#### **Last lecture**

Probability with equally likely outcomes (Ch 1.4)

Poker hands

Random Variables (RV) (Ch 2.1)

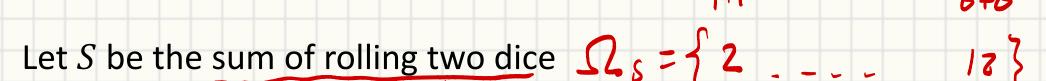
- Definition
- Probability Mass Function (pmf)

# Agenda

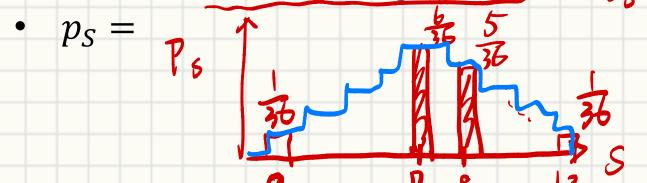
Random Variables (RV)

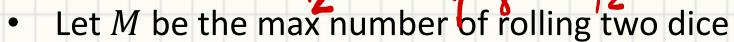
- Probability Mass Function (pmf)
- Mean and Variance (Ch 2.2)

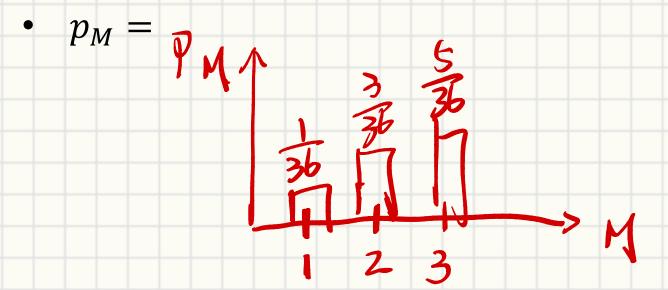
# **Probability Mass Function (PMF)**



6x6.

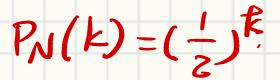






$$\begin{array}{lll}
(2,2) & (3,1) & (1,3) \\
(2,1) & (3,2) & (2,3) \\
(1,2) & (3,3)
\end{array}$$

# **Probability Mass Function (PMF)**



Let N be the # of toss until getting first tail



$$P_{N}(1) = \frac{1}{2} + \frac{1$$

$$\int_{M}^{2} \int_{M}^{2} \int_{M$$

# Mean and Variance (Ch 2.2)

#### Mean

### Do we need detailed $p_{height}$ ?

- In many cases, we just need mean  $\mu_X$  instead of  $p_X$
- $\mu_X = E[X] = \sum_i x_i p_X(x_i)$  probability
- X is the number for a die roll  $E[X] = \frac{1}{1+2+3+4+5}$

# **Function of RV - LOTUS**

X is RV uniformly sampled from  $\{-1, 0, 1, 2, 3\}$ 

$$P_{X}(x) = \begin{cases} \frac{1}{5} & x \in [-1, 3] \\ 0 & \text{else.} \end{cases}$$

$$Y = \begin{cases} x^{2} \\ \cdot & \mu_{y} = E[Y] = 0 \end{cases}$$

$$E[Y] = \begin{cases} 0 & \text{otherwise} \\ 0 & \text{otherwise} \end{cases}$$

$$E[Y] = \begin{cases} 1 & \text{otherwise} \\ 0 & \text{otherwise} \end{cases}$$

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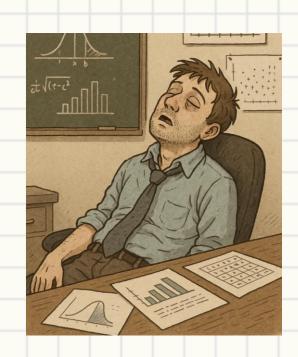
### **Function of RV - LOTUS**

X is RV uniformly sampled from  $\{-1,0,1,2,3\}$ ,  $Y=X^2$  But we can also compute E[Y] from  $p_X$ !

Mean of RV function g(X) is

$$E[g(X)] = \sum_{x} g(x_{x}) P_{x}(x_{x})$$

Law of the unconscious statistician (LOTUS)



# **LOTUS** examples

 $\chi$  is rolling a D6, Y is rolling a D8 (8-sided die)

$$E[XY] = ? \qquad \frac{1}{5} \times \frac{1}{8}$$

$$\sum_{i} \sum_{j} \chi_{i} y_{i} P_{XY}(\lambda, j) P_{XY}(\lambda, j$$





## **LOTUS** examples

#### Math magic trick

- Pick a number from 1-10 💃 🗙
- Multiply it by 3
- Subtract it by 2
- Divided it by 6 and keep the remainder
- Is it 1 or 4?

#### Digit sum

• 
$$Y = X - \text{sum of digits}(X)$$

$$Y = X - \text{sum of algits}(X)$$

$$Y = X - \text{Sum of algits}(X)$$

#### Variance and Standard Deviation

Do you want your salary to be

$$p_X(10K) = 1 \text{ or}$$
  
 $p_Y(0) = 0.99, p_Y(1000K) = 0.01$  Lottery

Mean is important... but not complete enough

• Variance Var(X) is how PMF spreads apart from  $\mu_X$ 

scalar

• 
$$Var(X) \triangleq E[(X - \mu_X)^2] = E[X^2 - \lambda \mu_X]X + \mu_X$$

$$= \sum_{i} (x_{i}^{2} - 2\mu_{i}x_{i} + \mu_{x}^{2}) P_{x}(x_{i})$$

