

Lecture 20

Topics:

I) Hypothesis Testing.

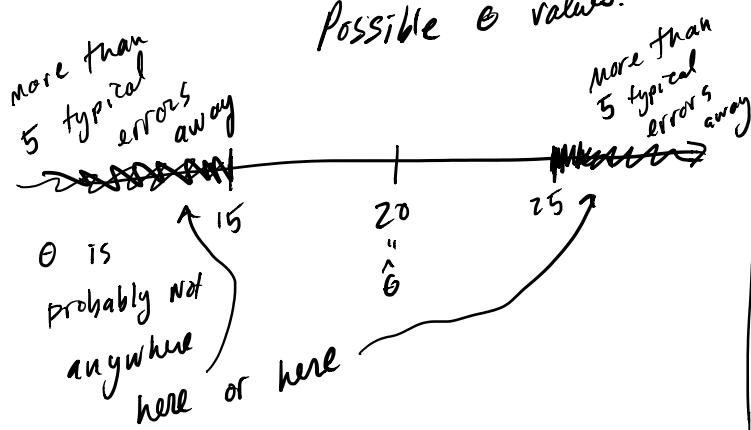
Hypothesis testing is essentially a formal way of ruling out possible values of a population parameter θ when observing an estimate $\hat{\theta}$.

e.g. You want to estimate some unknown θ . You collect data and get an estimate $\hat{\theta} = 20.0$

Suppose the $MSE(\hat{\theta}) = 1.0$

Then almost certainly the true θ isn't less than 15 or greater than 25

Possible θ values.



(1)

Note: the logic is a bit subtle

We know $\hat{\theta}$ is a random draw from a distribution that is close to θ (by about ± 1)

$$\uparrow$$

$$\sqrt{MSE}$$

e.g. suppose $\hat{\theta} \sim N(\theta, 2)$ where θ is unknown.

You sampled one draw from $N(\theta, 2)$ & got

$$\hat{\theta} = -105.7$$

Can you rule out some values for θ ?

Z-scores help you determine what values of θ are plausible:

$$Z\text{-score} = \frac{\hat{\theta} - E(\hat{\theta})}{\text{sd}(\hat{\theta})} = \frac{-105.7 - \theta}{\sqrt{2}}$$

Remember:

Z-score of ± 2 is Normal

Z-score of ± 3 is kinda unlikely

Z-score of ± 4 is rare

Z-score of ± 5 is extremely rare.

| hypothetical values of θ | z-score of $\hat{\theta} = -105.7$ | Plausible value of $\hat{\theta}$? |
|---------------------------------|---|-------------------------------------|
| $\theta = 1.2$ | $\frac{-105.7 - 1.2}{\sqrt{2}} = -7.5$ | No way |
| $\theta = -95.2$ | $\frac{-105.7 - (-95.2)}{\sqrt{2}} = -7.49$ | No way |
| $\theta = -110$ | $\frac{-105.7 - (-110)}{\sqrt{2}} = 3.04$ | Somewhat unlikely |
| $\theta = -108$ | $\frac{-105.7 - (-108)}{\sqrt{2}} = 1.62$ | plausible. |

(2)

Another way to phrase it:

(3)

- * $\theta = -95.2$ is inconsistent with the data $\hat{\theta} = -105.7$
- * $\theta = -108$ is consistent with the data $\hat{\theta} = -105.7$

To use z-scores for testing plausibility (or consistency) of a particular θ , you need to compute $E(\hat{\theta})$ & $sd(\hat{\theta})$ as a function of θ .

$E(\hat{\theta})$ & $sd(\hat{\theta})$ depend on the unknown θ .

to test a particular value of θ , just plug it in wherever it occurs in $E(\hat{\theta})$ & $sd(\hat{\theta})$.

e.g. Huge list of 0's & 1's:

$$x_1, x_2, \dots, x_{100000000}$$

You want to estimate the proportion of 1's in the list (call this θ). So you sample (w/rep) 2000 members from the list & get

$$X_1, X_2, \dots, X_{2000}$$

where

$$\hat{\theta} = \frac{X_1 + \dots + X_{2000}}{2000} = \bar{X} = 0.67.$$

Is it plausible that $\theta = 0.5$? (4)

$$\left(\begin{array}{l} \text{z-score for} \\ \hat{\theta} = 0.67 \text{ when} \\ \theta = 0.5 \end{array} \right) = \frac{0.67 - E(\hat{\theta})}{sd(\hat{\theta})}$$

$$= \frac{0.67 - \theta}{\sqrt{\frac{\theta(1-\theta)}{2000}}} \quad \begin{array}{l} \text{replace} \\ \theta \text{ with} \\ \text{the} \\ \text{possible} \\ \text{value} \\ \text{your} \\ \text{testing}, \\ 0.5 \text{ in} \\ \text{this case} \end{array}$$

$$= \frac{0.67 - 0.5}{\sqrt{\frac{0.5(1-0.5)}{2000}}}$$

$$= 15.2$$

$\therefore \theta = 0.5$ would be totally inconsistent with our observation that $\hat{\theta} = 0.67$

What if $E(\hat{\theta})$ or $sd(\hat{\theta})$ depend on other unknown population parameters?

In this case it often works to "plug in" an estimate of the other unknowns when computing the z-score.

e.g. Another huge list of numbers

(5)

$$y_1, y_2, \dots, y_N \text{ where } N = 5 \times 10^{22}$$

You want to estimate

$$\mu = \frac{1}{N} \sum_{i=1}^N y_i = \bar{y}$$

i.e. the list average.

Define the list s.d. as

$$\sigma = \sqrt{\frac{1}{N} \sum_{i=1}^N (y_i - \bar{y})^2}$$

Sometimes called the population s.d.

To construct $\hat{\mu}$ you randomly sample (w/rep) $n = 25$ numbers from the list

& get y_1, y_2, \dots, y_n

with

$$\hat{\mu} = \frac{1}{n} \sum_{i=1}^n y_i = 18.72$$

Sometimes called the sample s.d.

$$\hat{\sigma} = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y})^2} = 3.5$$

(one reason why we divide by $n-1$ for $\hat{\sigma}$ is that it makes $\hat{\sigma}^2$ an unbiased est of σ^2 ... this type of adjustment is more important when doing "regression". We do it here so you start getting used to it).

Question: Is $\mu = 17.0$ a

plausible value for μ ?

Ans:

(6)

The z-score for $\hat{\mu} = 18.72$ when $\mu = 17.0$ is given by

$$Z = \frac{\hat{\mu} - E(\hat{\mu})}{sd(\hat{\mu})}$$

$$= \frac{18.72 - \mu}{\sqrt{\frac{\sigma^2}{n}}}$$

$$= \frac{18.72 - 17}{\left(\frac{\sigma}{\sqrt{25}}\right)}$$

Now plug in
 $\hat{\sigma} = 3.5$ for
 σ

$$\approx \frac{18.72 - 17}{\left(\frac{3.5}{\sqrt{25}}\right)}$$

$$= 2.06$$

\therefore The z-score of our observation ($\hat{\mu} = 18.72, \hat{\sigma} = 3.5$) when $\mu = 17.0$ is totally plausible.

i.e. we can't rule out that the true value of μ is 17.0

e.g. Two huge lists of numbers

$$x_1, x_2, \dots,$$

$$y_1, y_2, \dots$$

Let μ_x = the Ave of the x_i 's
 μ_y = the Ave of the y_i 's.

σ_x = the "population s.d." of the x_i 's

σ_y = the "population s.d." of the y_i 's.

You want to estimate

$$\mu_x - \mu_y = ?$$

So you sample 25 numbers from the first list & get

$$X_1, X_2, \dots, X_{25}$$

also sample 20 numbers from the second list & get

$$Y_1, Y_2, \dots, Y_{20}.$$

Suppose this data results in

$$\hat{\mu}_x = \bar{X} = -2.5$$

$$\hat{\mu}_y = \bar{Y} = 1.23$$

$$\begin{aligned} \hat{\sigma}_x &= 10.2 \\ \hat{\sigma}_y &= 5.1 \end{aligned} \quad \left\{ \begin{array}{l} \text{sample standard deviations of} \\ \text{the two samples.} \end{array} \right.$$

Question: is it plausible that $\mu_x - \mu_y = 0$, i.e. that $\mu_x = \mu_y$?

Answer:

Let's compute the z-score of the estimate $\hat{\mu}_x - \hat{\mu}_y = -2.5 - 1.23$ under the hypothetical that $\mu_x - \mu_y = 0$.

Note: MF1 & MF2 give

$$E(\hat{\mu}_x - \hat{\mu}_y) = \mu_x - \mu_y$$

$$sd(\hat{\mu}_x - \hat{\mu}_y) = \sqrt{\frac{\sigma_x^2}{25} + \frac{\sigma_y^2}{20}}$$

$$\begin{aligned} \therefore z\text{-score} &= \frac{(\hat{\mu}_x - \hat{\mu}_y) - (\mu_x - \mu_y)}{\sqrt{\frac{\sigma_x^2}{25} + \frac{\sigma_y^2}{20}}} \\ &= \frac{(-2.5 - 1.23) - (0)}{\sqrt{\frac{\sigma_x^2}{25} + \frac{\sigma_y^2}{20}}} \quad \begin{array}{l} \text{hypothetical} \\ \text{value of} \\ \mu_x - \mu_y \end{array} \end{aligned}$$

$$\begin{aligned} &\approx \frac{(-2.5 - 1.23) - (0)}{\sqrt{\frac{10.2^2}{25} + \frac{5.1^2}{20}}} \quad \begin{array}{l} \text{plug in} \\ \hat{\sigma}_x = 10.2 \text{ &} \\ \hat{\sigma}_y = 5.1 \text{ for} \\ \sigma_x \text{ & } \sigma_y \end{array} \\ &= -1.59 \end{aligned}$$

$\therefore \mu_x - \mu_y = 0$ is totally consistent with the data.