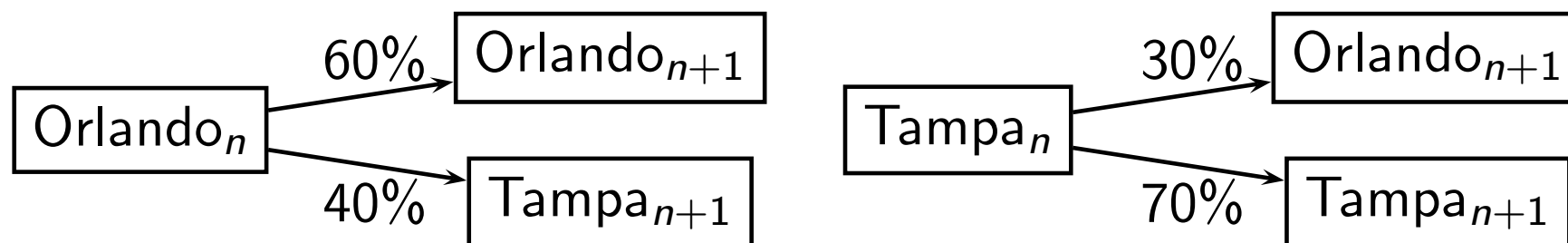


Markov Chains

A **Markov chain** is a sequence of random variables from some sample space, each corresponding to a successive time interval. From one time interval to the next, there is a *fixed* probability $a_{i,j}$ of transitioning from state j to state i . No transition depends on a past transition.

Keep track of these probabilities in an associated transition matrix A .

Example. Suppose you run a rental company based in Orlando and Tampa, Florida. People often drive between the cities; cars can be picked up and dropped off in either city. Suppose that **historically**,



What distribution of cars can the company expect in the long run?

Markov Chains

We will model this situation with a Markov Chain.

The historical data suggest that with a probability of **0.6**, a car in Orlando at time n will be in Orlando at time $n+1$. Use this and the other expected transition probabilities to form the transition matrix A .

		FROM:		
		Or	Tm	
TO:	Or	[= A,
	Tm			

- ▶ Let o_n be the **probability** that a car is in Orlando on day n
- ▶ Let t_n be the **probability** that a car is in Tampa on day n .

We can represent the distribution of cars at time n with the vector

$$\vec{x}_n = \begin{bmatrix} o_n \\ t_n \end{bmatrix}. \text{ And so, } \vec{x}_{n+1} = \begin{bmatrix} o_{n+1} \\ t_{n+1} \end{bmatrix} = A \cdot \begin{bmatrix} o_n \\ t_n \end{bmatrix} = A\vec{x}_n.$$

Given an initial distribution $\vec{x}_0 = \begin{bmatrix} o_0 \\ t_0 \end{bmatrix}$,

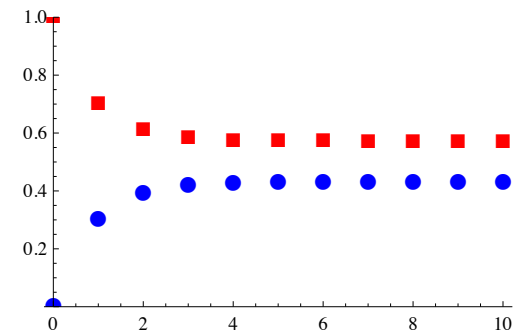
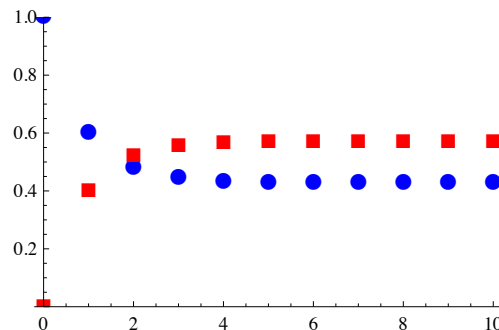
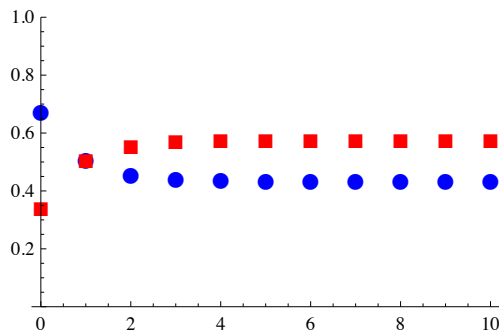
the expected distribution of cars at time n is $\vec{x}_n = \underline{\hspace{2cm}}$.

Markov Chains

For example, if the company starts off with twice as many cars in Orlando as in Tampa, then $\vec{x}_0 = \begin{bmatrix} 2/3 \\ 1/3 \end{bmatrix}$, so we expect

$$\vec{x}_1 = \begin{bmatrix} 0.6 & 0.3 \\ 0.4 & 0.7 \end{bmatrix} \begin{bmatrix} 2/3 \\ 1/3 \end{bmatrix} = \begin{bmatrix} \\ \end{bmatrix}.$$

$$\vec{x}_2 = \begin{bmatrix} 0.6 & 0.3 \\ 0.4 & 0.7 \end{bmatrix} \begin{bmatrix} \\ \end{bmatrix} = \begin{bmatrix} \\ \end{bmatrix}.$$



How do we determine the expected distribution in the long run?

Markov Chains

Definition: Given a Markov Chain with transition matrix A , an **equilibrium distribution** is a vector \vec{x}_{eq} that satisfies $A\vec{x}_{eq} = \vec{x}_{eq}$.
 [Linear Algebra: \vec{x}_{eq} is an eigenvector corresponding to $\lambda = 1$.]

In our example, the equilibrium distribution satisfies

$$\begin{bmatrix} 0.6 & 0.3 \\ 0.4 & 0.7 \end{bmatrix} \begin{bmatrix} o_{eq} \\ t_{eq} \end{bmatrix} = \begin{bmatrix} o_{eq} \\ t_{eq} \end{bmatrix}.$$

So solve: $0.6o_{eq} + 0.3t_{eq} = o_{eq}$ and $0.4o_{eq} + 0.7t_{eq} = t_{eq}$.
 Both equations reduce to $0.3t_{eq} = 0.4o_{eq}$, so $o_{eq} = \frac{3}{4}t_{eq}$.

Conclusion: If the company has 7000 cars in all, they would expect that in the long run, _____

In Markov Chains: ★ The sum of the entries in every column of A is 1, because the total probability of transitioning **from** state i is 1.

★ There is no general rule for what the row sum will be.