

18.440: Lecture 25
**Covariance and some conditional
expectation exercises**

Scott Sheffield

MIT

Covariance and correlation

Paradoxes: getting ready to think about conditional expectation

Covariance and correlation

Paradoxes: getting ready to think about conditional expectation

A property of independence

- ▶ If X and Y are independent then $E[g(X)h(Y)] = E[g(X)]E[h(Y)]$.
- ▶ Just write $E[g(X)h(Y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x)h(y)f(x, y)dx dy$.
- ▶ Since $f(x, y) = f_X(x)f_Y(y)$ this factors as $\int_{-\infty}^{\infty} h(y)f_Y(y)dy \int_{-\infty}^{\infty} g(x)f_X(x)dx = E[h(Y)]E[g(X)]$.

Defining covariance and correlation

- ▶ Now define covariance of X and Y by
$$\text{Cov}(X, Y) = E[(X - E[X])(Y - E[Y])].$$
- ▶ Note: by definition $\text{Var}(X) = \text{Cov}(X, X)$.
- ▶ Covariance (like variance) can also be written a different way. Write $\mu_X = E[X]$ and $\mu_Y = E[Y]$. If laws of X and Y are known, then μ_X and μ_Y are just constants.

- ▶ Then

$$\begin{aligned}\text{Cov}(X, Y) &= E[(X - \mu_X)(Y - \mu_Y)] = E[XY - \mu_X Y - \mu_Y X + \mu_X \mu_Y] = \\ &E[XY] - \mu_X E[Y] - \mu_Y E[X] + \mu_X \mu_Y = E[XY] - E[X]E[Y].\end{aligned}$$

- ▶ Covariance formula $E[XY] - E[X]E[Y]$, or “expectation of product minus product of expectations” is frequently useful.
- ▶ Note: if X and Y are independent then $\text{Cov}(X, Y) = 0$.

Basic covariance facts

- ▶ Using $\text{Cov}(X, Y) = E[XY] - E[X]E[Y]$ as a definition, certain facts are immediate.
- ▶ $\text{Cov}(X, Y) = \text{Cov}(Y, X)$
- ▶ $\text{Cov}(X, X) = \text{Var}(X)$
- ▶ $\text{Cov}(aX, Y) = a\text{Cov}(X, Y)$.
- ▶ $\text{Cov}(X_1 + X_2, Y) = \text{Cov}(X_1, Y) + \text{Cov}(X_2, Y)$.
- ▶ **General statement of bilinearity of covariance:**

$$\text{Cov}\left(\sum_{i=1}^m a_i X_i, \sum_{j=1}^n b_j Y_j\right) = \sum_{i=1}^m \sum_{j=1}^n a_i b_j \text{Cov}(X_i, Y_j).$$

- ▶ Special case:

$$\text{Var}\left(\sum_{i=1}^n X_i\right) = \sum_{i=1}^n \text{Var}(X_i) + 2 \sum_{(i,j): i < j} \text{Cov}(X_i, X_j).$$

Defining correlation

- ▶ Again, by definition $\text{Cov}(X, Y) = E[XY] - E[X]E[Y]$.
- ▶ **Correlation** of X and Y defined by

$$\rho(X, Y) := \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X)\text{Var}(Y)}}.$$

- ▶ Correlation doesn't care what units you use for X and Y . If $a > 0$ and $c > 0$ then $\rho(aX + b, cY + d) = \rho(X, Y)$.
- ▶ Satisfies $-1 \leq \rho(X, Y) \leq 1$.
- ▶ Why is that? Something to do with $E[(X + Y)^2] \geq 0$ and $E[(X - Y)^2] \geq 0$?
- ▶ If a and b are positive constants and $a > 0$ then $\rho(aX + b, X) = 1$.
- ▶ If a and b are positive constants and $a < 0$ then $\rho(aX + b, X) = -1$.

Important point

- ▶ Say X and Y are uncorrelated when $\rho(X, Y) = 0$.
- ▶ Are independent random variables X and Y always uncorrelated?
- ▶ Yes, assuming variances are finite (so that correlation is defined).
- ▶ Are uncorrelated random variables always independent?
- ▶ No. Uncorrelated just means $E[(X - E[X])(Y - E[Y])] = 0$, i.e., the outcomes where $(X - E[X])(Y - E[Y])$ is positive (the upper right and lower left quadrants, if axes are drawn centered at $(E[X], E[Y])$) balance out the outcomes where this quantity is negative (upper left and lower right quadrants). This is a much weaker statement than independence.

Examples

- ▶ Suppose that X_1, \dots, X_n are i.i.d. random variables with variance 1. For example, maybe each X_j takes values ± 1 according to a fair coin toss.
- ▶ Compute $\text{Cov}(X_1 + X_2 + X_3, X_2 + X_3 + X_4)$.
- ▶ Compute the correlation coefficient $\rho(X_1 + X_2 + X_3, X_2 + X_3 + X_4)$.
- ▶ Can we generalize this example?
- ▶ What is variance of number of people who get their own hat in the hat problem?
- ▶ Define X_i to be 1 if i th person gets own hat, zero otherwise.
- ▶ Recall formula
$$\text{Var}(\sum_{i=1}^n X_i) = \sum_{i=1}^n \text{Var}(X_i) + 2 \sum_{(i,j): i < j} \text{Cov}(X_i, X_j).$$
- ▶ Reduces problem to computing $\text{Cov}(X_i, X_j)$ (for $i \neq j$) and $\text{Var}(X_i)$.

Covariance and correlation

Paradoxes: getting ready to think about conditional expectation

Covariance and correlation

Paradoxes: getting ready to think about conditional expectation

Famous paradox

- ▶ Certain corrupt and amoral banker dies, instructed to spend some number n (of banker's choosing) days in hell.
- ▶ At the end of this period, a (biased) coin will be tossed. Banker will be assigned to hell forever with probability $1/n$ and heaven forever with probability $1 - 1/n$.
- ▶ After 10 days, banker reasons, "If I wait another day I reduce my odds of being here forever from $1/10$ to $1/11$. That's a reduction of $1/110$. A $1/110$ chance at infinity has infinite value. Worth waiting one more day."
- ▶ Repeats this reasoning every day, stays in hell forever.
- ▶ Standard punch line: this is actually what banker deserved.
- ▶ Fairly dark as math humor goes (and no offense intended to anyone...) but dilemma is interesting.

- ▶ **Paradox:** decisions seem sound individually but together yield worst possible outcome. Why? Can we demystify this?
- ▶ **Variant without probability:** Instead of tossing $(1/n)$ -coin, person deterministically spends $1/n$ fraction of future days (every n th day, say) in hell.
- ▶ **Even simpler variant:** infinitely many identical money sacks have labels $1, 2, 3, \dots$. I have sack 1. You have all others.
- ▶ You offer me a deal. I give you sack 1, you give me sacks 2 and 3. I give you sack 2 and you give me sacks 4 and 5. On the n th stage, I give you sack n and you give me sacks $2n$ and $2n + 1$. Continue until I say stop.
- ▶ Lets me get arbitrarily rich. But if I go on forever, I return every sack given to me. If n th sack confers right to spend n th day in heaven, leads to hell-forever paradox.
- ▶ I make infinitely many good trades and end up with less than I started with. “Paradox” is really just existence of 2-to-1 map from (smaller set) $\{2, 3, \dots\}$ to (bigger set) $\{1, 2, \dots\}$.

Money pile paradox

- ▶ You have an infinite collection of money piles with labeled $0, 1, 2, \dots$ from left to right.
- ▶ Precise details not important, but let's say you have $1/4$ in the 0th pile and $\frac{3}{8}5^j$ in the j th pile for each $j > 0$. Important thing is that pile size is increasing exponentially in j .
- ▶ Banker proposes to transfer a fraction (say $2/3$) of each pile to the pile on its left and remainder to the pile on its right. Do this simultaneously for all piles.
- ▶ Every pile is bigger after transfer (and this can be true even if banker takes a portion of each pile as a fee).
- ▶ Banker seemed to make you richer (every pile got bigger) but really just reshuffled your infinite wealth.

Two envelope paradox

- ▶ X is geometric with parameter $1/2$. One envelope has 10^X dollars, one has 10^{X-1} dollars. Envelopes shuffled.
- ▶ You choose an envelope and, after seeing contents, are allowed to choose whether to keep it or switch. (Maybe you have to pay a dollar to switch.)
- ▶ Maximizing conditional expectation, it seems it's always better to switch. But if you always switch, why not just choose second-choice envelope first and avoid switching fee?
- ▶ Kind of a disguised version of money pile paradox. But more subtle. One has to replace “ j th pile of money” with “restriction of expectation sum to scenario that first chosen envelop has 10^j ”. Switching indeed makes each pile bigger.
- ▶ However, “Higher expectation given amount in first envelope” may not be right notion of “better.” If S is payout with switching, T is payout without switching, then S has same law as $T - 1$. In that sense S is worse.

- ▶ Beware infinite expectations.
- ▶ Beware unbounded utility functions.
- ▶ They can lead to strange conclusions, sometimes related to “reshuffling infinite (actual or expected) wealth to create more” paradoxes.
- ▶ Paradoxes can arise even when total transaction is finite with probability one (as in envelope problem).

MIT OpenCourseWare
<http://ocw.mit.edu>

18.440 Probability and Random Variables

Spring 2014

For information about citing these materials or our Terms of Use, visit: <http://ocw.mit.edu/terms>.