

Cont. vs. Discrete RVs.

Discrete - counting, derive or infer mass function

Continuous - mostly cannot be derived from scratch
? assert a distribution (pdf)
Modeling assumptions

Statistics vs. Prob.

Prob - single events.

Statistics - tools for analyzing densities and distributions



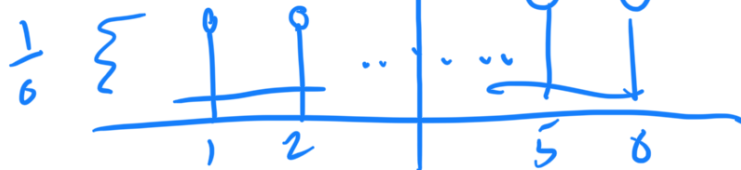
Expectation

RV X

$$E[X] = \sum_i a_i \Pr(X=a_i)$$
$$= \sum_i a_i f(a_i)$$

Roll of die...

$$\sum_{i=1}^6 i \cdot \frac{1}{6} = 3.5$$



Mean of distribution/density
 $f(a)$

$$\mu = \sum_i a_i f(a_i)$$

?*

Mean of distribution/density
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?*

Sum of 3 dice

$$E[X] = E[X_1 + X_2 + X_3] = E[X_1] + E[X_2] + E[X_3]$$
$$= 3 \times 3.5 = 10.5$$

EXPECTATION IS LINEAR.

Difference of 2 dice

$$E[X] = E[X_1 - X_2] = E[X_1] - E[X_2] = 0$$

↑ ↑
different RV's.

X = Difference of larger die minus smaller die

$$X \in \{0, 1, \dots, 5\}$$

$$\text{abs}(X_1 - X_2)$$

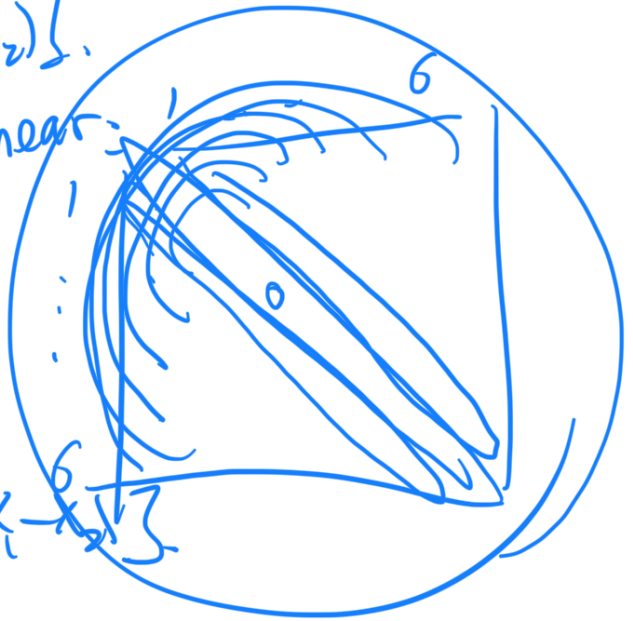
$$E[X] = E[\text{abs}(X_1 - X_2)]$$

not linear

$$\begin{aligned} f(6) &= \frac{6}{36} \\ f(1) &= \frac{10}{36} \\ f(2) &= \frac{8}{36} \end{aligned}$$

$$\sum_{i=0}^5 a_i f(a_i)$$

$$= E[\text{abs}(X_1 - X_2)]$$



$$0 \times \frac{6}{36} + 1 \times \frac{10}{36} + \dots$$

$$\sum_{i=0}^5 a_i f(a_i)$$

Binomial

n trials of Bernoulli RV $\{0, 1\}$. $P(X=1)=p$
prob of k success's (1's) in n trials

$$Pr(X=k) = \binom{n}{k} (p)^k (1-p)^{n-k}$$

$\binom{n}{k}$ \uparrow # of way to order k success's in an n -tuple

$(p)^k$ \uparrow prob of k success's

$(1-p)^{n-k}$ \uparrow prob of $n-k$ failure's

$$E[X] = E[X_1 + \dots + X_n] = E[X_1] + \dots + E[X_n]$$

\uparrow # success's

\uparrow p \uparrow p \uparrow p

$$= np$$

Variance - "spread" of a dist

$$\text{Var}[X] = E[(X - E[X])^2]$$

↑
mean, μ .

$$= \sum_i (a_i - \mu)^2 f(a_i)$$

↑
 $|a_i - \mu|$ ← mean deviation

aside: notice μ is not a Rv.

$\text{Var}[X]$ denotes σ^2 .

$$\sigma = \sqrt{\text{Var}[X]} = \text{std deviation}$$

Cont. form for expectation

$f(x)$ - prob density function
x - cont. RV

$$E[X] = \int_{-\infty}^{\infty} x f(x) dx$$

Annotations:
- x : RV
- $f(x)$: pdf at x
- dx : da

Uniform distribution $U(a, b)$

$$f(x) = \frac{1}{b-a}$$

$$\begin{aligned} \int_{-\infty}^{\infty} U(a, b) dx &= \int_a^b \frac{1}{b-a} dx = \frac{x^2}{2} \frac{1}{b-a} \Big|_a^b \\ &= \frac{b^2 - a^2}{2} \frac{1}{b-a} = \frac{b+a}{2} \left(\frac{b-a}{b-a} \right) = \frac{b+a}{2} \end{aligned}$$

Identitäten für Var

$$\text{Var}[aX+b] = a^2 \text{Var}[X].$$

RV konst.

X - Bernoulli

$$\mu = 0(1-p) + 1 \cdot p =$$

$$\text{Var}[X] = (0-p)^2(1-p) + (1-p)^2 \cdot p$$

$$= +p^2 - p^3 + (1-2p+p^2)p$$

$(a_i - \mu)$

$f(a_i)$

$$+p^2 + p^3 + p - 2p^2 + p^3$$

$$= p - p^2 = p(1-p).$$
