//(real evalue: 0.57; evector: (0.44, 0.56, 0.70) T)

What happens if we introduce a 4th species, and change the matrix to (1.5 -0.5 0 0) (0 1.5 -0.5 0) (0 0 1.5 -0.5) (-0.5 0 0 1.5)

real epairs: (1,(1,1,1,1)); (2,(1,-1,1,-1))

Remark: these matrices are not diagonalisable over the reals (although they are over the complex numbers! See later)

Example - network analysis and centrality:

Suppose we have a network consisting of some nodes and links from nodes to other nodes.

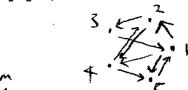
e.g. webpages and weblinks
 journals and citations
 tweeters and "following*

We want to determine which nodes are most "important" in the network.

Idea: an important node is one which is linked to from important nodes.

This seems circular! It is, but it can make sense anyway...

Say we have five webpages, with the following link graph:



Take a large number of people. Put each in front of a computer with a browser loaded to a random one of our five pages. Suppose that they're really just browsing aimlessly, and they click links at random.

Question: where will they end up?

Definition:

A column vector is _stochastic_ if each entry is non-negative and the sum of the entries is 1.

Idea: we can interpret a stochastic vector as giving the probability of being in each of ${\bf N}$ states, when we know we have to be in one of them.

A matrix is _stochastic_ if all its columns are.

Idea: the jth column represents the probabilities of something currently in the jth state changing to each of the states in the next step.

Fact [Perron-Frobenius for stochastic matrices] (not on syllabus):

If % is a stochastic square matrix and for some % all entries of % are

et
$$x_t$$
 be the 5-vector whose extres are the number of people of the various pages of ter following. I links so $x_0 = \begin{pmatrix} 0.2 \\ 0.2 \\ 0.2 \\ 0.2 \end{pmatrix}$

$$x_1 = \begin{pmatrix} 0.4 \\ 30.2 + 202 \\ 0.1 \end{pmatrix}$$

$$x_1 = Ax_0$$

$$x_1 = Ax_0$$

$$x_1 = Ax_0$$

$$x_1 = Ax_0$$

$$x_2 = Ax_1 = A^2x_0$$

$$x_2 = Ax_1 = A^2x_0$$

$$x_3 = A^2x_0$$

$$x_4 = A^4x_0$$

$$x_4 = A^4x_0$$

$$x_4 = A^4x_0$$

$$x_5 = A^4x_0$$

$$x_1 = A^2x_0$$

$$x_2 = Ax_1 = A^2x_0$$

$$x_3 = A^2x_0$$

$$x_4 = A^4x_0$$

$$x_5 = A^4x_0$$

$$x_6 = x_0$$

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$$x$$

TY FIND YY

positive, then it has a unique stochastic eigenvector v_1, which has f e-value 1, and for any stochastic f x $\lim_{n \to \infty} 1 = v_1$ Stochastic diagonalisable example: (0 1 2/3) (1/3 0 1/3} (2/3 0 I-Alx Ce.g. blinking links!) evalues and evectors are: 1, -2/3, -1/3 I A Court corresponding evectors (9, 5, 6) evalue 1 (1, 0, −1)[⊤] evalue (1, 1, -2)⁷ evalue So stochastic 1-e-vector is $v_1 = (9/20, 5/20, 6/20)^T$ So this is the limit behaviour. Any stochastic e-vector v can be written as $\mathbf{v} = \mathbf{v}_1 + \mathbf{x}_2 + \mathbf{x}_3$ where x_2 is a *** - evector and x_3 is a *** - evector. e.g. $(1/3,1/3,1/3) = v_1 + (1/12, 1/12, -1/6)^T + (1/5, 0, -1/5)^T$ So $A^n v = ...$ Another: (1/4 1/3 1/3) (/1/4 1/3 1/6) $D=d_{1}^{2}ag(1,1/6,-1/12)$ 1/2 1/3 1/2) evalues and evectors are: with corresponding evector (2/3/, 1/2, 1)evalue 1 -1 , 1) evalue 1/6 **(-3/2, 1/2,** 1)/ evalựe **-1/12** So stochastic 1-e-vector is (4/13 3/1/3 6/13) ^T So this is the limit behaviour.