

# 18.440: Lecture 32

## Strong law of large numbers and Jensen's inequality

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A story about Pedro

Strong law of large numbers

Jensen's inequality

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## Pedro's hopes and dreams

- ▶ Pedro is considering two ways to invest his life savings.
- ▶ One possibility: put the entire sum in government insured interest-bearing savings account. He considers this completely risk free. The (post-tax) interest rate equals the inflation rate, so the real value of his savings is guaranteed not to change.
- ▶ Riskier possibility: put sum in investment where every month *real* value goes up 15 percent with probability .53 and down 15 percent with probability .47 (independently of everything else).
- ▶ How much does Pedro make in expectation over 10 years with risky approach? 100 years?

## Pedro's hopes and dreams

- ▶ How much does Pedro make in expectation over 10 years with risky approach? 100 years?
- ▶ Answer: let  $R_i$  be i.i.d. random variables each equal to 1.15 with probability .53 and .85 with probability .47. Total value after  $n$  steps is initial investment times
$$T_n := R_1 \times R_2 \times \dots \times R_n.$$
- ▶ Compute  $E[R_1] = .53 \times 1.15 + .47 \times .85 = 1.009$ .
- ▶ Then  $E[T_{120}] = 1.009^{120} \approx 2.93$ . And
$$E[T_{1200}] = 1.009^{1200} \approx 46808.9$$

- ▶ How would you advise Pedro to invest over the next 10 years if Pedro wants to be completely sure that he doesn't lose money?
- ▶ What if Pedro is willing to accept substantial risk if it means there is a good chance it will enable his grandchildren to retire in comfort 100 years from now?
- ▶ What if Pedro wants the money for himself in ten years?
- ▶ Let's do some simulations.

## Logarithmic point of view

- ▶ We wrote  $T_n = R_1 \times \dots \times R_n$ . Taking logs, we can write  $X_i = \log R_i$  and  $S_n = \log T_n = \sum_{i=1}^n X_i$ .
- ▶ Now  $S_n$  is a sum of i.i.d. random variables.
- ▶  $E[X_1] = E[\log R_1] = .53(\log 1.15) + .47(\log .85) \approx -.0023$ .
- ▶ By the law of large numbers, if we take  $n$  extremely large, then  $S_n/n \approx -.0023$  with high probability.
- ▶ This means that, when  $n$  is large,  $S_n$  is *usually* a very negative value, which means  $T_n$  is *usually* very close to zero (even though its expectation is very large).
- ▶ Bad news for Pedro's grandchildren. After 100 years, the portfolio is probably in bad shape. But what if Pedro takes an even longer view? Will  $T_n$  converge to zero with probability one as  $n$  gets large? Or will  $T_n$  perhaps always *eventually* rebound?

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# Strong law of large numbers

- ▶ Suppose  $X_i$  are i.i.d. random variables with mean  $\mu$ .
- ▶ Then the value  $A_n := \frac{X_1+X_2+\dots+X_n}{n}$  is called the *empirical average* of the first  $n$  trials.
- ▶ Intuition: when  $n$  is large,  $A_n$  is typically close to  $\mu$ .
- ▶ Recall: **weak law of large numbers** states that for all  $\epsilon > 0$  we have  $\lim_{n \rightarrow \infty} P\{|A_n - \mu| > \epsilon\} = 0$ .
- ▶ The **strong law of large numbers** states that with probability one  $\lim_{n \rightarrow \infty} A_n = \mu$ .
- ▶ It is called “strong” because it implies the weak law of large numbers. But it takes a bit of thought to see why this is the case.

## Strong law implies weak law

- ▶ Suppose we know that the strong law holds, i.e., with probability 1 we have  $\lim_{n \rightarrow \infty} A_n = \mu$ .
- ▶ Strong law implies that for every  $\epsilon$  the random variable  $Y_\epsilon = \max\{n : |A_n - \mu| > \epsilon\}$  is finite with probability one. It has some probability mass function (though we don't know what it is).
- ▶ Note that if  $|A_n - \mu| > \epsilon$  for some  $n$  value then  $Y_\epsilon \geq n$ .
- ▶ Thus for each  $n$  we have  $P\{|A_n - \mu| > \epsilon\} \leq P\{Y_\epsilon \geq n\}$ .
- ▶ So  $\lim_{n \rightarrow \infty} P\{|A_n - \mu| > \epsilon\} \leq \lim_{n \rightarrow \infty} P\{Y_\epsilon \geq n\} = 0$ .
- ▶ If the right limit is zero for each  $\epsilon$  (strong law) then the left limit is zero for each  $\epsilon$  (weak law).

## Proof of strong law assuming $E[X^4] < \infty$

- ▶ Assume  $K := E[X^4] < \infty$ . Not necessary, but simplifies proof.
- ▶ Note:  $\text{Var}[X^2] = E[X^4] - E[X^2]^2 > 0$ , so  $E[X^2]^2 \leq K$ .
- ▶ The strong law holds for i.i.d. copies of  $X$  if and only if it holds for i.i.d. copies of  $X - \mu$  where  $\mu$  is a constant.
- ▶ So we may as well assume  $E[X] = 0$ .
- ▶ Key to proof is to bound fourth moments of  $A_n$ .
- ▶  $E[A_n^4] = n^{-4} E[S_n^4] = n^{-4} E[(X_1 + X_2 + \dots + X_n)^4]$ .
- ▶ Expand  $(X_1 + \dots + X_n)^4$ . Five kinds of terms:  $X_i X_j X_k X_l$  and  $X_i X_j X_k^2$  and  $X_i X_j^3$  and  $X_i^2 X_j^2$  and  $X_i^4$ .
- ▶ The first three terms all have expectation zero. There are  $\binom{n}{2}$  of the fourth type and  $n$  of the last type, each equal to at most  $K$ . So  $E[A_n^4] \leq n^{-4} \left( 6 \binom{n}{2} + n \right) K$ .
- ▶ Thus  $E[\sum_{n=1}^{\infty} A_n^4] = \sum_{n=1}^{\infty} E[A_n^4] < \infty$ . So  $\sum_{n=1}^{\infty} A_n^4 < \infty$  (and hence  $A_n \rightarrow 0$ ) with probability 1.

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## Jensen's inequality statement

- ▶ Let  $X$  be random variable with finite mean  $E[X] = \mu$ .
- ▶ Let  $g$  be a **convex** function. This means that if you draw a straight line connecting two points on the graph of  $g$ , then the graph of  $g$  lies below that line. If  $g$  is twice differentiable, then convexity is equivalent to the statement that  $g''(x) \geq 0$  for all  $x$ . For a concrete example, take  $g(x) = x^2$ .
- ▶ **Jensen's inequality:**  $E[g(X)] \geq g(E[X])$ .
- ▶ Similarly, if  $g$  is **concave** (which means  $-g$  is convex), then  $E[g(X)] \leq g(E[X])$ .
- ▶ If your utility function is concave, then you always prefer a safe investment over a risky investment with the same expected return.

## More about Pedro

- ▶ Disappointed by the strong law of large numbers, Pedro seeks a better way to make money.
- ▶ Signs up for job as “hedge fund manager”. Allows him to manage  $C \approx 10^9$  dollars of somebody else’s money. At end of each year, he and his staff get two percent of principle plus twenty percent of profit.
- ▶ Precisely: if  $X$  is end-of-year portfolio value, Pedro gets

$$g(X) = .02C + .2 \max\{X - C, 0\}.$$

- ▶ Pedro notices that  $g$  is a convex function. He can therefore increase his expected return by adopting risky strategies.
- ▶ Pedro has strategy that increases portfolio value 10 percent with probability .9, loses everything with probability .1.
- ▶ He repeats this yearly until fund collapses.
- ▶ With high probability Pedro is rich by then.



- ▶ The “two percent of principle plus twenty percent of profit” is common in the hedge fund industry.
- ▶ The idea is that fund managers have both guaranteed revenue for expenses (two percent of principle) and incentive to make money (twenty percent of profit).
- ▶ Because of Jensen’s inequality, the convexity of the payoff function is a genuine concern for hedge fund investors. People worry that it encourages fund managers (like Pedro) to take risks that are bad for the client.
- ▶ This is a special case of the “principal-agent” problem of economics. How do you ensure that the people you hire genuinely share your interests?

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