

## Problem 1

Assume the first customer arrives after  $T - s$  arrives at  $T - s + t_1$ , the second one arrives at  $T - s + t_1 + t_2$ . Then Joe achieves his goal iff  $t_1 < s$  and  $t_1 + t_2 > s$ .

$$\begin{aligned} f(s) &= \Pr [t_1 < s \wedge t_1 + t_2 > s] = \int_0^s \lambda e^{-\lambda t_1} \Pr [t_1 + t_2 > s] dt_1 \\ &= \int_0^s \lambda e^{-\lambda t_1} \Pr [t_2 > s - t_1] dt_1 \\ &= \int_0^s \lambda e^{-\lambda t_1} e^{-\lambda(s-t_1)} dt_1 \\ &= \lambda \int_0^s e^{-\lambda s} dt_1 \\ &= \lambda s e^{-\lambda s}. \end{aligned}$$

$$f'(s) = \lambda e^{-\lambda s}(1 - \lambda s) = 0 \Rightarrow s = \frac{1}{\lambda}, \text{Max Probability} = f(s) = \frac{1}{e}.$$

## Problem 2

1.

$$\begin{aligned} \Pr [X = \lambda + k] &= \frac{\lambda^{\lambda+k}}{(\lambda+k)!} e^{-\lambda}, \\ \Pr [X = \lambda - k - 1] &= \frac{\lambda^{\lambda-k-1}}{(\lambda-k-1)!} e^{-\lambda}. \end{aligned}$$

$$\begin{aligned} \frac{\Pr [X = \lambda + k]}{\Pr [X = \lambda - k - 1]} &= \frac{(\lambda - k - 1)!}{(\lambda + k)!} \lambda^{2k+1} \\ &= \frac{1}{\lambda + k} \frac{1}{\lambda + k - 1} \cdots \frac{1}{\lambda - k} \lambda^{2k+1} \\ &= \frac{\lambda^2}{\lambda^2 - k^2} \frac{\lambda^2}{\lambda^2 - (k-1)^2} \cdots \frac{\lambda^2}{\lambda^2 - 1^2} \geq 1 (= 1 \text{ only when } k = 0). \end{aligned}$$

$$\begin{aligned} 1 &= \sum_{x=0}^{\infty} \Pr [X = x] = \sum_{x=0}^{\lambda-1} \Pr [X = x] + \Pr [X \geq \lambda] \\ &\leq \sum_{x=\lambda}^{2\lambda-1} \Pr [X = x] + \Pr [X \geq \lambda] \\ &\leq \sum_{x=\lambda}^{\infty} \Pr [X = x] + \Pr [X \geq \lambda] = 2\Pr [X \geq \lambda]. \end{aligned}$$

2. Note that  $\sum_{i=1}^n Y_i \sim \text{Pois}(m)$ , so by 1., we have  $\Pr [\sum_{i=1}^n Y_i \geq m] \geq \frac{1}{2}$ .

$$\begin{aligned}
\mathbf{E} [f(Y_1, \dots, Y_n)] &= \sum_{k=0}^{\infty} \mathbf{E} \left[ f(Y_1, \dots, Y_n) \mid \sum_{i=1}^n Y_i = k \right] \Pr \left[ \sum_{i=1}^n Y_i = k \right] \\
&\geq \sum_{k=m}^{\infty} \mathbf{E} \left[ f(X_1, \dots, X_n) \mid \sum_{i=1}^n X_i = k \right] \Pr \left[ \sum_{i=1}^n Y_i = k \right] \\
&\geq \sum_{k=m}^{\infty} \mathbf{E} \left[ f(X_1, \dots, X_n) \mid \sum_{i=1}^n X_i = m \right] \Pr \left[ \sum_{i=1}^n Y_i = k \right] \\
&= \mathbf{E} [f(X_1, \dots, X_n)] \sum_{k=m}^{\infty} \Pr \left[ \sum_{i=1}^n Y_i = k \right] \\
&= \mathbf{E} [f(X_1, \dots, X_n)] \Pr \left[ \sum_{i=1}^n Y_i \geq m \right] \\
&\geq \frac{1}{2} \mathbf{E} [f(X_1, \dots, X_n)].
\end{aligned}$$

3. Assume  $X_i$  be the variables counting the number of students whose birthday is on the  $i$ 's day of a year ( $i = 1, 2, \dots, 365$ ). Then we have  $\sum_{i=1}^{365} X_i = 105$ , and for each  $X_i$ ,  $X_i \sim \text{Binom}(105, \frac{1}{365})$ . Define  $f(X_1, X_2, \dots, X_{365}) \triangleq \mathbf{1}[\exists X_i, X_i \geq 5]$ , and then  $\mathbf{E} [f(X_1, X_2, \dots, X_{365})] = \Pr [\exists X_i, X_i \geq 5]$ . Define  $Y_i \sim \text{Pois}(\frac{105}{365})$  with the condition that  $\sum_{i=1}^{365} Y_i = 105$ . Similarly, let  $f(Y_1, Y_2, \dots, Y_{365}) \triangleq \mathbf{1}[\exists Y_i, Y_i \geq 5]$ , and then  $\mathbf{E} [f(Y_1, Y_2, \dots, Y_{365})] = \Pr [\exists Y_i, Y_i \geq 5]$ . We have

$$\begin{aligned}
\Pr [\exists Y_i, Y_i \geq 5] &= 1 - \Pr [\forall Y_i, Y_i \leq 4] \\
&= 1 - \prod_{i=1}^{365} \Pr [Y_i \leq 4] \\
&= 1 - \left( e^{-\frac{105}{365}} \left( \frac{(\frac{105}{365})^0}{0!} + \frac{\frac{105}{365}}{1!} + \frac{(\frac{105}{365})^2}{2!} + \frac{(\frac{105}{365})^3}{3!} + \frac{(\frac{105}{365})^4}{4!} \right) \right)^{365} \\
&\approx 0.4707853\% < 0.5\%.
\end{aligned}$$

It is trivial that  $\mathbf{E} [f(X_1, \dots, X_{365})]$  is monotonically increasing in  $m$ . Thus, by the conclusion in 2.,  $\mathbf{E} [f(X_1, \dots, X_{365})] \leq 2\mathbf{E} [f(Y_1, \dots, Y_{365})] < 1\%$ .

## Problem 3

(a) We have:

1. If  $t = 0$ ,  $c^{-\frac{1}{2}}W(ct) = 0$ .

2. It holds the property called **independent increments** obviously.

3. **Stationary increments:**

$$\forall s, t > 0, c^{-\frac{1}{2}}W(c(s+t)) - c^{-\frac{1}{2}}W(cs) = c^{-\frac{1}{2}}(W(c(s+t)) - W(cs)) \sim \mathcal{N}(0, (c^{-\frac{1}{2}})^2 ct).$$

4.  $c^{-\frac{1}{2}}W(ct)$  is continuous almost surely.

Therefore,  $c^{-\frac{1}{2}}W(ct)$  is also a standard Brownian motion.

(b) We have:

1. If  $t = 0$ ,  $X(t) = W(c+t) - W(c) = 0$ .

2. **Independent increments:**  $\forall 0 \leq t_1 \leq t_2 \leq t_3 \leq t_4$ ,

$$X(t_1) - X(t_0) = W(c+t_1) - W(c+t_0), X(t_3) - X(t_2) = W(c+t_3) - W(c+t_2),$$

$\text{cov}(X(t_1) - X(t_0), X(t_3) - X(t_2)) = \text{cov}(W(c+t_1) - W(c+t_0), W(c+t_3) - W(c+t_2))$ , then it's clear that  $\text{cov}(W(c+t_1) - W(c+t_0), W(c+t_3) - W(c+t_2)) = 0$  if taking the

covariance apart.

### 3. Stationary increments:

$$\forall s, t > 0, X(s+t) - X(s) = W(c+s+t) - W(c+s) \sim \mathcal{N}(0, t).$$

4.  $X(t)$  is continuous almost surely.

Therefore,  $X(t)$  is also a standard Brownian motion.

Because  $\forall s \geq 0, 0 \leq t \leq c$ ,

$$\begin{aligned} \text{cov}(X(s), W(t)) &= \text{cov}(W(c+s) - W(c), W(t)) \\ &= \text{cov}(W(c+s), W(t)) + \text{cov}(W(c), W(t)) = 0, \end{aligned}$$

$\{X(t) : t \geq 0\}$  is independent of  $\{W(t) : 0 \leq t \leq c\}$ .

(c) Let  $c = \frac{1}{2}$ , we have  $\mathbf{Pr}[W(1) > 0 | W(1/2) > 0] = \mathbf{Pr}[W(1) < 2W(1/2) | W(1/2) > 0]$

$$\begin{aligned} 2\mathbf{Pr}[W(1) > 0 | W(1/2) > 0] &= \mathbf{Pr}[W(1) > 0 | W(1/2) > 0] + \mathbf{Pr}[W(1) < 2W(1/2) | W(1/2) > 0] \\ &= 1 + \mathbf{Pr}[0 < W(1) < 2W(1/2) | W(1/2) > 0] \\ &= 1 + \mathbf{Pr}[0 < W(1) < 2W(1/2)] \\ &= 1 + \mathbf{Pr}[0 < W(1/2) + X(1/2) < 2W(1/2)] \\ &= 1 + \mathbf{Pr}[-W(1/2) < X(1/2) < W(1/2)] \\ &= 1 + \mathbf{Pr}\left[-\sqrt{2}W(1/2) < \sqrt{2}X(1/2) < \sqrt{2}W(1/2)\right] \\ &= 1 + \Phi(\sqrt{2}W(1/2)) - \Phi(-\sqrt{2}W(1/2)) \text{ (by (b))} \\ &= 1 + \Phi(\sqrt{2}W(1/2)) - (1 - \Phi(\sqrt{2}W(1/2))) \\ &= 2\Phi(\sqrt{2}W(1/2)) \end{aligned}$$

Thus,  $\mathbf{Pr}[W(1) > 0 | W(1/2) > 0] = \Phi(\sqrt{2}W(1/2))$ .

## Problem 4

(a)  $W(t) \sim \mathcal{N}(0, t)$ ,  $\frac{W(t)}{\sqrt{t}} \sim \mathcal{N}(0, 1)$ , and  $\xi \sim \mathcal{N}(0, 1)$ .

$$\begin{aligned} \mathbf{Pr}[X(t) \leq \delta] &= \mathbf{Pr}[\mu t + \sigma W(t) \leq \delta] \\ &= \mathbf{Pr}\left[\frac{W(t)}{\sqrt{t}} \leq \frac{\delta - \mu \cdot t}{\sigma \sqrt{t}}\right] \\ &= \mathbf{Pr}\left[\xi \leq \frac{\delta - \mu \cdot t}{\sigma \sqrt{t}}\right]. \end{aligned}$$

(b)

$$\begin{aligned} \mathbf{E}[T] &= \mathbf{E}\left[\int_0^\infty \mathbb{1}[X(t) \in [0, \delta]] dt\right] \\ &= \int_0^\infty \mathbf{Pr}[X(t) \in [0, \delta]] dt \\ &= \int_0^\infty \mathbf{Pr}\left[-\frac{\mu\sqrt{t}}{\sigma} \leq \frac{W(t)}{\sqrt{t}} \leq \frac{-\mu t + \delta}{\sigma\sqrt{t}}\right] dt \\ &= \int_0^\infty \mathbf{Pr}\left[-\frac{\mu\sqrt{t}}{\sigma} \leq \xi \leq \frac{-\mu t + \delta}{\sigma\sqrt{t}}\right] dt \\ &= \int_0^\infty \mathbf{Pr}\left[\frac{\mu t - \delta}{\sigma\sqrt{t}} \leq \xi \leq \frac{\mu\sqrt{t}}{\sigma}\right] dt. \end{aligned}$$

(c) Solve the equation  $\frac{\mu t - \delta}{\sigma\sqrt{t}} = \xi$ . We can get  $f(\delta, \xi) = t = \frac{\xi^2\sigma^2 + 2\mu\delta + \xi\sigma\sqrt{\xi^2\sigma^2 + 4\mu\delta}}{2\mu^2}$ .

(d)

$$\mathbf{E}[f(\delta, \xi)] = \frac{\mathbf{E}[\xi^2\sigma^2 + 2\mu\delta + \xi\sigma\sqrt{\xi^2\sigma^2 + 4\mu\delta}]}{2\mu^2} = \frac{\sigma^2\mathbf{E}[\xi^2] + 2\mu\delta}{2\mu^2} = \frac{\sigma^2\mathbf{Var}[\xi] + 2\mu\delta}{2\mu^2} = \frac{\sigma^2 + 2\mu\delta}{2\mu^2}.$$

(e)

$$\begin{aligned}\mathbf{E}[T] &= \int_0^\infty \mathbf{Pr}\left[\frac{\mu t - \delta}{\sigma\sqrt{t}} \leq \xi \leq \frac{\mu\sqrt{t}}{\sigma}\right] dt \\ &= \int_0^\infty \mathbf{Pr}[f(0, \xi) \leq t \leq f(\delta, \xi)] dt \\ &= \mathbf{E}[f(\delta, \xi) - f(0, \xi)] \\ &= \mathbf{E}[f(\delta, \xi)] - \mathbf{E}[f(0, \xi)] \\ &= \frac{\delta}{\mu}.\end{aligned}$$