# Diocrete Multivariate Distributions REVIEW

### Multivariate Distributions

\* often, there is need to define multiple RVs in experiment For RVs X, Y, the joint / bivaciate distribution is

the collection of probabilities

- WELTION OF PROBABILITIES  $P((X,Y) \in B)$ ,  $\forall B \leq IR^2$  (note that B represents)

an arbitrary subset of  $R^2$ )

Exemple

$$\begin{array}{lll}
X & \text{ample} \\
S & = \{a, b, c, d\} \\
X & = \{b, c, d\} \\
X & = \{b, c, d\} \\
X & = \{b, c, d\} \\
X & = \{c, d$$

Note that only s= 2 results in (X(s), Y(s)) & B Then  $P((X,Y) \in B) = P(\{a\}) = \frac{|\{a\}|}{|a|} = \frac{1}{|a|}$ 

#### Joint PMF

For discrete RVs X,Y, the joint/biveriate PMF is defined as

etinel as
$$P_{XY}(x,y) = P(X=x,Y=y) = P(X=x,Y=y)$$

### Merginal PMF

Sometimes, you may went the distribution of single variable from joint distribution (called marginal distribution)

$$\rho_{\mathbf{x}}(\mathbf{x}) = P(\mathbf{x}_{=\mathbf{x}}) - \sum_{\mathbf{y}} P(\mathbf{x}_{=\mathbf{x}}, \mathbf{y}_{=\mathbf{y}}) = \sum_{\mathbf{y}} \rho_{\mathbf{x}, \mathbf{y}}(\mathbf{x}, \mathbf{y})$$

$$S = \{a, b, c, d\}$$

$$X(s) = \begin{cases} 1.5 = a, s = c \\ 3.5 = b, s = d \end{cases}$$

$$\sqrt{(5)} = \begin{cases} 1.8 = 0.5 = 0 \\ 3.9 = 0.6 = 0 \end{cases}$$

$$\begin{array}{l}
S = \{a, b, c, d\} \\
X(s) = \begin{cases}
3 \cdot 5 = b, 6 = d \\
0 \cdot 6 \cdot w
\end{cases}$$

$$\begin{array}{l}
Y(s) = \begin{cases}
1 \cdot 5 = 3, 5 = c \\
3 \cdot 5 = b, 6 = d
\end{cases}$$

$$\begin{array}{l}
Y(s) = \begin{cases}
1 \cdot 4 \cdot X = 1, Y = 1 \\
1 \cdot 4 \cdot X = 1, Y = 3
\end{cases}$$

$$\begin{array}{l}
Y(s) = \begin{cases}
1 \cdot 4 \cdot X = 3, Y = 3 \\
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### Marginal PMFs

$$P_{X}(1) = P(X=1) = \sum_{y} P(X=1,Y=y)$$
  
=  $P(X=1,Y=1) + P(X=1,Y=3)$ 

$$p_{x}(3) = P(X=3) = \sum_{y=1}^{\infty} P(X=3, Y=y)$$
  
=  $P(X=3, Y=1) + P(X=3, Y=3)$ 

$$7hus,$$

$$p_{X}(x) = \begin{cases} 1/2 : x=1 \\ 1/2 : x \ge 3 \\ 0 : o/w \end{cases}$$

#### Conditional PMF

For RVs X, Y, the conditional PMF of X given Y=y is

$$P_{XY}(x|y) = P(X=x|Y=y) = \frac{P(X=x,Y=y)}{P(Y=y)} = \frac{P_{X,Y}(x,y)}{P_{Y}(y)}$$

Remember: Px14 (xly) is just a function of x - don't be thrown off by notation

This means px14 (x14) must be proper PMF

$$S = \{a,b,c,d\}$$

$$P_{Y}(y) = \begin{cases} 1/2 & y = 1 \\ 1/2 & y = 3 \\ 0 & o/\omega \end{cases}$$

For 
$$y=1$$
 (similar for  $y=3$ )  

$$p_{XIY}(x|1) = \frac{p_{XIY}(x,1)}{p_{Y}(1)}$$

$$=\frac{1}{2} \cdot \rho_{X,Y}(x,1)$$

$$=\frac{1}{2} \cdot \rho_{X,Y}(x,1)$$

$$= \begin{cases} 2/4; & x = 1 \\ 2/4; & x = 3 \\ 0 & 0 \end{cases}$$

### Independent RVs

For discrete RVs X, Y, X and Y are independent when  $p_{xy}(x_1y) = P(X=x_1Y=y) = P(X=x_1) \cdot P(Y=y_1) = p_x(x_1) p_Y(y_1) \quad \forall x_1 y_1$ 

## Conditional Variables from Independent Variables

If X, Y are indepent  $p_{x|y}(x|y) = \frac{p_{x,y}(x,y)}{p_{y}(y)} = \frac{p_{x}(x)p_{y}(y)}{p_{y}(y)} = p_{x}(x)$ 

# Example

$$S = \{a, b, c, d\}$$

$$X(c) = \begin{cases} 1; 6 = e, s = c \\ 3 = e + b, s = d \\ 0; 0 \neq \omega \end{cases}$$

$$\begin{cases}
(5) = 6 \\
3:5 = 6 \\
0:6 \\
0
\end{cases}$$

$$p_{x}(x) = \begin{cases} 1/2 : x = 1 \\ 1/2 : x = 3 \\ 0 : o/w \end{cases}$$

Turns out  $X \perp Y$  (i.e.  $p_{x,y}(x,y) = p_x(x) p_y(x_y)$ )

We have
$$P_{x}(x) = \begin{cases} \frac{1}{2} \cdot x = 1 \\ 0 \cdot 0 / \omega \end{cases} \quad P_{x,y}(x,y) = \begin{cases} \frac{1}{4} \cdot x = 1, y = 1 \\ \frac{1}{4} \cdot x = 3, y = 1 \\ \frac{1}{4} \cdot x = 3, y = 1 \end{cases}$$

$$P_{x}(y) = \begin{cases} \frac{1}{2} \cdot y = 1 \\ \frac{1}{2} \cdot y = 3 \end{cases} \quad 0 \cdot 0 / \omega$$

Essy to see  $p_{x,y}(x,y) = p_x(x)p_y(y)$ 

This also means pary (xlg) = px(x)

$$P_{Q} = \frac{4}{52} = \frac{1}{13}$$

$$PK = \frac{1}{13}$$

$$P_0 = 1 - \frac{1}{13} - \frac{1}{13} = \frac{1}{13} = \frac{1}{13}$$

$$P_{Q} = \frac{4}{52} = \frac{1}{3}$$
(probability queen is drawn)
$$P_{K} = \frac{1}{13}$$
(probability king is drawn)
on single trial
$$P_{Q} = \frac{1}{13} = \frac{1}{13}$$
(probability neither is drawn)
on single trial

From Q6 (MULTINOMIAL DISTRIBUTION) Say you three categories (5 independent trials) [n = 5] Also note that we can convert [X+Y+Z=5] this into a two veriable problem Z=5-X-Y] 7.6) From 5 (MULTIVARIATE HYPERGEOMETRIC DISTRIBUTION) n=5 w/oreplacement M = 52c=44(8+6+c=52) (0ther) c=54-4-4) a=4 b=4 (Kings) (Queens)  $P(\chi=\chi,\gamma=y) = \frac{(\alpha)(\beta)(\beta-\chi-y)}{(\gamma)(\beta-\chi-y)}$ P(X=x,Y=y)  $\frac{(x)(y)^{y+t}}{(x)(5-x-y)}$   $\frac{(52)}{5}$ 

The when considering trials of replacement and multiple outcomes

3.  $P(X=x,Y=y) = \begin{cases} \frac{1}{2} & \frac{1}{2$