

Set theoretic notions have special
interpretations in prob. theory

⊛ Complement in Ω of event A , written
 A^c "NOT A "

⊛ UNION $A \cup B$ is event — at least
 A or B occurs

⊛ Intersection $A \cap B$ "Both A and B occur"

⊛ Inclusion $A \subseteq B$ "occurrence of
 A implies occurrence of B "



To a basic set of outcome, ω_i assign \mathbb{R} numbers, called probabilities,

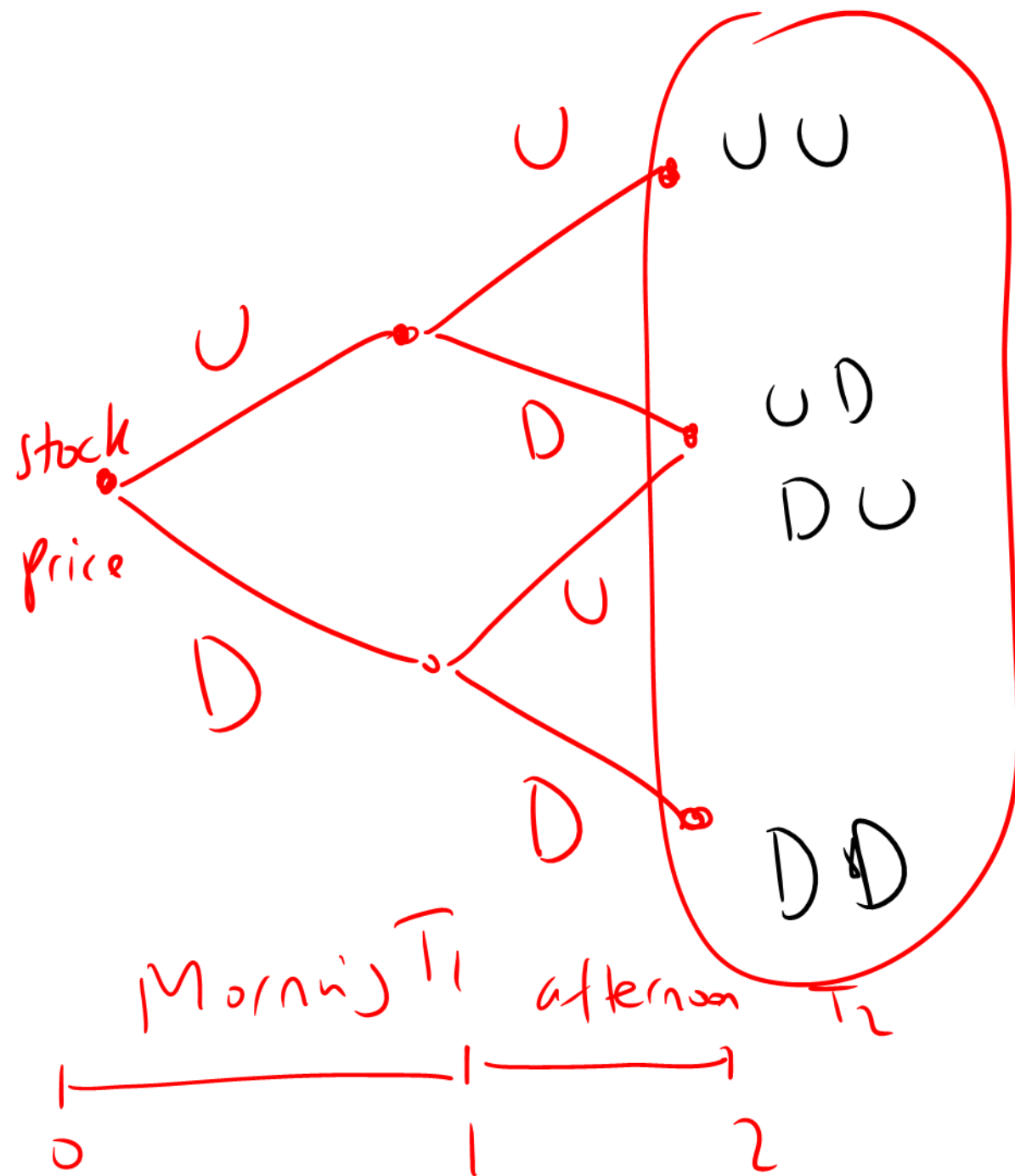
Written $P(\omega_i) = P_i$

Then for any event E_i

$$\textcircled{1} P(E) = \sum_{\omega_i \in E} P_i$$

$$\textcircled{2} 0 \leq P_i \leq 1$$

$$\textcircled{3} P(\Omega) = \sum_i P_i = 1$$

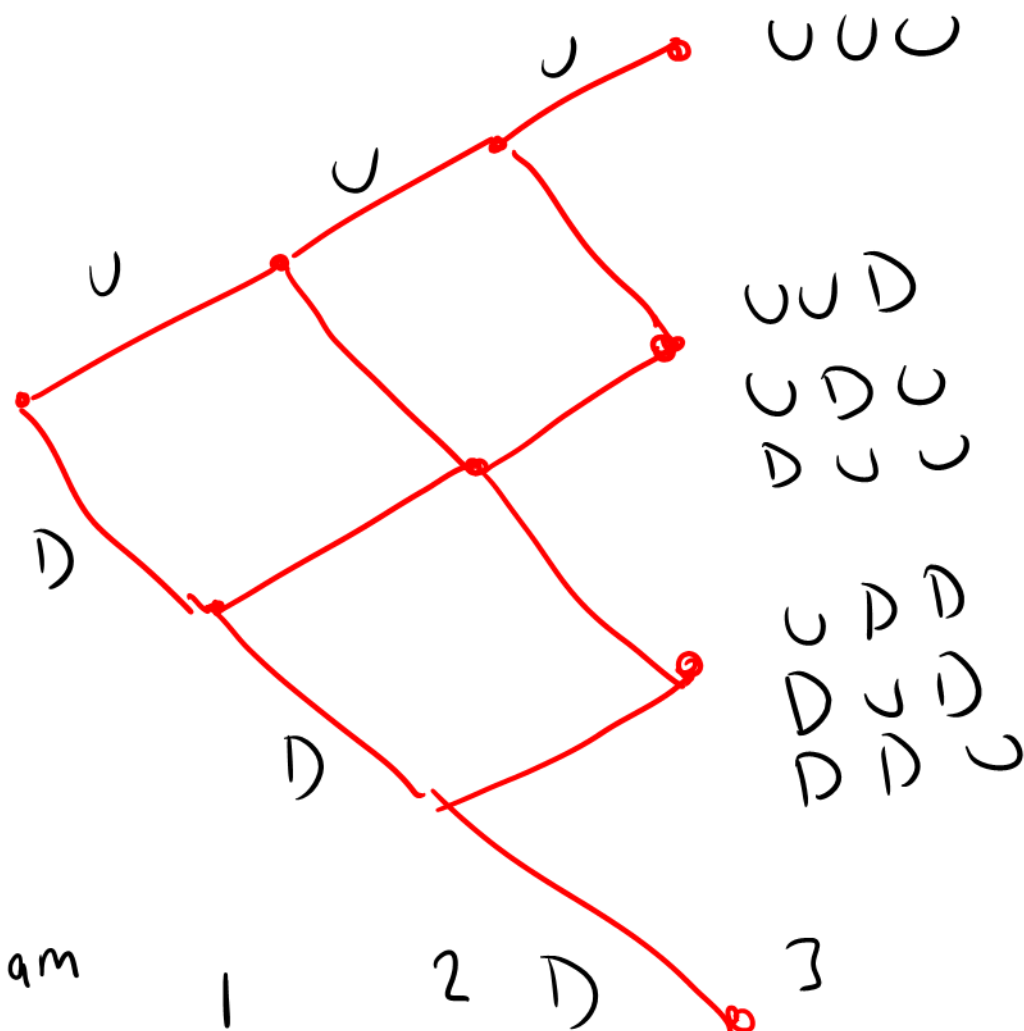


Sample Space Ω

Examples of Trajectories:

UU	ω_1
DD	ω_2
UD	ω_3
DU	ω_4

Seeing inside a filtration.

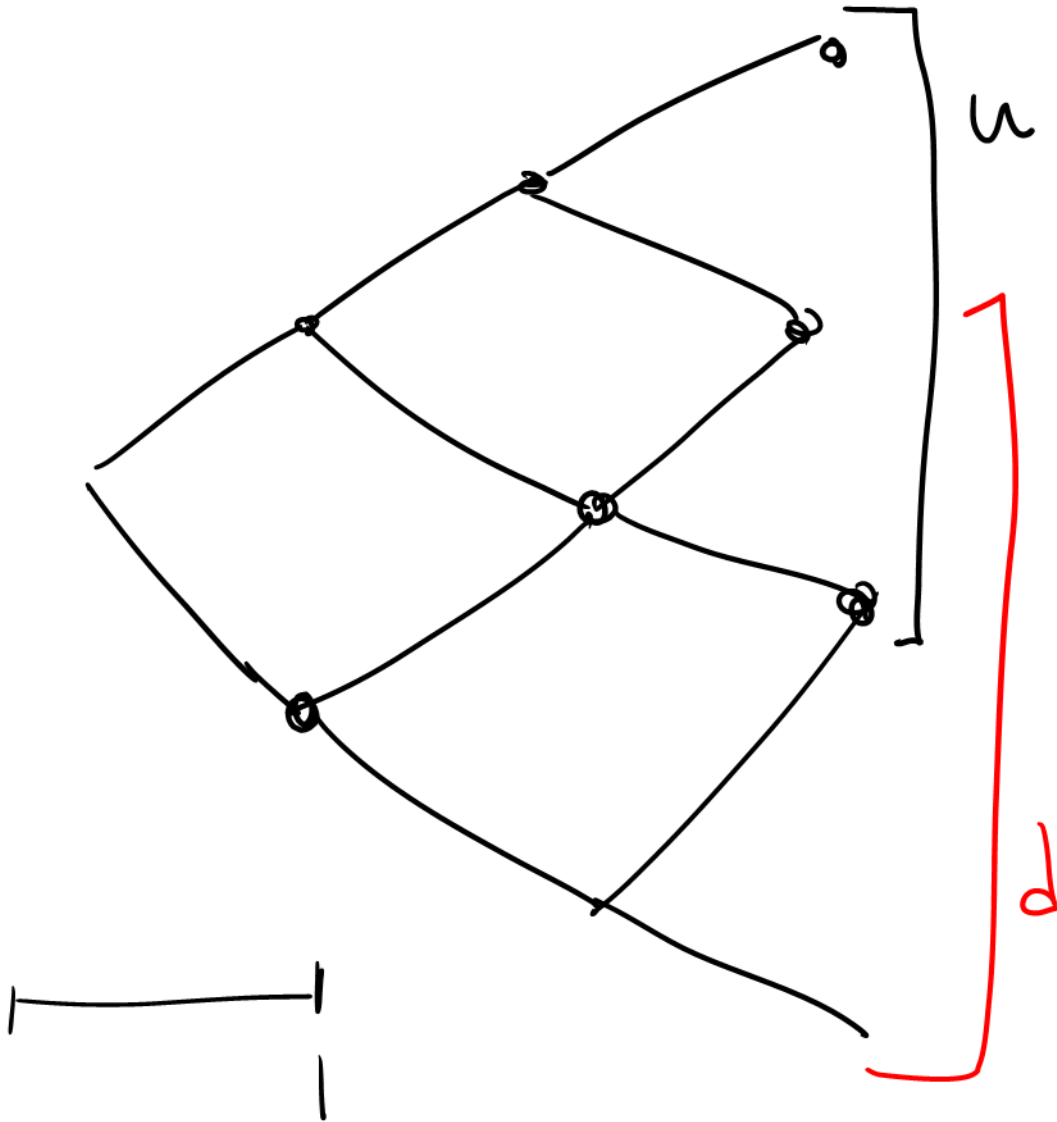


$$\Omega_3 = \{UUU, UUD, UDU, DUU, UDD, DUD, DDU, DDD\}$$

Period,	size of Ω
1	2
2	$2^2 = 4$
3	$2^3 = 8$
10	$2^{10} = 1024$
100	$2^{100} = 1.27 \times 10^{30}$

9 am 1 2 D 3
 11 am 1 pm
 DDDD sell

At time 1



$\Pi = \{0, 1, 2, \dots, n\}$ discrete time set

Then

$\Omega = \Omega_n$ is set of all outcomes of
n coin tosses; each sample path
of length n; written

$$\omega = \omega_1 \omega_2 \dots \omega_n$$

e.g. 3 time step binomial model
 $\longrightarrow \omega = \omega_1 \omega_2 \omega_3$ [length 3]

$$P(x) = P[X < x]$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{1}{2}s^2} ds$$

$M_X(\theta)$

M.G.F

TSE $e^x = \sum_{n=0}^{\infty} \frac{x^n}{n!}$

$$M_X(\theta) = \mathbb{E}[e^{\theta X}] = \int_{-\infty}^{\infty} e^{\theta x} p(x) dx$$

$$= \int_{\mathbb{R}} \left(1 + \theta x + \frac{\theta^2 x^2}{2!} + \dots \right) p(x) dx$$

$$= \int_{\mathbb{R}} p(x) dx + \theta \int_{\mathbb{R}} x p(x) dx + \frac{\theta^2}{2!} \int_{\mathbb{R}} x^2 p(x) dx$$

$$= 1 + \theta \mathbb{E}(X) + \frac{\theta^2}{2!} \mathbb{E}(X^2) + \dots + \frac{\theta^n}{n!} \mathbb{E}(X^n) + \dots$$

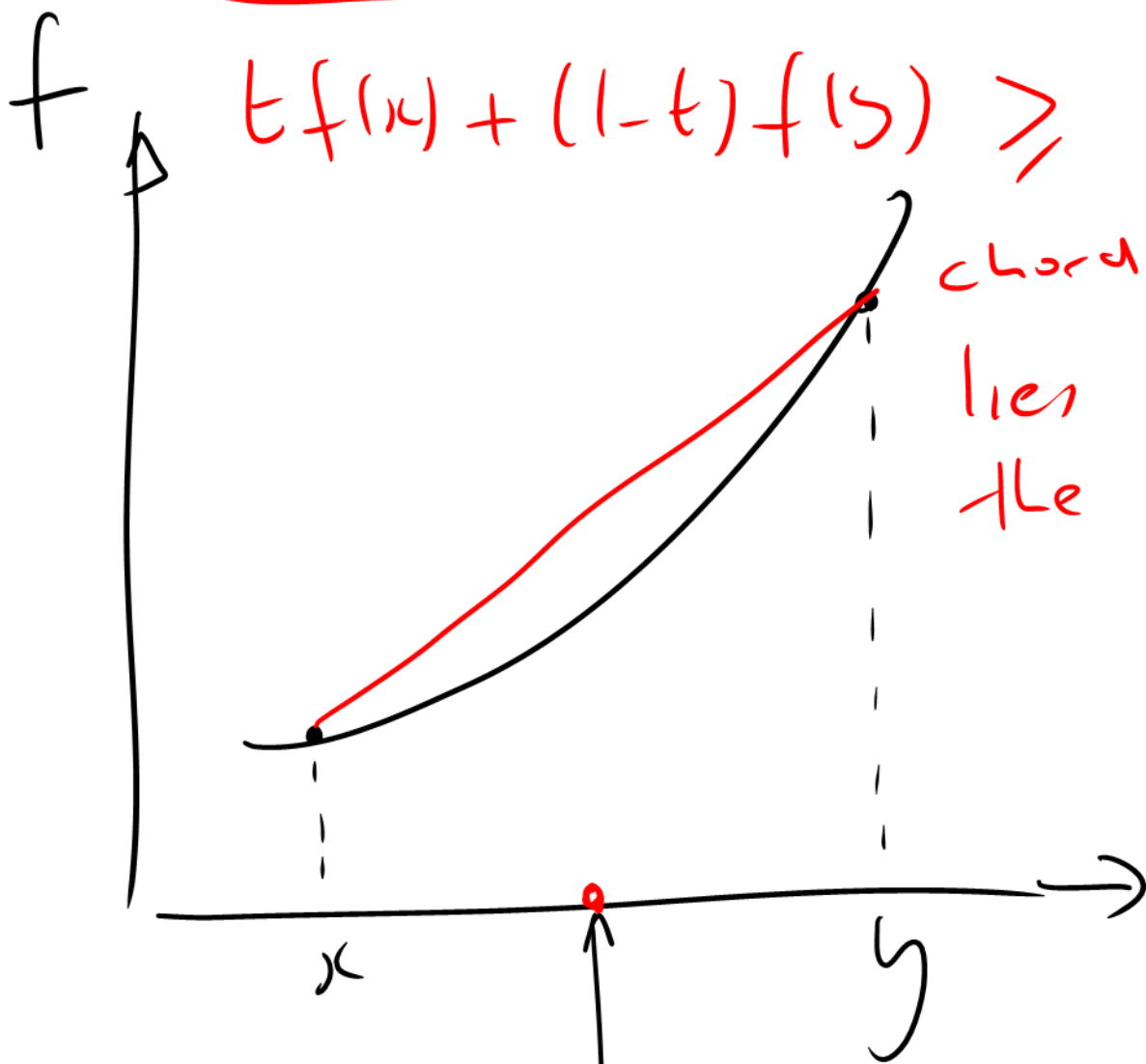
$$M_x(\theta) = \sum_0^{\infty} \frac{\theta^n}{n!} E[X^n]$$

for any k , the k^{th} moment write

$$m_k = \left. \frac{d^k}{d\theta^k} M_x(\theta) \right|_{\theta=0}$$

Convex Functions

$$tf(x) + (1-t)f(y) \geq f(tx + (1-t)y)$$



chord always lies above the curve

$$f(x) = x^2 \quad \cup$$

Variance, minimising

$$f(x) = e^x$$

returns

$$f(x) = |x| \quad \checkmark$$

integrability

$tx + (1-t)y$
 $0 < t < 1$ weighted average

