$P(X > Y) = \sum_{x \in R} \sum_{y \in R} P_{X,y}(x,y) \underbrace{1}_{x > y}$ XY H = E E M PS 11 x > y = S S (1-pxp1xeNo (-p)xp1yeNo 1x = y+1 = PS S (1-P) (1-P) 11 x z y = 1 = p2 5 (1-p) 5 (1-p) 1 x € 2 y+1, y  $= \rho^{2} \underbrace{\S(1-\rho)^{\vee}}_{\text{yeNo}} \underbrace{\S(1-\rho)^{\times}}_{\text{xeNo}} \underbrace{1}_{\text{xeSy+1,y+2,...}}$ = pr & (1-p)y & (1-p)x yen, xeny+1,...3  $= p^{2} \sum_{y \in \mathbb{N}_{\bullet}} (1-p)^{y} \sum_{x \in \mathbb{N}_{\bullet}} (1-p)^{x'+(y+1)}$ = pr & (1-p) (1-p) (1-p) & (1-p) x

it is not equal to 1/2 since X = Y sometimes but it should be slightly less than 1/2.

X, Y id beam(p) = (1-p) x p 1 x e N.

P(X > Y)

$$= \int_{X \in \mathbb{N}_{0}} \sum_{x \in \mathbb{N}_{0}} \sum_$$

Bag of fruits: apples, bananas and cantaloupes  $p_1$ : probability of drawing an apple  $p_2$ : probability of drawing a banana  $p_3$ : probability of drawing a cantaloupe Note:  $p_1 + p_2 + p_3 = 1$ Draw n fruits with replacement. Let  $X_1$  be the rv that counts the number of apples,  $X_2$  be the rv that counts the number of bananas and  $X_3$  be the rv that counts the number of

the number of apples, 
$$X_2$$
 be the rv that counts the number of bananas and  $X_3$  be the rv that counts the number of cantaloupes. Note:  $X_1 + X_2 + X_3 = n$ 

$$X_1 \sim \beta \ln (n, \rho_1) \quad X_2 \sim \beta \ln (n, \rho_2) \quad X_3 \sim \beta \ln (n, \rho_3)$$
What is the probability of drawing 2 apples, 2 bananas and 3 cantaloupes on a draw of  $n=7$ ?
$$\int_{X_1, X_2, X_3} (2, 2, 3) = \int_{X_1 + X_2 + X_3} (2, 2, 3) = \frac{7!}{2! 2! 3!} \int_{1}^{2} \int_{2}^{2} \int_{2}^{2$$

 $=\begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix} p_1^{\times} p_2^{\times} p_2^{\times}$ This is called the multinomial rv. In general consider K categories of items and probabilities of sampling each category  $p_1$ ,  $p_2$ , ...,  $p_K$  where  $p_1 + p_2 + ... + p_K = 1$ . The number of samples is n.  $\overrightarrow{X} \sim \mathcal{M}_{u}$  +inom  $\left(n, \overrightarrow{p}\right) = \left(x_{1}, x_{1}, \dots, x_{\kappa}\right) \underbrace{f_{1}^{x_{1}} f_{2}^{y_{2}} \dots f_{\kappa}^{x_{\kappa}}}_{\kappa}$ 

$$\begin{array}{lll}
\overrightarrow{X} & \sim Mul + i_{hom} \left( n, \overrightarrow{p} \right) = \left( x_{1}, x_{1}, \dots, x_{K} \right) \beta_{1}^{X_{1}} \beta_{2}^{X_{2}} \dots \beta_{K}^{X_{K}} \\
Supp\left[\overrightarrow{X}\right] &= \left\{ \overrightarrow{X} : \overrightarrow{X} \in \{0, \dots, n_{3}^{K}\} \in \overrightarrow{X} : \overrightarrow{1} = n_{3}^{K} \right\} \\
\text{Parameter space:} & n \in \mathbb{N}, \quad \overrightarrow{p} \in \left\{ \overrightarrow{v} : (0, 1)^{K} \notin \overrightarrow{v} : \overrightarrow{1} = 1_{3}^{K} \right\} \\
\text{Lea } K = 2 & x_{1} : (0, 1)^{K} \notin \overrightarrow{v} : \overrightarrow{1} = 1_{3}^{K} \\
\xrightarrow{X_{1} : (0, K_{1})^{1}} & x_{1} : (0, 1)^{K} \notin \overrightarrow{v} : \overrightarrow{1} = 1_{3}^{K} \\
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\xrightarrow{X_{1} : (0, K_{1})^{1}} & x_{1} : (0, K_{1})^{K} \notin \overrightarrow{v} : \overrightarrow{v} : (0, K_{1})^{K} \notin \overrightarrow{v} : (0, K_{1})^{K} \notin \overrightarrow{v} : (0, K_{1})^{K} \notin \overrightarrow{v} : (0, K_{1})^{K} \mapsto \overrightarrow{v}$$

0=P(X,=1 | Xz=n) + P(X,=1) = np(-p)-1

Some definitions for vector rvs  $\vec{\lambda} := E[\vec{X}] := \begin{bmatrix} E[X_1] \\ E[X_2] \end{bmatrix}$ 

definitions for vector 
$$\overrightarrow{VS}$$
  $\overrightarrow{A}$   $\overrightarrow{A}$ 

A matrix rv is a matrix whose entries are rv's

What is the "variance" of a vector rv? How do we define 
$$\bigvee_{n} \left[ \overrightarrow{X} \right] := E[(x - n)^{x}],$$
Recall the definition for a scalar rv.  $Var[X] := E[(x - n)^{x}],$ 
What is the "covariance" between two rv's  $X_{-1}$  and  $X_{-2}$ ?

$$Var[X] := E[(X_{-1}, X_{-1})^{x}],$$
Rules of covariances:
$$\bigvee_{n} Av \left[ X_{1} + X_{2} \right] := C_{1}^{n} + c_{1}^{n} + C_{1}^{n}$$

 $\mathbb{P}\left[\sum_{x\in X}[X,X]\right]=\bigvee_{x\in X}[X]\geq \mathbb{P}\left[X\right]$ 

(3) Cov [X1, X2] = Cov [X2, X1]  $(x_1 + x_1, x_3) = Cov[X_1, X_3] + Cov[X_2, X_3]$ 1 (ov [a, X, a, X,] = a, a, o,  $\Im \operatorname{Var} \left[ X_1 + X_2 + \ldots + X_n \right] = \underbrace{\widehat{S}}_{i} \underbrace{\widehat{S}}_{i} \operatorname{Cov} \left[ X_{i}, X_{j} \right]$ 

 $= \underbrace{\sum_{i=1}^{2} \sum_{j=1}^{2} Cov[X_{i}, X_{j}]}_{+ Cov[X_{i}, X_{i}] + Cov[X_{i}, X_{i}]} + \underbrace{Cov[X_{i}, X_{i}]}_{+ Cov[X_{i}, X_{i}] + Cov[X_{i}, X_{i}]}$ = 5,2 + 5,2 + 26,2