

9/12/17

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

$$\vec{\mu} = E(\vec{X}) = \begin{bmatrix} E(x_1) \\ \vdots \\ E(x_k) \end{bmatrix},$$

$$\Sigma := \text{Var}(\vec{X}) = \begin{bmatrix} \text{Var}(x_1) & \text{cov}(x_1, x_2) & \dots & \text{cov}(x_1, x_k) \\ \text{cov}(x_2, x_1) & \text{Var}(x_2) & \dots & \text{cov}(x_2, x_k) \\ \vdots & \vdots & \ddots & \vdots \\ \text{cov}(x_k, x_1) & \text{cov}(x_k, x_2) & \dots & \text{Var}(x_k) \end{bmatrix} = \left\{ \text{cov}(x_i, x_j) \right\}_{\substack{i=1 \dots k \\ j=1 \dots k}}$$

$$\Sigma_c = \text{Corr}(\vec{X}) = \begin{bmatrix} 1 & \dots & \text{corr}(x_1, x_k) \\ \vdots & \ddots & \vdots \\ \text{corr}(x_k, x_1) & \dots & 1 \end{bmatrix} = \left\{ \text{corr}(x_i, x_j) \right\}_{\substack{i=1 \dots k \\ j=1 \dots k}}$$

$$T = x_1 + \dots + x_k = \vec{1}^T \vec{X}$$

$$\vec{T}_k = \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} = \begin{bmatrix} 1 & \dots & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_k \end{bmatrix}$$

$$E(T) = \sum_{i=1}^k \mu_i = \vec{1}^T \vec{\mu}$$

$$\text{Var}(T) = \text{Var}[\vec{1}^T \vec{X}] = \text{Var}[x_1 + \dots + x_k]$$

$$= \text{cov}[x_1 + \dots + x_k, x_1 + \dots + x_k]$$

$$= \sum_{i=1}^k \sum_{j=1}^k \text{cov}[x_i, x_j]$$

$$\text{let } Y = \vec{c}^T \vec{X}, \text{ where } \vec{c} \in \mathbb{R}^k$$

$$E(Y) = \vec{c}^T \vec{\mu} \quad \text{Var}(Y) = \text{Var}[\vec{c}^T \vec{X}]$$

$$\text{If } A \in \mathbb{R}^{n \times n}, \vec{c} \in \mathbb{R}^n$$

$\vec{c}^T A \vec{c}$  "Quadratic Form"

$$= \vec{c}^T \begin{bmatrix} c_1 a_{11} + \dots + c_n a_{1n} \\ c_1 a_{21} + \dots + c_n a_{2n} \\ \vdots \\ c_1 a_{m1} + \dots + c_n a_{nm} \end{bmatrix} = c_1^2 a_{11} + c_1 c_2 a_{12} + \dots + c_1 c_n a_{1n} + \dots$$

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

$$= \text{Var}(c_1 x_1 + \dots + c_k x_k)$$

$$= \text{Cor}[c_1 x_1 + \dots + c_k x_k, c_1 x_1 + \dots + c_k x_k]$$

$$= \sum_{i=1}^k \sum_{j=1}^k \text{Cor}[c_i x_i, c_j x_j]$$

$$= \sum_{i=1}^k \sum_{j=1}^k \text{Cor} \cdot c_i c_j \text{Cor}[x_i, x_j]$$

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

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

$$\begin{bmatrix} \text{Var}(x_1) & \text{Cor}(x_1, x_2) \\ \text{Cor}(x_2, x_1) & \text{Var}(x_2) \end{bmatrix}$$

"Variance - Covariance matrix"

## Markowitz optimal Portfolio

let  $X_1, \dots, X_k$  be r.v models for the returns on  $k$  assets.

let  $w_1, \dots, w_k$  be the weights/allocations for each. note  $\vec{1}^T \vec{w} = 1$

$$V = \vec{w}^T \vec{X} \quad E(V) = \vec{w}^T \vec{\mu} = \mu_0$$

$$\text{Var}(V) = \vec{w}^T \Sigma \vec{w}$$

Given a  $\mu_0$ , min  $\vec{w}^T \Sigma \vec{w}$  st  $\vec{1}^T \vec{w} = 1$

or  $\{ \vec{w} : \vec{1}^T \vec{w} = 1 \}$ .

$\vec{X} \sim \text{multinomial}(n, \vec{p})$

$$E(\vec{X}) = \begin{bmatrix} E(X_1) \\ \vdots \\ E(X_k) \end{bmatrix} = \begin{bmatrix} np_1 \\ np_2 \\ \vdots \\ np_k \end{bmatrix} = n\vec{p}$$

$$\text{Var}(\vec{X}) = \begin{bmatrix} np_1(1-p_1) & \text{Cov}(X_1, X_2) & & \\ & np_2(1-p_2) & & \\ & & \ddots & \\ & & & np_k(1-p_k) \end{bmatrix}$$

$$\text{cov}(X_i, X_j) = E[X_i X_j] - \mu_i \mu_j$$

$$= \sum_{x_i \in \text{supp}(X_i)} \sum_{x_j \in \text{supp}(X_j)} x_i x_j P_{X_i X_j}(x_i, x_j) - \mu_i \mu_j$$

we don't know this.

Recall

$$X_1 \sim \text{Bin}(n, p_1) \quad X_1 = \sum_{i=1}^n X_{1i} \text{ s.t. } X_{11}, X_{12}, \dots, X_{1n} \stackrel{\text{iid}}{\sim} \text{Bern}(p_1)$$

$$X_k \sim \text{Bin}(n, p_k) \quad X_k = \sum_{i=1}^n X_{ik} \text{ s.t. } X_{1k}, X_{2k}, \dots, X_{nk} \stackrel{\text{iid}}{\sim} \text{Bern}(p_k)$$

$$\begin{bmatrix} X_{11} & X_{12} & \dots & X_{1n} \\ X_{21} & X_{22} & \dots & X_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ X_{k1} & X_{k2} & \dots & X_{kn} \\ \vdots & \vdots & \ddots & \vdots \\ X_{n1} & X_{n2} & \dots & X_{nn} \end{bmatrix}$$

$$\vec{X} \sim \text{multinomial}(n, \vec{p})$$

$$\vec{X} = \sum_{i=1}^n X_{i\cdot} \text{ s.t. } X_{1\cdot}, \dots, X_{n\cdot} \stackrel{\text{iid}}{\sim} \text{multinomial}(1, \vec{p})$$

$$\text{cov}[X_i, X_j]$$

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

$$= \sum_l \sum_h \text{cov}[X_{li}, X_{hj}]$$

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

$$= \sum_{y \in \{0,1\}} \sum_{x \in \{0,1\}} xy p(x,y) = 0 \quad \text{if } l=h.$$

$$= \sum E(X_{ei}, X_{ej}) - p_i p_j$$

$$= \sum_{e=1}^n -p_i p_j = -n p_i p_j$$

$$P(x,y)$$

	0	1
0	+	+
1	+	0

$$xy$$

	0	1
0	0	0
1	0	1

$$l \neq h.$$

$$E[X_{ei}, X_{hi}] = E(X_{ei}) E(X_{hj}).$$

$$= p_i p_j$$

$$\text{var}(\vec{X}) = n$$

$$\begin{bmatrix} p_1(1-p_1) & -p_1 p_2 & \dots & -p_1 p_k \\ -p_2 p_1 & p_2(1-p_2) & & \\ & & \ddots & \\ & & & p_k(1-p_k) \end{bmatrix}$$

For  $\vec{X} \sim \text{multinom}(n, \vec{p})$ .

Continuous r.v.'s  $X$  have CDF



$F(x)$  and PDF:

$$f(x) = F'(x)$$

Note

$$P(x) = 0 \quad \forall x$$

$$\text{Supp}(X) = \{x : f(x) > 0\}$$

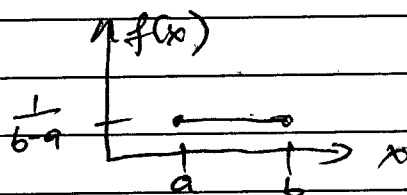
PMF...

$$|\text{Supp}(X)| \geq |\mathbb{R}|$$

$$\exists p \text{ s.t. } \sum_{x \in \text{Supp}(X)} p(x) < \infty$$

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

where  $a, b \in \mathbb{R}, b > a$ .

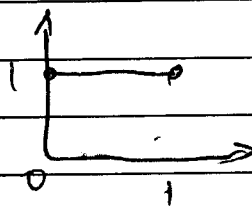
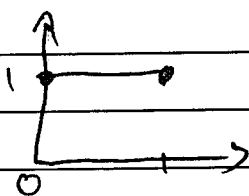


$$\text{Supp}(X) = [a, b]$$

let

$$a = 0, b = 1 \Rightarrow X \sim U(0, 1) \quad (\text{standard uniform})$$

let  $T_2 = X_1 + X_2$  so  $X_1, X_2 \sim \text{iid } U(0, 1)$ .



$$\text{Supp}(T) = [0, 2]$$

How far does  $T = 0$ ?  $X_1 = 0, X_2 = 0 \Rightarrow \text{rare}$

$T = 2$ ?  $X_1 = 1, X_2 = 1 \Rightarrow \text{rare}$

$T = 1$ ?  $X_1 = 0$  or  $X_1 = \frac{1}{3}, X_2 = \frac{2}{3} \Rightarrow \text{common}$   
 $X_2 = 1$

$$\begin{aligned}
 f(t) &= \int_{\substack{x_1 \\ x \in [0,1]}} f(x) \int_{\substack{x_2 \\ x \in [0,1]}} f(t-x) dx = \int_0^1 (1)(1) \underbrace{1_{\substack{t-x \in [0,1] \\ x-t \in [-1,0] \\ x \in [t-1,t]}}} dx \\
 &= \int_0^1 1_{x \in [t-1,t]} dx = \int_{\max\{0, t-1\}}^{\min\{1, t\}} 1 dx = (\min\{1, t\} - \max\{0, t-1\}) \cdot 1
 \end{aligned}$$

$$= \begin{cases} t & \text{if } t < 1 \\ 1 - (t-1) & \text{if } t \geq 1 \end{cases} \quad \text{for } t \in [0, 2]$$