

Math 621, Lec 5 9/12/17

$\vec{X}$  is a vector of r.v.'s s.t.  $\dim(\vec{X}) = k$ .

$$\vec{\mu} := E[\vec{X}] = \begin{bmatrix} E[X_1] \\ \vdots \\ E[X_k] \end{bmatrix}$$

$$\Sigma := \text{Var}(\vec{X}) = \begin{bmatrix} \text{Var}(X_1) & \text{Cov}(X_1, X_2) & \cdots \\ \text{Cov}(X_2, X_1) & \text{Var}(X_2) & \cdots \\ \vdots & \vdots & \ddots \\ \vdots & \vdots & \vdots & \text{Var}(X_k) \end{bmatrix} = \{ \text{Cov}(X_i, X_j) \}$$

$i=1 \dots k$   
 $j=1 \dots k$

$$\Sigma_0 := \text{Corr}(\vec{X}) = \begin{bmatrix} 1 & \text{Corr}(X_1, X_2) & \cdots \\ & 1 & \cdots \\ & & \ddots \\ & & & 1 \end{bmatrix} = \{ \text{Corr}(X_i, X_j) \}$$

$$\text{let } T = X_1 + \dots + X_k = \vec{1}^T \vec{X} = [1 \dots 1] \begin{bmatrix} X_1 \\ \vdots \\ X_k \end{bmatrix}$$

$$E[T] = \sum_{i=1}^k \mu_i = \vec{1}^T \vec{\mu} = \text{Cov}(\sum_{i=1}^k X_i, \sum_{i=1}^k X_i)$$

$$\text{Var}[T] = \text{Var}[\vec{1}^T \vec{X}] = \sum_{j=1}^k \sum_{i=1}^k \text{Cov}(X_i, X_j)$$

$$\text{let } Y = \vec{c}^T \vec{X}$$

$$E[Y] = \sum c_i \mu_i = \vec{c}^T \vec{\mu}$$

$$\text{Var}[Y] = \text{Var}[\vec{c}^T \vec{X}] = ?$$

If  $A \in \mathbb{R}^{n \times n}$ ,  $\vec{c} \in \mathbb{R}^n$   
 What is  $\vec{c}^T A \vec{c}$ ? Called a quadratic form...

$$= \vec{c}^T \begin{bmatrix} c_1 a_{11} + \dots + c_n a_{1n} \\ c_1 a_{21} + \dots + c_n a_{2n} \\ \vdots \\ c_1 a_{n1} + \dots + c_n a_{nn} \end{bmatrix} = \begin{aligned} &c_1^2 a_{11} + c_1 c_2 a_{12} + \dots + c_1 c_n a_{1n} + \\ &c_2 c_1 a_{21} + c_2^2 a_{22} + \dots + c_2 c_n a_{2n} + \\ &\vdots \\ &c_n c_1 a_{n1} + c_n c_2 a_{n2} + \dots + c_n^2 a_{nn} \end{aligned}$$

$$= \sum_{j=1}^n \sum_{i=1}^n c_i c_j a_{ij}$$

What is  $\text{Var}[\vec{c}^T \vec{X}]$ ?

$$\begin{aligned} &= \text{Var}[c_1 X_1 + \dots + c_n X_n] \\ &= \sum_{i=1}^n \sum_{j=1}^n \text{Cov}[c_i X_i, c_j X_j] \\ &= \sum_{i=1}^n \sum_{j=1}^n c_i c_j \text{Cov}[X_i, X_j] \\ &= \vec{c}^T \text{Var}(\vec{X}) \vec{c} \end{aligned}$$

↑ a 1x...  
 ↓ denotes the Var-cov matrix

Markowitz Portfolio Theory  
 $X_1, \dots, X_n$  are returns of  $K$  assets  
 $w_1, \dots, w_n$  are the allocation weights s.t.  $\vec{1}^T \vec{w} = 1$   
 $V = \vec{w}^T \vec{X} \Rightarrow M_V = \vec{w}^T \vec{\mu}$   
 $\sigma_V^2 = \vec{w}^T \text{Var}(\vec{X}) \vec{w}$   
 min  $\sigma_V^2$  s.t.  $M_V = M_0$  and  $\vec{1}^T \vec{w} = 1$   
 we will solve this later

$\vec{X} \sim \text{Multinomial}(n, \vec{p})$ ,  $E(\vec{X}) = \begin{bmatrix} E(X_1) \\ \vdots \\ E(X_n) \end{bmatrix} = \begin{bmatrix} n p_1 \\ \vdots \\ n p_n \end{bmatrix} = n \vec{p}$

Makes sense since

$X_1 \sim \text{Bin}(n, p_1), \dots, X_n \sim \text{Bin}(n, p_n)$  but not independent!

$\text{Var}(\vec{X}) = ?$

What is  $\text{Var}[X_1]$ ?  $= n p_1 (1 - p_1)$  See notes from prob. class

$$\text{Var}(\vec{X}) = \begin{bmatrix} n p_1 (1-p_1) & \text{Cov}(X_1, X_2) \\ \text{Cov}(X_2, X_1) & n p_2 (1-p_2) \\ \vdots & \ddots & \vdots \\ \vdots & \vdots & \vdots & n p_k (1-p_k) \end{bmatrix}$$

Need  $\text{Cov}[X_i, X_j]$

1st strategy

$$= E[X_i X_j] - n_i n_j$$

$$= \left( \sum_{X_i \in \text{supp}(X_i)} \sum_{X_j \in \text{supp}(X_j)} X_i X_j p(X_i, X_j) \right) - n p_i (1-p_i) n p_j (1-p_j)$$

Difficult to get!

This is the marginal prob of  $(X_i, X_j)$ . Disturbed... maybe on the!

Next strategy...

Recall if

$$X \sim \text{Bin}(n, p) \Leftrightarrow X = \sum_{i=1}^n X_i \quad \text{where } X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \text{Bern}(p) = \text{Bin}(1, p)$$

if

Since all marginals are binomial...

$$\begin{bmatrix} X_1 \\ \vdots \\ X_k \end{bmatrix} \sim \text{Multin}(n, \vec{p}) \Leftrightarrow \begin{aligned} X_1 &= \sum_{i=1}^n X_{i1} & X_{11}, \dots, X_{n1} &\stackrel{\text{iid}}{\sim} \text{Bern}(p_1) \\ \vdots & & & \\ X_k &= \sum_{i=1}^n X_{ik} & X_{1k}, \dots, X_{nk} &\stackrel{\text{iid}}{\sim} \text{Bern}(p_k) \end{aligned}$$

Since  $X_{i1}, X_{i2}, \dots, X_{ik}$  are not indep!

Further  $\vec{X} \sim \text{Multinomial}(n, \vec{p}) \Rightarrow \vec{X} = \sum_{i=1}^n \vec{X}_{i \cdot}$  s.t.  $\vec{X}_1, \dots, \vec{X}_n \stackrel{i.i.d.}{\sim} \text{Multinomial}(1, \vec{p})$  (4)

why OH + E HW!

Now...  $\text{Cov}[X_i, X_j] = \text{Cov}\left[\sum_{l=1}^n X_{li}, \sum_{h=1}^n X_{hj}\right]$

$$= \sum_{l=1}^n \sum_{h=1}^n \text{Cov}[X_{li}, X_{hj}]$$

$$= \sum_{l=1}^n \sum_{h=1}^n E[X_{li} X_{hj}] - p_i p_j$$

indep. across cols, dep. across rows

$X_{11}$	$X_{12}$	...	$X_{1n}$
$X_{21}$	$X_{22}$		$X_{2n}$
$\vdots$	$\vdots$		$\vdots$
$X_{K1}$	$X_{K2}$	...	$X_{Kn}$

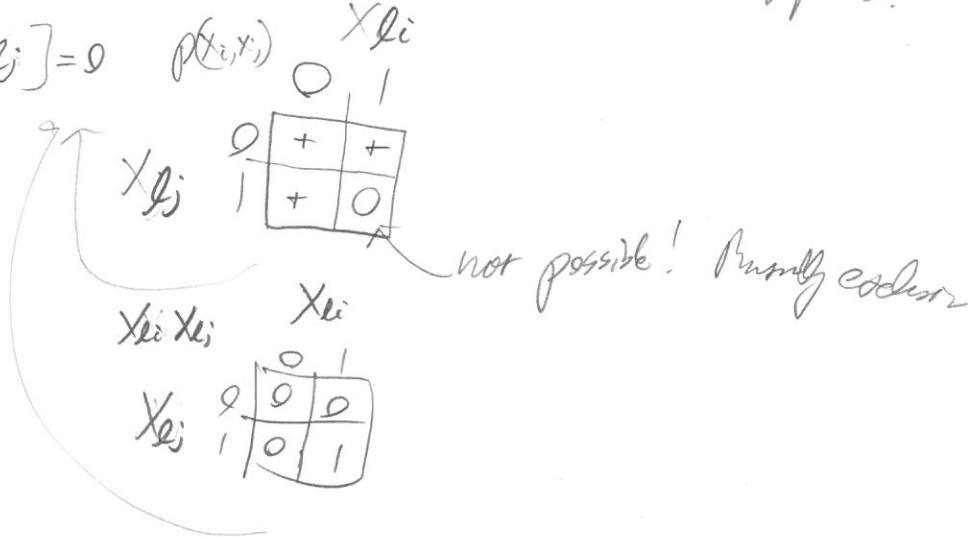
(All Bernoullis)

if  $l \neq h$ , this is a different multinomial model so  $E[X_{li} X_{hj}] = E[X_{li}] E[X_{hj}] = p_i p_j$

thus all these terms are zero. Now...  $l=h$ , what happens?

$$= \sum_{l=1}^n E[X_{li} X_{li}] - p_i p_j$$

$$E[X_{li} X_{li}] = 0 \quad p(X_{li}, X_{li})$$



How many terms where  $l=h$ ?  $\hookrightarrow \dots$

$$\Rightarrow \text{Cov}[X_i, X_j] = \underbrace{-n p_i p_j}_{\substack{\uparrow \uparrow \uparrow \\ + + +}}$$

make sense?

Yes if # C's  $\uparrow$

$\Rightarrow$  # A's  $\downarrow$

Since there's a fixed # of trials.

HW:  $\text{Cov}[X_i, X_j]$ ,  
matrix notation

Continuous r.v.'s have prob. density functions (PDF)  
 $f(x) = 0 \forall x$ . why?  $\nexists p.s.t. \sum_{x \in S} p(x) < \infty$  s.t.  $S$  is countable

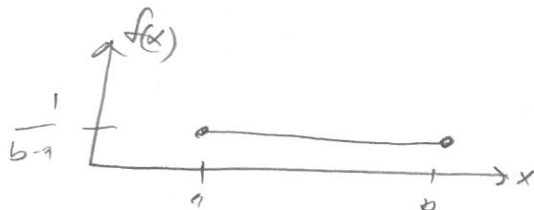
$$f(x) := F'(x) \text{ due } F(x) \text{ is the CDF}$$

The domain of  $f$  is the  $\text{supp}(f) := \{x : f(x) > 0\}$  s.t.  $f(x) = 0$

and  $|\text{supp}(f)| = |\mathbb{R}|$  i.e. countable, infinite

A basic continuous r.v. is the uniform r.v.

$$X \sim U(a, b) := \frac{1}{b-a}$$

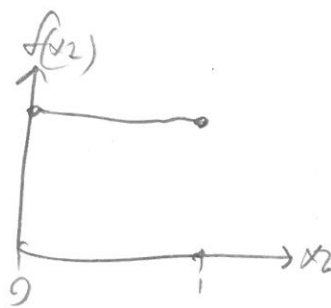
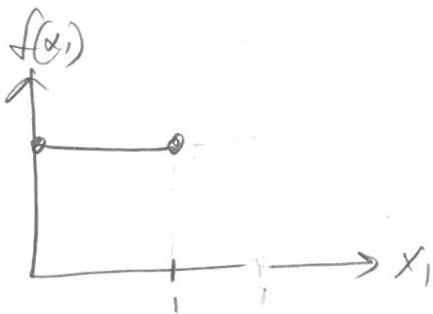


Params  $a, b \in \mathbb{R}$  s.t.  $a < b$   
 space:

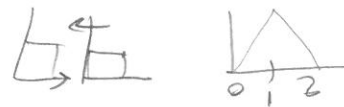
$$\text{supp}(X) = [a, b]$$

The "standard" uniform is  $a=0, b=1 \Rightarrow X \sim U(0, 1) := 1$

What does  $T_2 = X_1 + X_2 \sim ?$  If  $X_1, X_2 \stackrel{iid}{\sim} U(0, 1)$ ?



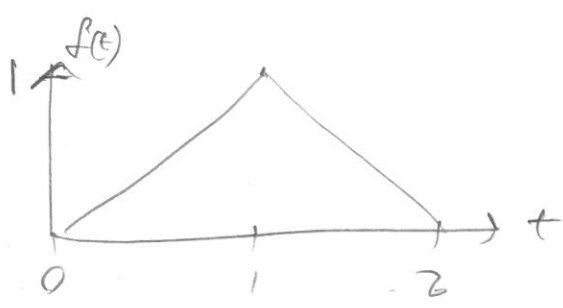
$$\text{supp}(T_2) = [0, 2]$$



Intuitively, how often could  $T_2 = 0$ ? well  $X_1 = 0 \& X_2 = 0 \Rightarrow$  rare common  
 ' ' '  $T_2 = 2$ ? well  $X_1 = 1 \& X_2 = 1 \Rightarrow$  rare  
 ' ' '  $T_2 = 1$ ?  $X_1 = 0 \& X_2 = 1, X_2 = 1 \& X_1 = 0, X_1 = 0.5 \& X_2 = 0.5$ , etc...  $X_1 = \frac{1}{4} \& X_2 = \frac{3}{4}$

$$T_2 = 1.9$$

$X_1 = 1$  &  $X_2 = 0.9$ ,  $X_1 = 0.95$  &  $X_2 = 0.95$ ,  $X_1 = 0.5$ ? No...  $\Rightarrow$  Rare



It seems this should be the triangular dist.

Proof...

$$f_{T_2}(t) = \int_{\text{supp}(X_1)} f_{X_1}(x) f_{X_2}(t-x) dx$$

$$x-t \in [-1, 0] \Rightarrow x \in [t-1, t]$$

$$f_{X_1}(x) * f_{X_2}(x) = \dots$$

$$= \int_0^1 \mathbb{1}_{x \in [0,1]} \mathbb{1}_{t-x \in [0,1]} dx$$

Convolutions for densities...  
Same concept as before...

$$= \int_0^1 \mathbb{1}_{x \in [t-1, t]} dx$$

$$= \int_{\max\{0, t-1\}}^{\min\{1, t\}} dx = x \Big|_{\max\{0, t-1\}}^{\min\{1, t\}} = \min\{1, t\} - \max\{0, t-1\}$$

this is the answer for  $t \in (0, 2)$

alternatively...  $f_{T_2}(t) = \begin{cases} \text{if } t < 1 & \Rightarrow t \\ \text{if } t \geq 1 & \Rightarrow 1 - (t-1) = 2-t \end{cases} \mathbb{1}_{t \in (0, 2)}$