ENGR 102 PROGRAMMING PRACTICE

WEEK 12



Searching & Ranking



Setting Up Database

Four dictionaries:

- urllist is the list of URLs that have been indexed.
 {url: outgoing_link_count}
- wordlocation is a list of the locations of words in the documents.
 {word: {url: [loc1, loc2, ..., locN]}}
- link stores two URL IDs, indicating a link from one page to another. {tourl: {fromUrl: None}}
- **linkwords** store words that are included in a link. {word: [(urlFrom I, urlTo I), ..., (urlFromN, urlToN)]}



Recording links

```
# Add a link between two pages
def addlinkref(self, urlFrom, urlTo, linkText):
    fromUrl = smart str(urlFrom)
    toUrl = smart str(urlTo)
    if fromUrl == toUrl: return False
    self.link.setdefault(toUrl, {})
    self.link[toUrl][fromUrl] = None
    words = self.separatewords(linkText)
    for word in words:
        word = smart str(word)
        if word in ignorewords: continue
        self.linkwords.setdefault(word, [])
        self.linkwords[word].append((fromUrl, toUrl))
    return True
```



Crawling pages

```
import mysearchengine
pagelist = ['http://sehir.edu.tr']
dbtables = { 'urllist': 'urllist.db',
            'wordlocation': 'wordlocation.db',
            'link': 'link.db',
            'linkwords': 'linkwords.db'}
crawler = mysearchengine.crawler(dbtables)
crawler.createindextables()
crawler.crawl (pagelist)
```



Search Engine

- I. Crawl to collect documents.
- 2. Index to improve search.
- 3. Query for a select set of documents.



Querying

return results, words

```
def getmatchingpages(self, q):
    results = {}
    # Split the words by spaces
    words = [smart str(word).lower for word in q.split()]
    if words[0] not in self.wordlocation:
        return results, words
    url set = set(self.wordlocation[words[0]].keys())
    for word in words[1:]:
        if word not in self.wordlocation:
            return results, words
        url set = url set.intersection(self.wordlocation[word].keys())
    for url in url set:
        results[url] = []
        for word in words:
            results [url].append(self.wordlocation[word][url])
```

Querying

- We've managed to retrieve pages that match the queries.
- **Problem**: In a large set of pages, you would be stuck sifting through <u>a lot of irrelevant content</u> for any mention of each of the query terms in order to find the pages that are **really** related to your search.



Ranking

```
import mysearchengine
pagelist = ['http://sehir.edu.tr']
dbtables = { 'urllist': 'urllist.db',
            'wordlocation': 'wordlocation.db',
            'link': 'link.db',
            'linkwords': 'linkwords.db'}
searcher = mysearchengine.searcher(dbtables)
# no ranking:
searcher.query('career')
# with ranking:
searcher.query('career', searcher.ALL)
```



Ranking

- Content-based Ranking
- Inbound-link Ranking



Ranking measures Content-based

- Word frequency: the number of times the words in the query appear in the document can help determine how relevant the document is.
- <u>Document location:</u> the main subject of a document will probably appear near the beginning of the document.
- **Word distance:** if there are multiple words in the query, they should appear close together in the document.



Word Frequency

- The word frequency metric scores a page based on how many times the words in the query appear on that page.
- Search for "python"
 - a page about Python (or pythons) with many mentions of the word python.

VS.

- a page about a musician that has a pet python.



Word Frequency

```
def frequencyscore(self, results):
    counts = {}
    for url in results:
        score = 1
        for wordlocations in results[url]:
            score *= len(wordlocations)
        counts[url] = score
    return self.normalizescores(counts, smallIsBetter=False)
```



Normalization

- In order to compare the results from different scoring methods, we need a way to normalize them.
- The normalization function: each score is scaled according to how close it is to the best result, which will always have a score of I.

normalizescores (scores, smallIsBetter=False)



Score Computation

```
def getscoredlist(self, results, words, ranking):
    totalscores = dict([(url, 0) for url in results])
    # This is where you'll put the scoring functions
    weights = []
    # word frequency scoring
    if (ranking & self.FREQ) != 0:
       weights.append((1.0, self.frequencyscore(results)))
    if (ranking & self.LOC) != 0:
       weights.append((1.0, self.locationscore(results)))
    if (ranking & self.LINK) != 0:
       weights.append((1.0, self.inboundlinkscore(results)))
    for (weight, scores) in weights:
        for url in totalscores:
            totalscores[url] += weight*scores.get(url, 0)
    return totalscores
```



Ranking word frequency - based

change the weights line in getscoredlist to this:

```
weights=[(1.0, self.frequencyscore(results))]
```



Document Location

- Search term's location in the page.
- If a page is relevant to the search term, it will appear closer to the top of the page
 - perhaps even in the title.
- Score results higher if the query term appears early in the document.
- wordlocation table:
 - the locations of the words were recorded.



Document Location

```
def locationscore(self, results):
    locations=dict([(url, 1000000) for url in results])
    for url in results:
        score = 0
        for wordlocations in results[url]:
            score += min(wordlocations)
        locations[url] = score
    return self.normalizescores(locations, smallIsBetter=True)
```



Ranking word location - based

change the weights line in getscoredlist to this:

```
weights=[(1.0, self.locationscore(results))]
```



Word Distance

- When a query contains multiple words, seek results in which the words in the query are close to each other in the page.
 - i.e., seek pages that conceptually relates search words.



Ranking word distance - based

Left as a take-home exercise!



Link information

- Used scoring metrics based on the content of the page.
- How about information that others have provided about the page?
 - who linked to the page?
 - what they said about it?

Pages created by spammers are less likely to be linked than pages with real content.



Link information

link stores two URL IDs, indicating a link from one page to another. {tourl: {fromUrl: None}}

linkwords store words that are included in a link. {word: [(urlFrom I, urlTo I), ..., (urlFromN, urlToN)]}

- The link table has the URLs for the source and target of every link that it has encountered.
- The linkwords table connects the words with the links.



Simple Count

- Count inbound links to each page and use the total number of links as a metric for the page.
- The scoring function based on this count:



Simple Count

 Using this metric by itself will simply return all the pages containing the search terms,

"ranked solely on how many inbound links they have."

 We need to combine the inbound-links metric with one of the content/relevance metrics we saw earlier.



What if?

 Someone can easily set up several sites pointing to a page whose score they want to increase.



What about ...?

- What about result pages that have attracted the attention of very **popular** sites?
- How to make <u>links from popular pages worth more</u> in calculating rankings?





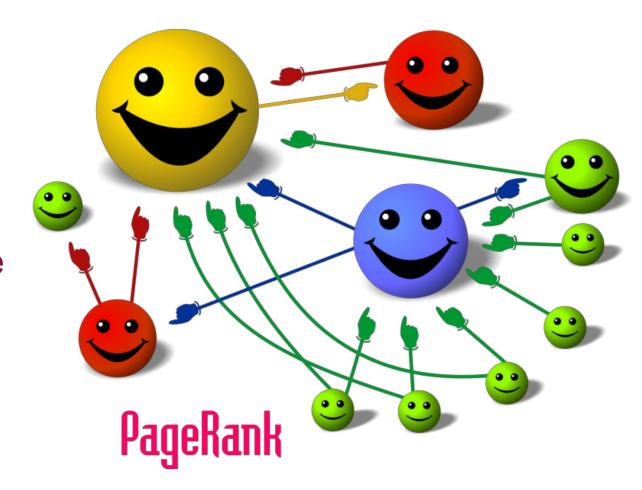
- PageRank (named after Larry Page) calculates the probability that someone randomly clicking on links will arrive at a certain page.
- The more inbound links the page has from other popular pages, the more likely it is that someone will end up there purely by chance.





Basic principle of PageRank.

The size of each face is proportional to the total size of the other faces which are pointing to it.



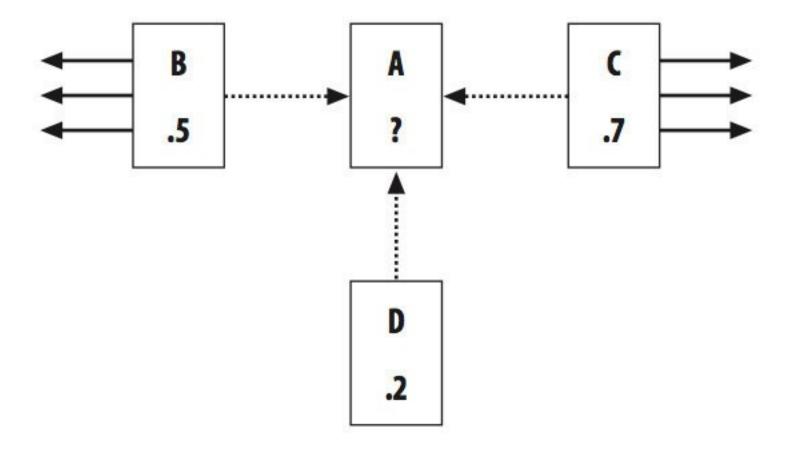


Source: http://en.wikipedia.org/wiki/PageRank

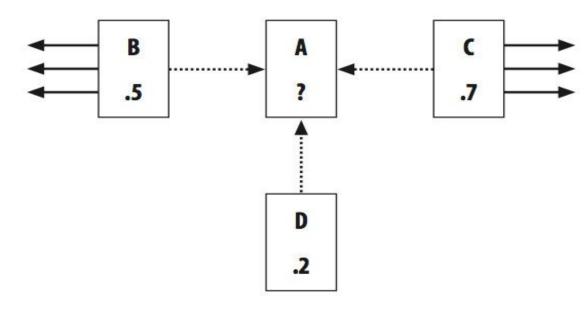
- If the user keeps clicking forever, they'll eventually reach every page, but most people stop surfing after a while.
- To capture this, PageRank also uses a damping factor of 0.85, indicating that

there is an 85% chance that a user will continue clicking on links at each page.









$$PR(A) = 0.15 + 0.85 *$$

$$(PR(B)/links(B) + PR(C)/links(C) + PR(D)/links(D))$$

$$= 0.15 + 0.85 * (0.5/4 + 0.7/4 + 0.2/1)$$
$$= 0.15 + 0.85 * (0.125 + 0.175 + 0.2)$$

$$= 0.15 + 0.85 * 0.5$$

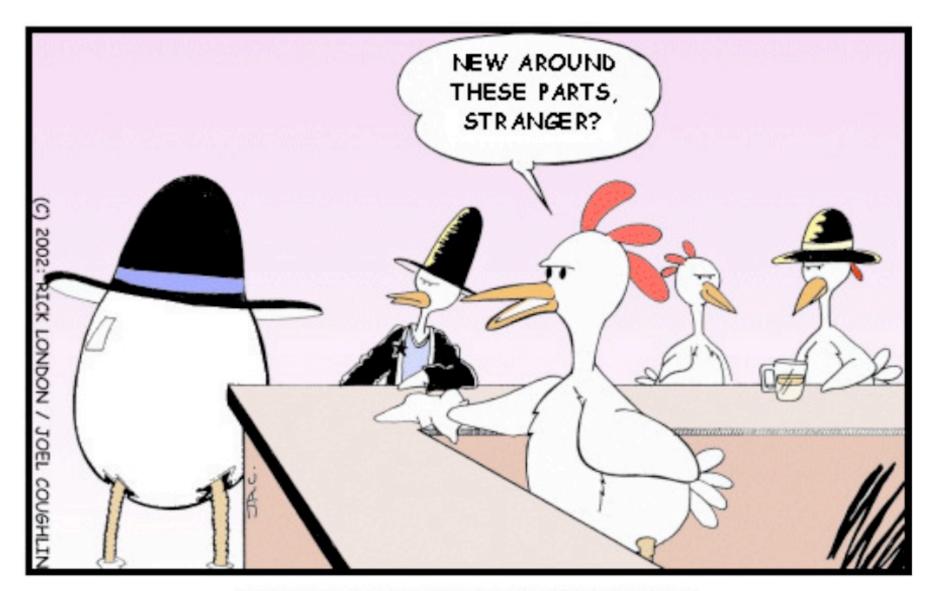
= 0.575



There is a small catch!

- All the pages linking to A already had PageRanks.
- You can't calculate a page's score until you know the scores of all the pages that link this page!
 - And, you can't calculate their scores without doing the same for all the pages that link to them.
- How is it possible to calculate PageRanks for a whole set of pages that don't already have PageRanks?



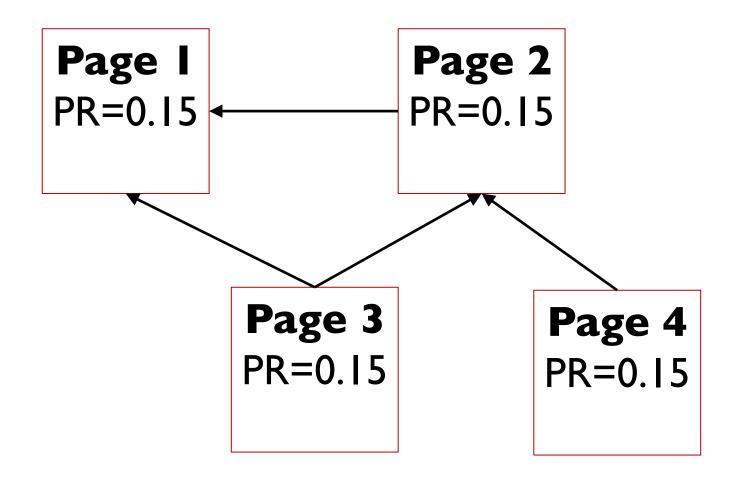


AND YET THE QUESTION REMAINED: "WHO CAME FIRST?"



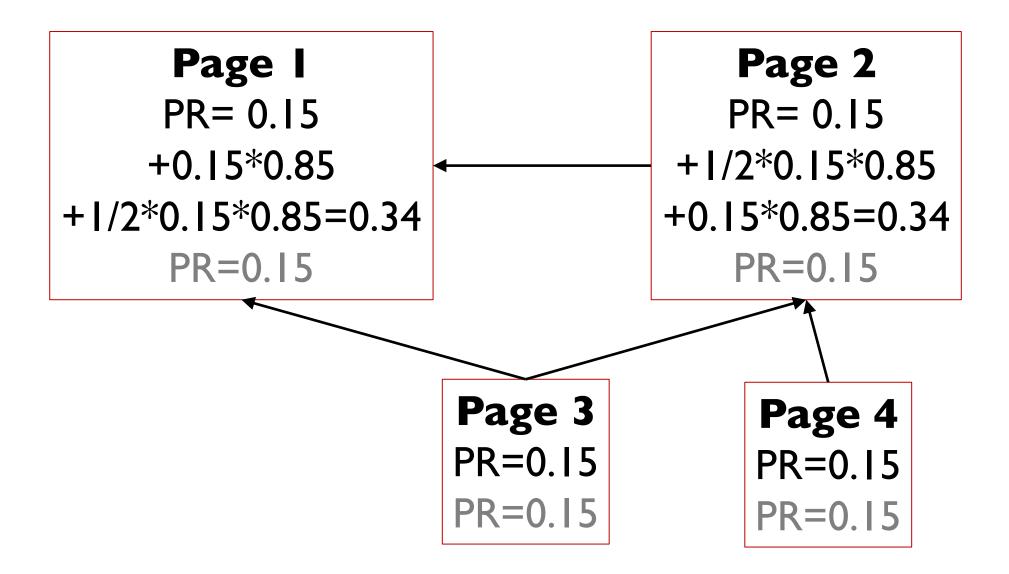
- The solution is to set all the PageRanks to an initial arbitrary value
 - (the code will use 1.0, but the actual value doesn't make any difference).
- Repeat the calculation over several iterations.
- After each iteration, the PageRank for each page gets closer to its true PageRank value.
- The number of iterations needed varies with the number of pages.

Iteration 0



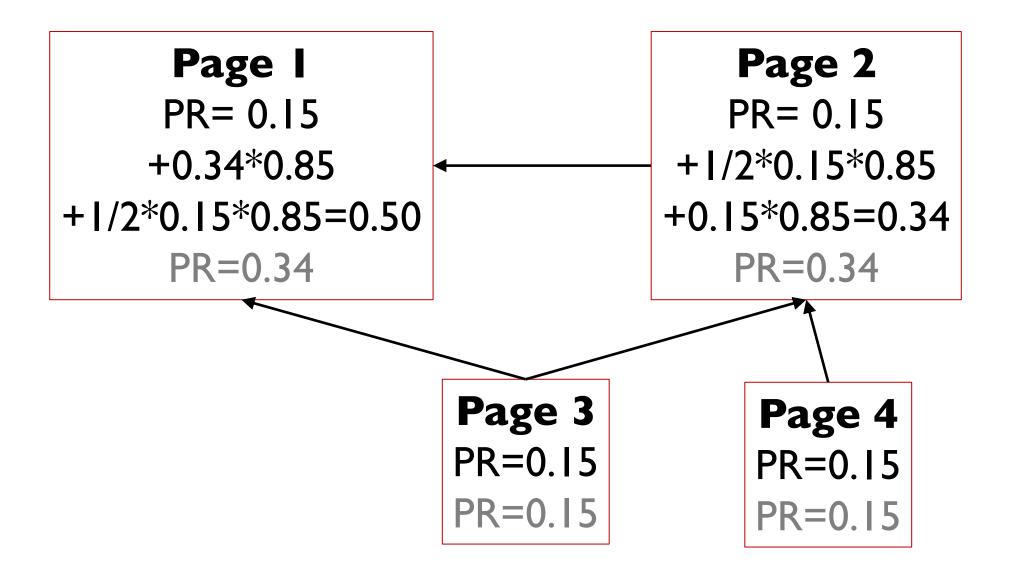


Iteration I



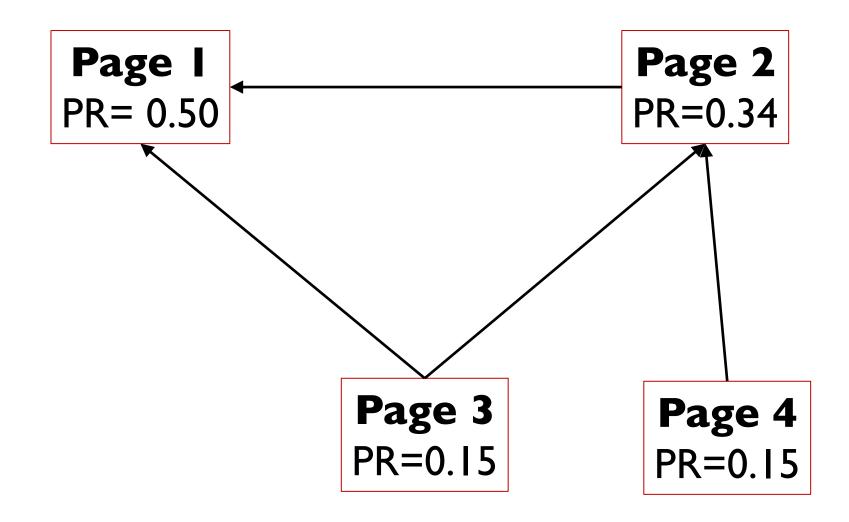


Iteration 2





Iteration 3+





PageRank computation

```
def calculatepagerank(self,iterations=20):
    self.pagerank = shelve.open(self.dbtables['pagerank'],
                                writeback=True, flag='n')
    # initialize every url with a page rank of 1
    for url in self.urllist:
        self.pagerank[url] = 1.0
    for i in range(iterations):
        print "Iteration %d" % (i)
        for url in self.urllist:
            pr = 0.15
            # Loop through all the pages that link to this one
            if url in self.link:
              for linker in self.link[url]:
                  linkingpr = self.pagerank[linker]
                  # Get the total number of links from the linker
                  linkingcount = self.urllist[linker]
                  pr += 0.85*(linkingpr/linkingcount)
```



self.pagerank[url] = pr

PageRank computation

import mysearchengine
crawler=searchengine.crawler(dbtables)
crawler.calculatepagerank()



Score Combination

