

## 7. Hypotheses testing

### 7.1 Significance test

#### Def 1: null and alternative hypotheses

null hypothesis  $H_0$ : the effect of interest is zero

alternative  $H_1$ : the effect of interest is not zero

$H_0$  represents an established theory that must be discredited in order to demonstrate some effect  $H_1$

#### Def 2: type I and type II errors

type I error = false positive: reject  $H_0$  when it's true

type II error = false negative: accept  $H_0$  when it's false

State of nature	Accept $H_0$ negative result	Reject $H_0$ positive result
$H_0$ is true	True negative specificity = $1 - \alpha$	False positive $\alpha = P(\text{reject } H_0   H_0)$
$H_1$ is true	False negative $\beta = P(\text{accept } H_0   H_1)$	True positive sensitivity = $1 - \beta$

#### Def 3: test statistic and rejection region

Test statistic = a function of the data

with distinct typical values under  $H_0$  and  $H_1$

Rejection region (RR) of a test

a set of values for the test statistic where  $H_0$  is rejected

If test statistic and sample size are fixed, then either  
 $(\alpha \nearrow \beta \searrow)$  or  $(\alpha \searrow \beta \nearrow)$ , when RR is changed

#### Def 4: significance test

fix an appropriate significance level  $\alpha$

find RR from  $\alpha = P(\text{test statistic} \in \text{RR} | H_0)$

using the null distribution of the test statistic

Common significance levels: 5%, 1%, 0.1%

Type I error is believed to have graver consequences

therefore we control  $\alpha$  and not  $\beta$

### 7.2 Large-sample test for proportion

Sample count  $Y \sim \text{Bin}(n, p)$ ,  $p$  = population proportion

simple null hypothesis  $H_0: p = p_0$

Test statistic  $Z = \frac{Y - np_0}{\sqrt{np_0q_0}}$   
approximate null distribution:  $Z \approx N(0,1)$

RRs for three composite alternative hypotheses

one-sided  $H_1: p > p_0$ ,  $\text{RR} = \{Z \geq z_\alpha\}$

one-sided  $H_1: p < p_0$ ,  $\text{RR} = \{Z \leq -z_\alpha\}$

two-sided  $H_1: p \neq p_0$ ,  $\text{RR} = \{Z \geq z_{\alpha/2} \text{ or } Z \leq -z_{\alpha/2}\}$

#### Ex 1: postponed deaths

$H_0$ : death can not be postponed

$H_1$ : death can be postponed untill after important date

Jewish data, California:  $n = 1919$  deaths including

$Y = 922$  deaths during the week before Passover

$Y \sim \text{Bin}(n, p)$ ,  $H_0: p = 0.5$ ,  $H_1: p < 0.5$

observed test statistic  $Z = -1.712$

If  $\alpha = 0.05$ , then  $z_\alpha = 1.645$  and  $Z < -z_\alpha$

reject  $H_0$  in favor of  $H_1$  at 5% level

Seasonal effect? Chinese and Japanese data (California):

$n = 852$ ,  $Y = 418$ ,  $n - Y = 434$ , observed  $Z = -0.548$

can not reject  $H_0$  in favor of  $H_1$  at 5% level

### Def 5: P-value of a test

$P$  = the smallest significance level at which  
the test rejects  $H_0$  for the observed data

shows how significantly observed data contradicts  $H_0$

If  $P \leq \alpha$ , reject  $H_0$  at the significance level  $\alpha$   
If  $P > \alpha$ , do not reject  $H_0$  at level  $\alpha$

### Ex 1: postponed deaths

Jewish data: one-sided  $P = P(Z < -1.712) = 0.043$

Chi-Jap data: one-sided  $P = P(Z < -0.548) = 0.292$

### Ex 2: aspirin treatment

placebo group: 11034 individuals, 189 heart attacks

aspirin group: 11037 individuals, 104 heart attacks

Test  $H_0 = \{\text{no aspirin effect}\}$  against

$H_1 = \{\text{aspirin reduces the risk of heart attack}\}$

Test statistic  $X = \#\{\text{heart attacks in the placebo group}\}$

$P = P(X \geq 189 | H_0) = 0.0000003$

### 7.3 Small-sample test for the proportion

Test statistic  $Y \sim \text{Bin}(n, p)$ ,  $H_0: p = p_0$

exact null distribution  $Y \sim \text{Bin}(n, p_0)$

if  $n$  is small, we can not use normal approximation

Significance tests

one-sided  $H_1: p > p_0$ ,  $\text{RR} = \{Y \geq y_\alpha\}$

one-sided  $H_1: p < p_0$ ,  $\text{RR} = \{Y \leq y'_\alpha\}$

two-sided  $H_1: p \neq p_0$ ,  $\text{RR} = \{Y \geq y_{\alpha/2} \text{ or } Y \leq y'_{\alpha/2}\}$

#### Def 6: test power

sensitivity of the test  $1 - \beta = P(\text{reject } H_0 | H_1)$

Powerless decision rule at 0% level: never reject  $H_0$

#### Ex 3: extrasensory perception (ESP)

ESP test: guess the suits of 20 cards

chosen at random with replacement from a deck

Number of cards guessed correctly  $Y \sim \text{Bin}(20, p)$

$H_0: p = 0.25$  (pure guessing)

$H_1: p > 0.25$  (ESP ability)

Bin(20,0.25) table: 

$y$	8	9	10	11
$P(Y \geq y)$	.101	.041	.014	0.004

Rejection region at 5% significance level =  $\{Y \geq 9\}$

exact significance level = 4.1%

Power function:  $\text{Pw}(p) = P[Y \geq 9 | Y \sim \text{Bin}(20, p)]$

$p$	0.27	0.3	0.4	0.5	0.6	0.7
$\text{Pw}(p)$	0.064	0.113	0.404	0.748	0.934	0.995

Warning for “fishing expeditions”: the number of false positives in  $k$  tests at level  $\alpha$  is  $\text{Pois}(k\alpha)$

## Planning of sample size

given  $\alpha$  and  $\beta$  for  $H_0: p = p_0$ ,  $H_1: p = p_1$   
 choose sample size  $n$  such that  $\sqrt{n} = \frac{z_\alpha \sqrt{p_0 q_0} + z_\beta \sqrt{p_1 q_1}}{p_1 - p_0}$

### Ex 1: postponed deaths

Planning of sample size for  $H_0: p = 0.50$ ,  $H_1: p = 0.48$

$\alpha = 0.05$ ,  $\beta = 0.10$  requires  $n = 5351$  observations

$\alpha = \beta = 0.05$  requires  $n = 6773$  observations

Larger power requires larger sample

## 7.4 Large-sample test for mean

PD is not necessarily normal, test  $H_0: \mu = \mu_0$

Test statistic  $T = \frac{\bar{X} - \mu_0}{s_{\bar{X}}}$   
 approximate null distribution  $T \approx N(0,1)$

### Ex 4: radon level in home

Swedish official limit of the radon level in home:

year average = 400 disintegrations per second and  $\text{m}^3$

Data: 36 measurements in your home:  $\bar{X} = 450$ ,  $s = 180$

PD is non-normal, test  $H_0: \mu = 400$  vs  $H_1: \mu \geq 400$

Observed test statistic  $T = \frac{450 - 400}{\frac{180}{\sqrt{36}}} = 1.67$

one-sided  $P = 0.048$ , reject  $H_0$  at  $\alpha = 5\%$

## 7.5 One-sample t-test

used for small  $n$ , PD must be normal

$$H_0: \mu = \mu_0, \text{ test statistic: } T = \frac{\bar{X} - \mu_0}{s_{\bar{X}}}$$
$$\text{exact null distribution: } T \sim t_{n-1}$$

### Ex 5: measuring fat content

Two methods of measuring in % the fat content of meat

pairwise diff. for 16 hotdogs:  $\bar{X} = 0.53\%$ ,  $s = 1.06\%$

Test  $H_0: \mu = 0$  against  $H_1: \mu \neq 0$

$s_{\bar{X}} = 0.265$ , observed test statistic  $T = 2.0$

one-sided  $P = P(T < 2 | T \sim t_{15}) = 0.032$

Two-sided  $P = 2 \cdot 0.032 = 0.064$

do not reject  $H_0: \mu = 0$  in favor of  $H_1: \mu \neq 0$

Choose  $H_1$  before seeing the data

### CI method of hypotheses testing

accept  $H_0: \mu = \mu_0$  at 5% level if a 95% CI covers  $\mu_0$

reject  $H_0$  at 5% level if a 95% CI does not cover  $\mu_0$

CI is more informative than a test result  
wider CI indicates less power of the test

### Ex 5: measuring fat content

Exact 95% CI for the mean difference  $\mu$  is  $(-0.03, 1.08)$

do not reject  $H_0: \mu = 0$  in favor of  $H_1: \mu \neq 0$

note that approximate 95% CI is  $(0.01, 1.05)$