

More about midterm 2

Problem 5:

The first step is to use the equation

$$Y = 2 + X_1 + 3X_2 + \varepsilon$$

Notice that the mis-specified model includes an ε' . You are not told anything about ε' . In particular, you do not know that $\text{Cov}(X_1, \varepsilon') = 0$.

After the first step, a common error was to write

$$\text{Cov}(X_2, X_1) = \text{something. not right.}$$

We know : $\text{Cor}(X_1, X_2) = \rho$ $\text{Var}(X_1) = \sigma_1^2$
 $\text{Var}(X_2) = \sigma_2^2$.

Which formula for $\text{Cor}(X_1, X_2)$ in terms of $\text{Cor}(X_1, X_2)$ is correct ?

- $\text{Cov}(X_1, X_2) = \rho$
- $\text{Cov}(X_1, X_2) = \rho \sigma_1^2 \sigma_2^2$
- $\text{Cov}(X_1, X_2) = \rho \sqrt{\sigma_1^2 + \sigma_2^2}$
- $\text{Cov}(X_1, X_2) = \frac{\rho}{\sigma_1 \sigma_2}$
- $\text{Cov}(X_1, X_2) = \rho \sigma_1 \sigma_2$ ← Correct!
- $\text{Cov}(X_1, X_2) = \rho^2 \sigma_1^2 \sigma_2^2$

Note analogy with

$$\langle v, w \rangle = \|v\| \|w\| \cos \theta$$

A model of the stock market

$X_1, \dots, X_N$: % return on stock i
over the next year

$\text{Var}(X_i) = \sigma^2$: all stocks have the same
volatility.

$\text{Cor}(X_i, X_j) = \rho$: for all $i \neq j$.

Think : $N = 1000$ $\sigma = 0.25 = 25\%$
 $\rho = 0.2$

⚠ of course the real stock market is more complex.

Two portfolio managers: P, Q .

$P = \frac{1}{n} (X_{i_1} + \dots + X_{i_n})$ ← return on portfolio P
consisting of $X_{i_1}, \dots, X_{i_n}$

$Q = \frac{1}{n} (X_{j_1} + \dots + X_{j_n})$ ← return on Q .

Think : $n = 100$. P, Q have no stocks in
common.

Q: What is the correlation of the return on P and the return on Q?

You might think: the correlation ρ of two different stocks is low, and P, Q have nothing in common, so $\text{Cor}(P, Q)$ is low.

This is WRONG.

A calculation at the Math 447 level shows that the correlation is high.

$$\begin{aligned}\text{Cov}(P, Q) &= \text{Cov}\left(\frac{1}{n} \sum_{i=1}^n X_{i1}, \frac{1}{n} \sum_{m=1}^n X_{jm}\right) \\ &= \frac{1}{n^2} \sum_i \sum_m \text{Cov}(X_{i1}, X_{jm}) \\ &= \frac{1}{n^2} \cdot n^2 \cdot \rho \sigma^2 = \rho \sigma^2\end{aligned}$$

$$\text{Cor}(P, Q) = \frac{1}{\sigma_P \sigma_Q} \text{Cov}(P, Q)$$

$$\sigma_P = \sqrt{\text{Cov}(P, P)}$$

$$\sigma_p^2 = \frac{1}{n^2} \sum_l \sum_m \text{Cor}(X_{il}, X_{im})$$

$$= \frac{1}{n^2} \left(\sum_{l \neq m} \rho \sigma^2 + \sum_{l=m} \sigma^2 \right)$$

$$= \frac{1}{n^2} \left[(n^2 - n) \rho \sigma^2 + n \sigma^2 \right]$$

$$= \sigma^2 \left(\frac{1}{n} + \rho \left(1 - \frac{1}{n} \right) \right)$$

$$\sigma_Q^2 = \sigma_P^2, \text{ so } \sigma_P \sigma_Q = \sigma_P^2.$$

$$\text{and } \text{Cor}(P, Q) = \frac{n\rho}{1 + (n-1)\rho}$$

With $n = 100$, $\rho = 0.2$, this is ≈ 0.96

High, not low.

This idea has meaningful consequences:

- Much of the performance of a portfolio of stocks is explained by the performance of the stock market as a whole.

- The Capital Asset Pricing Model
(Sharpe, Nobel 1990) (Also Treynor, Linter, Mossin)
- The "index fund": we don't need to pay a fund manager, we just buy the "index", own all stocks in proportion to market capitalization.

Math 532 May 12.

Bayesians, Frequentists, Likelihood Principle

The most important thing to know: in most applications, nobody cares.

The ultimate validation in business: We made money.

Arguing philosophy of statistics does not make money.

So take the following with a grain of salt.

An example of Savage:

We are testing a coin with probability of heads θ for bias. We consider

$$H_0: \theta = \frac{1}{2} \quad \text{vs.} \quad H_1: \theta > \frac{1}{2}.$$

Experiment A: We flip the coin 12 times and get 9 heads.

X , the number of heads is $X \sim \text{Bin}(12, \theta)$

$$\text{and } P(X \geq 9 | H_0) = \sum_{x=9}^{12} \binom{12}{x} \left(\frac{1}{2}\right)^x \left(1 - \frac{1}{2}\right)^{12-x} \approx 0.073.$$

So H_0 is NOT rejected at $\alpha = 0.05$

Experiment B: We flip the coin until 3 tails are observed and count X , the number of heads. We observe $X = 9$.

$X \sim NB(3, 1-\theta)$ (negative binomial)

$$P(X \geq 9 | H_0) = \sum_{x=9}^{\infty} \binom{3+x-1}{2} \left(\frac{1}{2}\right)^x \left(\frac{1}{2}\right)^3$$

$$\approx 0.033$$

So H_0 is rejected at $\alpha = 0.05$.

OK, fine: these are different experiments.

Different rules governing collection of data.

However: in both cases we observed 9 heads and 3 tails.

The Likelihood Principle: after we observe the data, all relevant information is contained in the likelihood function.

What are the likelihood functions in A and B?

$$A: f(x|\theta) = \binom{n}{x} \theta^x (1-\theta)^{n-x} = \binom{12}{9} \theta^9 (1-\theta)^3 \\ = 220 \theta^9 (1-\theta)^3$$

$$B: f(x|\theta) = \binom{\alpha+x-1}{\alpha-1} (1-\theta)^\alpha \theta^x = \binom{3+9-1}{3-1} (1-\theta)^3 \theta^9 \\ = 55 \theta^9 (1-\theta)^3.$$

Note that multiplying the likelihood function by a constant (here 4) does not change anything.

If we are strictly Bayesian and accept the likelihood principle, then the evidence of experiments A and B is the same.

In other words, the rules governing when data collection stops are not relevant to interpretation of the data.

Many people have a hard time accepting this.

The Chow-Robbins Game.

We flip a fair coin repeatedly, and can stop whenever we like. If, upon stopping, we have h heads in n flips, we are paid $\frac{h}{n}$ dollars.

Example 1: We flip THHH and elect to stop: we are paid 75¢.

Example 2: We flip H and elect to stop. We are paid \$1.

Example 3: We flip ~~TH~~ TT and elect to stop (bad strategy): we are paid \$0.

The exact value of the game, and a perfect strategy, is unknown.

Partial results, based on the law of the iterated logarithm, are known.

Easy proof that the value of the game is at least 75¢:

Simple strategy: If the first flip is H, stop.
otherwise flip 1000 times.

We get about $75¢ = \frac{1}{2} \cdot (\$1) + \frac{1}{2} \cdot (50¢)$.

To see the difficulty, show that it is
correct with 2 H and 1 T to continue
but with 5 H and 3 T to stop.

Now what if we have 66 H and 59 T?

This gives most people the intuition that the
rules of data collection do matter.

Again: in the workplace none of this matters.

Very simple ideas can be a major contribution.

I will post a link to a NY Times article
about statistics in ~~the~~ advertising.