bar{x} - \mu_0}{SEM}$$
- $SEM = \sigma/\text{sqrt}(N) = 15/\text{sqrt}(9) = 5$
- $Z_{stat} = (112.8 - 100)/5 = 2.56$
- look up Z table  $\rightarrow p = 0.0052 = 0.52\%$

<http://www.sjsu.edu/faculty/gerstman/StatPrimer/hyp-test.pdf>

---

# Example

## Standard Normal Probabilities

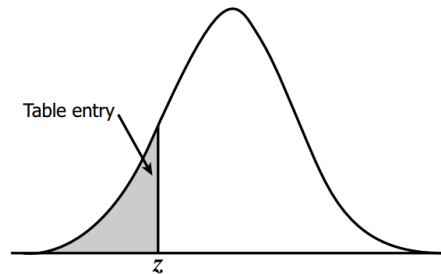
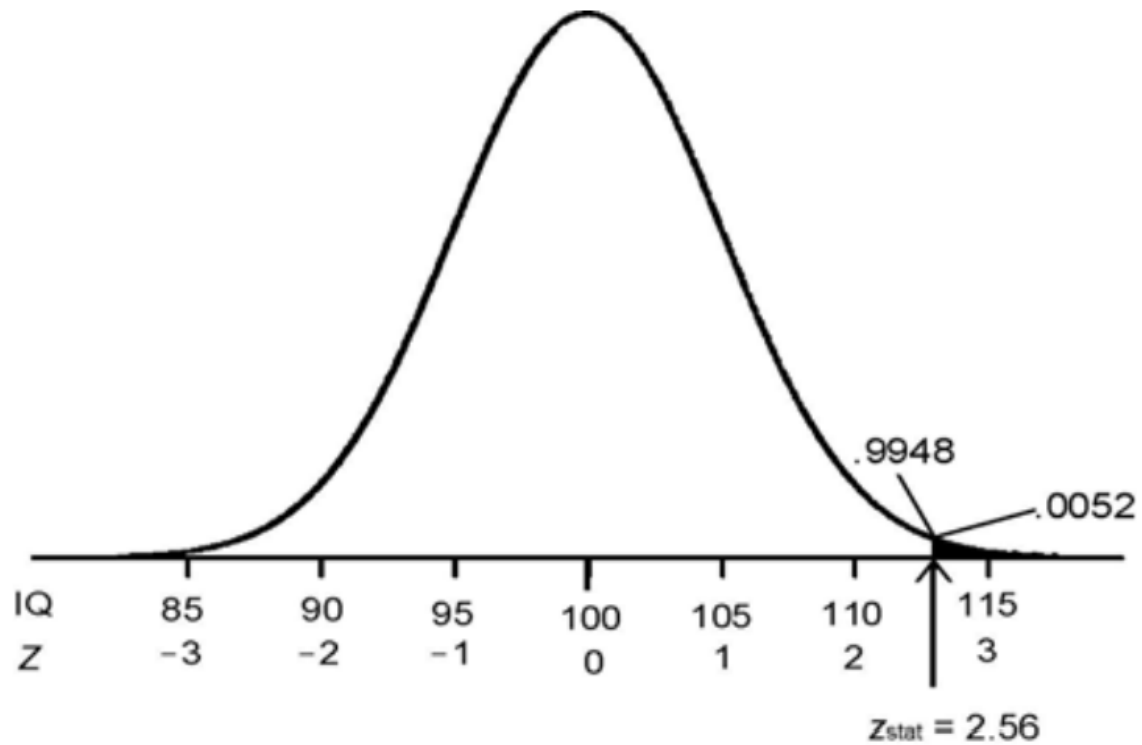


Table entry for  $z$  is the area under the standard normal curve to the left of  $z$ .

$z$	.00	.01	.02	.03	.04	.05	.06	.07	.08	.09
-3.4	.0003	.0003	.0003	.0003	.0003	.0003	.0003	.0003	.0003	.0002
-3.3	.0005	.0005	.0005	.0004	.0004	.0004	.0004	.0004	.0004	.0003
-3.2	.0007	.0007	.0006	.0006	.0006	.0006	.0006	.0005	.0005	.0005
-3.1	.0010	.0009	.0009	.0009	.0008	.0008	.0008	.0008	.0007	.0007
-3.0	.0013	.0013	.0013	.0012	.0012	.0011	.0011	.0011	.0010	.0010
-2.9	.0019	.0018	.0018	.0017	.0016	.0016	.0015	.0015	.0014	.0014
-2.8	.0026	.0025	.0024	.0023	.0023	.0022	.0021	.0021	.0020	.0019
-2.7	.0035	.0034	.0033	.0032	.0031	.0030	.0029	.0028	.0027	.0026
-2.6	.0047	.0045	.0044	.0043	.0041	.0040	.0039	.0038	.0037	.0036
-2.5	.0062	.0060	.0059	.0057	.0055	.0054	.0052	.0051	.0049	.0048
-2.4	.0082	.0080	.0078	.0075	.0073	.0071	.0069	.0068	.0066	.0064
-2.3	.0107	.0104	.0102	.0099	.0096	.0094	.0091	.0089	.0087	.0084
-2.2	.0139	.0136	.0132	.0129	.0125	.0122	.0119	.0116	.0113	.0110
-2.1	.0179	.0174	.0170	.0166	.0162	.0158	.0154	.0150	.0146	.0143
-2.0	.0228	.0222	.0217	.0212	.0207	.0202	.0197	.0192	.0188	.0183
-1.9	.0287	.0281	.0274	.0268	.0262	.0256	.0250	.0244	.0239	.0233

# Example



z-stat-one-sided.ai

# Example

---

- Define conventions:
  - When  $p \text{ value} > .10 \rightarrow$  the observed difference is “not significant”
  - When  $p \text{ value} \leq .10 \rightarrow$  the observed difference is “marginally significant”
  - When  $p \text{ value} \leq .05 \rightarrow$  the observed difference is “significant”
  - When  $p \text{ value} \leq .01 \rightarrow$  the observed difference is “highly significant”

# How is it related to us?

---

- Test Calibration
  - Test Constraints
  - QA audit
  - Actually test some hypothesis...
-



# How can we use it?

---

- Data Acquisition:
    - Input side:
      - A sensor input value
    - Output side:
      - VSCADA
      - raw sensor data is retrievable and transformable
  - Post analysis:
    - apply suitable test statistic using favorable tools
-