

X a sample space, P probability
 f is a random variable

$$f: X \rightarrow \mathbb{R}.$$

$$\text{Def: } E[f] = \int_X f(x) dP(x)$$

- X is discrete

$$E[f] = \sum_{x \in X} f(x) P(x)$$

Bernoulli: parameter p $X = \{H, T\}$ $f(H) = 1$
 $f(T) = 0$

$$E(f) = f(H)p + f(T)(1-p)$$

$$= p$$

$$E[f] = p.$$

Continuous

$$X = \mathbb{R} \quad P(u \in \mathbb{R}) = \frac{1}{\sigma\sqrt{2\pi}} \int_u e^{-x^2/(2\sigma^2)} dx$$

$$u = [-s, s]$$

$$P([-s, s]) = \frac{1}{\sigma\sqrt{2\pi}} \int_{-s}^s e^{-x^2/(2\sigma^2)} dx$$

$$x: X \rightarrow \mathbb{R}$$

$$E[x] = \frac{1}{\sigma\sqrt{2\pi}} \int_{x=\mathbb{R}} x e^{-x^2/(2\sigma^2)} dx \leftarrow E[x] = 0$$

$$E[f] = \frac{1}{\sigma\sqrt{2\pi}} \int f(x) e^{-x^2/(2\sigma^2)} dx$$

Prop: $E[af+bg] = aE[f] + bE[g]$
 f, g random variables, a, b constants

$$E[af+bg] = \sum_{x \in X} [af(x) + bg(x)] P(x) = a \sum_{x \in X} f(x) P(x) + b \sum_{x \in X} g(x) P(x).$$

Continuous case

$$E[af+bg] = \int_X (af(x) + bg(x)) \overbrace{P(x) dx}^{\text{density}} = aE[f] + bE[g].$$

Def: Recall that A, B are independent if $P(A \cap B) = P(A)P(B)$.
Discrete f and g are independent random variables if

$$P(f=a \text{ and } g=b) = P(f=a)P(g=b) \text{ for all values } a, b.$$

$X = \{H, T\}^N$ f_1 is 1 if first flip is heads, 0 otherwise. f_2 is 1 if 2nd flip is heads, 0 otherwise.

$$P(f_1 = a \text{ and } f_2 = b) = P(f_1 = a)P(f_2 = b)$$

$$E[fg] = E[f]E[g].$$

★ What is $E[f]$ if f is binomial with parameters N, p .

N Bernoulli's with parameter

$$f = f_1 + \dots + f_N$$

$$E[f] = NE[f_i] = Np.$$

Variance: f is a random variable.

$$E[(f - E[f])^2] = \sigma^2, \text{ the variance.}$$

Lemma $\sigma^2(f) = E[f^2] - E[f]^2$

$$(f - E[f])^2 = f^2 - 2fE[f] + E[f]^2$$

$$E[(f - E[f])^2] = E[f^2] - 2E[f]^2 + E[f]^2 \\ = E[f^2] - E[f]^2.$$

Bernoulli:

$$f^2 = f \quad f^2(H) = 1 = f(H) \\ f^2(T) = 0 = f(T)$$

$$E[f^2] = E[f] = p$$

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

Maximum σ^2 at $p = 1/2$.

Binomial.

$$f = (f_1 + \dots + f_N) \quad N(N-1) \\ f^2 = f_1^2 + \dots + f_N^2 + \sum_i f_i f_j$$

$$E[f^2] = NE[f_i^2] = Np + E[\sum_i f_i f_j] \\ = Np + N(N-1)p^2$$

$$E[f]^2 = N^2 p^2$$

$$\sigma^2(f) = Np + N^2 p^2 - Np^2 - N^2 p^2 \\ = Np(1-p).$$

Continuous case
X normal random variable

$$E[X] = 0.$$

$$E[X^2] = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} \underbrace{x^2 e^{-x^2/(2\sigma^2)}}_{p(x)dx} dx$$

$$\frac{d}{d\sigma} \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-x^2/(2\sigma^2)} dx = \frac{d}{d\sigma} 1$$

... get answers ...
 $\sigma^2(x) = \sigma^2$

