## Functions of random variables, Expectation

Putting a value on random variables

Michael Psenka



"Random variable": notes Oet A "random variable" is a function X: 52 -SR  $\mathcal{L} = \{\mathcal{H}, \mathcal{T}\}^{\times \mathsf{S}}$ X, "number of hads" Q = 591, 2, 3, 9, 53 ∈ IR R= "vif. over 2"  $P = \frac{1}{2^{s}} \begin{pmatrix} s \\ 1 \end{pmatrix} = \frac{1}{2^{s}} \begin{pmatrix} s \\ 2 \end{pmatrix} =$ X: {H,T} ->  $\begin{pmatrix} S \\ k \end{pmatrix} = \frac{1}{2^{k}}$ (H,T,T,H,T)

#### Functions induce random variables

### Numeric random variables A roman variable (e) 's numeric P 2 CR

**Distribution** (graphical) for function fik > R, the Graph of f is {(x, f(x))} The graph of a distribution is the set  $\xi(\omega, P(\omega))^{2}_{\omega \in \Omega}$  o.s coin, breed towards e.s.



e.g. com, p=0.5, versited com, p=0.2

**Examples:** Binomial 'n Bernaulli r.v., all v' probability p' Q = {0,1}x {0,1}x ... x {0,1} (n copies), det P  $Q_2 = f(Q_1), \quad f := "sum of indkeg" = <math>\sum_{i=1}^{n} \omega_i$  $P(X=k) = \sum_{a \in E} P(a), E := \xi a | exactly k | s z$  $= \sum_{q \in E} p^{k} (1-p)^{n-k} = \binom{n}{k} p^{k} (1-p)^{n-k} \in \binom{k}{k} p^{k} (1-p)^{n-k} \in \binom{n}{k} p^{k} (1-p)^{n-k} = \binom{n}{k} p^{k} (1-p)^{n-k} p^{k} (1-p)^{n-k} = \binom{n}{k} p^{k} (1-p)^{n-k} p^{k$ 

(Numerical) joint distributions  $\frac{1}{0.05} \in P(X=0, Y=0)$ YOI N 0.05 0.8 0.8 0.03 0.1 a boint distribution M(XXY) E.c. 2 "entropled colong" Two r.v. X,Y, they have calns F.g. I, the Indepent

Marginal distributions P(x=k, Y=1)XY  $P(X=k) = P(X=k|Y=1)P(Y=1) + P(X=k|Y=2)P(Y=2) + \dots$ = P(x-le, x=1) + P(x=le, y=2)+...  $= \sum_{i \in \mathcal{A}_{v}} P(x_{i} \mid x_{i}, y_{i} = i)$ 

Independence P(X=k, Y=() X,Y Def: XY indep if V k, ( P(X=k,Y=l) = P(X=k)P(Y=l)→ P(X=E, Y=F)=IP(XEE)P(YEF) P(X=k,Y=l) > P(XEE,Y=l) > P(XEE,Y=F)

#### Expectation

**Definition.** The expectation of a numeric random variable X = $(\Omega, \mathbb{P})$  is given by the following:

$$\mathbf{E}[X] \coloneqq \sum_{x \in \Omega} x \mathbb{P}(X = x).$$

e.g. Banaulli of 
$$p=0.3$$
,  $E[(x) = 0.0.5 + 1.0.5 = 0.5]$   
 $G_{nompol}$ ,  $E[x] = \sum_{x''} \frac{1}{p} \frac{1}{(1-p)^{n+k}} \binom{n}{k}$ 

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#### Linearity of expectation

**Theorem.** Let X, Y be two (numeric) random variables, and their joint distribution given by  $\mathbb{P}(X = x, Y = y)$ . The expectation is a "linear operator": that is,

1. E[X + Y] = E[X] + E[Y],2. E[cX] = cE[X] for any fixed  $c \in \mathbb{R}$ .

Linearity of expectation AF Willing out E[X+Y] = Z(X+Y)P(X=X, Y=Y)=  $\sum_{x \in Q_{x}} (x+y) P(x=x, Y=y)$ xeQ\_x yeQ\_y  $= \underbrace{\sum}_{x \neq y} P(x_{x,y}, y_{x,y}) + \underbrace{\sum}_{y \neq y} P(x_{y,y}, y_{x,y}) + \underbrace{\sum}_{y \neq y} P(x_{y,y}, y_{x,y})$  $= \sum_{x} \left( \sum_{y} \left[ P(X=y, Y=y) \right] + \sum_{y} \left( \sum_{x} \left[ P(X=y, Y=y) \right] \right) \right)$  $= \sum_{x} |P(X = x) + \sum_{y} P(Y = y) = E[X] \cdot E[Y].$ 

#### Product of independent expectations

**Theorem.** Let X, Y be two (numeric) random variables, and their joint distribution given by  $\mathbb{P}(X = x, Y = y)$ . If X, Y are independent, then expectation factors out their product:

 $\mathbf{E}[XY] = \mathbf{E}[X]\mathbf{E}[Y].$ 

# Product of independent expectations ECXIJ = ZZ XyP(X=x, Y=y) $= Z \sum_{x,y} P(X = x) P(Y = y) \neq$ $= \underbrace{\leq}_{x} P(X=x) \left( \underbrace{\leq}_{y} P(Y=y) \right)$ $= \mathbb{E}[Y] \left( \sum_{x \in X} \mathbb{P}(X : x) = \mathbb{E}[Y] \mathbb{E}[X] \right)$