

①

$$\begin{aligned}
 \text{D) a) } P(A \cup B) &= P(A) + P(B) - P(A \cap B) \quad \textcircled{*} \\
 &= P(A) + P(B) - P(A|B)P(B) \\
 &= P(A) + P(B)(1 - P(A|B)) \\
 &= P(A) + P(B)P(\bar{A}|B) \\
 \underline{P(A \cup B)} &= \underline{P(A) + P(\bar{A} \cap B)} \quad \text{as required.}
 \end{aligned}$$

Alternative: note that $P(B) = P(A \cap B) + P(\bar{A} \cap B)$ by Partition Rule,
 So $P(B) - P(A \cap B) = P(\bar{A} \cap B)$.

Thus by $\textcircled{*}$, $\underline{P(A \cup B) = P(A) + P(\bar{A} \cap B)}$.

$$\begin{aligned}
 \text{b) } P(A \cup B) &= P(A) + P(\bar{A} \cap B) \quad \text{from (a).} \quad \textcircled{**} \\
 \text{By the Partition Rule, } P(\bar{A} \cap B) &+ P(\bar{A} \cap \bar{B}) = P(\bar{A}). \\
 \text{So } P(\bar{A} \cap B) &= P(\bar{A}) - P(\bar{A} \cap \bar{B}).
 \end{aligned}$$

Substituting into $\textcircled{*}$: $P(A \cup B) = P(A) + P(\bar{A}) - P(\bar{A} \cap \bar{B})$
 $\therefore \underline{P(A \cup B) = 1 - P(\bar{A} \cap \bar{B})}$ as stated.

2) $X \sim \text{Uniform}(0, 5)$, so p.d.f. of X is $f_X(x) = \frac{1}{5}$ ($0 < x < 5$),
 Given information: $P(\text{see ball} | \text{ball is at } x) = e^{-x}$.

$$\begin{aligned}
 \text{Thus } P(\text{see ball}) &= E_X \{ P(\text{see ball} | X) \} \quad \left(\begin{array}{l} \text{Probability as a} \\ \text{Conditional} \\ \text{Expectation;} \\ \text{Section 3.8} \end{array} \right) \\
 &= \int_0^5 P(\text{see ball} | X=x) f_X(x) dx \\
 &= \int_0^5 e^{-x} \cdot \frac{1}{5} dx
 \end{aligned}$$

②

$$\begin{aligned}
 \text{2 cont.) } P(\text{see ball}) &= \frac{1}{5} [-e^{-x}]_0^5 \\
 \underline{P(\text{see ball})} &= \underline{\frac{1}{5} (1 - e^{-5})} \quad (0.199).
 \end{aligned}$$

$$\begin{aligned}
 \text{3a) } G_Y(s) &= E(s^Y) \\
 &= E(s^{aX+b}) \quad \text{as } Y = aX + b \\
 &= E(s^{aX} s^b) \\
 &= s^b E(s^{aX}) \quad (s^b \text{ is a constant}) \\
 &= s^b E((s^a)^X) \\
 \underline{G_Y(s)} &= \underline{s^b G_X(s^a)} \quad \text{as stated.}
 \end{aligned}$$

b) $X \sim \text{Binomial}(5, 0.4)$, so $P(X=x) = \binom{5}{x} 0.4^x 0.6^{5-x}$
 for $x=0, \dots, 5$.

$$\begin{aligned}
 \text{Thus: } G_X(s) &= E(s^X) = \sum_{x=0}^5 s^x P(X=x) \\
 &= \sum_{x=0}^5 \binom{5}{x} (0.4)^x (0.6)^{5-x} \cdot s^x \\
 &= \sum_{x=0}^5 \binom{5}{x} (0.4s)^x (0.6)^{5-x} \\
 &= \underline{(0.4s + 0.6)^5} \quad \text{by the Binomial Theorem} \\
 &\quad \text{(see Attachment).}
 \end{aligned}$$

The working is valid for all $s \in \mathbb{R}$,
 so radius of convergence = ∞ .

c) $Y = 2X + 1 \Rightarrow G_Y(s) = s^1 G_X(s^2)$ by part (a):
 $a=2, b=1$.

So $\underline{G_Y(s) = s \{0.4s^2 + 0.6\}^5}$ for $s \in \mathbb{R}$.

4) a) $Y \sim \text{Binomial}(2, \frac{3}{4})$, so $G(s) = (\frac{3}{4}s + \frac{1}{4})^2$.

b) $G_2(s) = G(G(s))$

$= (\frac{3}{4}G(s) + \frac{1}{4})^2$

$\Rightarrow G_2(s) = \left\{ \frac{3}{4}(\frac{3}{4}s + \frac{1}{4})^2 + \frac{1}{4} \right\}^2$ *

c) $P(Z_4=0) = G_4(0)$

$= G_2(G_2(0))$ (or $G(G(G(G(0))))$)

$= G_2(0.088)$ by substitution in *

$= \underline{\underline{0.106}}$ as stated.

d) γ is the minimal non-negative solution to the equation

$G(s) = s$
 $\Rightarrow (\frac{3}{4}s + \frac{1}{4})^2 = s$

$\Rightarrow \frac{9}{16}s^2 + \frac{6}{16}s + \frac{1}{16} = s$

$\Rightarrow 9s^2 - 10s + 1 = 0$

$(s-1)(9s-1) = 0$ (know $s=1$ is a solution)

$\Rightarrow \underline{\underline{s=1, s=1/9}}$

So γ is the smaller solution, $\Rightarrow \underline{\underline{P(\text{extinction}) = \gamma = 1/9}}$

e) If $Z_6=8$, we have 8 independent branching processes beginning at time 6.

$P(\text{extinction}) = P(\text{all 8 become extinct}) = \gamma^8 = (\frac{1}{9})^8 = 2.3 \times 10^{-8}$

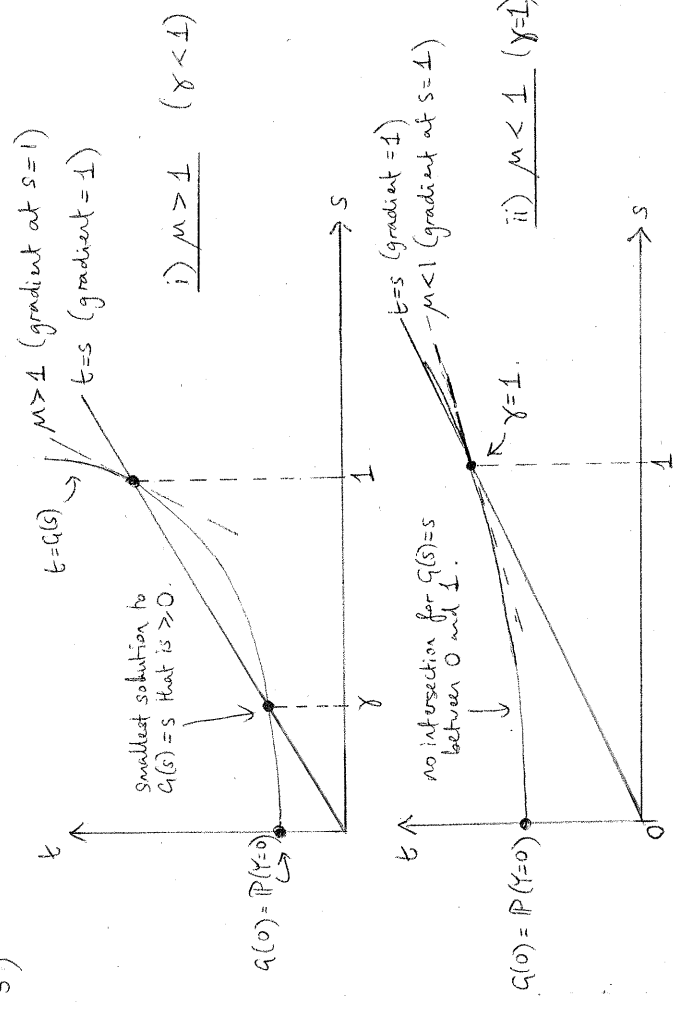
4f) We have 8-independent branching processes, beginning in generation 6.

For each process, the # individuals in generation 10 has the same distribution as Z_4 (starting from 1 individual at $t=0$).

So $P(\text{descendants still alive at } t=10) = 1 - P(Z_4=0)$
 $= 1 - 0.106$
 $= \underline{\underline{0.894}}$ for each strain.

Thus $P(5 \text{ out of } 8 \text{ still alive}) = \binom{8}{5} (0.894)^5 (0.106)^3$
 $= \underline{\underline{0.038}}$

5)



γ is the minimal solution ≥ 0 to the equation $G(s)=s$.
 μ is the gradient of $G(s)$ at $s=1$ ($\mu = G'(1)$).
 The graph $G(s)$ cuts the t-axis at $P(Y=0)$, which is ≥ 0 .
 Thus, when $\mu > 1$, the curve $G(s)$ is forced below the line $t=s$ at $s=1$, so (by convex shape) must cross the line $t=s$ again to meet the t-axis, giving an intersection at $\gamma < 1$. When $\mu < 1$, no such intersection is possible.

5

a) $X_0 \sim \text{Bin}(2, 0.4)$ (sample space is $\{1, 2, 3\}$ not $\{0, 1, 2, 3\}$)

So $P(X_0=1) = \binom{2}{0} (0.4)^0 (0.6)^2 = 0.36$ [uses Bin. formula for 0]
 $P(X_0=2) = \binom{2}{1} (0.4)(0.6) = 0.48$
 $P(X_0=3) = \binom{2}{2} (0.4)^2 (0.6)^0 = 0.16$

Thus $X_0 \sim (0.36, 0.48, 0.16)^T P$

So $X_1 \sim (0.36, 0.48, 0.16)^T P$
 $= (0.36 \ 0.48 \ 0.16) \begin{pmatrix} 0.6 & 0.2 & 0.2 \\ 0.4 & 0 & 0.6 \\ 0 & 0.8 & 0.2 \end{pmatrix}$
 $X_1 \sim (0.408, 0.2, 0.392)$

b) $P(X_1=1, X_2=2, X_3=3 | X_0=2)$
 $= P(X_1=1 | X_0=2) P(X_2=2 | X_1=1) P(X_3=3 | X_2=2)$
 $= p_{21} \cdot p_{12} \cdot p_{23}$
 $= 0.4 * 0.2 * 0.6$
 $= 0.048$

c) $P(X_t=3 | X_0=1) = (P^t)_{13}$

Now if P has distinct eigenvalues $\lambda_1=1, \lambda_2, \lambda_3$, then we can find a general solution

$(P^t)_{13} = c_1 + c_2 \lambda_2^t + c_3 \lambda_3^t$,

where constants c_1, c_2, c_3 are to be determined.

6

bc cont.) Find the eigenvalues of P :

$\det(P - \lambda I) = \begin{vmatrix} 0.6-\lambda & 0.2 & 0.2 \\ 0.4 & -\lambda & 0.6 \\ 0 & 0.8 & 0.2-\lambda \end{vmatrix} = 0$

$\Rightarrow (0.6-\lambda) \{-\lambda(0.2-\lambda) - 0.48\}$

$- 0.2(0.4(0.2-\lambda)) + 0.2(0.32) = 0$

$-\lambda(0.2-\lambda)(0.6-\lambda) - 0.48(0.6-\lambda)$

$- 0.016 + 0.08\lambda + 0.064 = 0$

$-\lambda(0.12 - 0.8\lambda + \lambda^2) - 0.24 + 0.56\lambda = 0$

$-\lambda^3 + 0.8\lambda^2 + 0.44\lambda - 0.24 = 0$

$\Rightarrow \lambda^3 - 0.8\lambda^2 - 0.44\lambda + 0.24 = 0$

$(\lambda-1)(\lambda^2 + 0.2\lambda - 0.24) = 0$

$\lambda=1$, or $\lambda = \frac{-0.2 \pm \sqrt{(0.2)^2 + 4(0.24)}}{2}$
 must be a factor

$\Rightarrow \lambda = 0.4, \lambda = -0.6$

Thus the matrix has distinct eigenvalues, $\lambda_1=1, \lambda_2=0.4, \lambda_3=-0.6$

General solution:

$(P^t)_{13} = c_1 + c_2(0.4)^t + c_3(-0.6)^t$

Initial conditions:

$(P^0)_{13} = 0 \Rightarrow c_1 + c_2 + c_3 = 0$

$(P^1)_{13} = 0.2 \Rightarrow c_1 + 0.4c_2 - 0.6c_3 = 0.2$

7

6c cont.) For $t=2$, we need to find $(P^2)_{13}$:

$$P^2 = \begin{pmatrix} 0.6 & 0.2 & 0.2 \\ \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots \end{pmatrix} \begin{pmatrix} 0.2 \\ 0.6 \\ 0.2 \end{pmatrix} = \begin{pmatrix} \dots & 0.28 \\ \vdots & \vdots \\ \vdots & \vdots \end{pmatrix}$$

$$\text{So } (t=2) \quad (P^2)_{13} = 0.28 = c_1 + (0.4)^2 c_2 + (-0.6)^2 c_3$$

The initial equations are therefore:

$$c_1 + c_2 + c_3 = 0 \quad (1)$$

$$c_1 + 0.4c_2 - 0.6c_3 = 0.2 \quad (2)$$

$$c_1 + 0.16c_2 + 0.36c_3 = 0.28 \quad (3)$$

Using $c_1 = -(c_2 + c_3)$ from (1) in (2), (3) \Rightarrow

$$(2) \quad -c_2 - c_3 + 0.4c_2 - 0.6c_3 = 0.2$$

$$\Rightarrow -0.6c_2 - 1.6c_3 = 0.2 \quad (2^*)$$

$$(3) \quad -c_2 - c_3 + 0.16c_2 + 0.36c_3 = 0.28$$

$$\Rightarrow -0.84c_2 - 0.64c_3 = 0.28 \quad (3^*)$$

$$(2^*) - 2.5^*(3^*) \Rightarrow 1.5c_2 = -0.5 \Rightarrow c_2 = \underline{\underline{-\frac{1}{3}}}$$

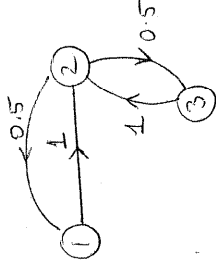
$$\text{In } (2^*) \Rightarrow c_3 = \underline{\underline{0}}$$

$$\text{Thus } c_1 = \frac{1}{3}, \quad c_2 = -\frac{1}{3}, \quad c_3 = 0.$$

The general solution is therefore:

$$(P^t)_{13} = \underline{\underline{\frac{1}{3} - \frac{1}{3}(0.4)^t}} \quad \text{for } t=0, 1, 2, \dots$$

7) a)



b) Equilibrium distribution π satisfies $\pi^T P = \pi^T$

$$\text{and } \sum_{i=1}^3 \pi_i = 1.$$

$$\text{Now } (\pi_1 \quad \pi_2 \quad \pi_3) \begin{pmatrix} 0 & 1 & 0 \\ 0.5 & 0 & 0.5 \\ 0 & 1 & 0 \end{pmatrix} = (\pi_1 \quad \pi_2 \quad \pi_3)$$

$$\Rightarrow 0.5 \pi_2 = \pi_1 \quad (1)$$

$$\pi_1 + \pi_3 = \pi_2 \quad (2)$$

$$0.5 \pi_2 = \pi_3 \quad (3)$$

$$(1), (3) \Rightarrow \pi_1 = \pi_3, \text{ so } (2) \Rightarrow \pi_2 = 2\pi_1.$$

$$\text{Then } \pi_1 + \pi_2 + \pi_3 = 1 \Rightarrow \pi_1 + 2\pi_1 + \pi_1 = 1$$

$$\Rightarrow \pi_1 = \underline{\underline{\frac{1}{4}}}.$$

So $\pi^T = \left(\frac{1}{4}, \frac{1}{2}, \frac{1}{4} \right)$ is an equilibrium distribution for P .

c) The matrix P is irreducible, but it is not aperiodic (period of all states is 2).

Thus the chain does not converge to the equilibrium distⁿ π .

8

8) a) Consider $\underline{h} = (h_0, \dots, h_{10})$, where h_k is the probability of hitting state 10, given that gambler starts at state k .

The hitting probabilities are the minimal non-negative solutions to the equations:

$$h_{10} = 1$$

$$h_k = \sum_{j=0}^{10} p_{kj} h_j \quad \text{for } k \neq 10.$$

From the diagram, state 0 is an absorbing state, so $h_0 = 0$.

Otherwise,
$$h_k = \frac{1}{2} h_{k-1} + \frac{1}{2} h_{k+1} \quad (k=1, 2, \dots, 9)$$

So
$$h_1 = \frac{1}{2} h_0 + \frac{1}{2} h_2 = \frac{1}{2} * 0 + \frac{1}{2} h_2$$

$$\Rightarrow h_1 = \frac{1}{2} h_2$$

Also,
$$h_2 = \frac{1}{2} h_1 + \frac{1}{2} h_3 = \frac{1}{2} \cdot \frac{1}{2} h_2 + \frac{1}{2} h_3$$

$$\Rightarrow \frac{3}{4} h_2 = \frac{1}{2} h_3$$

$$\Rightarrow h_2 = \frac{2}{3} h_3$$

The hypothesis that $h_k = \frac{k}{k+1} h_{k+1}$ is true for $k=1, 2$.

Prove by induction.

Suppose it is true up to $k-1$:

Then
$$h_k = \frac{1}{2} h_{k-1} + \frac{1}{2} h_{k+1}$$

$$= \frac{1}{2} \cdot \frac{k-1}{k} h_k + \frac{1}{2} h_{k+1}$$

$$\Rightarrow h_k \left(1 - \frac{k-1}{2k}\right) = \frac{1}{2} h_{k+1}$$

$$h_k \frac{(k+1)}{2k} = \frac{1}{2} h_{k+1}$$

$$\Rightarrow \frac{h_k}{\left(\frac{k}{k+1}\right) h_{k+1}}$$

So the hypothesis is true for k \Rightarrow proof by induction.

8 a cont) Thus $h_0 = 0, h_{10} = 1$,

and
$$h_k = \left(\frac{k}{k+1}\right) h_{k+1} \quad \text{for } k=1, \dots, 9.$$

b) We have
$$h_{10} = 1$$

$$h_9 = \frac{9}{10} h_{10} = \frac{9}{10}$$

$$h_8 = \frac{8}{9} \cdot \frac{9}{10} = \frac{8}{10}$$

$$h_7 = \frac{7}{8} \cdot \frac{8}{10} = \frac{7}{10}$$

...

Clearly,
$$h_1 = \frac{1}{10} \quad \text{i.e. } P(\text{win } \$10, \text{ starting at } \$1) = \frac{1}{10}.$$

c) The gambler might never reach \$10, starting from \$1.

Thus,
$$E(T_1) = \infty.$$

9) X_t = maximum score after t tosses $\Rightarrow X_t$ can be $1, 2, \dots, 6$.

Suppose $X_t = 1$ (i.e. maximum score is currently 1):
Then the next toss gives maximum $1, 2, \dots, 6$ with equal prob,

i.e.
$$P(X_{t+1} = k | X_t = 1) = \frac{1}{6} \quad \text{for } k=1, \dots, 6.$$

Suppose $X_t = 2$:

the maximum on the next toss will remain 2 if the next toss is 1 or 2, (prob. $\frac{2}{6}$), otherwise it is 3, 4, 5, or 6 each with probability $\frac{1}{6}$.

Suppose $X_t = k$:

The maximum on the next toss is k with prob. $\frac{k}{6}$ (if 1, 2, ..., k is thrown), otherwise it is $k+1, k+2, \dots, 6$ each with prob. $\frac{1}{6}$.

9 cont.) Thus the transition matrix is:

$$\begin{matrix}
 & \begin{matrix} 1 & 2 & 3 & 4 & 5 & 6 \end{matrix} \\
 \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \end{matrix} & \begin{pmatrix}
 1/6 & 1/6 & 1/6 & 1/6 & 1/6 & 1/6 \\
 0 & 2/6 & 1/6 & 1/6 & 1/6 & 1/6 \\
 0 & 0 & 3/6 & 1/6 & 1/6 & 1/6 \\
 0 & 0 & 0 & 4/6 & 1/6 & 1/6 \\
 0 & 0 & 0 & 0 & 5/6 & 1/6 \\
 0 & 0 & 0 & 0 & 0 & 1
 \end{pmatrix}
 \end{matrix}$$

10) Possibilities:

Person on Floor 1	Person on Floor 2	X (first stop)	Probability
going \rightarrow 2 (prob $\frac{1}{3}$)	\rightarrow 3 (prob $\frac{1}{6}$)	2	$\frac{1}{3} * \frac{1}{6} = \frac{1}{18}$
\rightarrow 2 ($\sim \frac{1}{3}$)	\rightarrow 1 ($\sim \frac{7}{8}$)	2	$\frac{1}{3} * \frac{7}{8} = \frac{7}{24}$
\rightarrow 3 ($\sim \frac{2}{3}$)	\rightarrow 3 ($\sim \frac{1}{8}$)	2	$\frac{2}{3} * \frac{1}{8} = \frac{2}{24}$
\rightarrow 3 ($\sim \frac{2}{3}$)	\rightarrow 1 ($\sim \frac{7}{8}$)	3	$\frac{2}{3} * \frac{7}{8} = \frac{14}{24}$

Thus X has the following distribution:

x	2	3
P(X=x)	$\frac{10}{24}$	$\frac{14}{24}$