

$$A \in \{t-2k \mid k \in \mathbb{N}, k \leq t\}$$

(2) (a) For given $t \in \{1, 2, \dots, T\}$

① $A \in \{t-2k \mid 0 \leq k \leq t\}$ kGN

$$P(S_t = A) = \binom{t}{\frac{t-A}{2}} \left(\frac{1}{2}\right)^{\frac{t-A}{2}} \left(\frac{1}{2}\right)^{\frac{t+A}{2}} \cdot \binom{t}{\frac{t-A}{2}} \left(\frac{1}{2}\right)^{\frac{t-A}{2}} \left(\frac{1}{2}\right)^{\frac{t+A}{2}} \cdot \left(\frac{1}{2}\right)^t$$

$$= \binom{t}{\frac{t-A}{2}} \left(\frac{1}{2}\right)^{\frac{t-A}{2}} \left(\frac{1}{2}\right)^{\frac{t+A}{2}} = \binom{t}{\frac{t-A}{2}} \left(\frac{1}{2}\right)^t$$

② $A \notin \{t-2k \mid 0 \leq k \leq t\}$ kGN

$$P(S_t = A) = 0$$

(b)

① $A \in \{t-2k \mid 0 \leq k \leq t\}$ kGN

$$P(S_t = A, \min_{t' \leq t} S_{t'} \leq B) = P(S_t = A) - P(S_t = A, \min_{t' \leq t} S_{t'} > B)$$

since $\min_{t' \leq t} S_{t'} \leq S_t = A \leq B$, $P(S_t = A, \min_{t' \leq t} S_{t'} > B) = 0$

$$P(\min_{t' \leq t} S_{t'} \leq B \mid S_t = A) = 1$$

$$P(S_t = A, \min_{t' \leq t} S_{t'} \leq B) = P(\min_{t' \leq t} S_{t'} \leq B \mid S_t = A) P(S_t = A)$$

$$= P(S_t = A) = \binom{t}{\frac{t-A}{2}} \left(\frac{1}{2}\right)^t$$

② $A \notin \{t-2k \mid 0 \leq k \leq t\}$ kGN

$$P(S_t = A) = 0 \Rightarrow P(S_t = A, \min_{t' \leq t} S_{t'} \leq B) = 0$$

(c) Define $\hat{s}_t = \begin{cases} s_t & \text{if } t \leq t^* \\ 2B - s_t & \text{if } t > t^* \end{cases}$

where $t^* = \min \{t \mid s_t = B\}$

For $t > t^*$

$$\hat{s}_t = 2B - s_t = \overset{s_{t^*}}{B} + B - s_t = B + \sum_{i=t^*+1}^t \epsilon_i - \sum_{i=1}^{t^*} \epsilon_i$$

$$= \cancel{B} + \sum$$

$$= B - \sum_{i=t^*+1}^t \epsilon_i$$

$$= s_{t^*} - \sum_{i=t^*+1}^t \epsilon_i$$

while $s_t = s_{t^*} + \sum_{i=t^*+1}^t \epsilon_i$

Let ϵ_t^0 be the realization of ϵ_t . s_t^0 be the realization of s_t

~~$P(s_t = s_t^0 \mid s_{t^*} = s_{t^*}^0)$~~

$$P(s_t = s_t^0 \mid s_{t^*} = s_{t^*}^0) = \cancel{P(s_{t^*+1} = \dots = s_t = s_{t^*}^0 + \epsilon_{t^*+1} + \dots + \epsilon_t = s_t^0 - s_{t^*}^0)} \\ \left(\frac{t - t^*}{t - t^* - |s_t^0 - s_{t^*}^0|} \right) \left(\frac{1}{2} \right)^{\frac{t - t^* - |s_t^0 - s_{t^*}^0|}{2}} \left(\frac{1}{2} \right)^{\frac{t - t^* + (s_t^0 - s_{t^*}^0)}{2}}$$

$$P(\hat{s}_t = \hat{s}_t^0 \mid s_{t^*} = s_{t^*}^0) = \left(\frac{t - t^*}{t - t^* - |\hat{s}_t^0 - s_{t^*}^0|} \right) \left(\frac{1}{2} \right)^{\frac{t - t^* - |\hat{s}_t^0 - s_{t^*}^0|}{2}} \left(\frac{1}{2} \right)^{\frac{t - t^* + |\hat{s}_t^0 - s_{t^*}^0|}{2}}$$

$$= \left(\frac{t - t^*}{t - t^* - |s_t^0 - s_{t^*}^0|} \right) \left(\frac{1}{2} \right)^{\frac{t - t^* - |s_t^0 - s_{t^*}^0|}{2}} \left(\frac{1}{2} \right)^{\frac{t - t^* + |s_t^0 - s_{t^*}^0|}{2}}$$

$$= P(s_t = s_t^0 \mid s_{t^*} = s_{t^*}^0)$$

$$t - t^* + |s_t^0 - s_{t^*}^0|$$

if $B > 0$, then

$$\cancel{P(s_t = A)} P(s_t = A, \min_{t^* \leq t} s_t \leq B) \\ = P(s_t = A)$$

so we ~~only~~ consider $B \leq 0$

$$P(\hat{s}_t = s_t^* \mid s_{t^*} = s_{t^*}^*) = \begin{pmatrix} t - t^* \\ \frac{t - t^* - |s_t^* - s_{t^*}^*|}{2} \end{pmatrix} \left(\frac{1}{2}\right)^{\frac{t - t^* - |s_t^* - s_{t^*}^*|}{2}} \left(\frac{1}{2}\right)^{\frac{t - t^* + |s_t^* - s_{t^*}^*|}{2}}$$

$$= P(s_t = s_t^* \mid s_{t^*} = s_{t^*}^*)$$

Since $\hat{s}_t = s_t$ for $t \leq t^*$

We have { ① \hat{s}_t has symmetric distribution around B

② \hat{s}_t has the same distribution as s_t .

$$P(s_t = A, \min_{t' \leq t} s_{t'} \leq B) = P(\hat{s}_t = 2B - A)$$

$$= \begin{cases} \begin{pmatrix} t \\ \frac{t - |2B - A|}{2} \end{pmatrix} \left(\frac{1}{2}\right)^{\frac{t - |2B - A|}{2}} \left(\frac{1}{2}\right)^{\frac{t + |2B - A|}{2}} & \text{if } \frac{t - |2B - A|}{2} \in \{0, 1, \dots, t\} \\ 0 & \text{if } \frac{t - |2B - A|}{2} \notin \{0, 1, \dots, t\} \end{cases}$$

$$(d). P(\min_{t' \leq t} s_{t'} \leq B) = P(s_t > B, \min_{t' \leq t} s_{t'} \leq B) + P(s_t \leq B, \min_{t' \leq t} s_{t'} \leq B)$$

$$= P(\hat{s}_t < B) + P(s_t \leq B) \quad (*)$$

$$= P(s_t < B) + P(s_t \leq B)$$

$$= 2P(s_t \leq B) - \cancel{2P(s_t = B)} P(s_t = B)$$

(*) is because \hat{s}_t has the same distribution as s_t by definition
and $P(s_t \leq B, \min_{t' \leq t} s_{t'} \leq B) = P(s_t \leq B)$

(c) The distribution of S_t is that.

$$P(S_t = A) = \begin{cases} \binom{t}{\frac{t-A}{2}} \left(\frac{1}{2}\right)^{\frac{t-A}{2}} \left(\frac{1}{2}\right)^{\frac{t+A}{2}} = \left(\frac{1}{2}\right)^t \binom{t}{\frac{t-A}{2}} & A \in \{-t, -(t-1), \dots, 0, 1, \dots, t-1, t\} \\ 0 & A \notin \{-t, -(t-1), \dots, 0, 1, \dots, t-1, t\} \end{cases}$$

$$\begin{aligned} \frac{1}{\sqrt{t}} S_{\sqrt{t}} &= \frac{\sqrt{t}}{\sqrt{t}} \cdot \sqrt{t} \frac{1}{\sqrt{t}} S_{\sqrt{t}} \\ &= \frac{\sqrt{t}}{\sqrt{t}} \sqrt{t} \frac{1}{\sqrt{t}} (\xi_1 + \xi_2 + \dots + \xi_{\sqrt{t}}) \end{aligned}$$

ξ_i are i.i.d. $E(\xi_i) = 0$
 $\text{Var}(\xi_i) = 1 < \infty$, by CLT,

$$\begin{aligned} \sqrt{t} \frac{1}{\sqrt{t}} (\xi_1 + \xi_2 + \dots + \xi_{\sqrt{t}}) &\xrightarrow[t \rightarrow \infty]{} \mathcal{N}(0, 1) \\ \frac{\sqrt{t}}{\sqrt{t}} &\xrightarrow[t \rightarrow \infty]{} \sqrt{t} \end{aligned}$$

$$\Rightarrow \frac{1}{\sqrt{t}} S_{\sqrt{t}} \xrightarrow[t \rightarrow \infty]{} \sqrt{t} \mathcal{N}(0, 1) = \mathcal{N}(0, t)$$

(d) given sample $\{\xi_1, \xi_2, \dots, \xi_t\}$

$$\frac{1}{\sqrt{t}} S_{\sqrt{t}} \xrightarrow[t \rightarrow \infty]{} \mathcal{N}(0, t) = W(t)$$

$$\text{Let } I(t) = \min_{0 \leq t \leq T} g(t)$$

① f is continuous

Proof: $\forall \epsilon > 0$, $\exists \delta = \frac{\epsilon}{2}$, for $\forall g_1, g_2$, $d_{\infty}(g_1, g_2) \leq \delta$, $g_1, g_2 \in C[0, T]$ \Rightarrow

$$\text{Let } g_1(t_1^*) = \min_{0 \leq t \leq T} g_1(t) \quad g_2(t_2^*) = \min_{0 \leq t \leq T} g_2(t)$$

(b) suppose $g_2(t_2^*) < g_1(t_1^*) - \delta$, then

$$g_2(t_2^*) < g_1(t_1^*) - \delta \leq g_1(t_1^*) - \delta \Rightarrow g_1(t_1^*) - g_2(t_2^*) \geq \delta$$

Contradicts that $d_{\infty}(g_1, g_2) \leq \delta$

(2) Suppose $g_1(z_1^*) > g_1(z_1^*) + \delta$, then

$$g_1(z_1^*) \geq g_1(z_1^*) + \delta \Rightarrow g_1(z_1^*) - g_1(z_1^*) > \delta$$

that contradicts that $\sup_{z \in \mathcal{Z}} |g_1(z) - g_2(z)| \leq \delta$

$$\delta \|g_1(z_1^*) - g_1(z_1^*)\| \leq \delta \Rightarrow \left\| \min_{z \in \mathcal{Z}} g_1(z) - \min_{z \in \mathcal{Z}} g_2(z) \right\| \leq \delta = \frac{\epsilon}{2} < \epsilon.$$

$$\Rightarrow \|F(g_1(z)) - F(g_2(z))\| < \epsilon$$

③ F is bounded.

$$\|F(g)\| = \left| \min_{z \in \mathcal{Z}} g(z) \right| \leq \frac{\|g\|}{1} = \sup_{z \in \mathcal{Z}} |g(z)|$$

$$\frac{1}{\sqrt{T}} S_{[T]} \xrightarrow[T \rightarrow \infty]{} W(t), \text{ by FCLT}$$

$$\min_{z \in \mathcal{Z}} \frac{1}{\sqrt{T}} S_{[T]} \xrightarrow[T \rightarrow \infty]{} \min_{z \in \mathcal{Z}} W(t)$$

$$P\left(\min_{z \in \mathcal{Z}} \frac{1}{\sqrt{T}} S_{[T]} \in B\right) \xrightarrow[T \rightarrow \infty]{} P\left(\min_{z \in \mathcal{Z}} W(t) \in B\right)$$

$$P\left(\min_{z \in \mathcal{Z}} \frac{1}{\sqrt{T}} S_{[T]} \in B\right) = P\left(\min_{z \in \mathcal{Z}} S_{[T]} \in B\right)$$

$$= P\left(\min_{z \in \mathcal{Z}} S_{[T]} \in B\right)$$

$$= 2P\left(S_{[T]} \in B\right) - P\left(S_{[T]} = 0\right) = 2P\left(\frac{1}{\sqrt{T}} S_{[T]} \in B\right) - P\left(\frac{1}{\sqrt{T}} S_{[T]} = 0\right)$$

$$= 2 \sum_{A=-L \vee A}^{\sqrt{B}} \left(\frac{1}{2}\right)^{\sqrt{B}} \left(\frac{A}{\sqrt{B}}\right) - \left(\frac{1}{2}\right)^{\sqrt{B}} \left(\frac{A}{\sqrt{B}}\right)$$

$$= 2 \sum_{A=-L \vee A}^{\sqrt{B}} \left(\frac{1}{2}\right)^{\sqrt{B}} \left(\frac{\sqrt{B}}{2}\right) - \left(\frac{1}{2}\right)^{\sqrt{B}} \left(\frac{\sqrt{B}}{2}\right)$$

$$\xrightarrow[T \rightarrow \infty]{} 2 \int_{-\infty}^B \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt - \frac{1}{\sqrt{2\pi}} e^{-\frac{B^2}{2}}$$

$S_n \approx A$ $\min (1, n) = n/2$

$$2P(S_n \leq b) - P(S_n = b) = 2P\left(\frac{1}{\sqrt{n}} S_n \leq b\right) - P\left(\frac{1}{\sqrt{n}} S_n = b\right).$$

$$\xrightarrow{T \rightarrow \infty} 2P(W(r) \leq b) - P(W(r) = b) \quad (W(r) \sim \sqrt{2r})$$

$$= 2 \cdot \int_{-\infty}^b \frac{1}{\sqrt{2\pi r}} e^{-\frac{x^2}{2r}} dx - 0.$$

$$P\left(\min_{0 \leq t \leq r} W(t) \leq b\right) = 2 \int_{-\infty}^b \frac{1}{\sqrt{2\pi r}} e^{-\frac{x^2}{2r}} dx$$

1