

18.600: Lecture 9

Expectations of discrete random variables

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Outline

Defining expectation

Functions of random variables

Motivation

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- ▶ Represents weighted average of possible values X can take, each value being weighted by its probability.

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- ▶ Roll a standard six-sided die. What is the expectation of number that comes up?
- ▶ Answer: $\frac{1}{6}1 + \frac{1}{6}2 + \frac{1}{6}3 + \frac{1}{6}4 + \frac{1}{6}5 + \frac{1}{6}6 = \frac{21}{6} = 3.5$.

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- ▶ Example: toss two coins. If X is the number of heads, what is $E[X]$?
- ▶ State space is $\{(H, H), (H, T), (T, H), (T, T)\}$ and summing over state space gives $E[X] = \frac{1}{4}2 + \frac{1}{4}1 + \frac{1}{4}1 + \frac{1}{4}0 = 1$.

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- ▶ In principle, yes... We only say expectation is defined when $\sum_{s \in S} P(\{s\})|X(s)| < \infty$, in which case it turns out that the sum does not depend on the order.

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- ▶ Generally, $E[aX + b] = aE[X] + b = a\mu + b$.

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- ▶ Alternatively, use symmetry. Expected number of heads should be same as expected number of tails.
- ▶ This implies $E[X] = E[n - X]$. Applying $E[aX + b] = aE[X] + b$ formula (with $a = -1$ and $b = n$), we obtain $E[X] = n - E[X]$ and conclude that $E[X] = n/2$.

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- ▶ Can extend to more variables
$$E[X_1 + X_2 + \dots + X_n] = E[X_1] + E[X_2] + \dots + E[X_n].$$

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- ▶ Can write total number with own hat as
$$X = X_1 + X_2 + \dots + X_n.$$
- ▶ Linearity of expectation gives
$$E[X] = E[X_1] + E[X_2] + \dots + E[X_n] = n \times 1/n = 1.$$

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- ▶ **Comes up everywhere** probability is applied.

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- ▶ Let's assume $u(0) = 0$ and $u(1) = 1$. Then $u(x) = y$ means that you are indifferent between getting 1 dollar no matter what and getting x dollars with probability $1/y$.