

Recall :-

Theorem 3.12 :  $\cdot X_1, X_2, \dots, X_n$ , i.i.d  $X$  with  $\sum_{i=1}^n x_i = S_n$

$\cdot \exists \delta > 0$   $\sigma \in (-\delta, \delta)$ ,  $E e^{\sigma X} < \infty$

$\cdot T$  be a stopping time wrt  $\{X_n\}_{n \geq 1}$

$$E[e^{rS_T - T\eta(r)}] = 1 \quad \text{where } r \in (-\delta, \delta)$$

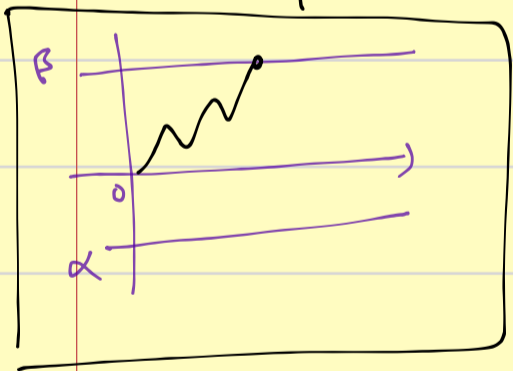
$$\text{and } \eta(r) = \log(E e^{rX}).$$

Proof:- [Optional Sampling Theorem] :  $Z_n = e^{rS_n - n\eta(r)}$   
under  $|\text{Range}(X)| < \infty$

Lemma 3.13 let  $\{X_n\}_{n \geq 1}$  i.i.d  $X \in X \not\equiv 0$

For each  $n \geq 1$ ,  $S_n = \sum_{i=1}^n X_i$

$\beta < 0 < \alpha$



$$T = \min\{n \geq 1 \mid S_n \geq \alpha \text{ or } S_n \leq \beta\}$$

Threshold stopping times

$$\Rightarrow P(T < \infty) = 1 \quad \& \quad E[T^n] < \infty \quad \forall n \geq 1$$

Theorem 3.14 : let  $\{X_n : n \geq 1\}$  be a sequence <sup>(of Discrete)</sup> IID s.i.s

$X$ . For  $n \geq 1$  let  $S_n = \sum_{i=1}^n X_i$ . Assume :-

(M1) -  $E[X] < \infty$ .

(S1) -  $T$  be a stopping time for  $\{X_n\}_{n \geq 1}$   $E(T) < \infty$

$$\text{then } E(S_T) = E[X] E(T).$$

# Proof of Theorem 3.14 :-

(Discrete Fubini) Result from Analysis:-  $\{a_{jk}\}_{j \geq 1, k \geq 1}$   $a_{jk} \in \mathbb{R}$

If  $\sum_{j=1}^{\infty} \sum_{k=1}^{\infty} |a_{jk}| < \infty$  then  $\sum_{j=1}^{\infty} \sum_{k=1}^{\infty} a_{jk} = \sum_{k=1}^{\infty} \sum_{j=1}^{\infty} a_{jk}$

As  $E(J) < \infty \Rightarrow P(J < \infty) = 1$

$\therefore S_J \equiv$  is well defined.

$$E(|S_J|) = \sum_{k \in R_S} |k| P(S_J = k) \quad \text{where}$$

$R_S \equiv$  countable set  $\supseteq \bigcup_{n=1}^{\infty} \text{Range}(S_n)$

Ex: Requires applying at finite sums & limits

$$= \sum_{k \in R_S} |k| \sum_{n=1}^{\infty} P(S_J = k, J=n)$$

[Rearrangement non-negative]  $\Leftarrow = \sum_{n=1}^{\infty} \sum_{k \in R_S} |k| P(S_n = k, J=n)$

[Fix  $n \geq 1$  and consider  $S_n \mathbb{1}(J=n)$ ]  $\Leftarrow = \sum_{n=1}^{\infty} E(|S_n| \mathbb{1}(J=n))$

$$= \sum_{n=1}^{\infty} E\left(\left|\sum_{m=1}^n X_m\right| \mathbb{1}(J=n)\right)$$

(Apply D.C. inequality at finite sums and take limits)  $\leq \sum_{n=1}^{\infty} \sum_{m=1}^n E(|X_m| \mathbb{1}(J=n)) \quad \text{--- } (*)$

[interchange n and m]

Discrete Fubini

(order of sums.)

$$\sum_{m=1}^{\infty} \sum_{n=m}^{\infty} E[X_m \mathbb{1}(J=n)]$$

$$= \sum_{m=1}^{\infty} E[X_m \mathbb{1}(J \geq m)]$$

By definition: independent of  $X_1, \dots, X_{n-1}$

"function of  $X_1, \dots, X_{n-1}$ "

J stopping time wrt  $\{X_n\}_{n \geq 1}$

$\mathbb{1}(J \geq m) \in \mathcal{A}_{n-1}$

$(J < m)^c$

Independence

$$= \sum_{m=1}^{\infty} E[X_m] P(J \geq m)$$

$\{X_n\}_{n \geq 1}$  iid  $X$

$$= \left[ \sum_{m=1}^{\infty} P(J \geq m) \right] E[X]$$

(Discrete Fubini)

$$= E|X| E[J] < \infty$$

(m1) and (S1)

Repeat above calculation without 1-1 :-

$$E(S_J) = \sum_{k \in \mathbb{R}_+} k P(S_n = k)$$

(inequality at  $\otimes \equiv$ )

$$\sum_{n=1}^{\infty} \sum_{m=1}^n E(X_m, J=n)$$

Reorder of sums

(Discrete Fubini)

$$= \sum_{m=1}^{\infty} E[X_m \sum_{n=m}^{\infty} \mathbb{1}(J=n)]$$

$$\{X_m\}_{m \geq 1} \text{ i.i.d. } \leftarrow E[X] = \sum_{m=1}^{\infty} P(J \geq m)$$

$$= E[X] E[J] \quad \square$$

- "Strategies for when to stop are not really effective as far as mean is concerned".

Corollary 3.15 :-

Let  $\{X_n\}_{n \geq 1}$  i.i.d.  $X \in X \neq 0$

For each  $n \geq 1$ ,  $S_n = \sum_{i=1}^n X_i$

$\beta < 0 < \alpha$

$$J = \min\{n \geq 1 \mid S_n \geq \alpha \text{ or } S_n \leq \beta\}$$

Threshold stopping times

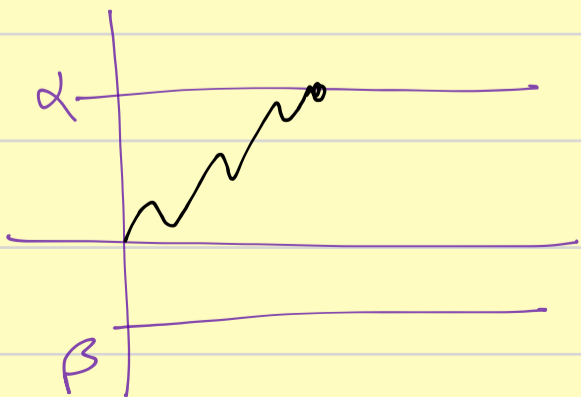
•  $E[X] < 0$

•  $\exists \delta > 0$

$$E[e^{\delta X}] = 1$$

Then  $P(S_J > \alpha) \leq \exp(-\delta \alpha)$

"exit  $(\alpha, \beta)$  at  $\alpha$ "



Proof :- Theorem 3.12 (Wald's Identity)  
lemma 3.17

$$\Rightarrow E[S_J^m] < \infty \quad \forall m \geq 1 ; E[e^{\delta S_J}] < \infty$$

$$\therefore E[e^{r S_J - \sum_{i=1}^J \eta(r)}] = 1$$

$$\Rightarrow \text{At } r = \delta ; E e^{\delta X} = 1 \Rightarrow \eta(\delta) = 0$$

$$\Rightarrow E[e^{\delta S_J}] = 1$$

$$\Rightarrow 1 = E[e^{\delta S_J} (S_J \geq \alpha)] + E[e^{\delta S_J} (S_J \leq \alpha)]$$

$$\geq P(S_J \geq \alpha) \underbrace{E(e^{\delta S_J} | S_J \geq \alpha)}_{\geq 1} + 0$$

$$\Rightarrow P(S_J \geq \alpha) \leq \frac{1}{E(e^{\delta S_J} | S_J \geq \alpha)}$$

$$\Rightarrow P(S_J \geq \alpha) \leq e^{-\delta \alpha} \quad \square$$