

Nov 8th

Law of Large Numbers → if you repeat an experiment independently a large number of times, we obtain something to the expected value

↳ two versions — weak law of large numbers (WLLN)
— strong " " " " (SLLN)

↳ difference is mostly theoretical

Let: $X_i \ i=1, \dots, n$ i.i.d. (independent, identical distributions)

$$\bar{X} = \frac{X_1 + X_2 + \dots + X_n}{n} \quad (\text{Sample mean})$$

WLLN Let $X_i \ i=1, \dots, n$ be i.i.d. random variables ($\mu < \infty$, finite mean). Then for any $\epsilon > 0$,

$$\lim_{n \rightarrow \infty} P(|\bar{X} - \mu| \geq \epsilon) = 0$$

the prob that the difference btwn mean of the sample (\bar{X}) and the expected value of r.v. (μ) tends to zero

Proof assume $\text{Var}(X_i) = \sigma^2$ finite

$$P(|\bar{X} - \mu| \geq \epsilon) \leq \frac{\text{Var}(\bar{X})}{\epsilon^2} \quad (\text{Chebyshov})$$

$$= \frac{\text{Var}(X)}{n \cdot \underline{\epsilon^2}} \quad \#$$

as $n \rightarrow \infty$: $\lim_{n \rightarrow \infty} P(|\bar{X} - \mu| \geq \epsilon) = 0$



Where does $\text{Var}(\bar{X}) = \frac{\text{Var}(X)}{n}$ come from?

$$\begin{aligned} E[\bar{X}] &= E\left[\frac{X_1 + \dots + X_n}{n}\right] = \underbrace{E[X_1]}_n + \underbrace{E[X_2]}_n + \dots + \underbrace{E[X_n]}_n && \text{linearity} \\ &= \underbrace{E[X] + E[X] + \dots + E[X]}_{n \text{ times}} && \text{i.i.d} \\ &= \frac{n \cdot E[X]}{n} = E[X] \end{aligned}$$

$$\begin{aligned} \text{Var}(\bar{X}) &= \text{Var}\left(\frac{X_1 + \dots + X_n}{n}\right) = \frac{\text{Var}(X_1 + \dots + X_n)}{n^2} \\ &= \frac{\text{Var}(X_1) + \text{Var}(X_2) + \dots + \text{Var}(X_n)}{n^2} \\ &\simeq \frac{\text{Var}(X) + \text{Var}(X) + \dots + \text{Var}(X)}{n^2} \\ &= \frac{n \text{Var}(X)}{n^2} = \frac{\text{Var}(X)}{n} \end{aligned}$$

CLT Suppose $X_i \ i=1, \dots, n$ are i.i.d r.v

with $E[X_i] = \mu < \infty$ $\text{Var}(X_i) = \sigma^2 < \infty$ will tend to the standard normal as $n \rightarrow \infty$.

$$P\left(\frac{X_1 + X_2 + \dots + X_n - n\mu}{\sqrt{n}\sigma} < c\right) \xrightarrow{\text{Standard normal}} \underline{\Phi}(c)$$

ex There are 100 people on a plane. Let X_i be the weight ^{continuous} of the i th person.

Suppose X_i 's i.i.d and $E[X_i] = \mu = 170$ and $\sigma = 30$. Find the prob that the total weight exceeds 18,000 pounds.

Let $W = X_1 + X_2 + \dots + X_{100}$

$$E[W] = n\mu = 100 \cdot 170 \approx 17,000$$

$$\sqrt{n}\sigma = 10 \cdot 30 = 300$$

$$P(W > 18000) = 1 - P(W \leq 18000)$$

$$= 1 - P\left(\frac{W - 17000}{300} \leq \frac{18000 - 17000}{300}\right)$$

$$= 1 - P\left(Z \leq \frac{10}{3}\right)$$

$$= 1 - \Phi\left(\frac{10}{3}\right) \xrightarrow{3.33}$$

$$= 1 - .9996 = .0004$$

Let X_1, X_2, \dots, X_{25} be i.i.d with pmf

$$P_X(k) = \begin{cases} 0.6 & k=1 \\ 0.4 & k=-1 \\ 0 & \text{Otherwise} \end{cases} \quad \text{discrete}$$

estimate $P(4 \leq Y \leq 6)$ if $Y = X_1 + \dots + X_{25}$

$$\text{we have } E[X_i] = 1(0.6) + (-1)(0.4) = \frac{1}{5} = \mu$$

$$E[X_i^2] = 1(0.6) + (-1)^2(0.4) = 1$$

$$\text{Var}(X_i) = 1 - \left(\frac{1}{5}\right)^2 = \frac{24}{25} = \sigma^2$$

$$\sigma = \frac{\sqrt{24}}{5}$$

$$\Rightarrow n\mu = 25 \cdot \frac{1}{5} = 5 \quad \sqrt{n} \cdot \sigma = \sqrt{25} \cdot \sqrt{24} = \sqrt{24}$$

cont. correction \rightarrow discrete

$$P(4 \leq Y \leq 6) = P(3.5 \leq Y \leq 6.5)$$

$$= P\left(\frac{3.5-5}{\sqrt{24}} \leq Z \leq \frac{6.5-5}{\sqrt{24}}\right)$$

$$\approx P(-.31 \leq Z \leq .31)$$

$$= P(Z \leq .31) - P(Z \leq -.31)$$

$$= P(Z \leq .31) - (1 - P(Z \leq -.31))$$

$$= 2 \Phi(.31) - 1$$

$$\approx 0.24$$

Monte Carlo \rightarrow named after casino in Monaco

randomness

chance

\hookrightarrow simulation, computational method that uses

sampling to estimate numerical results

↓
random

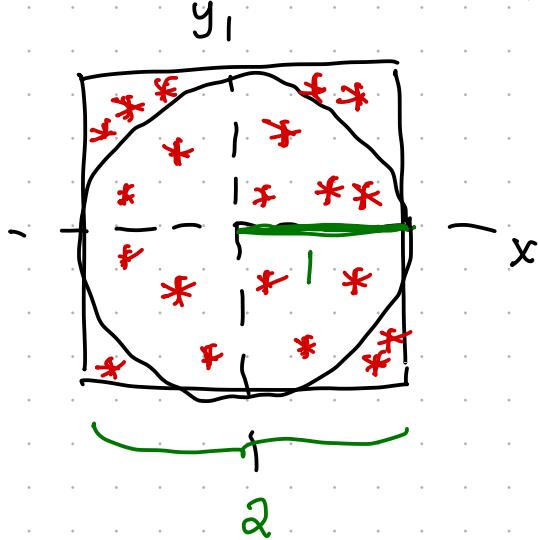
\hookrightarrow by performing a large # of simulations with varying results, method provides

probabilistic outcomes close to what we hope for

↳ usually used in place of other methods that can be analytically hard

Estimate π (by using computer)

↳ pick a random point within a square



Probability that the point $*$ lies in the

$$\text{circle: } P = \frac{\text{A circle}}{\text{A square}} = \frac{\pi r^2}{2^2} = \frac{\pi}{4}$$

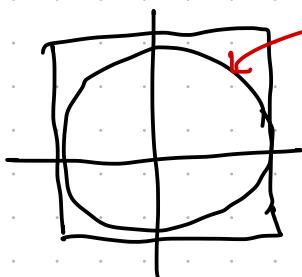
$$\text{Ratio} = \frac{N \text{ of pt inside circle}}{N \text{ of total pts used}} = \frac{\pi}{4}$$

$$\rightarrow \pi \approx 4 \cdot \text{Ratio}$$

ex: if $N = 1,000,000$ (total) but inside the circle there are 785,398

$$\pi \approx 4 \cdot \frac{785,398}{1000000} \approx 4 \cdot 0.785398 \approx 3.141592$$

How to count/check if * is in the circle?



check if x^2+y^2 of * is ≤ 1 . If so, count it.

① Computer generates random pt

② Check if it is in the circle \rightarrow count

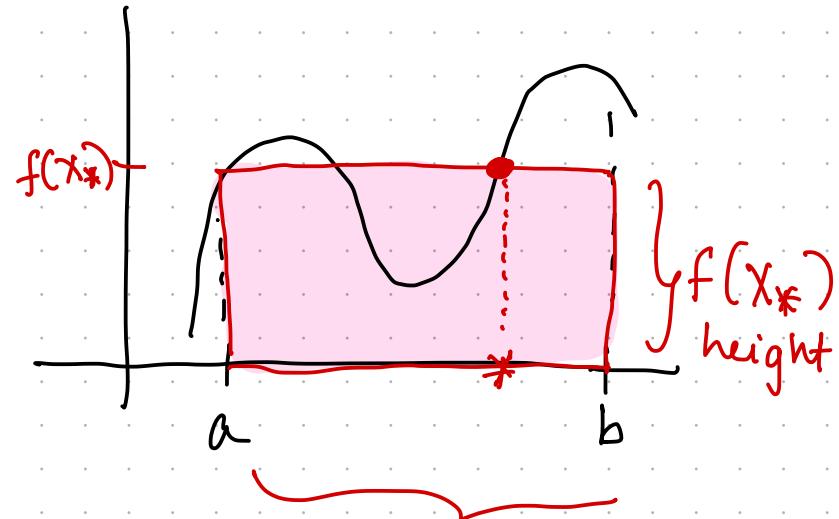
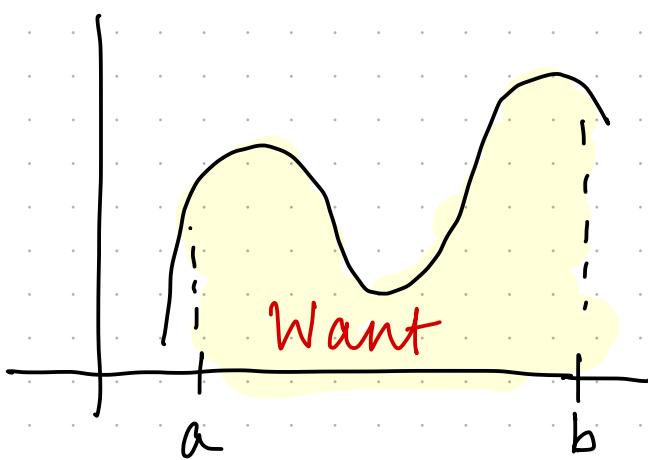
③ $\pi \approx 4 \cdot \frac{N_{\text{inside}}}{N_{\text{total}}} \xrightarrow{\quad} x^2+y^2 \leq 1 ?$

Monte Carlo for Integrations \rightarrow used to compute

integrals that are high dimensional or complex to solve by hand

Given $f(x)$ on a domain (D)

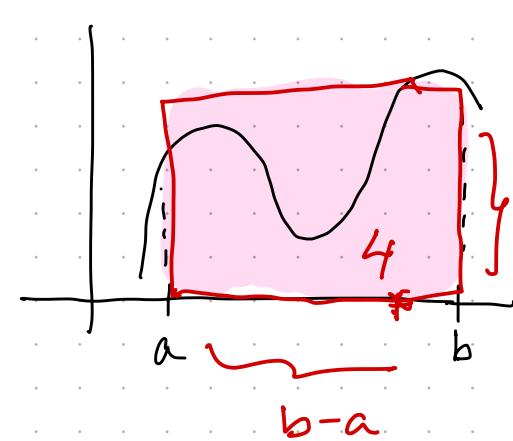
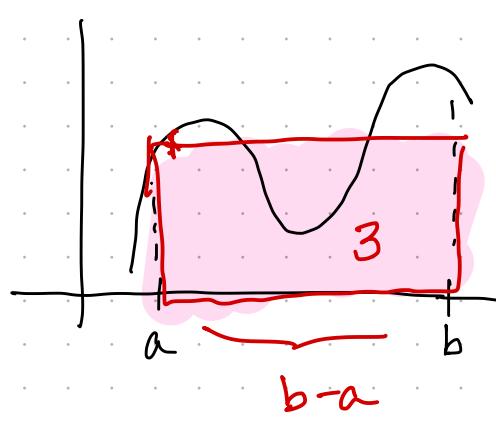
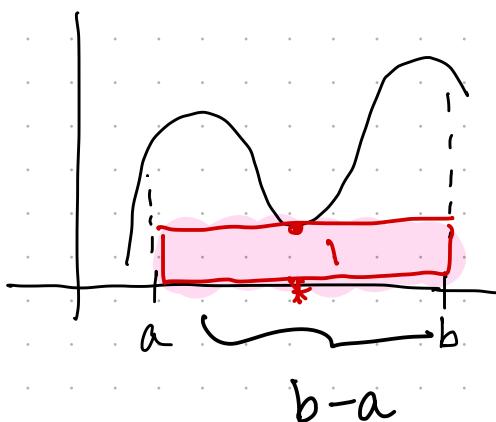
$$I = \int_D f(x) dx = \int_a^b f(x) dx$$



$b-a$ width

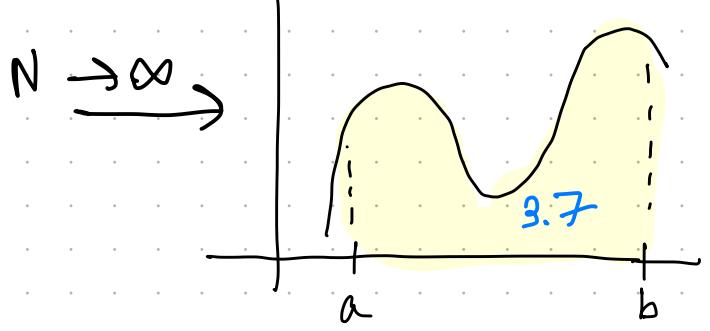
$$\text{area: } (b-a)f(x^*)$$

Repeat the process



Avg:

$$\frac{1}{3} \left(\begin{array}{c} 1 \\ + \\ 3 \\ + \\ 4 \end{array} \right)$$



$$\begin{aligned}
 \int_a^b f(x) dx &\approx \frac{1}{N} \sum_{i=1}^N (b-a) f(x_i) \\
 &= \frac{1}{N} (b-a) \sum_{i=1}^N f(x_i) \\
 &= (b-a) \cdot \frac{1}{N} \sum_{i=1}^N f(x_i)
 \end{aligned}$$

width height
 ↓
 random pts

length of avg function
 interval

ex estimate $\int_0^1 e^{-x^2} dx$ * Note $\int_a^b e^{-x^2} dx$ does not have a closed-form solution

D: $[0, 1] \rightarrow (b-a) = 1$

- ① Generate N random points in $[0, 1]$
- ② Compute $f(x_i)$ (height)
- ③ $\int_0^1 e^{-x^2} dx \approx 1 \cdot \frac{1}{N} \sum_{i=1}^N e^{-x_i^2}$

mean/avg function

→ exact value is approx 0.7468

Markov Chains: example of stochastic (random) process changing over time

Before: Sequence R.V. $X_1, X_2, X_3 \dots$ and assumed to be i.i.d

↳ WLLN, CLT

Now: R.V. can have time dependence

↳ can get very complicated if we allowed for the dependence of an infinite amt of parameters

Compromise: Markov chain are "one step" beyond i.i.d

Applications: Weather
Stock prices
genetics

Computer science → Google PageRank

* can extend to dependence of more than one dependence → need to look at single dependence

Let X_n be the state of the system

at (discrete) time n & takes on finite # of states

↳ process that is bouncing around randomly from state to state

ex Imagine a simple weather model where it is either sunny (S) or rainy (R)

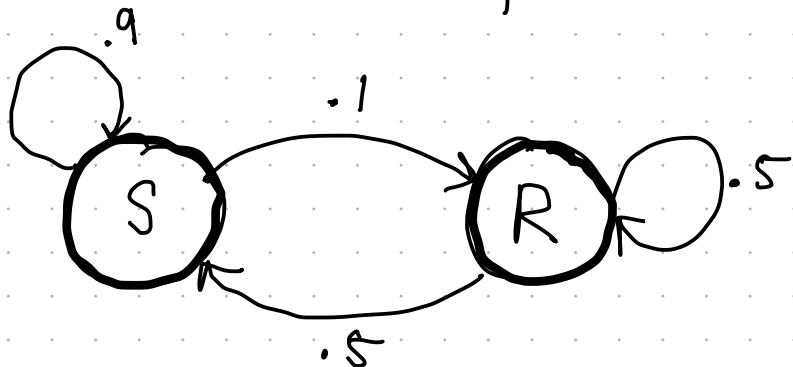
States

The weather tomorrow only depends on the weather today.

- If it is sunny today, there is a 90%

chance it is sunny tomorrow & 10% chance " " rainy "

- If it is rainy today, there is a 50% chance it is sunny tomorrow & 50% " " " rainy tomorrow



Transition Matrix (Q, OR P, OR M)

$$Q = \begin{matrix} & \begin{matrix} S & R \end{matrix} \\ \begin{matrix} S \\ R \end{matrix} & \begin{bmatrix} .9 & .1 \\ .5 & .5 \end{bmatrix} \end{matrix}$$

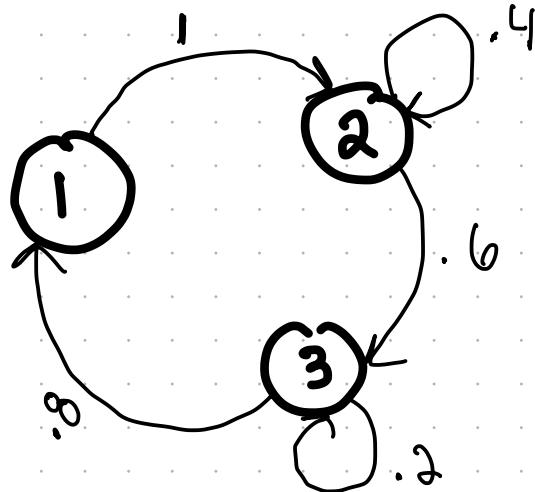
Board Game Movement : 3 spaces {1,2,3}

The player's ^{next} position only depends on their current position.

From space 1 → move to space 2 w/ prob 1

" space 2 → " " space 3 w/ prob .6
stays at space 2 w/ prob .4

From space 3 → move to space 1 w/ prob .8
stay at space 3 w/ prob .2



$$Q = \begin{bmatrix} 0 & 1 & 0 \\ 0 & .4 & .6 \\ .8 & 0 & .2 \end{bmatrix}$$

all rows add up
to 1

* NOT true for columns

pmf for each
stage

each row is a pmf for the
prob you will end up at
the other stages in the
next time step

Markov Property (mathematically)

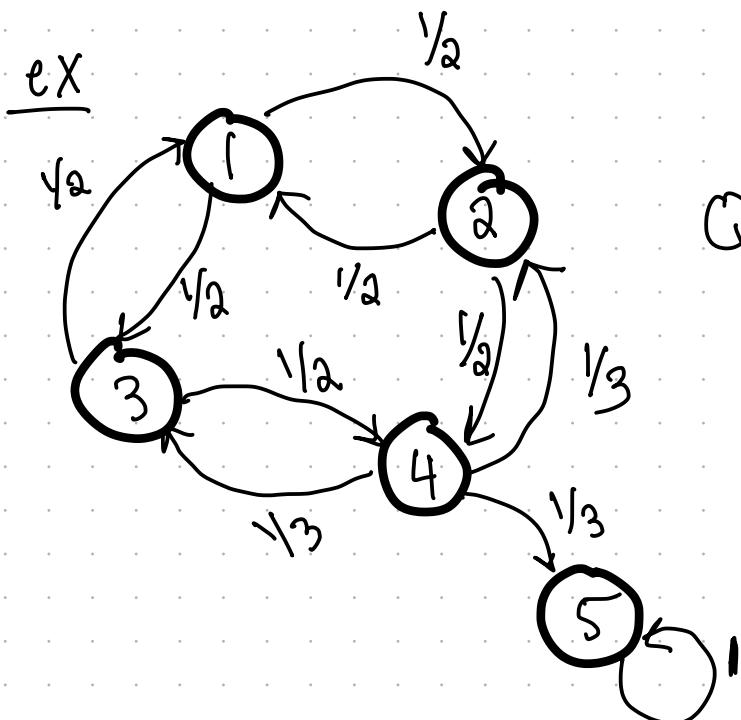
Conditional prob statement if we think of
 X_n as the current time, ^{we} want to
know one step in the future

$$P(X_{n+1} = j \mid X_n = i_n, X_{n-1} = i_{n-1}, X_{n-2} = i_{n-2}, \dots)$$

↓ and ↓ and
 all previous knowledge

Markovian property

$$= P(X_{n+1} \mid X_n = i_n)$$



$$Q = \begin{bmatrix} 0 & 1/2 & 1/2 & 0 & 0 \\ 1/2 & 0 & 0 & 1/2 & 0 \\ 1/2 & 0 & 0 & 1/2 & 0 \\ 0 & 1/3 & 1/3 & 0 & 1/3 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\text{Find } P(X_2 = 5 \mid X_0 = 3)$$

Want to find prob we end up at state 5 after 2 timesteps, starting from state 3.

Intuitively $\rightarrow \frac{1}{2} \cdot \frac{1}{3}$ "follow" path

↪ this is just the total law of prob!
 TLOP

TLOP says if B_i 's partition a sample space

$$P(A) = \sum_i P(A|B_i) P(B_i)$$

What if we need $P(A|C)$? We just condition everything on C to get

$$P(A|C) = \sum_i P(A|B_i \cap C) P(B_i|C)$$

So: $P(X_2=5|X_0=3) = \sum_{i=1}^5 P(X_2=5 | X_0=3, X_1=i) P(X_1=i | X_0=3)$

irrelevant

$= \sum_{i=1}^5 P(X_2=5 | X_1=i) P(X_1=i | X_0=3)$

$= P(X_2=5 | X_1=1) P(X_1=1 | X_0=3) + P(X_2=5 | X_1=2) P(X_1=2 | X_0=3)$

$+ P(X_2=5 | X_1=3) P(X_1=3 | X_0=3) + P(X_2=5 | X_1=4) P(X_1=4 | X_0=3)$

$+ P(X_2=5 | X_1=5) P(X_1=5 | X_0=3)$

$\stackrel{=0}{=} \quad \stackrel{=0}{=} \quad \stackrel{=0}{=} \quad \stackrel{=0}{=} \quad \stackrel{=0}{=} \quad \stackrel{=0}{=} \quad \stackrel{=0}{=}$

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$= \frac{1}{3} \cdot \frac{1}{2} = \frac{1}{6} \quad \checkmark$

General Pattern :

$$P(X_{n+1} = j | X_n = i) = q_{ij}$$

$(i,j)^{\text{th}}$ entry in Q

What about 2 step?

$$P(X_{n+2} = j | X_n = i)$$

missing link/info X_{n+1}

$$= \sum_K P(X_{n+2} = j | X_{n+1} = K, X_n = i) P(X_{n+1} = K | X_n = i)$$

previously: $j=5$ $K=1-5$ $i=3$ irrelevant

Markov

$$\underset{\text{prop}}{=} \sum_K P(X_{n+2} = j | X_{n+1} = K) P(X_{n+1} = K | X_n = i)$$

$$= \sum_K q_{kj} \cdot q_{ik} = \sum_K q_{ik} q_{kj}$$

$(i,j)^{\text{th}}$ entry of Q^2

transition
matrix
multiplied
twice

$$P(X_{n+m} = j | X_n = i) = (i,j)^{\text{th}} \text{ entry of } Q^m$$

ex Consider a system w/ 2 states {0, 1}

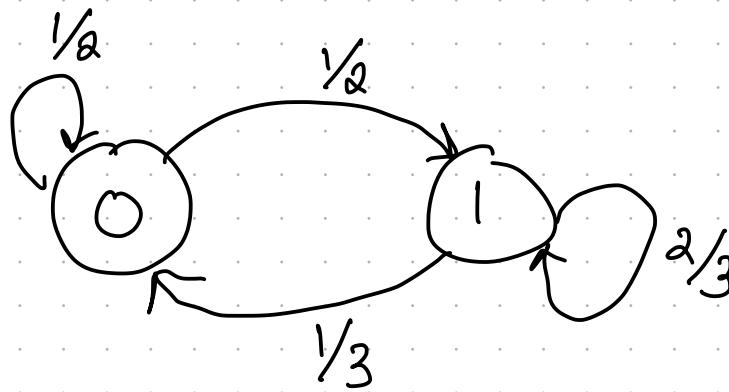
w/

$$Q = \begin{bmatrix} 1/2 & 1/2 \\ 1/3 & 2/3 \end{bmatrix}$$

Suppose system is in state 0 at time n=0.

Find the prob that the system is in state 1

at n=3.



initial starting $\vec{s}^{(0)} = [P(X_0=0) \ P(X_0=1)]$

OR $\vec{\pi}^{(0)} = [1 \ 0]$

$$\vec{s}^{(1)} = \vec{s}^{(0)} Q = [1 \ 0] \begin{bmatrix} 1/2 & 1/2 \\ 1/3 & 2/3 \end{bmatrix}$$

$$= \left[1 \cdot \frac{1}{2} + 0 \cdot \frac{1}{3} \quad 1 \cdot \frac{1}{2} + 0 \cdot \frac{2}{3} \right]$$

$$= \begin{bmatrix} 1/2+0 & 1/2+0 \end{bmatrix} = \begin{bmatrix} 1/2 & 1/2 \end{bmatrix}$$

$$\vec{S}^{(1)} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \end{bmatrix} \rightarrow \vec{S}^{(2)} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{3} & \frac{2}{3} \end{bmatrix} \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{3} & \frac{2}{3} \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{2} \cdot \frac{1}{2} + \frac{1}{2} \cdot \frac{1}{3} & \frac{1}{2} \cdot \frac{1}{2} + \frac{1}{2} \cdot \frac{2}{3} \\ \frac{1}{4} + \frac{1}{6} & \frac{1}{4} + \frac{1}{3} \end{bmatrix}$$

$$\vec{S}^{(3)} = \vec{S}^{(2)} Q = \begin{bmatrix} \frac{5}{12} & \frac{7}{12} \end{bmatrix} \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{3} & \frac{2}{3} \end{bmatrix}$$

adds to 1 ✓

↓

$$\begin{bmatrix} \frac{5}{12} \cdot \frac{1}{2} + \frac{7}{12} \cdot \frac{1}{3} & \frac{5}{12} \cdot \frac{1}{2} + \frac{7}{12} \cdot \frac{2}{3} \\ \frac{5}{24} + \frac{7}{36} & \frac{5}{24} + \frac{14}{36} \end{bmatrix}$$

$$\begin{bmatrix} \frac{29}{72} & \frac{43}{72} \end{bmatrix}$$

T
adds to 1 ✓