

Eigenvalues and Search Engines

Alessandro Panconesi
Informatica, La Sapienza
via Salaria 113, piano III
00198 Roma, Italy

These notes complement Kleinberg's paper by giving a full proof of the convergence of the method.

1 The convergence of Hits

Let A be the adjacency matrix of the hubs/authorities graph, and let

$$B := A^T A$$

Exercise 1 *Show that B is symmetric.*

Starting from a unit vector \hat{a}_0 the Hits algorithm computes the normalized sequences

$$\hat{h}_k = \frac{A\hat{a}_{k-1}}{\|A\hat{a}_{k-1}\|}.$$

and

$$\hat{a}_k = \frac{A^T \hat{h}_k}{\|A^T \hat{h}_k\|}$$

so that

$$\hat{a}_k = \frac{B\hat{a}_{k-1}}{\|B\hat{a}_{k-1}\|}$$

Exercise 2 *Prove or disprove: $AA^T = A^T A$.*

We want to show that, unless we are very unlucky with our choice of \hat{a}_0 , the sequence \hat{a}_k *always* converge (rather quickly) to the principal eigenvector of B of unit norm.

We will work with the the n eigenvalues of B and consider them as an *ordered* sequence

$$\lambda_1, \lambda_2, \dots, \lambda_n$$

where

$$|\lambda_1| \geq |\lambda_2| \geq \dots \geq |\lambda_n|$$

and will denote the corresponding eigenvectors as

$$b_1, b_2, \dots, b_n.$$

These eigenvectors can be assumed to be of *unit norm*.

Exercise 3 *Show that if x is an eigenvector of a matrix A then so is cx , for any c .*

Exercise 4 Show that eigenvectors corresponding to different eigenvalues are linearly independent.

Exercise 5 Show that eigenvectors corresponding to different eigenvalues of a symmetric matrix are orthonormal.

Exercise 6 Show that the eigenvalues of a symmetric matrix are real.

In the rest of the section we shall prove the following.

Theorem 1 If the eigenvalues of B are all different, then \hat{a}_k converges to b_1 , provided that \hat{a}_0 and b_1 are not orthogonal.

In the next section we will show the same convergence under the weaker assumption $|\lambda_1| > |\lambda_2|$.

Consider the non normalized sequence

$$a_k = Ba_{k-1} = B^k a_0.$$

where $a_0 = \hat{a}_0$. By Exercise 4 the n eigenvalues form a basis. Thus,

$$a_0 = \sum_{i=1}^n c_i b_i$$

for some vector $c := (c_1, \dots, c_n)$. Let us start by unfolding a_1 ,

$$a_1 = Ba_0 = B \sum_{i=1}^n c_i b_i = \sum_{i=1}^n c_i B b_i = \sum_{i=1}^n c_i \lambda_i b_i$$

and then a_2 ,

$$a_2 = Ba_1 = B \sum_{i=1}^n c_i \lambda_i b_i = \sum_{i=1}^n c_i \lambda_i B b_i = \sum_{i=1}^n c_i \lambda_i^2 b_i$$

Thus in general, by a trivial induction,

$$a_k = \sum_{i=1}^n c_i \lambda_i^k b_i.$$

This vector tends to a vector parallel to b_1 . To see this, consider the vector

$$v_k := \frac{a_k}{(\lambda_1)^k} = c_1 b_1 + \sum_{i=2}^n c_i \left(\frac{\lambda_i}{\lambda_1} \right)^k b_i.$$

By our assumption that a_0 is not perpendicular to b_1 , $c_1 \neq 0$. Therefore v_k tends to $c_1 b_1$ when k goes to infinity.

Note: the above is nothing else than the well-known power method for computing the eigenvalues of a matrix.

Thus the non normalized sequence converges to a vector parallel to b_1 . Let us show now that the normalized sequence converges to b_1 . The normalized sequence is defined as

$$\begin{aligned} \hat{a}_0 &= a_0 \\ \hat{a}_k &= \frac{B\hat{a}_{k-1}}{\|B\hat{a}_{k-1}\|} \end{aligned}$$

We show by induction that

$$\hat{a}_k = \frac{a_k}{\|a_k\|}$$

i.e., \hat{a}_k is a unit vector parallel to a_k . The basis holds trivially. For the inductive step,

$$\hat{a}_k = \frac{B\hat{a}_{k-1}}{\|B\hat{a}_{k-1}\|} = \frac{Ba_{k-1}}{\|a_{k-1}\| \|B\hat{a}_{k-1}\|} = \frac{Ba_{k-1} \|a_{k-1}\|}{\|a_{k-1}\| \|Ba_{k-1}\|} = \frac{Ba_{k-1}}{\|Ba_{k-1}\|} = \frac{a_k}{\|a_k\|}.$$

2 Extending the method

We now show a more general form of Theorem 1, namely,

Theorem 2 *If $|\lambda_1| > |\lambda_2|$ then a_k converges to b_1 , provided that a_0 and b_1 are not orthogonal.*

This result too is well-known in computational linear algebra. To establish the result we will show that under the current hypothesis we can find a set of eigenvectors that form an orthonormal basis. Then the result follows from the analysis of the previous section.

We will use a well-known result of linear algebra that says that *any* square matrix can be decomposed as the product of an upper-triangular matrix with a unitary matrix and its inverse.

Definition 1 *A matrix U is unitary if (a) its columns are orthogonal, unit vectors and (b) $U^T = U^{-1}$, i.e., $U^T U = I$.*

Exercise 7 *Show that the product of unitary matrices is unitary.*

Exercise 8 *Let A and B be square matrices. Show that $(AB)^T = B^T A^T$. What if the matrices are not square?*

Lemma 1 (Schur's Normal Form) *Any square matrix A can be decomposed as*

$$A = UTU^T$$

where U is unitary and T is upper-triangular.

Proof. The proof is by induction on n , the number of rows of A . The base case ($n = 1$) is trivial. For $n > 1$, let x_1 be the unit-norm eigenvector corresponding to λ_1 (every non null matrix has at least one eigenvalue). Let y_2, \dots, y_n be unit vectors such that x_1, y_2, \dots, y_n form a basis. The matrix

$$U_1 := [x_1 | y_2 | \dots | y_n]$$

is unitary. Consider the matrix

$$B := U_1^T A U_1.$$

This matrix is of the form

$$B = \begin{bmatrix} \lambda_1 & v_1 \\ z & A_1 \end{bmatrix}$$

where z is a vector of $n - 1$ 0's, v_1 is some real vector with $n - 1$ components, and A_1 is a $(n - 1) \times (n - 1)$ real matrix. We now apply the induction hypothesis to the matrix A_1 obtaining

$$A_1 = UTU^T$$

where U is unitary and T upper triangular. If we define

$$U_2 := \begin{bmatrix} 1 & z^T \\ z & U \end{bmatrix}$$

where z is a vector of $n - 1$ zeroes, then U_2 is unitary. Now,

$$A = U_1 B U_1^T = U_1 U_2 \begin{bmatrix} \lambda_1 & v'_1 \\ 0 & T \end{bmatrix} U_2^T U_1^T.$$

The claim follows from Ex. 7. ⊛

Theorem 3 *Let A be a symmetric matrix. Then we can find n orthogonal eigenvectors of unit norm.*

Proof. By Lemma 1, $A = UTU^T$, but A is symmetric hence,

$$UTU^T = A = A^T = (UTU^T)^T = UT^T U^T.$$

Thus $T = T^T$, but this implies that T is a diagonal matrix. Since U is unitary, T then has the n eigenvalues of A along its diagonal. ⊛