Supplemental problems: §5.6

1. Suppose the internet has four pages in the following manner. Arrows represent links from one page towards another. For example, page 1 links to page 4 but not vice versa.



- a) Write the importance matrix and the Google matrix for this internet using damping constant p = 0.15. You don't need to simplify the Google matrix.
- **b)** The steady-state vector for the Google matrix is (approximately)

$$\begin{pmatrix} 0.23 \\ 0.23 \\ 0.23 \\ 0.31 \end{pmatrix}.$$

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What is the top-ranked page?

Solution.

(a) The importance matrix is

$$A = \begin{pmatrix} 0 & 0 & 1 & 0 \\ 1/3 & 0 & 0 & 1/2 \\ 1/3 & 0 & 0 & 1/2 \\ 1/3 & 1 & 0 & 0 \end{pmatrix}$$

The Google matrix is

- (b) From the steady-state vector we see page 4 has the highest rank.
- **2.** The companies X, Y, and Z fight for customers. This year, company X has 40 customers, Company Y has 15 customers, and Z has 20 customers. Each year, the following changes occur:
 - X keeps 75% of its customers, while losing 15% to Y and 10% to Z.
 - Y keeps 60% of its customers, while losing 5% to X and 35% to Z.

• Z keeps 65% of its customers, while losing 15% to X and 20% to Y.

Write a stochastic matrix A and a vector x so that Ax will give the number of customers for firms X, Y, and Z (respectively) after one year. You do not need to compute Ax.

Solution.

$$A = \begin{pmatrix} 0.75 & 0.05 & 0.15 \\ 0.15 & 0.6 & 0.20 \\ 0.1 & 0.35 & 0.65 \end{pmatrix} \qquad x = \begin{pmatrix} 40 \\ 15 \\ 20 \end{pmatrix}.$$

3. Suppose *p* and *q* are real numbers on the open interval (0, 1), and

$$A = \begin{pmatrix} p & 1-q \\ 1-p & q \end{pmatrix}$$

(1) Is A a positive stochastic matrix? Why?

(2) Does *A* have unique steady state vector? Why?

(3) Without computation, give an eigenvalue of A.

(4) Compute the steady-state vector of *A*.

Solution.

(1) Yes: columns sum to 1, all entries strictly positive

- (2) Yes: A is a positive stochastic matrix, so the Perron-Frobenius theorem applies.
- (3) $\lambda = 1$
- (4) Solving (A I)v = 0 and scaling v to get the steady-state vector w, we get

$$w = \frac{1}{2-p-q} \begin{pmatrix} 1-q\\ 1-p \end{pmatrix}.$$

Supplemental problems: Chapter 6

- **1.** True or false. If the statement is always true, answer true. Otherwise, answer false. Justify your answer.
 - a) Suppose $W = \text{Span}\{w\}$ for some vector $w \neq 0$, and suppose v is a vector orthogonal to w. Then the orthogonal projection of v onto W is the zero vector.
 - **b)** Suppose *W* is a subspace of \mathbb{R}^n and *x* is a vector in \mathbb{R}^n . If *x* is not in *W*, then $x x_W$ is not zero.
 - c) Suppose W is a subspace of \mathbb{R}^n and x is in both W and W^{\perp} . Then x = 0.
 - **d)** Suppose \hat{x} is a least squares solution to Ax = b. Then \hat{x} is the closest vector to *b* in the column space of *A*.

Solution.

- a) True. Since $v \in W^{\perp}$, its projection onto W is zero.
- **b)** True. If x is not in W then $x \neq x_W$, so $x x_W$ is not zero.
- c) True. Since x is in W and W^{\perp} it is orthogonal to itself, so $||x||^2 = x \cdot x = 0$. The length of x is zero, which means every entry of x is zero, hence x = 0.
- **d)** False: $A\hat{x}$ is the closest vector to *b* in Col *A*.

2. Let
$$W = \text{Span}\{v_1, v_2\}$$
, where $v_1 = \begin{pmatrix} -1 \\ 2 \\ 1 \end{pmatrix}$ and $v_2 = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$.

a) Find the closest point *w* in *W* to $x = \begin{pmatrix} 0 \\ 14 \\ -4 \end{pmatrix}$.

Let
$$A = \begin{pmatrix} -1 & 1 \\ 2 & 2 \\ 1 & 3 \end{pmatrix}$$
. We solve $A^{T}Av = A^{T}x$.
 $A^{T}A = \begin{pmatrix} 6 & 6 \\ 6 & 14 \end{pmatrix} \qquad A^{T}\begin{pmatrix} 0 \\ 14 \\ -4 \end{pmatrix} = \begin{pmatrix} 24 \\ 16 \end{pmatrix}$.
We find $\begin{pmatrix} 6 & 6 \\ 6 & 14 \\ 16 \end{pmatrix} \xrightarrow{RREF} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -1 \end{pmatrix}$, so $v = \begin{pmatrix} 5 \\ -1 \end{pmatrix}$ and therefore
 $w = Av = \begin{pmatrix} -1 & 1 \\ 2 & 2 \\ 1 & 3 \end{pmatrix} \begin{pmatrix} 5 \\ -1 \end{pmatrix} = \begin{pmatrix} -6 \\ 8 \\ 2 \end{pmatrix}$.

b) Find the distance from *w* to $\begin{pmatrix} 0\\ 14\\ -4 \end{pmatrix}$.

$$||x - w|| = \left| \left| \begin{pmatrix} 0 \\ 14 \\ -4 \end{pmatrix} - \begin{pmatrix} -6 \\ 8 \\ 2 \end{pmatrix} \right| = \left| \left| \begin{pmatrix} 6 \\ 6 \\ -6 \end{pmatrix} \right| = \sqrt{36 + 36 + 36} = \sqrt{108} = 6\sqrt{3}.$$

c) Find the standard matrix for the orthogonal projection onto $\text{Span}\{v_1\}$.

$$B = \frac{1}{\nu_1 \cdot \nu_1} \nu_1 \nu_1^T = \frac{1}{(-1)^2 + 2^2 + 1^2} \begin{pmatrix} -1 \\ 2 \\ 1 \end{pmatrix} \begin{pmatrix} -1 & 2 & 1 \end{pmatrix} = \frac{1}{6} \begin{pmatrix} 1 & -2 & -1 \\ -2 & 4 & 2 \\ -1 & 2 & 1 \end{pmatrix}$$

d) Find the standard matrix for the orthogonal projection onto *W*.

Let $A = \begin{pmatrix} -1 & 1 \\ 2 & 2 \\ 1 & 3 \end{pmatrix}$. Since the columns of *A* are linearly independent, our projection matrix is $A(A^TA)^{-1}A^T$. We already computed A^TA in part (a), so our matrix is $\begin{pmatrix} -1 & 1 \\ 1 \end{pmatrix} (f_1 - f_2)^{-1} (f_1 - f_2) = f_2 (f_1 - f_2)^{-1} (f$

$$A(A^{T}A)^{-1}A^{T} = \begin{pmatrix} -1 & 1\\ 2 & 2\\ 1 & 3 \end{pmatrix} \begin{pmatrix} 6 & 6\\ 6 & 14 \end{pmatrix}^{-1} \begin{pmatrix} -1 & 2 & 1\\ 1 & 2 & 3 \end{pmatrix} = \begin{pmatrix} -1 & 1\\ 2 & 2\\ 1 & 3 \end{pmatrix} \begin{pmatrix} 6 & 6\\ 6 & 14 \end{pmatrix}^{-1} \begin{pmatrix} -1 & 2 & 1\\ 1 & 2 & 3 \end{pmatrix}$$
$$= \frac{1}{48} \begin{pmatrix} -1 & 1\\ 2 & 2\\ 1 & 3 \end{pmatrix} \begin{pmatrix} 14 & -6\\ -6 & 6 \end{pmatrix} \begin{pmatrix} -1 & 2 & 1\\ 1 & 2 & 3 \end{pmatrix} = \frac{1}{3} \begin{pmatrix} 2 & -1 & 1\\ -1 & 2 & 1\\ 1 & 1 & 2 \end{pmatrix}.$$

3. Find the least-squares line y = Mx + B that approximates the data points

$$(-2, -11), (0, -2), (4, 2).$$

Solution.

If there were a line through the three data points, we would have:

$$(x = -2) \qquad B + M(-2) = -11$$
$$(x = 0) \qquad B + M(0) = -2$$
$$(x = 4) \qquad B + M(4) = 2.$$
This is the matrix equation
$$\begin{pmatrix} 1 & -2 \\ 1 & 0 \\ 1 & 4 \end{pmatrix} \begin{pmatrix} B \\ M \end{pmatrix} = \begin{pmatrix} -11 \\ -2 \\ 2 \end{pmatrix}.$$

Thus, we are solving the least-squares problem to Av = b, where

$$A = \begin{pmatrix} 1 & -2 \\ 1 & 0 \\ 1 & 4 \end{pmatrix} \qquad b = \begin{pmatrix} -11 \\ -2 \\ 2 \end{pmatrix}.$$

We solve $A^{T}A\hat{x} = A^{T}b$, where $\hat{x} = \begin{pmatrix} B \\ M \end{pmatrix}.$
$$A^{T}A = \begin{pmatrix} 1 & 1 & 1 \\ -2 & 0 & 4 \end{pmatrix} \begin{pmatrix} 1 & -2 \\ 1 & 0 \\ 1 & 4 \end{pmatrix} = \begin{pmatrix} 3 & 2 \\ 2 & 20 \end{pmatrix},$$
$$A^{T}b = \begin{pmatrix} 1 & 1 & 1 \\ -2 & 0 & 4 \end{pmatrix} \begin{pmatrix} -11 \\ -2 \\ 2 \end{pmatrix} = \begin{pmatrix} -11 \\ 30 \end{pmatrix}.$$
$$\begin{pmatrix} 3 & 2 \\ 2 & 20 \\ 30 \end{pmatrix} \xrightarrow{R_{1} \leftrightarrow R_{2}} \begin{pmatrix} 2 & 20 \\ 3 & 2 \\ -11 \end{pmatrix} \xrightarrow{R_{2} = R_{2} - \frac{3R_{1}}{2}} \begin{pmatrix} 1 & 10 \\ 0 & -28 \\ -56 \end{pmatrix} \xrightarrow{R_{2} = -\frac{R_{2}}{R_{1} = R_{1} - 10R_{2}}} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -2 \end{pmatrix}.$$
So $\hat{x} = \begin{pmatrix} -5 \\ 2 \end{pmatrix}$. In other words, $y = -5 + 2x$, or $y = 2x - 5$.