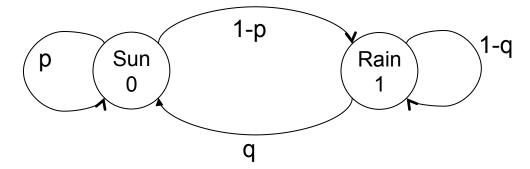
Markov Chains (Part 5)

Estimating Probabilities and Absorbing States

Weather Example: How to estimate *p* and *q*?

State Transition Diagram



Probability Transition Matrix

$$\begin{array}{cccc}
 & 0 & 1 \\
Sunny & 0 & p & 1-p \\
Rainy & 1 & q & 1-q
\end{array}$$

Weather Example: Estimation from Data

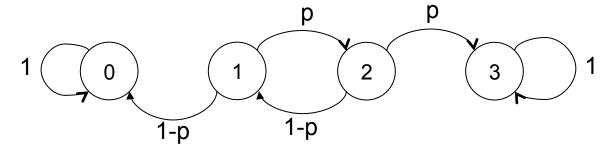
Estimate transition probabilities from data
 Weather data for 1 month (R – rainy, S – sunny)

Sun	M	Т	W	Th	F	Sat
R	R	R	S	S	R	R
R	S	S	S	S	R	R
R	S	S	R	R	R	R
S	S	R	S	R	R	R

$$P = \begin{cases} Sunny & Rainy \\ Sunny & 0 \\ Rainy & 1 \end{cases} \begin{bmatrix} 6/& 5/\\ 11 & /11 \\ 5/& 11/\\ 16 & /16 \end{bmatrix}$$

Gambler's Ruin Example: How to estimate *p*?

- Probability p of winning on any turn
- State Transition Diagram



Probability Transition Matrix

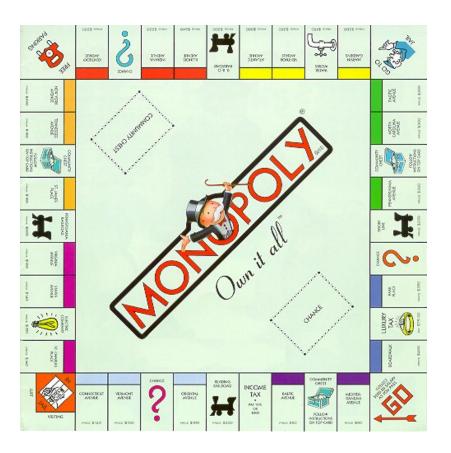
$$\begin{array}{c|ccccc}
0 & 1 & 2 & 3 \\
0 & 1 & 0 & 0 & 0 \\
1 & -p & 0 & p & 0 \\
2 & 0 & 1-p & 0 & p \\
3 & 0 & 0 & 0 & 1
\end{array}$$

Gambler's Ruin Example: How to estimate *p*?

- Here, p can be estimated by analyzing the game.
- For example, a roll of the dice can be analyzed directly.
- A poker game may need more data, and p may depend on who is at the table, and what the stakes are.
- Probability Transition Matrix

$$\begin{array}{c|ccccc}
0 & 1 & 2 & 3 \\
0 & 1 & 0 & 0 & 0 \\
1 & -p & 0 & p & 0 \\
2 & 0 & 1-p & 0 & p \\
3 & 0 & 0 & 0 & 1
\end{array}$$

Monopoly Example



- You roll a pair of dice to advance around the board
- If you land on the "Go To Jail" square, you must stay in jail until you roll doubles or have spent three turns in jail
- Let X_t be the location of your token on the Monopoly board after t dice rolls
 - Can a Markov chain be used to model this game?
 - If not, how could we transform the problem such that we can model the game with a Markov chain?

... more to come ...

Absorbing States

- Recall a state i is an absorbing state if p_{ii}=1
- Example: Gambler's ruin has two absorbing states

$$P = \begin{bmatrix} 0 & 1 & 2 & 3 \\ 1 & 0 & 0 & 0 \\ 1-p & 0 & p & 0 \\ 0 & 1-p & 0 & p \\ 3 & 0 & 0 & 0 & 1 \end{bmatrix}$$

Absorbing States

- If p_{kk}=1 (that is, once the chain visits state k, it remains there forever), then we may want to know:
 the probability of absorption, denoted f_{ik}
- These probabilities are important because they provide answers to questions such as
 - If I start out with a dollar, what is the probability that I will go broke in the gambling game? What is the probability I get \$3?
- Three observations:
 - $-f_{kk}=1$
 - If *i* is recurrent, and *i* is different from k, then $f_{ik}=0$
 - We are most interested in f_{ik} when i is a transient state and k is an absorbing state

Linear Equations for Absorption Probabilities

 As before, we condition on the first transition of the Markov chain to get

$$f_{ik} = \sum_{j=0}^{M} p_{ij} f_{jk}$$
, for $i = 0, 1, ..., M$

with f_{kk} =1 for absorbing state k, and f_{ik} =0 when j is a recurrent state

Another Gambling Example

• Two players A and B, each having \$2, agree to keep betting \$1 at a time until one of them goes broke. The probability that A wins a bet is 1/3. So B wins a bet with probability 2/3. We model the evolution of the number of dollars that A has as a Markov chain. Note that A can have 0,1,2,3, or 4 dollars. The transition probability matrix is given by

States	\$0	\$1	\$2	\$3	\$4
\$0	1	0	0	0	0
\$1	2/3	0	1/3	0	0
\$2	0	2/3	0	1/3	0
\$3	0	0	2/3	0	1/3
\$4	0	0	0	0	1

Another Gambling Example

- Find the probability that A will lose her \$2. That is, find f₂₀.
- Use, f_{kk}=1 for absorbing k, and f_{jk}=0 for recurrent state j

$$f_{20} = \sum_{j=0}^{4} p_{2j} f_{j0} = p_{20} f_{00} + p_{21} f_{10} + p_{22} f_{20} + p_{23} f_{30} + p_{24} f_{40}$$

$$= (0)(1) + (2/3) f_{10} + (0) f_{20} + (1/3) f_{30} + (0)(0)$$

$$= (2/3) f_{10} + (1/3) f_{30}$$

$$f_{10} = \sum_{j=0}^{4} p_{1j} f_{j0} = p_{10} f_{00} + p_{11} f_{10} + p_{12} f_{20} + p_{13} f_{30} + p_{14} f_{40}$$

$$= (2/3)(1) + (0) f_{10} + (1/3) f_{20} + (0) f_{30} + (0)(0)$$

$$= (2/3) + (1/3) f_{20}$$

$$f_{30} = \sum_{j=0}^{4} p_{3j} f_{j0} = p_{30} f_{00} + p_{31} f_{10} + p_{32} f_{20} + p_{33} f_{30} + p_{34} f_{40}$$

$$= (0)(1) + (0) f_{10} + (2/3) f_{20} + (0) f_{30} + (1/3)(0)$$

$$= (2/3) f_{20}$$

Another Gambling Example

Solve three equations, with three unknowns

$$f_{20} = (2/3)f_{10} + (1/3)f_{30}$$

$$f_{10} = (2/3) + (1/3)f_{20}$$

$$f_{30} = (2/3)f_{20}$$

- Substitute, to get $f_{20} = (2/3)((2/3) + (1/3)f_{20}) + (1/3)((2/3)f_{20})$ = $(4/9) + (2/9)f_{20} + (2/9)f_{20}$
- Solving, yields $f_{20} = 4/5$, and $f_{24} = 1 f_{20} = 1/5$ $f_{10} = 14/15$, and $f_{14} = 1/15$

$$f_{30} = 8/15$$
, and $f_{34} = 7/15$

Credit Evaluation Example

- Every month, credit accounts are checked to determine the state of each customer:
 - State 0: Fully paid
 - State 1: Payment on account is 1 month (1-30 days) overdue
 - State 2: Payment on account is 2 months (31-60 days) overdue
 - State 3: Account is written off as bad debt
- The transition probability matrix is given as

$$P = \begin{bmatrix} 0 & 1 & 2 & 3 \\ 1 & 0 & 0 & 0 \\ 0.7 & 0.2 & 0.1 & 0 \\ 2 & 0.5 & 0.1 & 0.2 & 0.2 \\ 3 & 0 & 0 & 0 & 1 \end{bmatrix}$$

Credit Evaluation Example

 What is the probability an account 1 month overdue eventually gets fully paid? Becomes a bad debt?

$$f_{00} = 1$$

$$f_{10} = 0.7f_{00} +0.2f_{10} +0.1f_{20}$$

$$f_{20} = 0.5f_{00} +0.1f_{10} +0.2f_{20} +0.2f_{30}$$

$$f_{30} = 0$$

- Solving, yields $f_{00} = 1$ $f_{10} = 0.968$, and $f_{13} = 0.032$ $f_{20} = 0.746$, and $f_{23} = 0.254$ $f_{30} = 0$
- The probability an account 1 month overdue eventually gets fully paid is $f_{10} = 0.968$
- Becomes a bad debt? $f_{13} = 0.032$