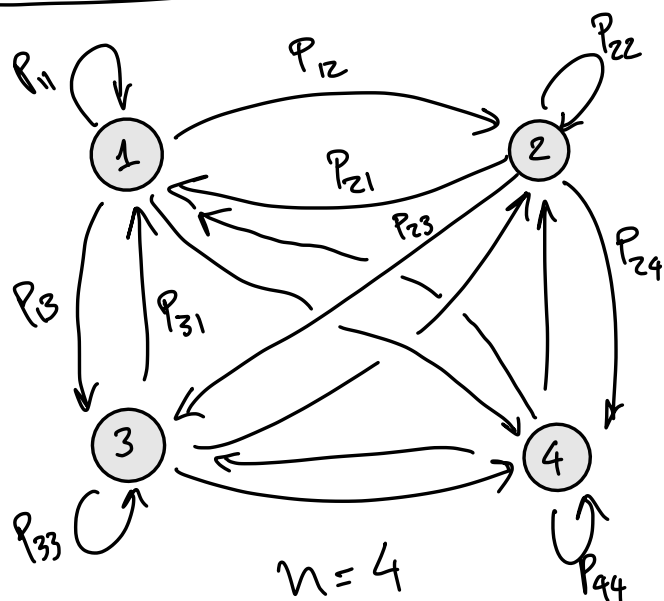


Markov Chains: Suppose that a system can be in one of  $n$  different states at each given time.



state:	2	1	3	...	1	...
Time:	1	2	3	...	$k$	...
random variables:	$X_1$	$X_2$	$X_3$	...	$X_k$	...

If the system is in state  $i$ , the probability that it transitions to state  $j$  in the next unit time is given by:

Transition probabilities from  $i$  to  $j$ .

$$P_{ij} = P(\underbrace{X_{k+1} = j}_{\text{next state being } j} \mid \underbrace{X_k = i}_{\text{current state is } i})$$

Markov hypothesis

$$= P(X_{k+1} = j \mid \underbrace{X_k = i}_{\text{current state}}, \underbrace{X_{k-1} = i_{k-1}, X_{k-2} = i_{k-2}, \dots, X_0 = i_0}_{\text{past}}), \forall k.$$

Past is irrelevant to determine current transition probabilities.

Organize transition probabilities in a matrix, as follows:

	next state 1	next state 2	next state 3	...	next state n	
	↓	↓	↓		↓	
	1	2	3	...	n	
current state 1 →	$P_{11}$	$P_{12}$	$P_{13}$	...	$P_{1n}$	1
current state 2 →	$P_{21}$	$P_{22}$	$P_{23}$	...	$P_{2n}$	2
current state n →	$P_{n1}$	$P_{n2}$	$P_{n3}$	...	$P_{nn}$	n

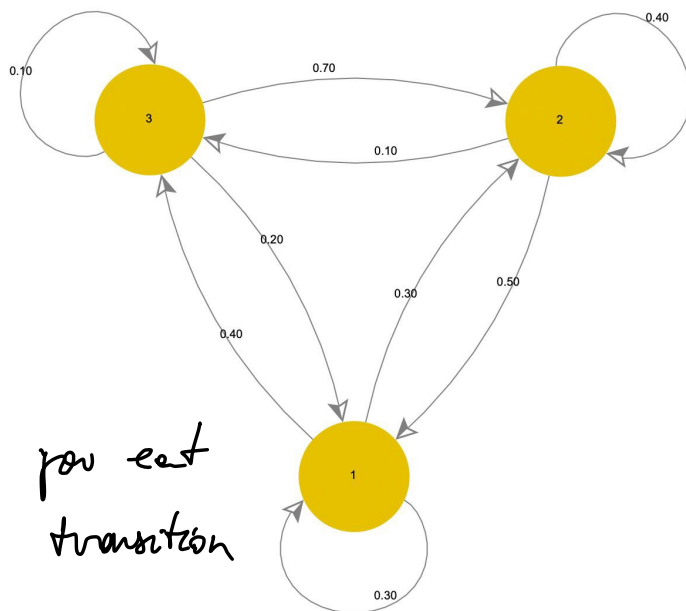
Clearly, the sum of probabilities in each line of  $P$  is 1.

$$\sum_{j=1}^n P_{ij} = 1.$$

This means that  $P$  is a "stochastic matrix."

Example. Suppose you eat 1 of 3 desserts every day:

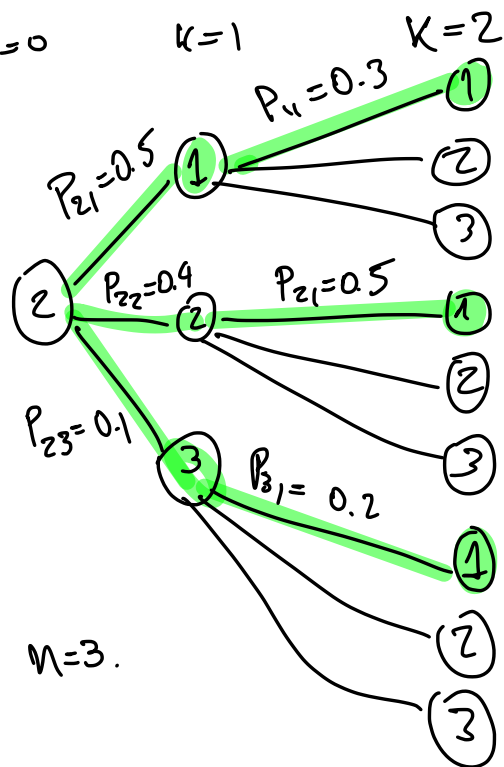
1. Ice Cream
2. Cake
3. Chocolate



Assume the sequence of desserts you eat form a Markov chain with transition probabilities as indicated.

$$P = \begin{pmatrix} 0.3 & 0.3 & 0.4 \\ 0.5 & 0.4 & 0.1 \\ 0.2 & 0.7 & 0.1 \end{pmatrix}$$

Q: What is the probability that you eat Ice Cream in 2 days, given that you ate Cake today?



$$P_{21}^{(2)} := P(X_2=1 | X_0=2) = ?$$

$P_{ij}^{(n)} = \sum_{k=1}^n P_{ik} P_{kj}$  ← entry (i,j) of matrix  $P^n$ .

$$\begin{aligned} P_{21}^{(2)} &= P_{21} \cdot P_{11} + P_{22} \cdot P_{21} + P_{23} \cdot P_{31} \\ &= 0.5 \cdot 0.3 + 0.4 \cdot 0.5 + 0.1 \cdot 0.2 \\ &= 0.15 + 0.20 + 0.02 \\ &= 0.37 \\ &= 37\% \end{aligned}$$

More generally,  $P_{ij}^{(N)} = (i,j)$  entry of matrix  $P^N$ .

Q: What is the probability that you eat Ice Cream in 200 days, given that you ate Cake today?

Using Linear Algebra,

$$P^{200} \approx \begin{pmatrix} 0.36 & 0.43 & 0.21 \\ \boxed{0.36} & 0.43 & 0.21 \\ 0.36 & 0.43 & 0.21 \end{pmatrix}$$

$N=200$

$$P_{21}^{(200)} \approx 0.36 = 36\%$$

# Chapman-Kolmogorov Equation

$$P_{ij}^{(N)} = \sum_{k=1}^n P_{ik}^{(r)} P_{kj}^{(N-r)}, \quad \text{for any } 0 < r < N.$$

Probability of going from state  $i$  to state  $j$  after  $N$  steps.

In particular, if  $r=1$ , then

$$P_{ij}^{(N)} = \sum_{k=1}^n P_{ik} \cdot P_{kj}^{(N-1)} \leftarrow \left( \begin{array}{l} \text{Recursively} \\ \text{defining the} \\ (i,j)\text{-entry of} \\ P^N. \end{array} \right)$$

$$(P_{ij}^{(1)} = P_{ij})$$

Pr.  $P_{ij}^{(N)} = P(X_N = j \mid X_0 = i)$

$$= \sum_{k=1}^n P(X_N = j, X_r = k \mid X_0 = i)$$

$$\frac{P(X_N = j, X_r = k, X_0 = i)}{P(X_0 = i)}$$

$$= \sum_{k=1}^n \underbrace{P(X_N = j \mid X_r = k, X_0 = i)}_{\frac{P(X_N = j, X_r = k, X_0 = i)}{P(X_r = k, X_0 = i)}} \cdot \underbrace{P(X_r = k \mid X_0 = i)}_{\frac{P(X_r = k, X_0 = i)}{P(X_0 = i)}}$$

$$\frac{P(X_N = j, X_r = k, X_0 = i)}{P(X_r = k, X_0 = i)}$$

$$\frac{P(X_r = k, X_0 = i)}{P(X_0 = i)}$$

$$P(X_N = j \mid X_r = k)$$

$$P_{kj}^{(N-r)}$$

By Markov hypothesis,  $P(X_N = j \mid X_r = k, X_0 = i) = P(X_N = j \mid X_r = k) = P_{kj}^{(N-r)}$

$$\dots = \sum_{k=1}^n P_{kj}^{(N-r)} \cdot P_{ik}^{(r)} = \sum_{k=1}^n P_{ik}^{(r)} \cdot P_{kj}^{(N-r)}$$

Q1: Over the long run, with what frequency is state  $j$  visited?

Q2: As time goes to  $+\infty$ , in what state is the system most likely to be?

Q1':  $\pi_j := \lim_{N \rightarrow \infty} P_{ij}^{(N)} = ?$  Does it exist?  
How to compute it?

Q2': Which  $\pi_j$ ,  $1 \leq j \leq n$ , is the largest?

Def: A Markov chain is ergodic if there exists  $N > 0$  s.t. all entries of  $P^N$  are strictly positive, i.e.,  $P_{ij}^{(N)} > 0$ , for all  $1 \leq i, j \leq n$ .

(i.e. one can reach every state from any other state) within  $N$  steps.

Thm: If a Markov chain is ergodic, then the limits  $\pi_j = \lim_{N \rightarrow \infty} P_{ij}^{(N)}$  exist for all  $1 \leq j \leq n$ , and can be

computed as the unique non-negative solutions to

$$\sum_{j=1}^n \pi_j = 1 \quad \text{and} \quad \sum_{k=1}^n \pi_k P_{kj} = \pi_j.$$

i.e.  $P^t \vec{\pi} = \vec{\pi}$  is an eigenvector of  $P^t$  w/ eigenvalue 1, whose entries are  $\geq 0$  and add up to 1.

In the example from before, with transition matrix

$$P = \begin{pmatrix} 0.3 & 0.3 & 0.4 \\ 0.5 & 0.4 & 0.1 \\ 0.2 & 0.7 & 0.1 \end{pmatrix} = \begin{pmatrix} P_{11} & P_{12} & P_{13} \\ P_{21} & P_{22} & P_{23} \\ P_{31} & P_{32} & P_{33} \end{pmatrix}$$

We have:

- Markov chain is ergodic  $P_{ij} > 0$ ,  $1 \leq i, j \leq 3$ .
- $\pi_j$  can be computed as the nonnegative solution to:

$$\begin{cases} \pi_1 + \pi_2 + \pi_3 = 1 \leftarrow \langle \vec{\pi}, \vec{1} \rangle = 1 \\ \pi_1 = \pi_1 P_{11} + \pi_2 P_{21} + \pi_3 P_{31} \\ \pi_2 = \pi_1 P_{12} + \pi_2 P_{22} + \pi_3 P_{32} \\ \pi_3 = \pi_1 P_{13} + \pi_2 P_{23} + \pi_3 P_{33} \end{cases} \quad \vec{\pi} = (\pi_1, \pi_2, \pi_3)$$

$P^t \vec{\pi} = \vec{\pi}$

In other words, using Linear Algebra, the above is equivalent to finding an eigenvector  $\vec{\pi} = (\pi_1, \pi_2, \pi_3)$  for the matrix  $P^t$  with eigenvalue 1, whose entries are nonnegative and add up to 1.

$$\vec{\pi} = (\pi_1, \pi_2, \pi_3) = \left( \frac{47}{129}, \frac{55}{129}, \frac{9}{43} \right) \cong (0.36, 0.43, 0.21)$$

(cf. with entries of  $P^{\text{pro}}$ )