

CSE 312

Foundations of Computing II

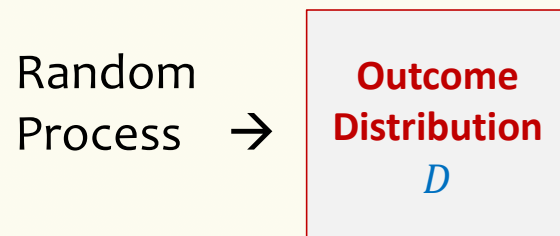
Lecture 26: Markov Chain



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Slide Credit: Based on Stefano Tessaro's slides for 312 19au
incorporating ideas from Alex Tsun's and Anna Karlin's slides for 312 20su and 20au

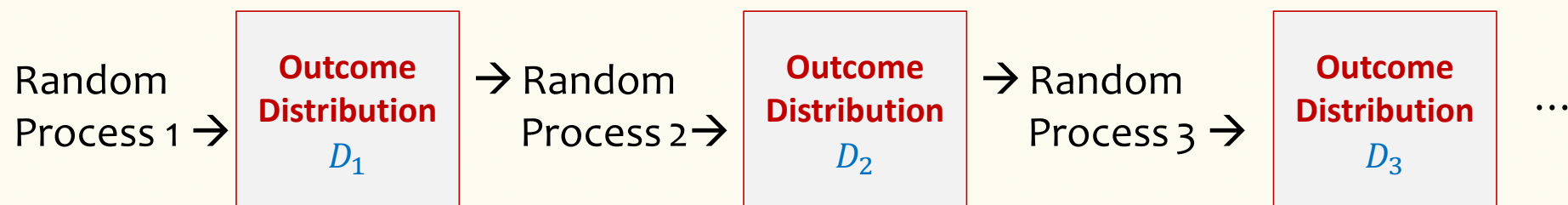
So far, a single-shot random process



Today:

See a very special type of DTSP called **Markov Chains**

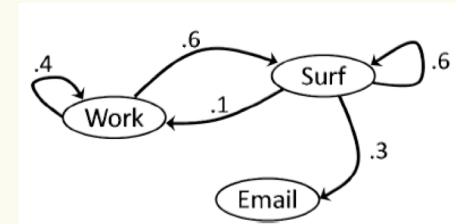
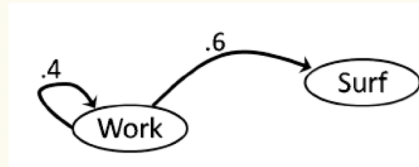
Many-step random process



Definition: A **discrete-time stochastic process** (DTSP) is a sequence of random variables $X^{(0)}, X^{(1)}, X^{(2)}, \dots$ where $X^{(t)}$ is the value at time t .

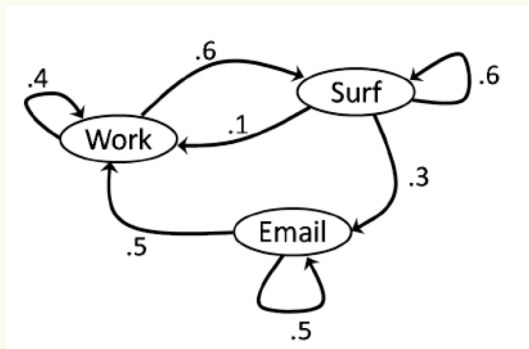
A boring day in pandemic

$t = 0$



This graph defines a **Markov Chain**

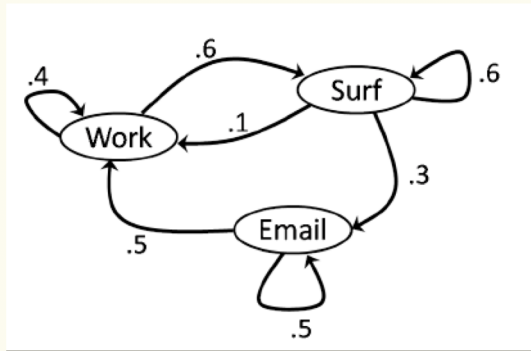
Special Property: The random process at each step is the same!



For ANY $t \geq 0$,
if I was working at time t , then at $t+1$
with probability 0.4 I continue working
with probability 0.6, I move to surfing, and
with probability 0, I move to emailing

This is called History Independent (similar to memoryless)

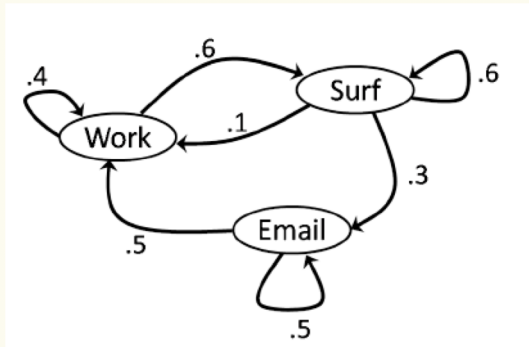
A boring day in pandemic



Many interesting questions?

1. What is the probability that I work at time 1?
2. What is the probability that I work at time 2?
3. What is the probability that I work at time $t=100$?
4. What is the probability that I work at time $t \rightarrow \infty$?
Does it always converge?

Formalizing Markov Chain

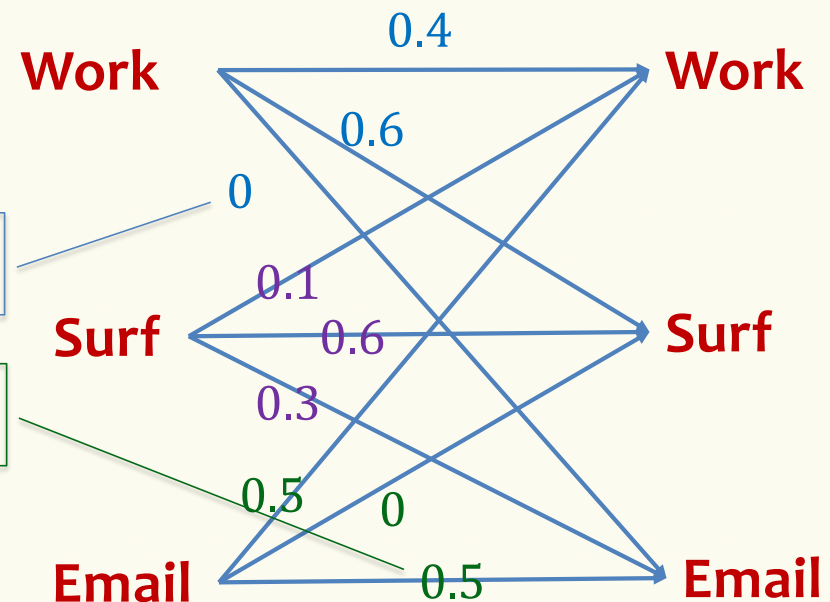


$X^{(t)}$ my state at t

X^{t+1} my state at t+1

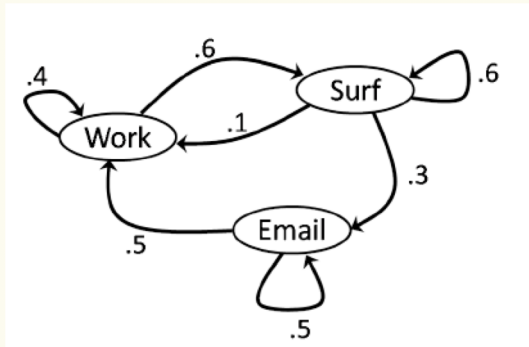
$$p_{WE} = P(X^{(t+1)} = E \mid X^t = W) = 0$$

$$p_{EE} = P(X^{(t+1)} = E \mid X^t = E) = 0.5$$



- What is the probability that I work at time 1? $p_W^{(1)} = P(X^{(1)} = W)$
- What is the probability that I work at time 2? $p_W^{(2)} = P(X^{(2)} = W)$

Formalizing Markov Chain

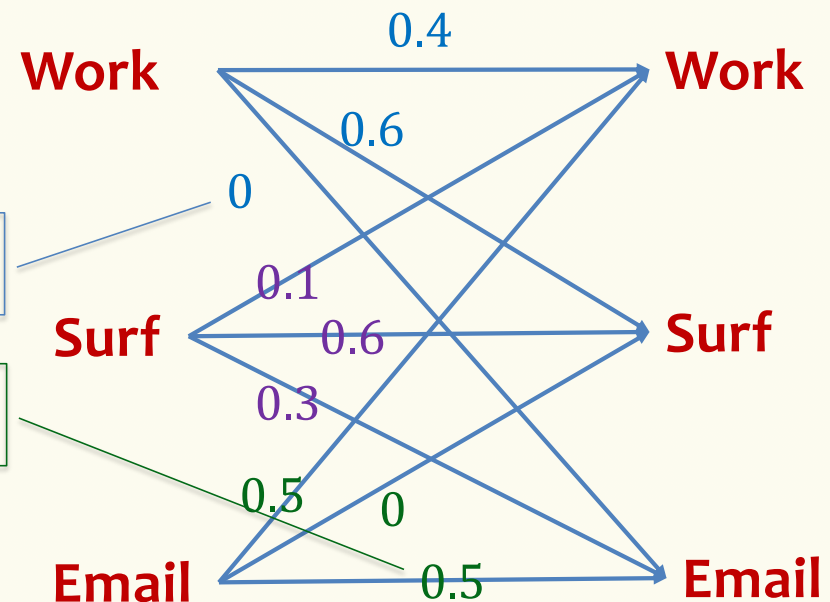


$X^{(t)}$ my state at t

X^{t+1} my state at t+1

$$p_{WE} = P(X^{(t+1)} = E \mid X^t = W) = 0$$

$$p_{EE} = P(X^{(t+1)} = E \mid X^t = E) = 0.5$$



3. What is the prob that I work at t+1 = 100?

By LTP: $p_W^{(t+1)} = P(X^{(t+1)} = W) = \sum_{U \in \{W, S, E\}} P(X^{(t+1)} = W \mid X^{(t)} = U) P(X^{(t)} = U)$

Vectors and Matrixes

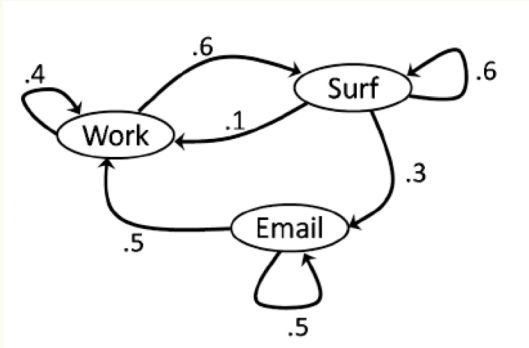
$$\begin{bmatrix} a_1 & a_2 & a_3 \\ a_4 & a_5 & a_6 \\ a_7 & a_8 & a_9 \end{bmatrix} \begin{bmatrix} b_1 \\ b_4 \\ b_7 \end{bmatrix} \begin{matrix} b_2 \\ b_5 \\ b_8 \end{matrix} \begin{matrix} b_3 \\ b_6 \\ b_9 \end{matrix} = \begin{bmatrix} c_1 & c_2 & c_3 \\ c_4 & c_5 & c_6 \\ c_7 & c_8 & c_9 \end{bmatrix}$$

Vectors and Matrixes

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix}, \quad \mathbf{B} = \begin{pmatrix} b_{11} & b_{12} & \cdots & b_{1p} \\ b_{21} & b_{22} & \cdots & b_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ b_{n1} & b_{n2} & \cdots & b_{np} \end{pmatrix}$$

$$\mathbf{C} = \begin{pmatrix} a_{11}b_{11} + \cdots + a_{1n}b_{n1} & a_{11}b_{12} + \cdots + a_{1n}b_{n2} & \cdots & a_{11}b_{1p} + \cdots + a_{1n}b_{np} \\ a_{21}b_{11} + \cdots + a_{2n}b_{n1} & a_{21}b_{12} + \cdots + a_{2n}b_{n2} & \cdots & a_{21}b_{1p} + \cdots + a_{2n}b_{np} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1}b_{11} + \cdots + a_{mn}b_{n1} & a_{m1}b_{12} + \cdots + a_{mn}b_{n2} & \cdots & a_{m1}b_{1p} + \cdots + a_{mn}b_{np} \end{pmatrix}$$

Transition Matrix



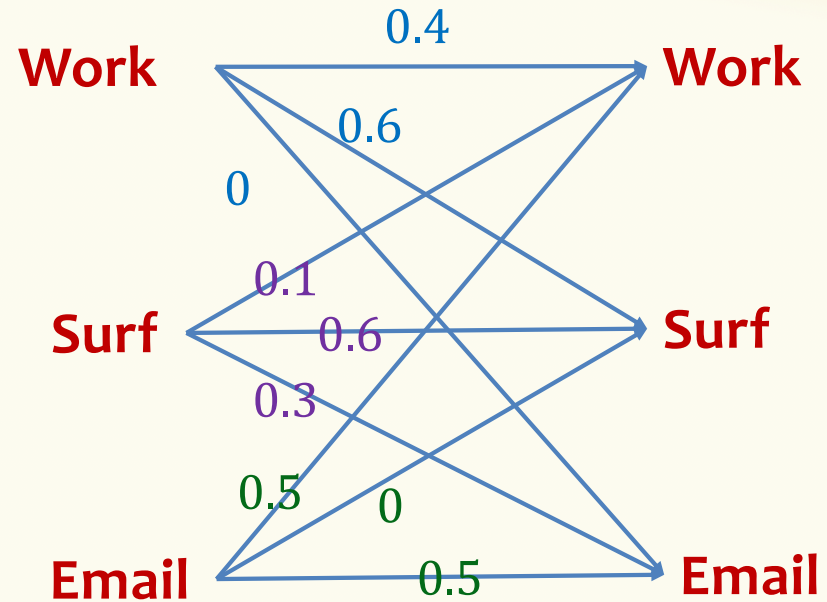
$$P = \begin{pmatrix} .4 & .6 & 0 \\ .1 & .6 & .3 \\ .5 & 0 & .5 \end{pmatrix}$$

$$X^{(t)} = (p_w^{(t)} \quad p_s^{(t)} \quad p_e^{(t)})$$

By LTP: $p_w^{(t+1)} = P(X^{(t+1)} = W) = \sum_{U \in \{W, S, E\}} P(X^{(t+1)} = W | X^{(t)} = U) P(X^{(t)} = U)$

$$\Rightarrow X^{(t+1)} = X^{(t)} P$$

$X^{(t)}$ my state at t X^{t+1} my state at t+1



Transition Matrix

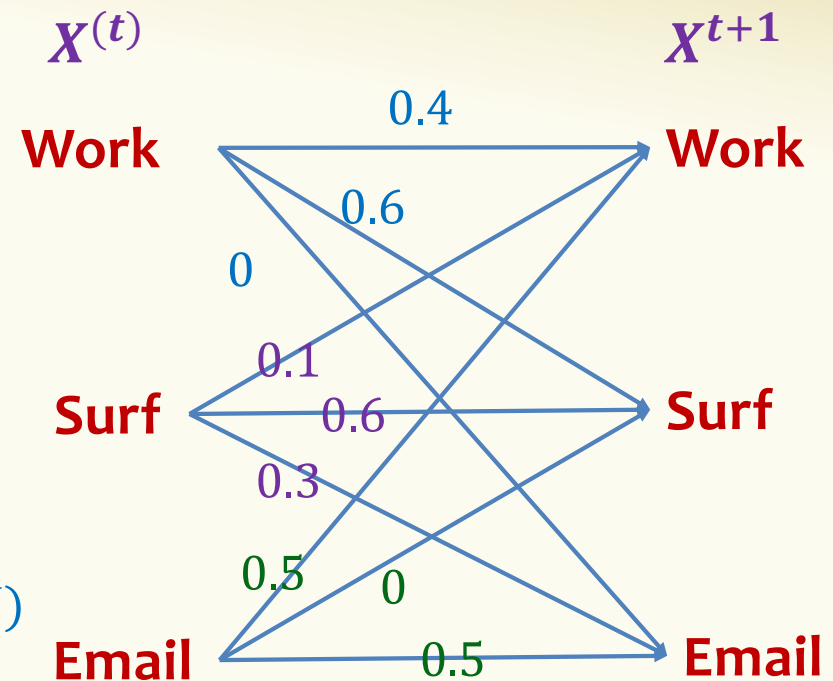
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$$X^{(t)} = (p_w^{(t)} \quad p_s^{(t)} \quad p_E^{(t)})$$

$$\begin{aligned} \text{LTP: } p_w^{(t+1)} &= P(X^{(t+1)} = W) \\ &= \sum_{U \in \{W, S, E\}} P(X^{(t+1)} = W | X^{(t)} = U) P(X^{(t)} = U) \end{aligned}$$

$$\rightarrow X^{(t+1)} = X^{(t)} P$$

$$(p_w^{(t+1)} \quad p_s^{(t+1)} \quad p_E^{(t+1)}) = (p_w^{(t)} \quad p_s^{(t)} \quad p_E^{(t)}) \begin{pmatrix} .4 & .6 & 0 \\ .1 & .6 & .3 \\ .5 & 0 & .5 \end{pmatrix}$$



Transition Matrix

$$P = \begin{pmatrix} .4 & .6 & 0 \\ .1 & .6 & .3 \\ .5 & 0 & .5 \end{pmatrix}$$

$$X^{(t)} = (p_w^{(t)} \quad p_s^{(t)} \quad p_e^{(t)})$$

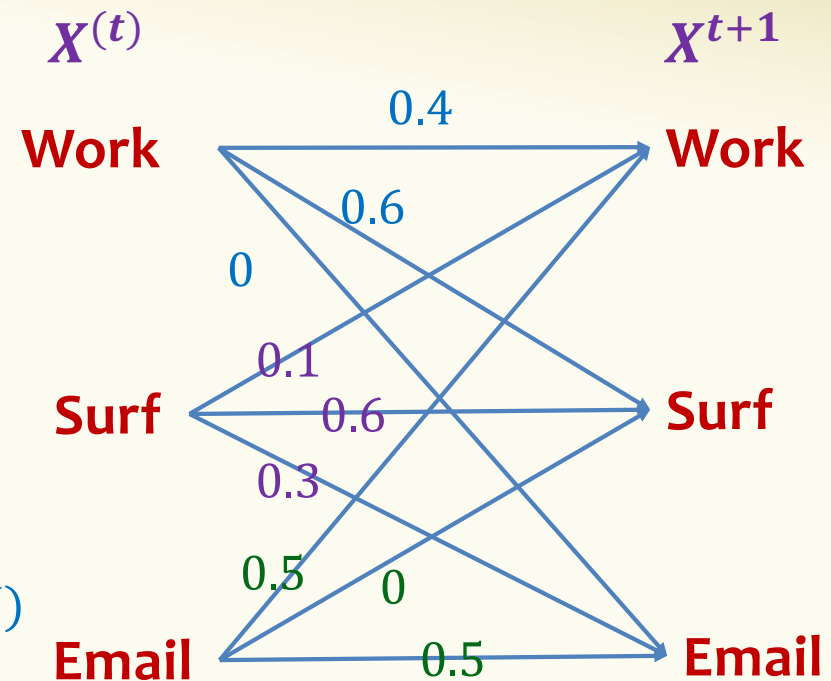
$$\begin{aligned} \text{LTP: } p_w^{(t+1)} &= P(X^{(t+1)} = W) \\ &= \sum_{U \in \{W, S, E\}} P(X^{(t+1)} = W | X^{(t)} = U) P(X^{(t)} = U) \end{aligned}$$

$$\rightarrow X^{(t+1)} = X^{(t)} P$$

$$\rightarrow X^{(t)} = X^{(0)} P^t$$

3. What is the prob that I work at $t = 100$?

Closed formula: $p_w^{(t)} = X^{(t)}[1] = (X^{(0)} P^t)[1]$



Transition Matrix

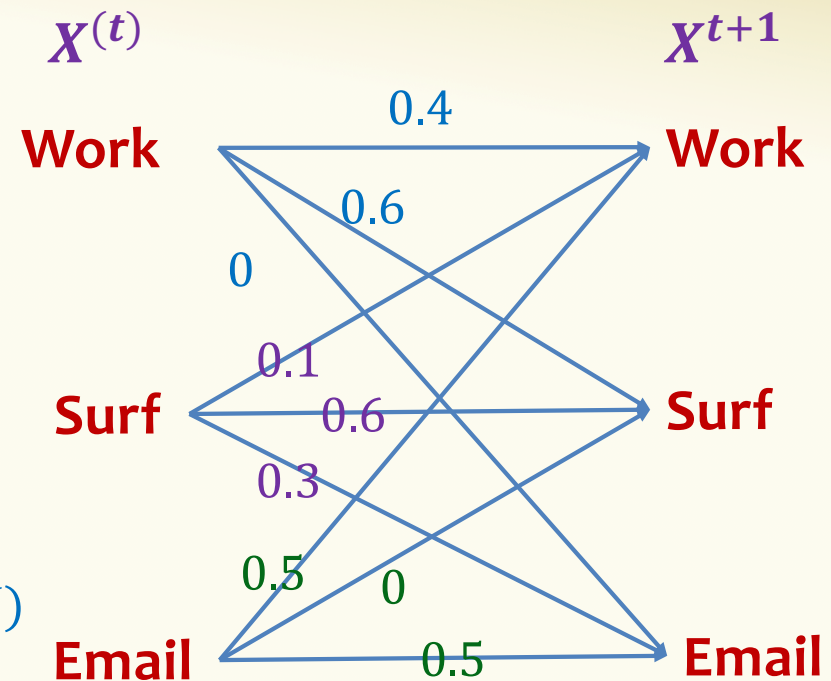
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$$\rightarrow X^{(t+1)} = X^{(t)} P$$

$$\rightarrow X^{(t)} = X^{(0)} P^t$$



4. What is the probability that I work at time $t \rightarrow \infty$?
Does it always converge?

Poll: A. Yes it converges (orderly universe)

B. No, it does not converge (anarchic universe)

Stationary Distribution

$$P = \begin{pmatrix} .4 & .6 & 0 \\ .1 & .6 & .3 \\ .5 & 0 & .5 \end{pmatrix}$$

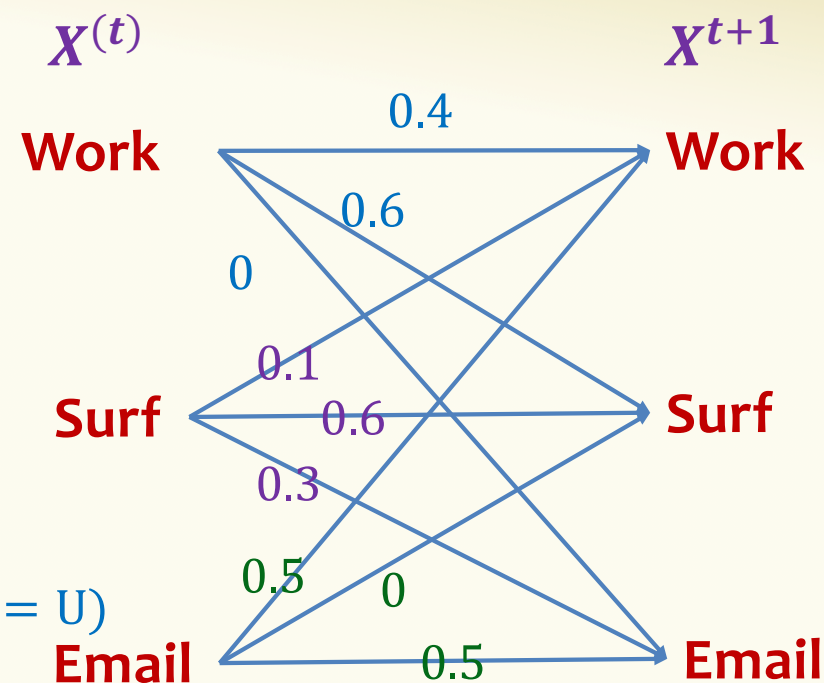
$$X^{(t)} = (p_w^{(t)} \quad p_s^{(t)} \quad p_e^{(t)})$$

$$\begin{aligned} \text{LTP: } p_w^{(t+1)} &= P(X^{(t+1)} = W) \\ &= \sum_{U \in \{W, S, E\}} P(X^{(t+1)} = W | X^{(t)} = U) P(X^{(t)} = U) \end{aligned}$$

$$\rightarrow X^{(t+1)} = X^{(t)} P$$

$$\rightarrow X^{(t)} = X^{(0)} P^t$$

$$\rightarrow \text{As } t \rightarrow \infty, \quad X^{(t)} \rightarrow \pi \quad \text{the stationary distribution}$$



4. What is the probability that I work at time $t \rightarrow \infty$?
Does it always converge?

Solving for Stationary Distribution

$$P = \begin{pmatrix} .4 & .6 & 0 \\ .1 & .6 & .3 \\ .5 & 0 & .5 \end{pmatrix}$$

$$\begin{aligned} \text{LTP: } p_W^{(t+1)} &= P(X^{(t+1)} = W) \\ &= \sum_{U \in \{W, S, E\}} P(X^{(t+1)} = W \mid X^{(t)} = U) P(X^{(t)} = U) \end{aligned}$$

$$\rightarrow X^{(t+1)} = X^{(t)} P$$

$$\rightarrow X^{(t)} = X^{(0)} P^t$$

$$\rightarrow \text{As } t \rightarrow \infty, \quad X^{(t)} \rightarrow \pi \quad \text{the stationary distribution}$$

Constraints on Stationary Distribution

- $\pi = \pi P$
- $\pi[1] + \pi[2] + \pi[3] = 1$

$$\rightarrow \pi[1] = \frac{15}{34}, \quad \pi[2] = \frac{10}{34}, \quad \pi[3] = \frac{9}{34}$$

General Markov Chain

- A set of n states $\{1, 2, 3, \dots, n\}$
- The state at time t $X^{(t)}$
- A transition matrix P , dimension $n \times n$
$$P[i, j] = \Pr(X^{(t+1)} = j \mid X^{(t)} = i)$$
- Transition: LTP $\rightarrow X^{(t+1)} = X^{(t)} P \Rightarrow X^{(t)} = X^{(0)} P^t$
- A stationary distribution π is the solution to:

$$\pi = \pi P, \text{ normalized so that } \sum_{i \in [n]} \pi[i] = 1$$

The Fundamental Theorem of Markov Chain

If a Markov Chain is **Irreducible** and **aperiodic**, then it has a unique stationary distribution.

Moreover, $t \rightarrow \infty$, for all i, j , $P^t[i, j] \rightarrow \pi[j]$