Hypothesis Testing Procedures for One-sample Data

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The one-sample Z-test

 rationale, steps, examples

 The one-sample T-test

 rationale, steps, examples



So far we've used the following procedure for hypothesis testing:

- 1) Propose null and alternative hypotheses
- 2) Find a model for the null distribution (via simulation or CLT)
- 3) Compare the outcome observed in the real data against the null distribution to find the *p*-value

This procedure can be made more general using *test statistics*, or *Z*-scores describing how many standard errors (*SE*) the observed outcome is above or below what we'd expect under H_0



Proceduralized hypothesis testing

Randomization test:

The Z-test:

▶ Propose H_0

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Randomization test:

The Z-test:

Propose H₀
 Find a null distribution specific to your study

► Propose *H*₀

 Use CLT formulas to calculate an SE, then find Z = Observed-Null SE



Randomization test:

- ▶ Propose H_0
- Find a null distribution specific to your study
- Locate the observed outcome in the null distribution and count the simulated outcomes that are at least as extreme

The Z-test:

- Propose H₀
- ► Use CLT formulas to calculate an *SE*, then find $Z = \frac{Observed - Null}{SE}$
- Locate Z in the Standard Normal curve and find the area outside it



$$SE = \sqrt{\frac{p(1-p)}{n}}$$

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 The observed outcome was â = 14/16 = 0.875 or 14 of 16

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2) Then, $Z = \frac{\hat{p}-p_0}{SE} = \frac{0.875-0.5}{0.125} = 3$



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2) Then, $Z = \frac{\hat{p}-p_0}{SE} = \frac{0.875-0.5}{0.125} = 3$
3) Comparing $Z = 3$ against a Standard Normal curve, the one-sided *p*-value is 0.0013 (two-sided is 0.0026)



A 2009 study investigated the behavior of 120 players in a rock-paper-scissors tournament. In their first move, 66 players chose "rock", 39 chose "paper", and 15 chose "scissors". Do these data provide statistically compelling evidence that players in this tournament were not choosing their first move randomly?

- 1) Propose a null hypothesis and an alternative hypothesis
- 2) Use CLT to find a standard error, then calculate a Z-value measuring how standard errors the observed outcome is from the value specified in H_0
- 3) Compare Z against the Standard Normal distribution to find the *p*-value



- H₀: p = 0.33, where p is the proportion that choose "rock", vs. H₀: p ≠ 0.33
 SE = √(0.33(1-0.33)/120) = 0.043, then notice we observed p̂ = 66/120 = 0.55, so Z = (0.55-0.33)/0.043 = 5.12
 The p-value is nearly zero, indicating overwhelming evidence
- 3) The p-value is nearly zero, indicating overwhelming evidence that players in this tournament are not choosing randomly, they're playing "rock" significantly more often than expected (55% of the time).



For a single mean, recall that CLT suggests:

$$\bar{x} \sim N(\mu, \frac{\sigma}{\sqrt{n}})$$

However, estimating σ (an unknown population parameter) via s (the sample standard deviation) introduces additional uncertainty that necessitates the *t*-distribution



Using $SE = \frac{s}{\sqrt{n}}$, the one-sample *T*-test calculates a *T*-value:

$$T = \frac{\bar{x} - \mu}{SE}$$

Then compares this *T*-value against a *t*-distribution with df = n - 1 to find the *p*-value



Radon is a toxic gas and the second leading cause of lung cancer. The EPA's federal action limit is 4 pCi/L, but they advise residents to take action if levels are above 0.4 pCi/L. In the basement of an lowa home, radon levels are tested on 8 randomly selected dates over the course a month, the measurements are:

 $\{.2, .7, .3, .9, .5, .3, .7, .6\}$

- 1) Propose a null hypothesis and an alternative hypothesis
- 2) The mean of these data is $\bar{x} = 0.525$ and s = 0.243, use this information to find a *T*-value measuring how standard errors the observed outcome is from the value specified in H_0
- 3) Compare T against the proper *t*-distribution to find the *p*-value

1)
$$H_0: \mu = 0.4$$
 vs. $H_a: \mu > 0.4$
2) $T = \frac{0.525 - 0.4}{0.243/\sqrt{8}} = 1.455$
3) The one-sided *p*-value using $df = 7$ is 0.09

The one-sided *p*-value using *df* = 7 is 0.094, so we conclude there's borderline evidence of Radon levels that exceed 0.4. Since a Type II error is more costly than a Type I error in application, we might still choose to reject *H*₀.



In 2010, new international rules were created to regulate swimsuit coverage and material after an inordinate amount of records were set at the 2008 Olympics by swimmers wearing a suit known as the LZR Racer. The wetsuits data, which is available on our course website the contains 1500m swim velocities of 12 competitive swimmers with and without a scientifically designed wetsuit.

- 1) Use StatKey to find the sample mean and sample standard deviation of the variable "difference"
- 2) Propose null and alternative hypotheses these researchers should evaluate using the variable "difference"
- 3) Find the *T* pertaining to these data, then compare it against the proper *t*-distribution to find the *p*-value.



- 1) Using the Descriptive Statistics and Graphs section of StatKey, $\bar{x}=0.077$ and s=0.022
- 2) $H_0: \mu = 0$, or the average improvement in velocity when wearing the wetsuit is zero, vs. $H_a: \mu \neq 0$
- 3) $T = \frac{0.077-0}{0.022/\sqrt{12}} = 12.12$, the *p*-value is nearly zero, indicating overwhelming evidence that the wetsuit improves swim velocity

- The wetsuits study is an example of a paired design, a type of design where each subject serves as their own control
- Paired designs have a number of statistical advantages over other designs, with an important one being the elimination of confounding variables

This presentation streamlines hypothesis testing into a procedure involving a consistent set of steps that can be applied to nearly any scenario:

- 1) Propose a null and alternative hypothesis
- 2) Calculate a test statistic (a Z-value or a T-value)
- Compare the test statistic against a null distribution (a Standard Normal curve or a *t*-distribution) to find the *p*-value
- 4) Use the *p*-value to make a decision

We focused on two specific applications:

• Testing
$$H_0: p = p_0$$
 using $Z = \frac{\hat{p} - p_0}{\sqrt{p_0(1 - p_0)/n}}$

• Testing
$$H_0: \mu = p_0$$
 using $T = \frac{\bar{x} - \mu}{s/\sqrt{n}}$

