

Computational Data Mining

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- The eigenvectors of A don't change direction when you multiply them by A
- Ax is on the same line as the input vector x

x = **eigenvector of A**
 λ = **eigenvalue of A**

$$Ax = \lambda x$$

$$A^2x = \lambda^2x.$$

how eigenvalues and eigenvectors are useful

$$A^2 \mathbf{x} = \lambda^2 \mathbf{x}.$$

$$A^k \mathbf{x} = \lambda^k \mathbf{x} \text{ for all } k = 1, 2, 3, \dots \text{ And } A^{-1} \mathbf{x} = \frac{1}{\lambda} \mathbf{x} \text{ provided } \lambda \neq 0.$$

- Most n by n matrices have n independent eigenvectors \mathbf{x}_1 to \mathbf{x}_n with n different eigenvalues λ_1 to λ_n
- every n -dimensional vector \mathbf{v} will be a combination of the eigenvectors

Every \mathbf{v}	$\mathbf{v} = c_1 \mathbf{x}_1 + \cdots + c_n \mathbf{x}_n$
Multiply by A	$A\mathbf{v} = c_1 \lambda_1 \mathbf{x}_1 + \cdots + c_n \lambda_n \mathbf{x}_n$
Multiply by A^k	$A^k \mathbf{v} = c_1 \lambda_1^k \mathbf{x}_1 + \cdots + c_n \lambda_n^k \mathbf{x}_n$

how eigenvalues and eigenvectors are useful

Every v	$v = c_1 \mathbf{x}_1 + \cdots + c_n \mathbf{x}_n$
Multiply by A	$A\mathbf{v} = c_1 \lambda_1 \mathbf{x}_1 + \cdots + c_n \lambda_n \mathbf{x}_n$
Multiply by A^k	$A^k \mathbf{v} = c_1 \lambda_1^k \mathbf{x}_1 + \cdots + c_n \lambda_n^k \mathbf{x}_n$

If $|\lambda_1| > 1$ then the component $c_1 \lambda_1^n \mathbf{x}_1$ will grow as n increases

If $|\lambda_2| < 1$ then that component $c_2 \lambda_2^n \mathbf{x}_2$ will steadily disappear

The powers of Q don't grow or decay

Eigenvalues and eigenvectors properties

Example $S = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$ has eigenvectors $S \begin{bmatrix} 1 \\ 1 \end{bmatrix} = 3 \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ and $S \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$

(Trace of S) The sum $\lambda_1 + \lambda_2 = 3 + 1$ equals the diagonal sum $2 + 2 = 4$

(Determinant) The product $\lambda_1 \lambda_2 = (3)(1) = 3$ equals the determinant $4 - 1$

(Real eigenvalues) Symmetric matrices $S = S^T$ always have real eigenvalues

(Orthogonal eigenvectors) If $\lambda_1 \neq \lambda_2$ then $\mathbf{x}_1 \cdot \mathbf{x}_2 = 0$.

Here $(1, 1) \cdot (1, -1) = 0$

Computing the Eigenvalues (by hand)

$Ax = \lambda x$ is the same as $(A - \lambda I)x = 0$

$A - \lambda I$ is not invertible:

determinant of $A - \lambda I$ must be zero

equation $\det(A - \lambda I) = 0$ has n roots

$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ has two eigenvalues

Question

Find the eigenvalues and eigenvectors of $A = \begin{bmatrix} 8 & 3 \\ 2 & 7 \end{bmatrix}$

If A is shifted to $A + sI$, what happens to the x 's and λ 's?

Similar Matrices

- For every invertible matrix B :
 - eigenvalues of BAB^{-1} are the same as the eigenvalues of A

$$\text{If } A\mathbf{x} = \lambda\mathbf{x} \text{ then } (BAB^{-1})(B\mathbf{x}) = BA\mathbf{x} = B\lambda\mathbf{x} = \lambda(B\mathbf{x}).$$

The matrices BAB^{-1} (for every invertible B) are “similar” to A : same eigenvalues.

Question

- Show that the eigenvalue of BAB^{-1} is equal to the eigenvalue of A
- Show that the eigenvalue of AB equals the eigenvalue of BA

The eigenvalues of any triangular matrix $\begin{bmatrix} a & b \\ 0 & d \end{bmatrix}$ are $\lambda_1 = a$ and $\lambda_2 = d$.

Diagonalizing a Matrix

- A has a full set of n independent eigenvectors
- Put those eigenvectors $\mathbf{x}_1, \dots, \mathbf{x}_n$ into an invertible matrix X
- Multiply AX column by column

$$A \begin{bmatrix} \mathbf{x}_1 & \dots & \mathbf{x}_n \end{bmatrix} = \begin{bmatrix} A\mathbf{x}_1 & \dots & A\mathbf{x}_n \end{bmatrix} = \begin{bmatrix} \lambda_1\mathbf{x}_1 & \dots & \lambda_n\mathbf{x}_n \end{bmatrix} = \begin{bmatrix} \mathbf{x}_1 & \dots & \mathbf{x}_n \end{bmatrix} \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix}.$$

Λ = diagonal eigenvalue matrix

$$A = X\Lambda X^{-1}$$

X = invertible eigenvector matrix

$$A^2 = (X\Lambda X^{-1})(X\Lambda X^{-1}) = X\Lambda^2 X^{-1}$$

$$A = \begin{bmatrix} 0.8 & 0.3 \\ 0.2 & 0.7 \end{bmatrix} \quad \lambda_1 = 1 \text{ and } \lambda_2 = \frac{1}{2}$$

- A is a Markov matrix, with columns adding to 1

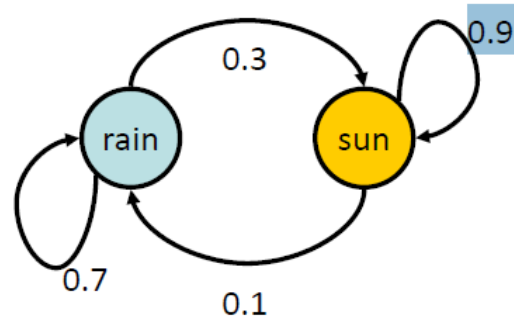
$$A^k \mathbf{v} = c_1 (1)^k \mathbf{x}_1 + c_2 \left(\frac{1}{2}\right)^k \mathbf{x}_2$$

As k increases, $A^k \mathbf{v}$ approaches $c_1 \mathbf{x}_1 = \text{steady state}$

- We can follow each eigenvector separately. Its growth or decay depends on the eigenvalue

- CPT $P(X_t | X_{t-1})$:

X_{t-1}	X_t	$P(X_t X_{t-1})$
sun	sun	0.9
sun	rain	0.1
rain	sun	0.3
rain	rain	0.7



- What is the probability distribution after one step?

$$P(X_2 = sun) = \sum_{x_1} P(x_1, X_2 = sun) = \sum_{x_1} P(X_2 = sun | x_1) P(x_1)$$

From initial observation of sun

$$\begin{array}{ccccccc} \left\langle \begin{array}{c} 1.0 \\ 0.0 \end{array} \right\rangle & \left\langle \begin{array}{c} 0.9 \\ 0.1 \end{array} \right\rangle & \left\langle \begin{array}{c} 0.84 \\ 0.16 \end{array} \right\rangle & \left\langle \begin{array}{c} 0.804 \\ 0.196 \end{array} \right\rangle & \longrightarrow & \left\langle \begin{array}{c} 0.75 \\ 0.25 \end{array} \right\rangle \\ P(X_1) & P(X_2) & P(X_3) & P(X_4) & & P(X_\infty) \end{array}$$

From initial observation of rain

$$\begin{array}{ccccccc} \left\langle \begin{array}{c} 0.0 \\ 1.0 \end{array} \right\rangle & \left\langle \begin{array}{c} 0.3 \\ 0.7 \end{array} \right\rangle & \left\langle \begin{array}{c} 0.48 \\ 0.52 \end{array} \right\rangle & \left\langle \begin{array}{c} 0.588 \\ 0.412 \end{array} \right\rangle & \longrightarrow & \left\langle \begin{array}{c} 0.75 \\ 0.25 \end{array} \right\rangle \\ P(X_1) & P(X_2) & P(X_3) & P(X_4) & & P(X_\infty) \end{array}$$

From yet another initial distribution $P(X_1)$:

$$\begin{array}{ccc} \left\langle \begin{array}{c} p \\ 1-p \end{array} \right\rangle & \dots & \longrightarrow \left\langle \begin{array}{c} 0.75 \\ 0.25 \end{array} \right\rangle \\ P(X_1) & & P(X_\infty) \end{array}$$

$$\begin{aligned} P(x_t) &= \sum_{x_{t-1}} P(x_{t-1}, x_t) \\ &= \sum_{x_{t-1}} P(x_t | x_{t-1}) P(x_{t-1}) \end{aligned}$$

Forward simulation

Factor these two matrices into $A = X\Lambda X^{-1}$:

$$A = \begin{bmatrix} 1 & 2 \\ 0 & 3 \end{bmatrix} \quad \text{and} \quad A = \begin{bmatrix} 1 & 1 \\ 3 & 3 \end{bmatrix}.$$

If $A = X\Lambda X^{-1}$ then $A^3 = (\quad)(\quad)(\quad)$ and $A^{-1} = (\quad)(\quad)(\quad)$.