

Hubs and Authorities

HITS (Hyperlink-Induced Topic Search)

- It is developed by John Kleinberg in 1999.
- Every web page is assigned two scores, one is called Hub score and other is its authority score.

→ For any query, we compute two ranked lists i.e. hub score list & authority score list.

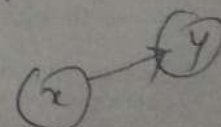
→ A good hub page is one which points to many good authorities & a good authority page is one which is pointed to by many good hub pages.



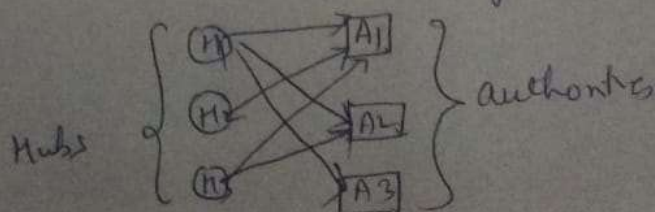
→ Let us take a subset of webgraph and then compute its hub & authority score.

for a webpage x , x has a hyperlink that points to y .

$$\text{Hub score of node } x = h(x) = \sum_{x \rightarrow y} a(y)$$



$$\text{authority score of node } y = a(y) = \sum_{y \leftarrow x} h(x)$$



Algorithm :-

1. Assign each node an authority and hub score of webgraph.

$$\therefore \text{initially } h(x) = a(x) = 1/N$$

2. Apply the Authority update rule: Each

node's authority score is the sum of indegree hub scores of each node that it points to.

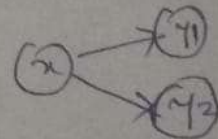
$$a(x) = \sum_{y \rightarrow x} h(y)$$



3. Apply the Hub update rule: Each node's

hub score is sum of authority score of outdegree each node that it points to.

$$h(x) = \sum_{x \rightarrow y} a(y)$$



4. Normalize authority and hub scores:

$$auth(i^*) = \frac{auth(i^*)}{\sum_{i^* \in H} auth(i^*)}$$

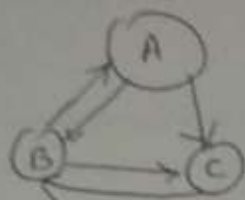
$$hub(j^*) = \frac{hub(j^*)}{\sum_{i^* \in H} hub(i^*)}$$

5. Repeat 2, 3, 4 until convergence condition is met. i.e. score becomes constant.

Example :-

Step 1

Let $a_{10} = h_{10} = 1$



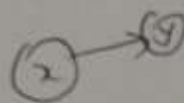
Step 2

	old A	old H	old New A	New H
A	1	1	1 (B)	2 (B & C)
B	1	1	2 (A & C)	2 (A & C)
C	1	1	2 (A & B)	1 (B)

$$\text{Authority (indegree)}(x) = \sum_{y \rightarrow x} h(y)$$



$$\text{Hub}(x) = \sum_{x \rightarrow y} a(y)$$



Normalize :- $\text{Authority}(A) = \frac{A(A)}{A(A) + A(B) + A(C)} = \frac{1}{5}$

Similarly for other nodes.

	new A	new H	new A	new H
A	$\frac{1}{5}$	$\frac{2}{5}$	$\frac{H(B)}{2/5} = \frac{1}{5}$	$\frac{A(B) + A(C)}{1/5} = \frac{4}{5}$
B	$\frac{2}{5}$	$\frac{2}{5}$	$\frac{H(A) + H(C)}{2/5} = \frac{3}{5}$	$\frac{A(A) + A(C)}{2/5} = \frac{3}{5}$
C	$\frac{2}{5}$	$\frac{1}{5}$	$\frac{H(A) + H(B)}{1/5} = \frac{4}{5}$	$\frac{A(B)}{2/5} = \frac{1}{5}$

and now normalize and do it iteratively till stopping condition is met.

Matrix Notation :-

Let vector $a = (a_1, a_2, \dots, a_n)$ be a authority vector

$h = (h_1, h_2, \dots, h_n)$ be a hub vector

n = no of nodes in the web graph.

Step 1: initialize $a_i = h_i = \frac{1}{\sqrt{n}}$ or
 $a_i = h_i = 1$

Step 2: Repeat until stopping condition is met (or it converges)

(i) $h = A \cdot a \Rightarrow h = (A A^T) \cdot h$

(ii) $a = A^T \cdot h \Rightarrow a = (A^T A) \cdot a$

(iii) Normalize a & h . (Use L_1 or L_2 norm)

Step 3:- If convergence criterion is met, stop else go to step 2.

L_2 Norm n :- $\sum_{i \in N} (a_i)^2 = 1$ & $\sum_{i \in N} (h_i)^2 = 1$

Hence, at every iteration, normalize each value of a & h by

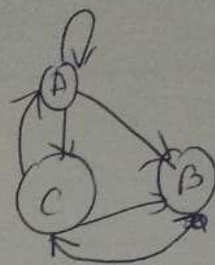
$$a_i^{(t)} = \frac{a_i^{(t)}}{\sqrt{\sum_{j=1}^n a_j^{(t)2}}} \quad \text{and} \quad h_i^{(t)} = \frac{h_i^{(t)}}{\sqrt{\sum_{j=1}^n h_j^{(t)2}}}$$

$$a_i^{(t)} = \frac{a_i^{(t)}}{\sqrt{a_1^2 + a_2^2 + \dots + a_n^2}} \quad \text{and} \quad h_i^{(t)} = \frac{h_i^{(t)}}{\sqrt{h_1^2 + h_2^2 + \dots + h_n^2}}$$

Example :- $[A] = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix}$

$$a = (A^T \times A \cdot a)$$

$$h = (A \times A^T \cdot h)$$



$$A^T A = \begin{bmatrix} 2 & 2 & 1 \\ 2 & 2 & 1 \\ 1 & 1 & 2 \end{bmatrix}$$

$$A A^T = \begin{bmatrix} 3 & 1 & 2 \\ 1 & 1 & 0 \\ 2 & 0 & 2 \end{bmatrix}$$

$$a = [A^T A] \cdot a = \begin{bmatrix} 2 & 2 & 1 \\ 2 & 2 & 1 \\ 1 & 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 5 \\ 5 \\ 4 \end{bmatrix}$$

$$h = [A A^T] h = \begin{bmatrix} 3 & 1 & 2 \\ 1 & 1 & 0 \\ 2 & 0 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 6 \\ 2 \\ 4 \end{bmatrix}$$

next iteration:

$$a = \begin{bmatrix} 24 \\ 24 \\ 18 \end{bmatrix}$$

$$h = \begin{bmatrix} 28 \\ 8 \\ 20 \end{bmatrix}$$

and so on till it converges.