# INFORMATION RETRIEVAL

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Lecture 2

# **IMPROVING THE QUALITY OF RETRIEVAL**

## BUILDING AN INVERTED INDEX MAJOR STEPS

- 1. Collect the documents to be indexed
- 2. Tokenize the text, turning each document into a list of tokens
- 3. Do linguistic preprocessing, producing a list of normalized tokens, which are the indexing terms
- 4. Index the documents that each term occurs in by creating an inverted index, consisting of a dictionary and postings.

# LECTURE OUTLINE

Array, linked lists, and skip lists

Tokenization, Stop Words, Normalization, Stemming & Lemmatization

> PRACTICAL PART A PYTHON IMPLEMENTATION OF A SIMPLE BOOLEAN RETRIEVAL SYSTEM

Phrase Queries, Biwords, positional postings

# CHOOSING A DOCUMENT UNIT

- First step in the indexing process is to decide what is the granularity of the indexing
- For very long documents, the issue of indexing granularity arises, e.g. emails and attachments and zip files, html files, returning chapters or paragraphs instead of entire books, individual sentences as minidocument, ...
- There is a precision/recall tradeoff
- For now, we will henceforth assume that a suitable size document unit has been chose
- The problems with large document units can be alleviated by use of explicit or implicit proximity search

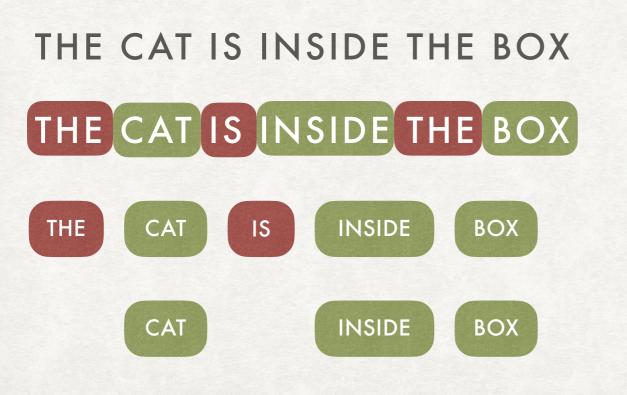
# TERMINOLOGY (4) THIS TIME FOR TOKENIZATION

- Token: instance of a sequence of characters
- Type: collection of all tokens with the same character sequence

Text

**Tokens** 

• Term: a type that is inserted into the dictionary



Types (notice only one instance of "the")

Terms (after removal of common words and normalization)

# TOKENIZATION SPLITTING THE TEXT IN WORDS

- The second step is to split a text sequence into tokens.
- In some cases deciding where to split the text sequence is simple...
- ...but in many others it is not, even in English.
- For others languages it might not even be clear where a word ends and the next one starts.

# **EXAMPLES OF PROBLEMATIC TOKENIZATION**

Text	Possible tokenizations
New York	[New] [York]
File-system	[File] [system], [File-system]
555-1234 567	[555] [1234] [567], [555-1234] [567], [555-1234 567]
Upper case	[Upper][case]
Uppercase	[Uppercase]
O'Hara	[O] [Hara], [O'Hara]
Aren't	[Aren][t], [Aren't]

Possible (partial) solutions:

- use the same tokeniser for the documents and the queries
- use a collection of heuristics to decide where to split words

# STOP WORDS DROPPING COMMON TERMS

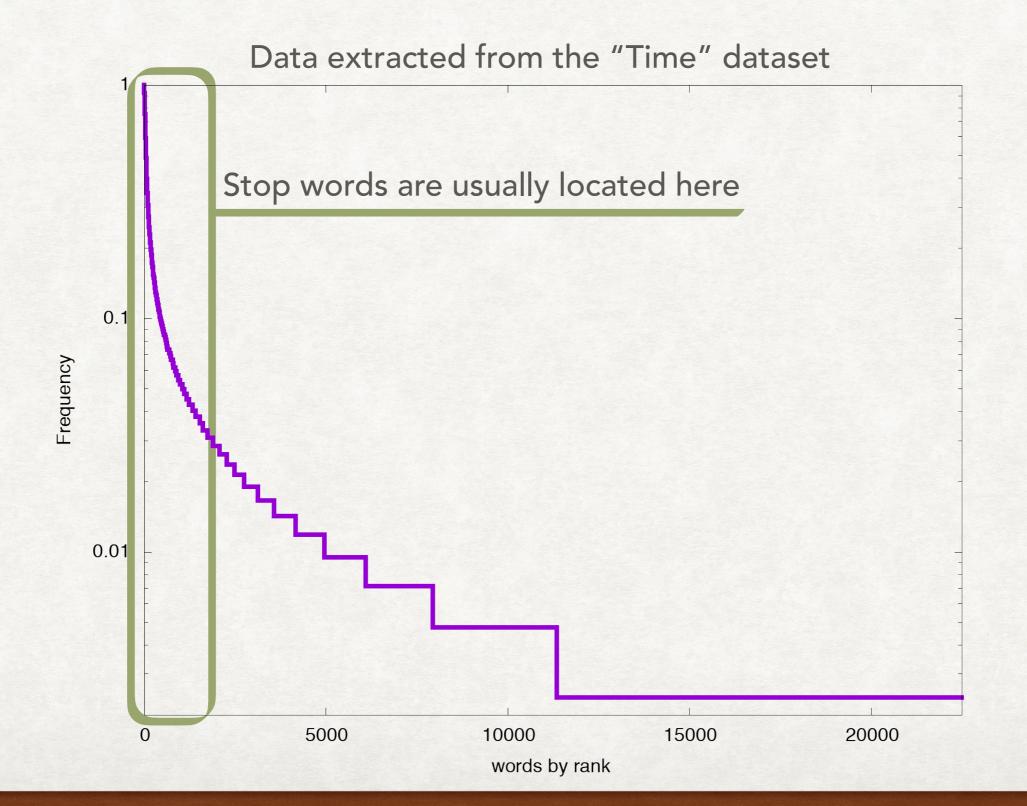
As anticipated before:



Some terms are not useful: "A" is in all the documents!

- Stop words: common words that do not help in selecting a document. They are discarded from the indexing and querying processes
- Stop list: list of stop words. Specific for a language/corpus. Usually consists of the most frequent words, curated for their semantic.

# DISTRIBUTION OF WORDS FREQUENCIES OF WORDS IN A CORPUS



# STOP WORDS FOR THE ENGLISH LANGUAGE AND STOP WORDS FOR SPECIFIC TOPICS

- You can find multiple lists of stop words for the English language. They usually include words like:
  - a, about, above, after, again...
  - ... the, their, theirs, ..., your, yours, yourself, yourselves.
- The list of stop words is language specific: stop words in Italian are different (additional challenge: you might need to infer the language of a document).
- Stop lists can be specific by topic. E.g., in a "books on cats" corpus, the word "cat" might be a stop word.

# DROPPING STOP WORD SOMETIMES IT IS USEFUL

- Using a stop list significantly reduces the number of postings that a system has to store
- And a lot of the time not indexing stop words does little harm
- but..

### PROBLEMS WITH STOP WORDS SOMETIMES STOP WORDS ARE USEFUL

- You now have a IR system that removes all stop words.
- You receive the queries:
  - To be or not to be
  - Dr Who
  - Do it yourself
  - Let it be
- Removing stop words can reduce the recall.

### PROBLEMS WITH STOP WORDS SOMETIMES STOP WORDS ARE USEFUL

- A single stop word alone can usually be removed...
- ... but in a phrase search it might be important
- The trend has been from large stop lists (200–300 terms) to very small stop lists (7-12 terms) or no stop word list but:
  - Use compression techniques to reduce the storage requirements
  - Use weighting to limit the impact of stop words
  - Use specific algorithms to limit the runtime impact of stop words

# NORMALIZATION REMOVING SUPERFICIAL DIFFERENCES

- The same word can be written in different ways and it must be normalized to allow the matching to occur.
- The idea is to define equivalence classes of terms, for example:
  - By ignoring capitalization (e.g., "HOME", "home", "HoMe"), always or just at the beginning of a sentence and titles
  - By removing accents and diacritics (e.g., cliché is considered the same as cliche).
  - Other normalization steps specific to the language, like ignoring spelling differences (e.g., "colors" vs "colours").

# **RELATIONS BETWEEN UNNORMALIZED TOKENS** AN ALTERNATIVE TO EQUIVALENCE CLASSES

Sometimes capitalization and other features are important

This can be solved by saving (possibly asymmetric) relations between token in a *query expansion list* 

Query Term	Equivalent terms
Windows	Windows
windows	Windows, windows, window
window	windows, window

# **RELATIONS BETWEEN UNNORMALIZED TOKENS** AN ALTERNATIVE TO EQUIVALENCE CLASSES

Sometimes capitalization and other features are important

Or it can be solved performing the expansion during index construction.

Term	Indexing
automobile	car, automobile
car	automobile, car

# PRO AND CONTRO WHICH IS IT BETTER?

- Equivalence classing reduce information but has less postings to store and merge
- A query expansion list adds a query expansion dictionary and requires more processing at query time; the second method requires more space for storing postings.
- But they are more flexible because the expansion lists can overlap while not being identical (e.g. asymmetry in expansion)
- Traditionally, expanding the space required for the postings lists was seen as more disadvantageous, but with modern storage costs, the increased flexibility that comes from distinct postings lists is appealing.

### **STEMMING AND LEMMATIZATION** REDUCE WORDS TO A COMMON BASE FORM



Two main ways: stemming and lemmatization

**Based on heuristics** 

Uses a vocabulary and morphological analysis

# PORTER STEMMER

### MOST USED STEMMER FOR THE ENGLISH LANGUAGE

Invented in 1979 (published 1980) by Martin Porter, it is one of the most common stemmers for the English language

Five stages applied sequentially.

Each stage consists of a series of rewriting rules for words, an example is given here

Rule	
SSES → SS	caressess → caress
IES → I	poinies → poni
SS → SS	caress → caress
$S \rightarrow$	cats → cat

Porter Stemmer implementations: <u>https://tartarus.org/martin/PorterStemmer/</u> (or you can read the original paper and the BCLP implementation)

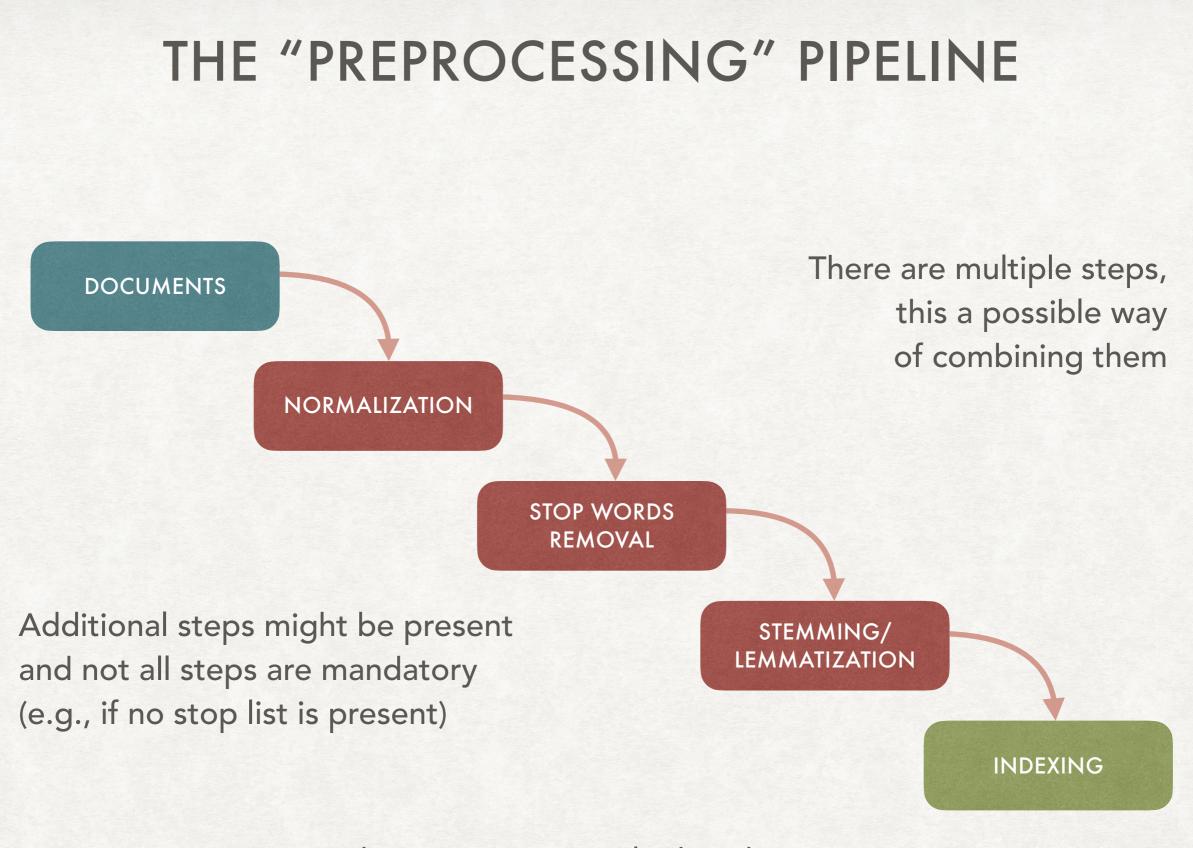
# WHAT CAN GO WRONG

- While it helps a lot for some queries, it equally hurts performance a lot for others
- Stemming increases recall while harming precision
- e.g. the Porter stemmer stems all of the following words:

operate operating operates operation operative operatives operational -> oper

losing precision on queries such as:

operational AND research, operating AND system, operative AND dentistry



The same steps applied to the queries!

# **ANSWERING PHRASE QUERIES**

# OUR GOAL EXTENDING THE QUERY LANGUAGE

- We want to be able to ask queries consisting of multiple consecutive words:
  - "calico cat"
  - "University of Trieste"
- A common syntax for this kind of queries is to enclose the words in double quotes.
- Two approaches shown: biword indexes and positional indexes.

# BIWORD INDEXES WORKING ON PAIRS OF WORDS

THE CAT IS INSIDE THE BOX

Text



Terms

- The terms are pairs of words
- Queries need to be "rewritten":

"inside the box" \_\_\_\_\_\_ "inside the" AND "the box"

# **BIWORD INDEXES POSSIBLE PROBLEMS**

INSIDE THE HOUSE THERE IS THE BOX Text:

Original Query: "inside the box"

**No Match** 

Rewritten Query: "inside the" AND "the box"

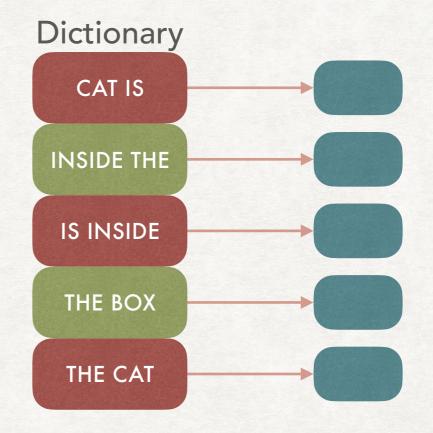
Match

Rewriting the query might generate false positives (but it works quite well in practice)

### BIWORD INDEXES POSSIBLE PROBLEMS



To answer the query we would need to find all terms containing "cat"



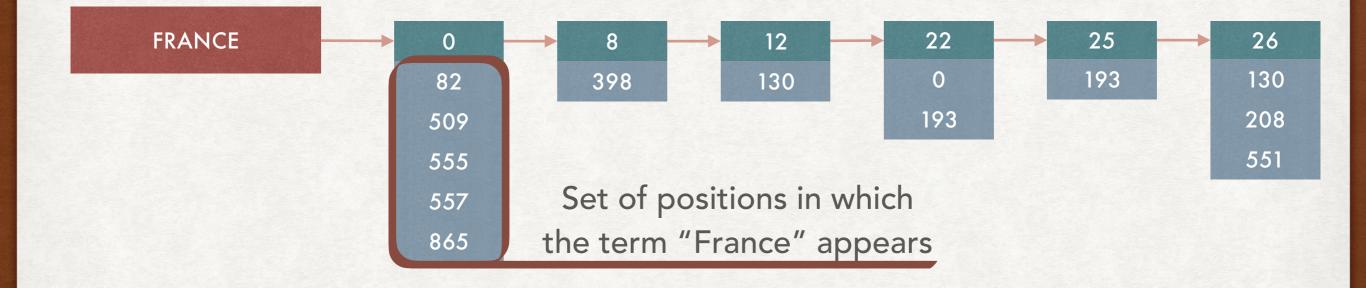
We also need an index of single-word terms!

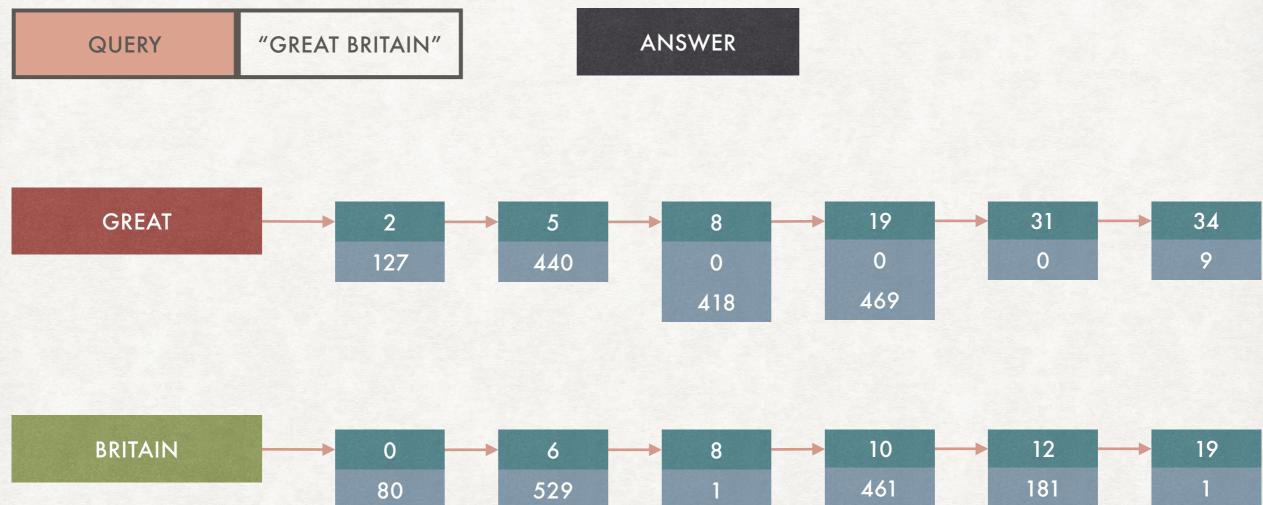
# **BIWORD INDEXES** EXTENSIONS AND FURTHER OBSTACLES

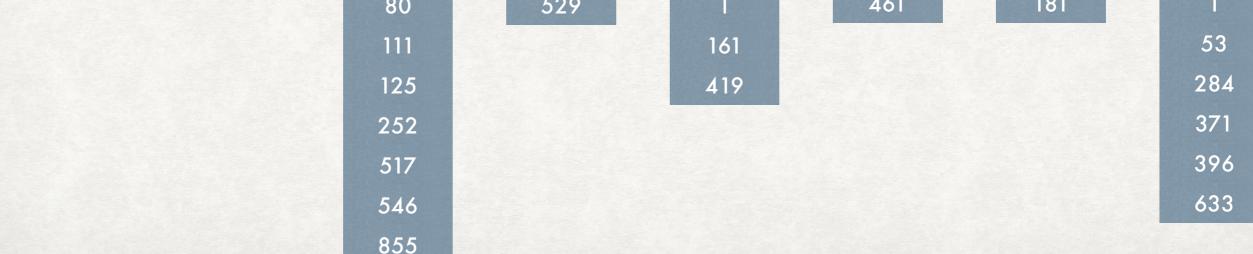
- The idea of using pair of words as terms can be extended to any length, reducing the risk of false positives...
- ...but increasing the amount of space needed.
- If the number of words in a term is variable it is called *phrase index*.
- It is also possible to "tag" the part of speech (i.e., names, verbs, articles, prepositions, etc.) to add pairs of names separated by articles and prepositions to the index.
  - E.g., in "door at the entrance", "door entrance" is considered a term

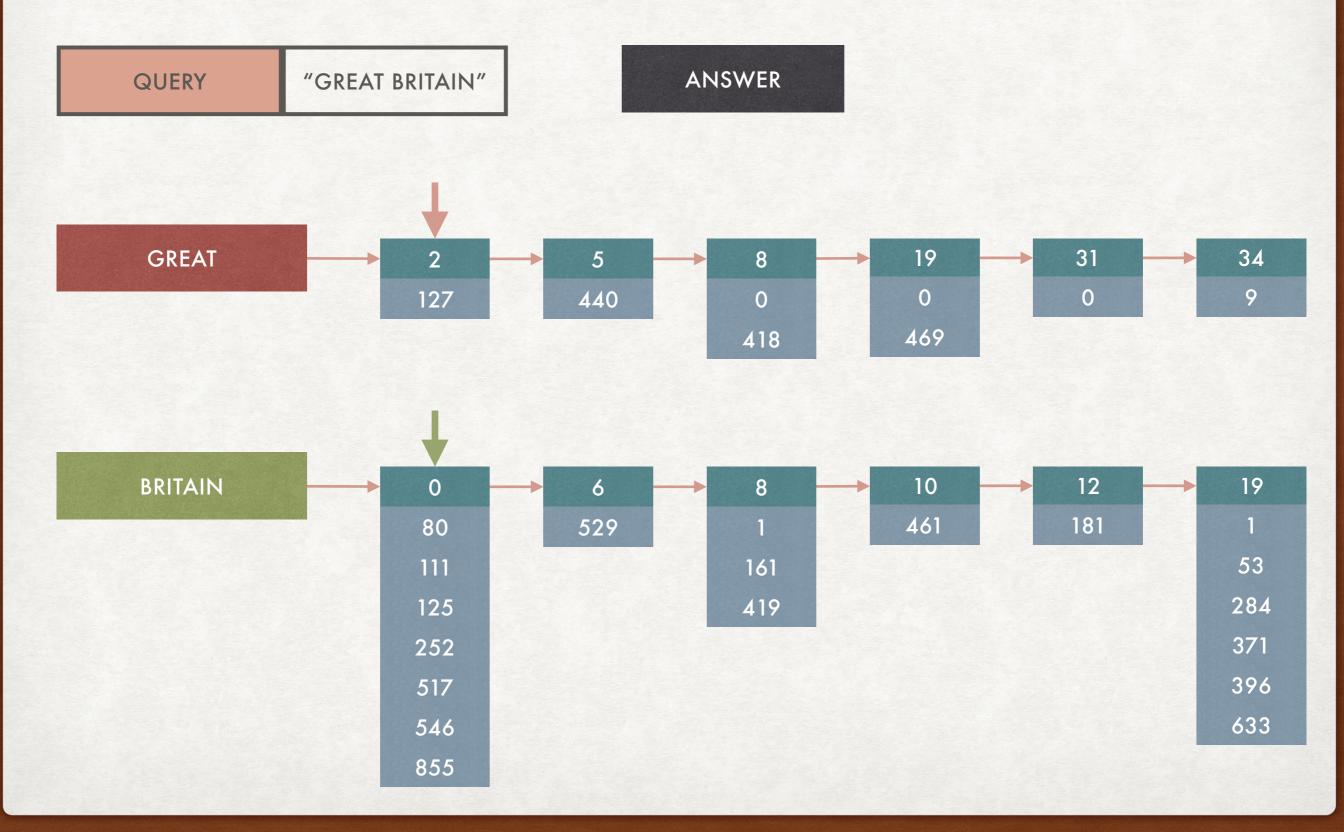
# **POSITIONAL INDEXES** ADDING POSITIONS TO THE POSTINGS

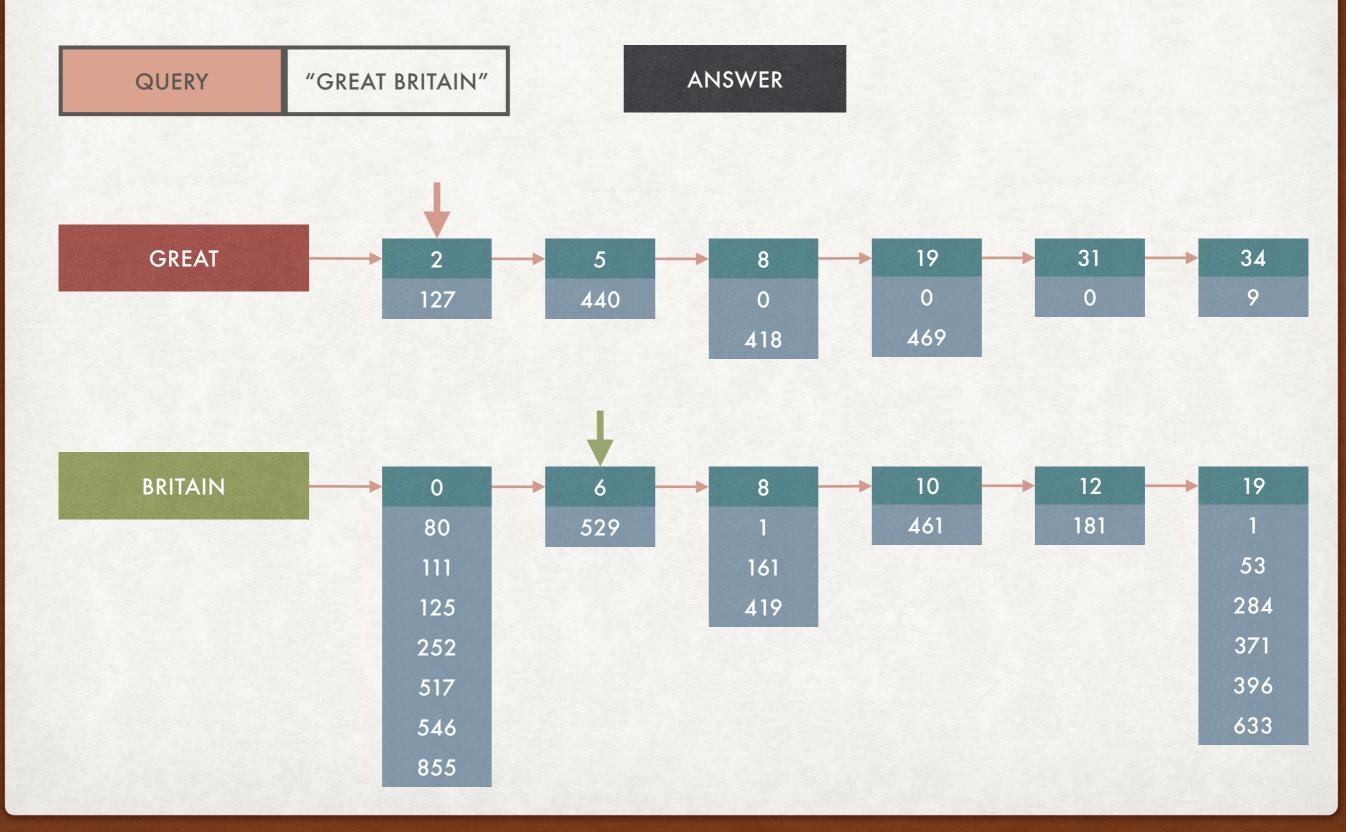
One way to answer a phrase query is to add, for each posting, the set of positions in which the term appear in the document.

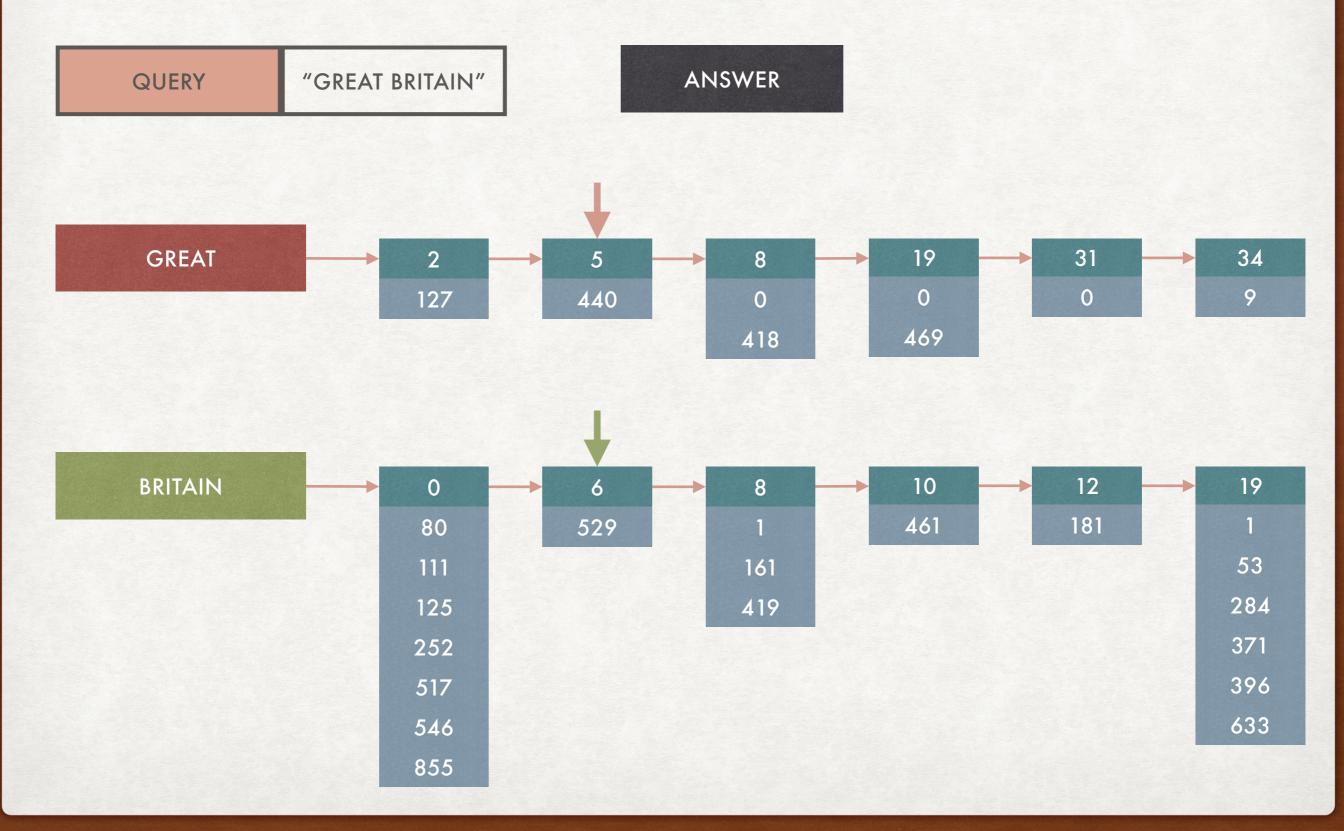


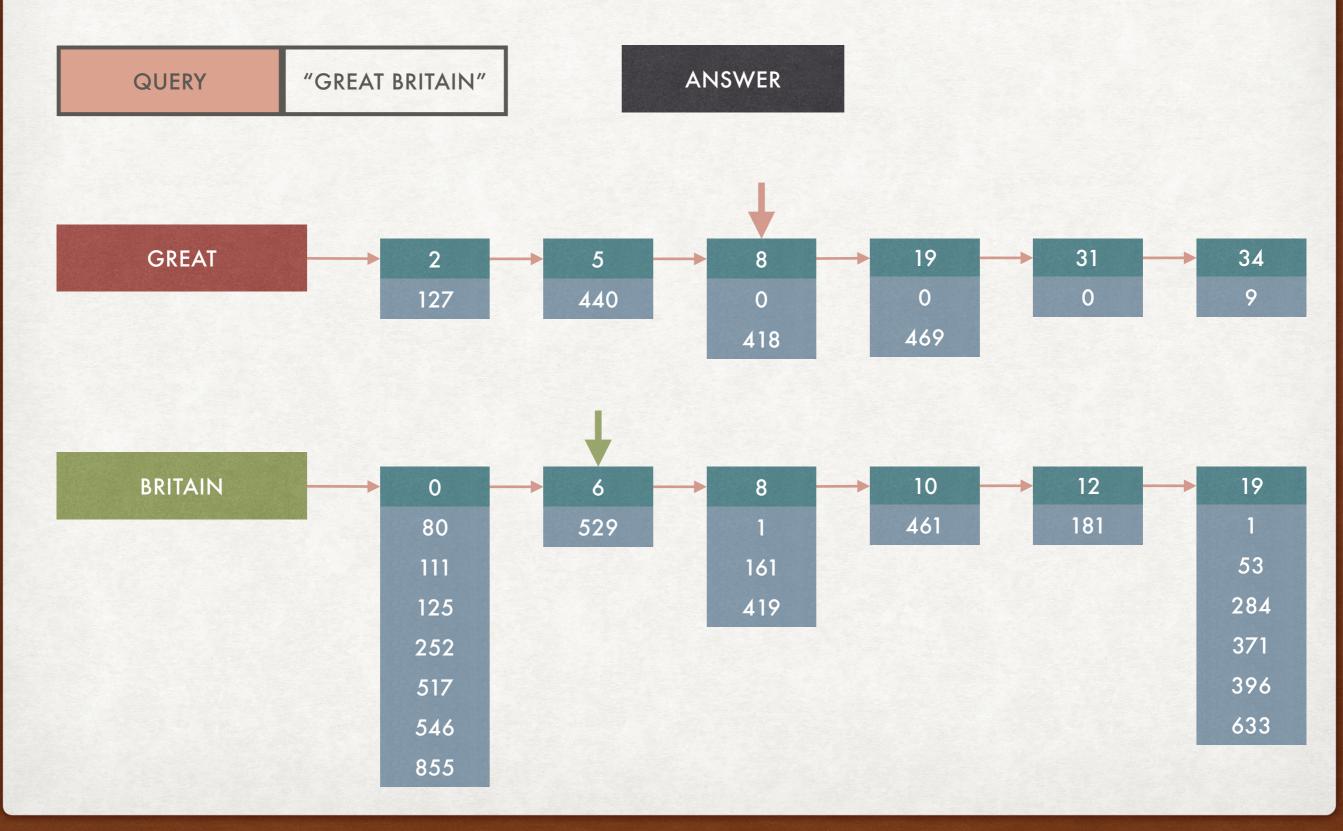


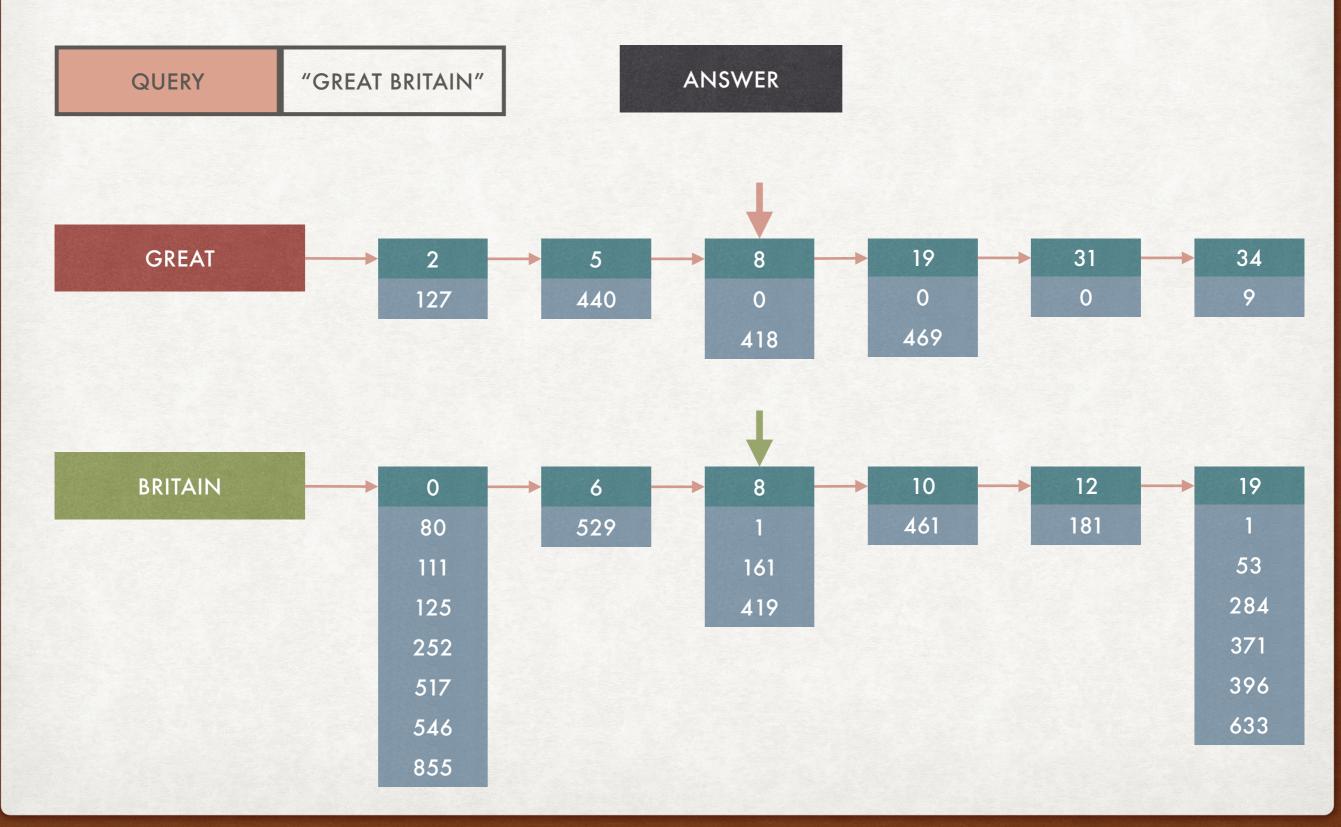


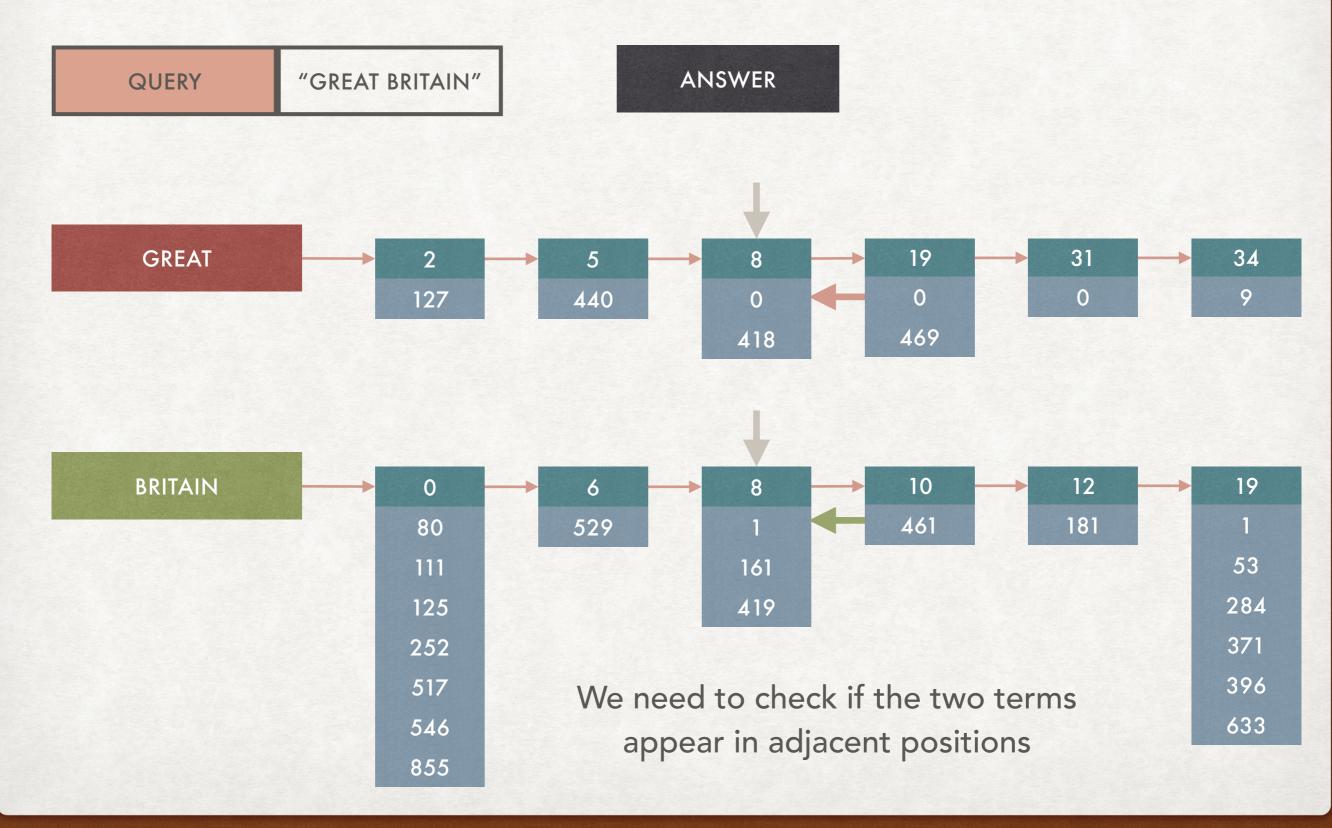


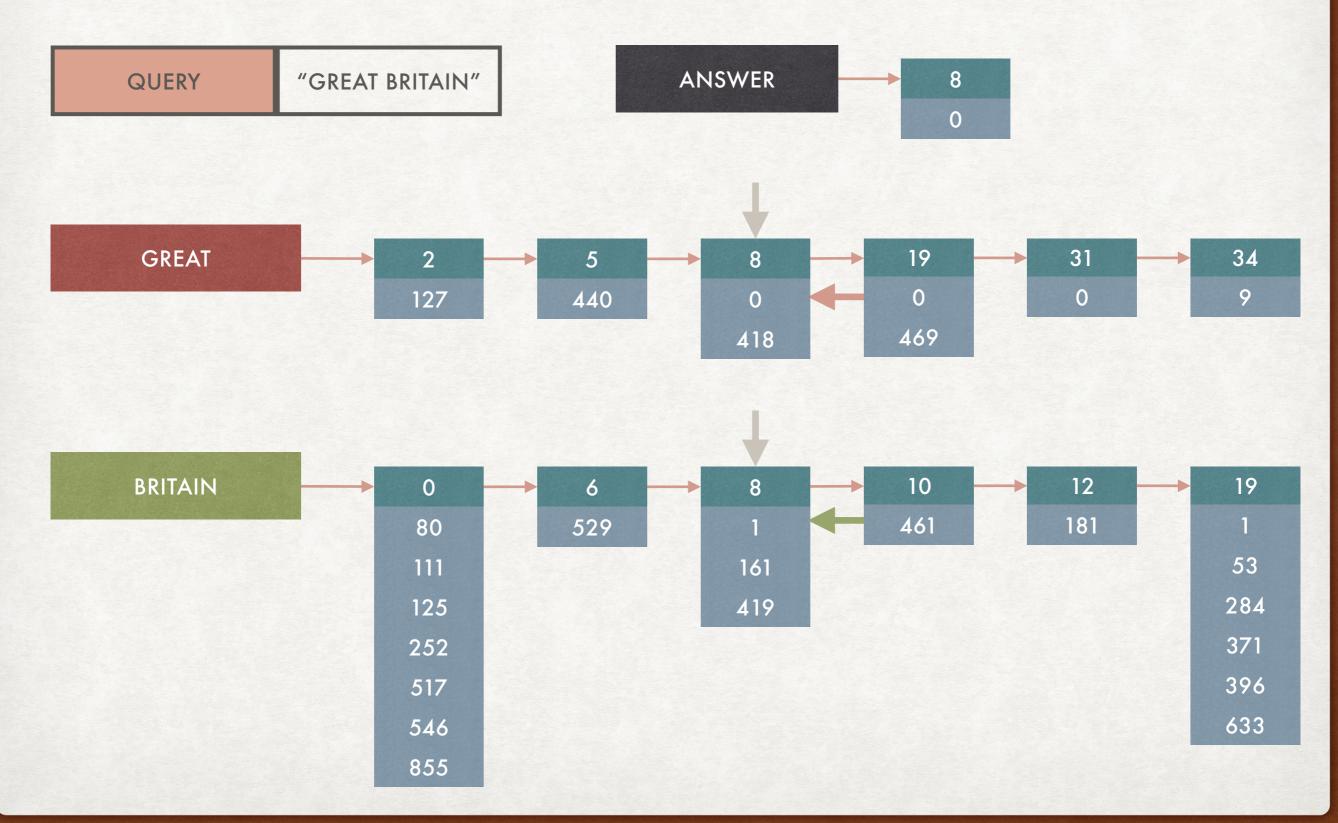


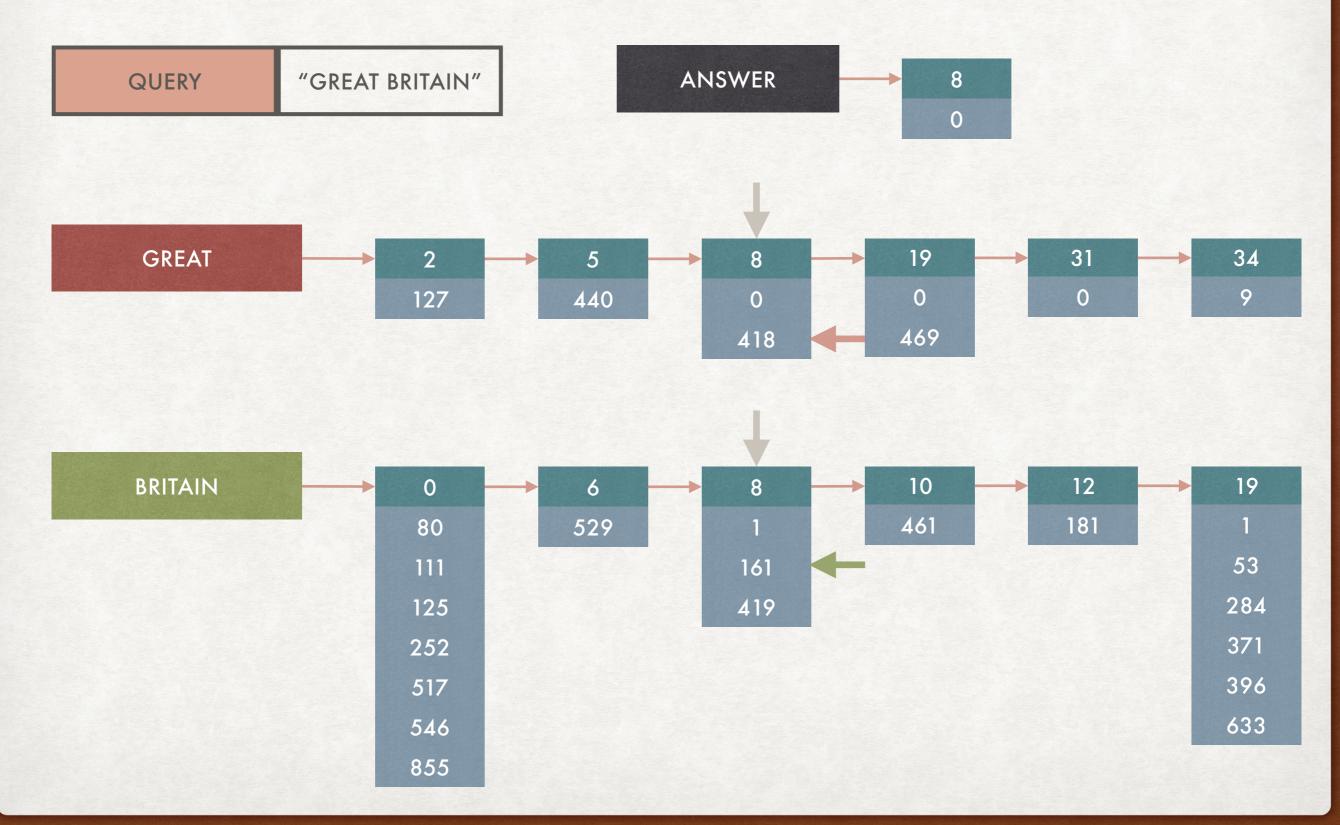


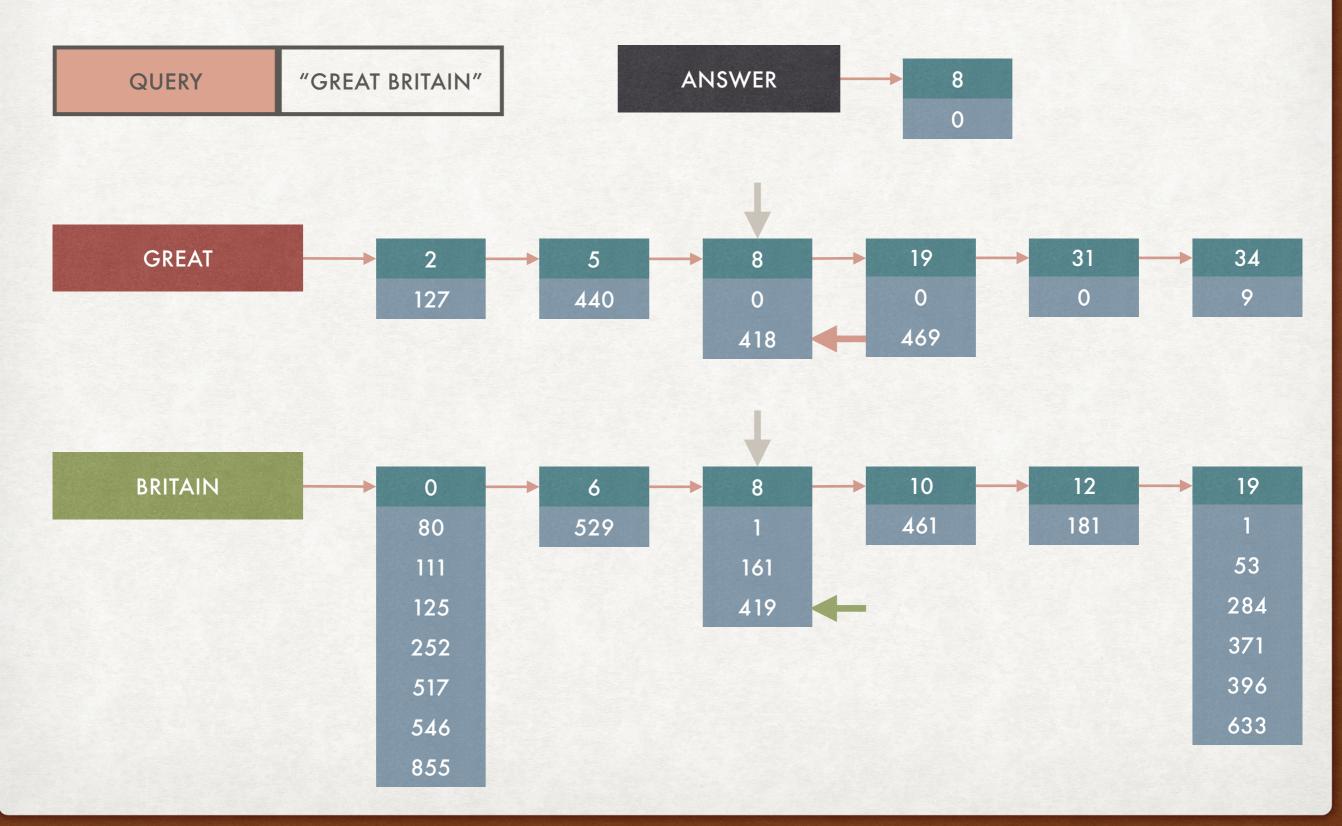


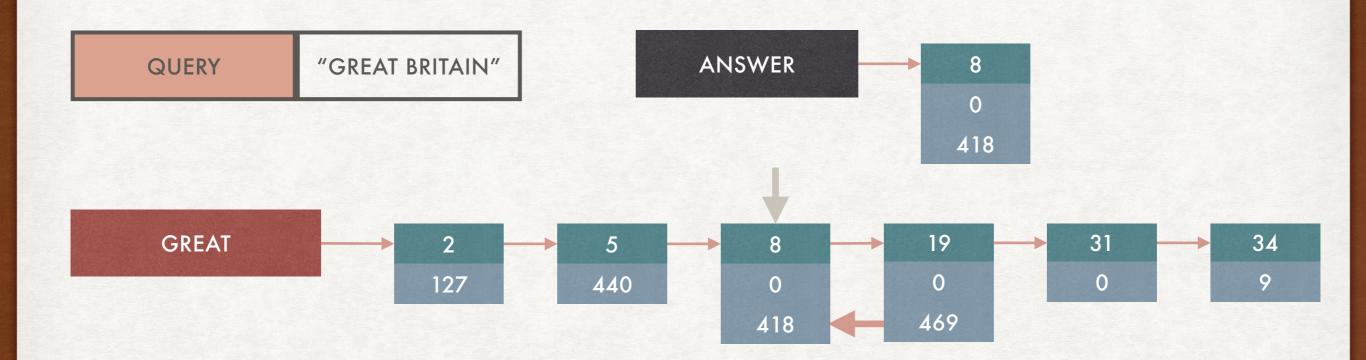


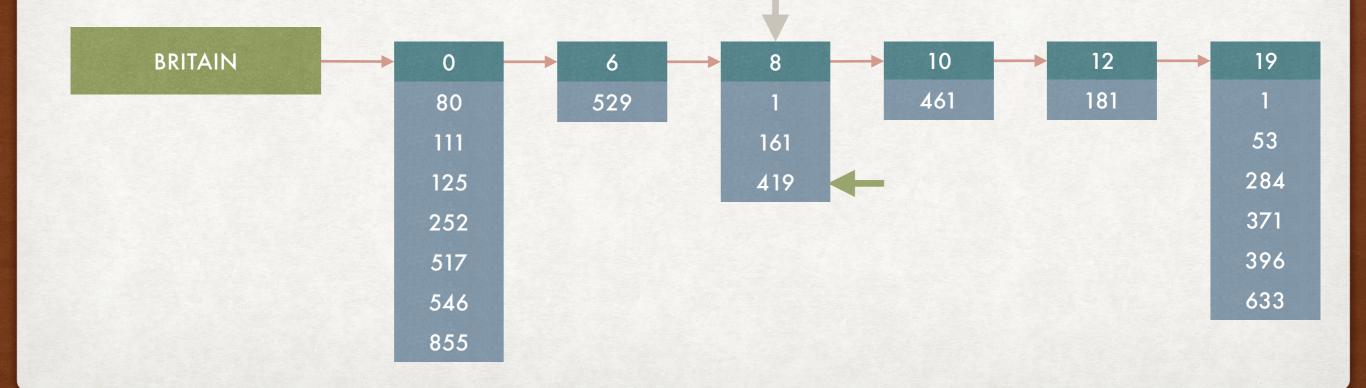


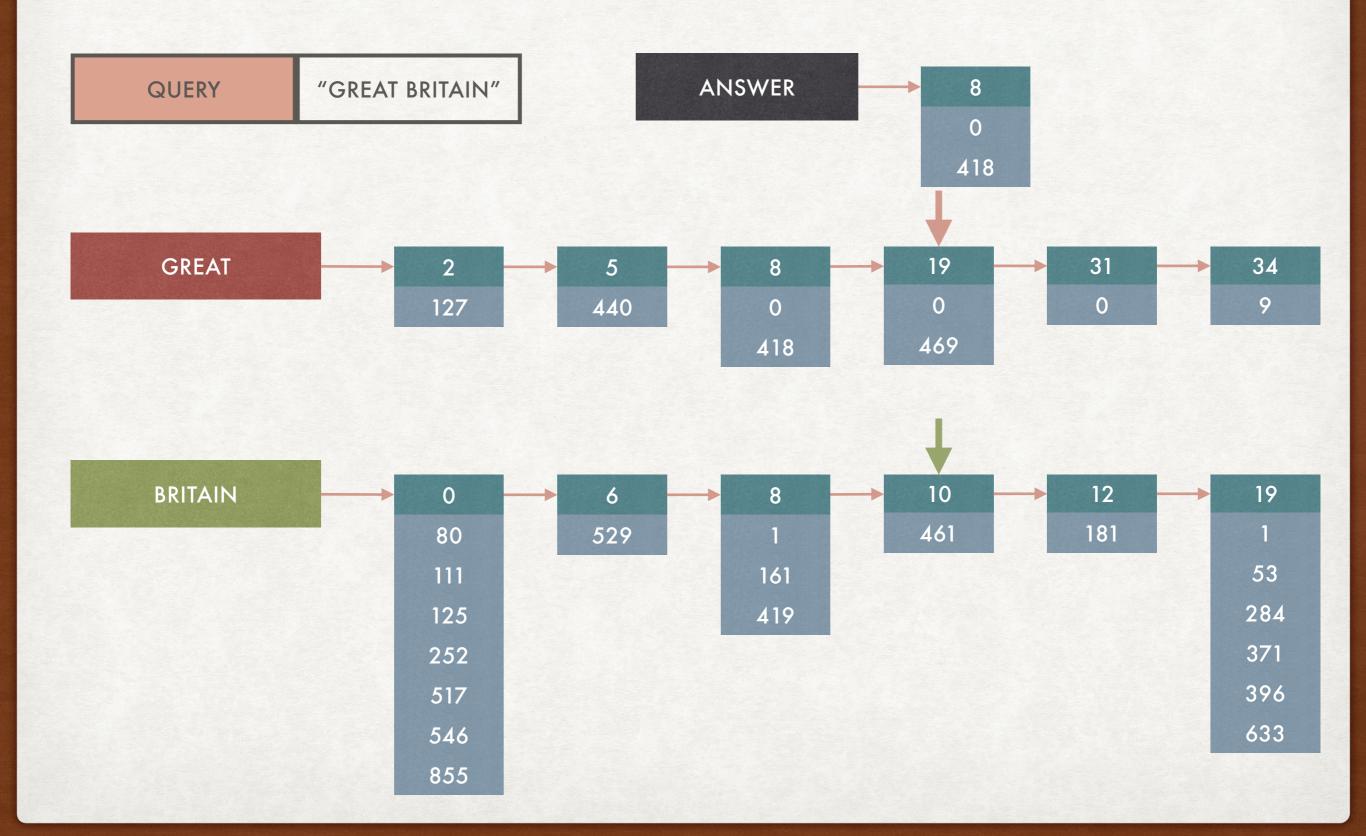


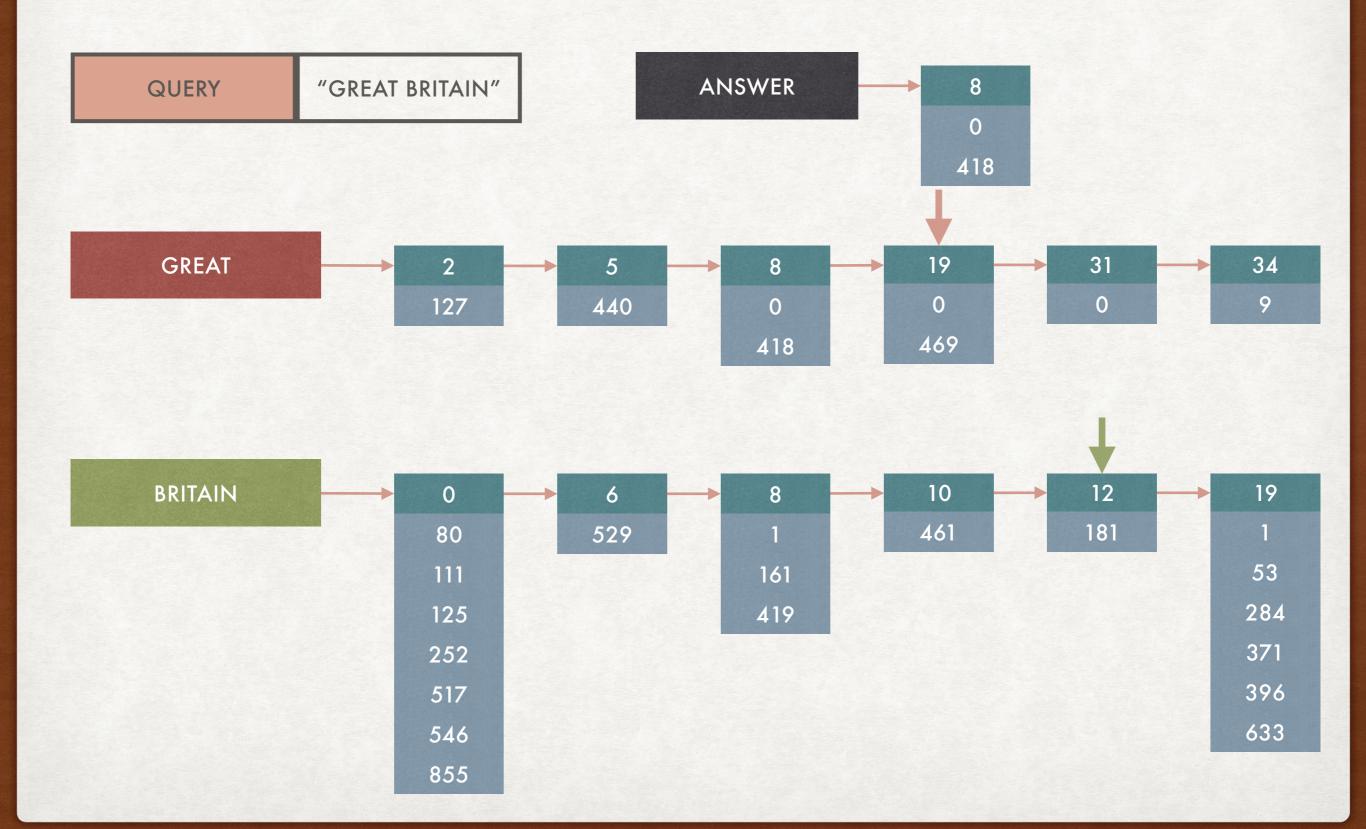


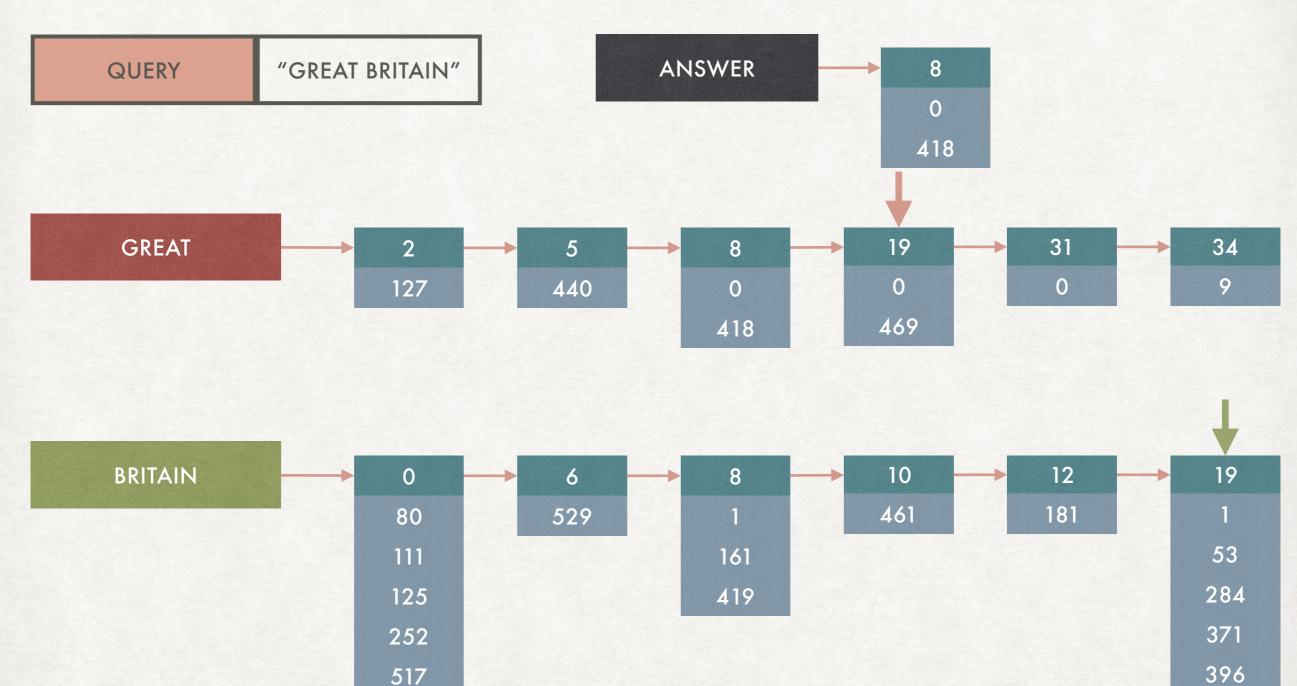








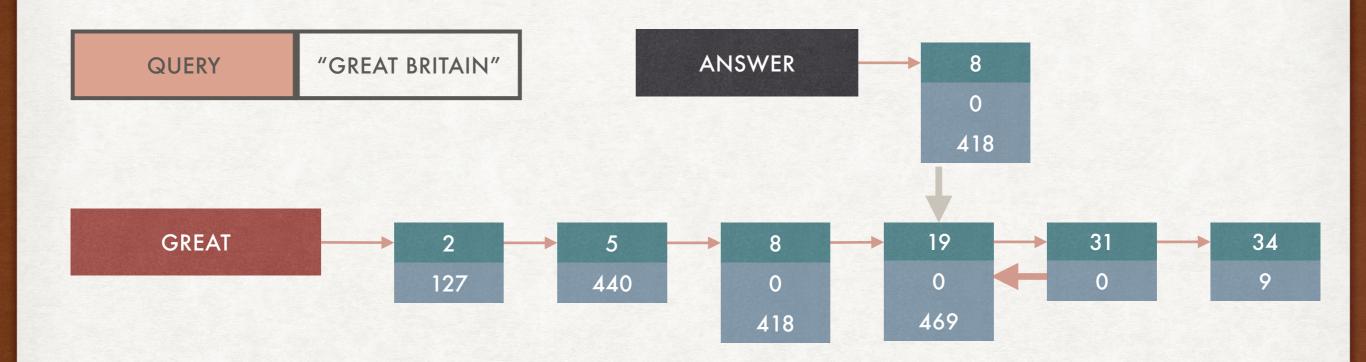


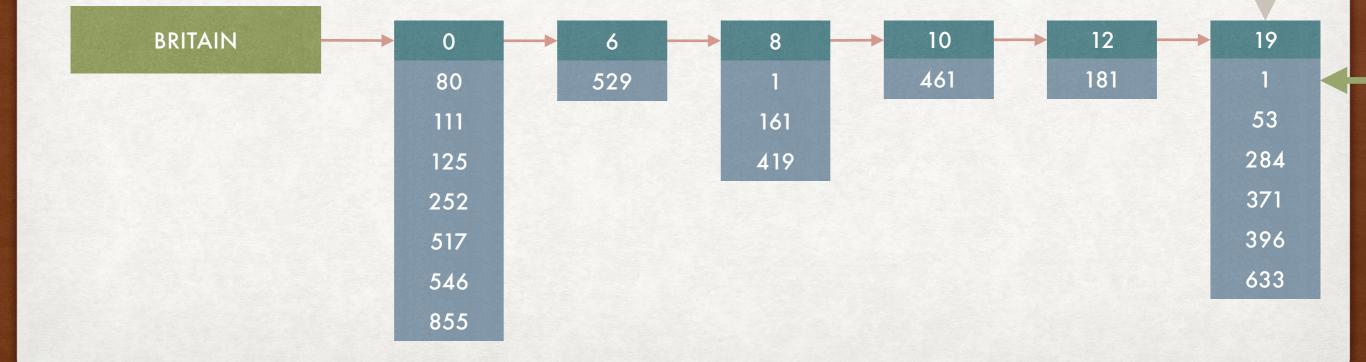


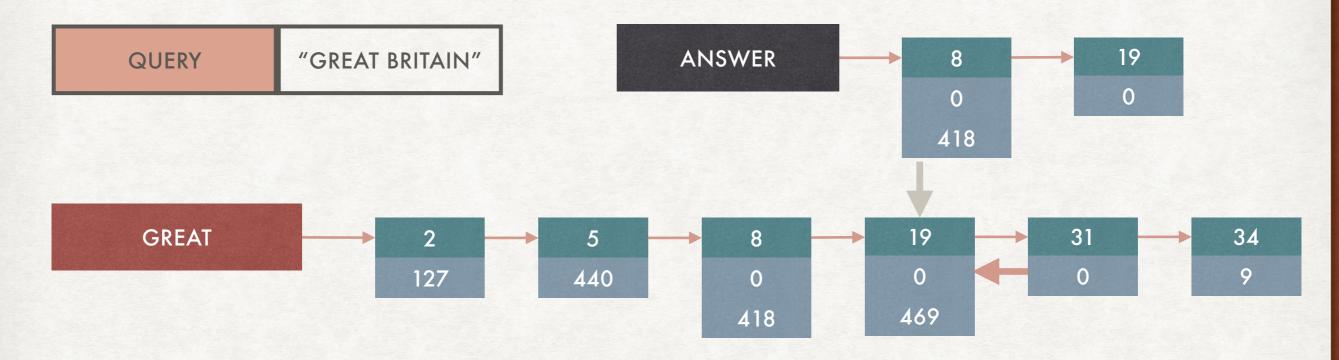
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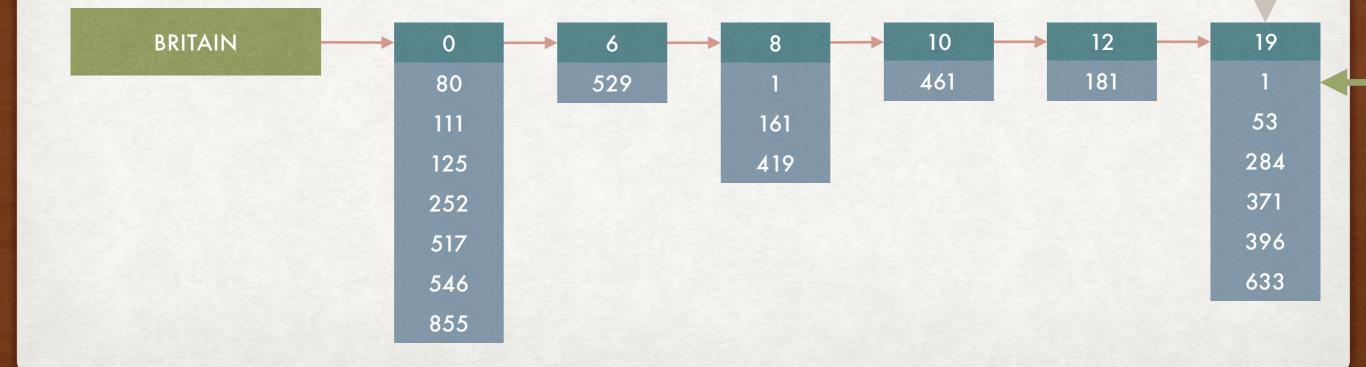
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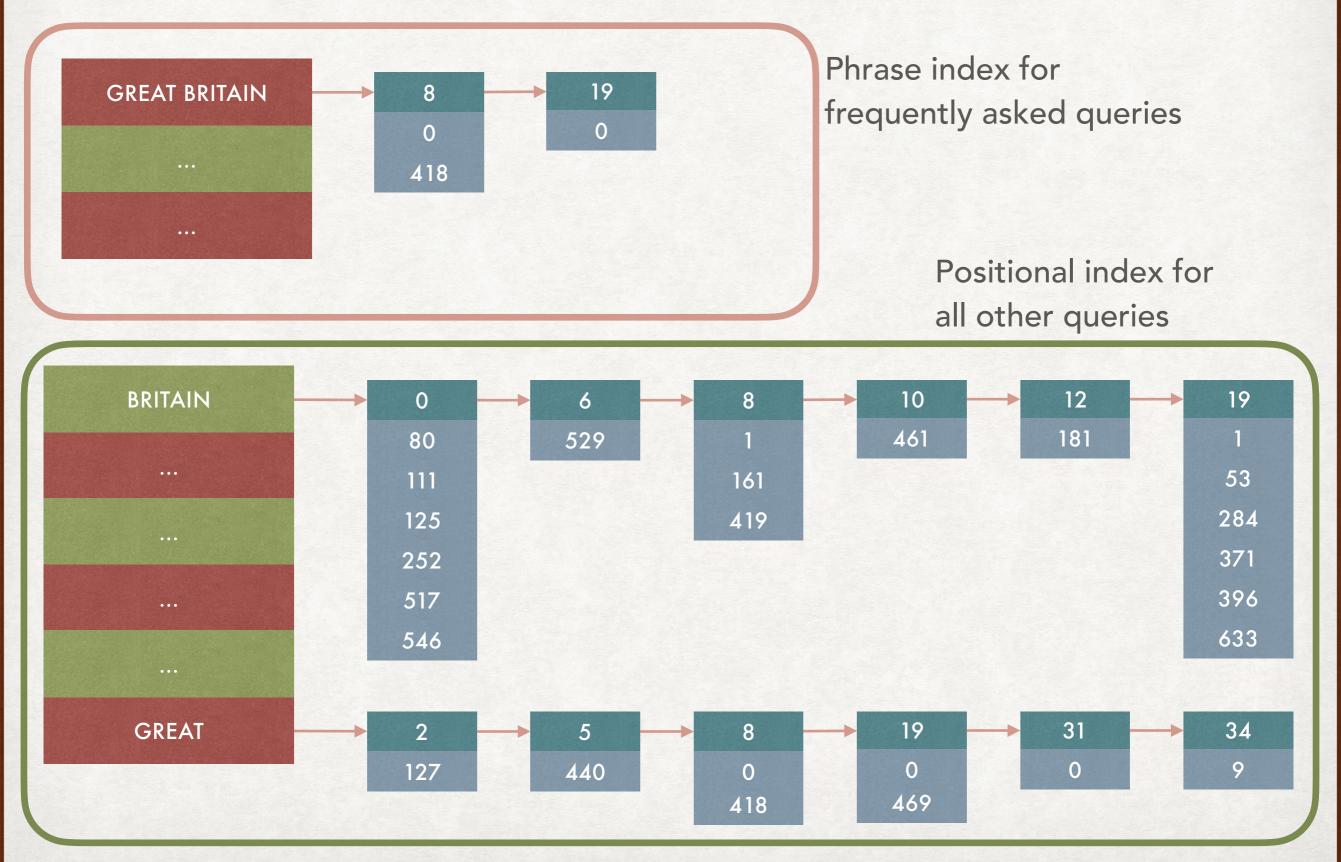




## **POSITIONAL INDEXING: SUMMARY** THE GOOD, THE BAD, AND THE UGLY

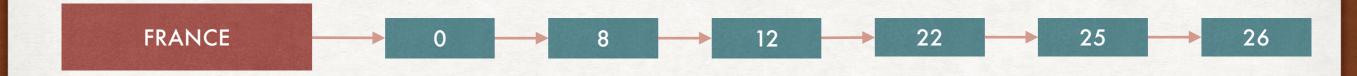
- The positional index can be used to support the operators of the form "term<sub>1</sub> /k term<sub>2</sub>" with k an integer indicating the maximum number of words that can be between term<sub>1</sub> and term<sub>2</sub>.
- The complexity of performing a query is not bounded anymore by the number of documents, but by the number of terms
- The size of the index now depends on the average document size.

## COMBINING BIWORD AND POSITIONAL INDEXES

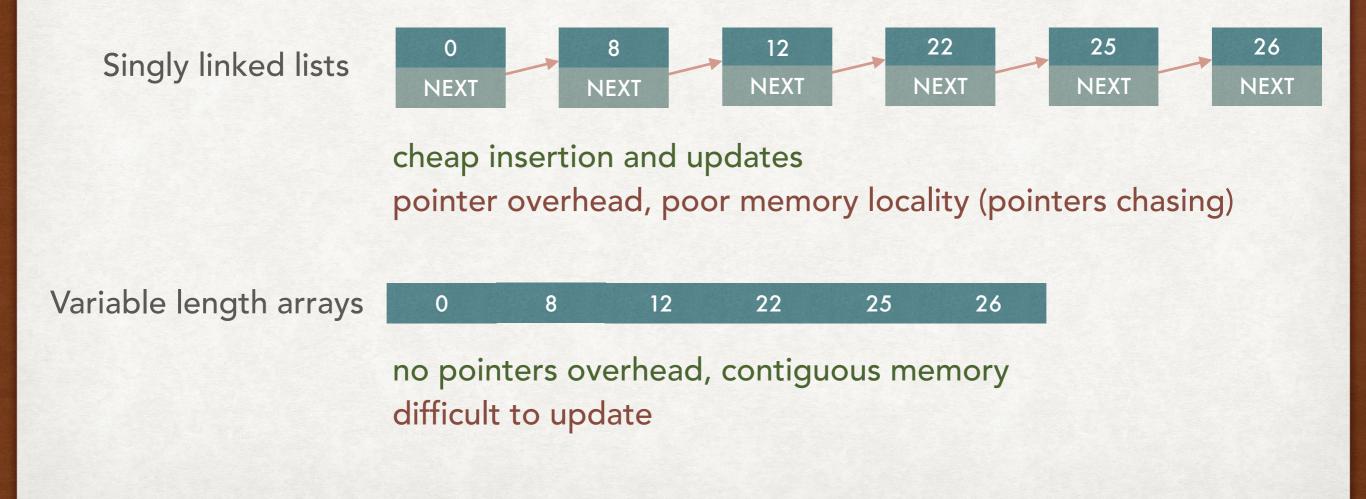


## **IMPROVING THE INVERTED INDEX**

## ARRAYS OR LINKED LISTS? WHAT TO USE FOR THE POSTING LISTS?

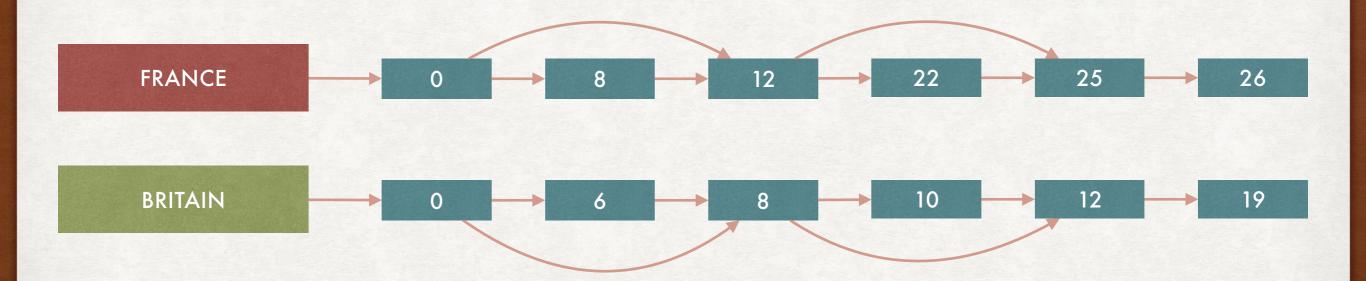


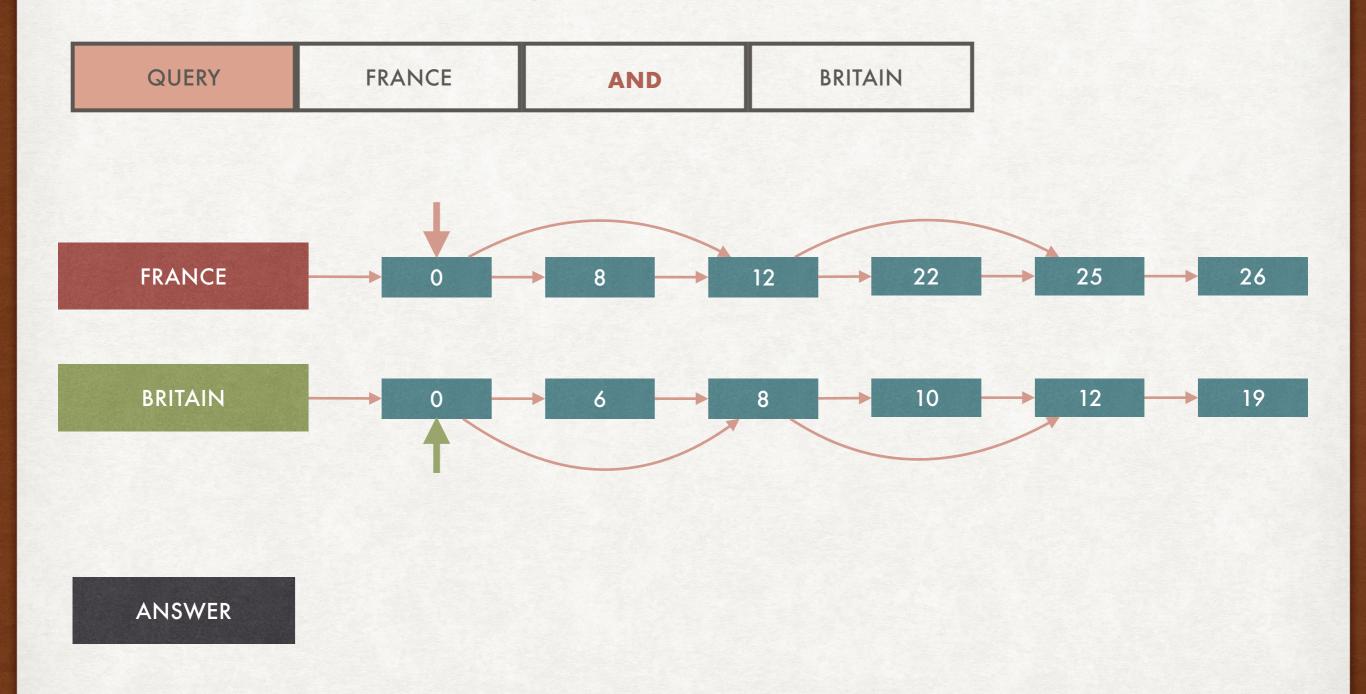
Which data structures should we actually use for the postings list?

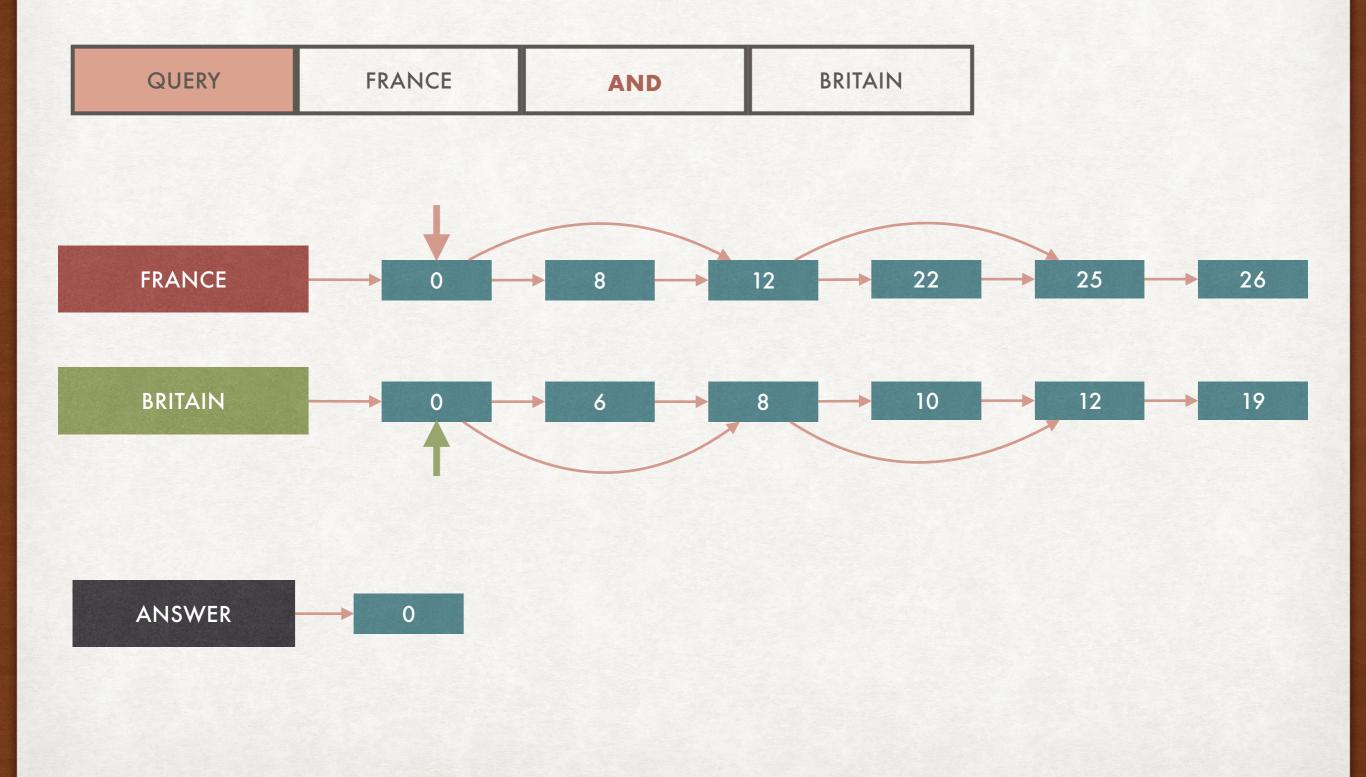


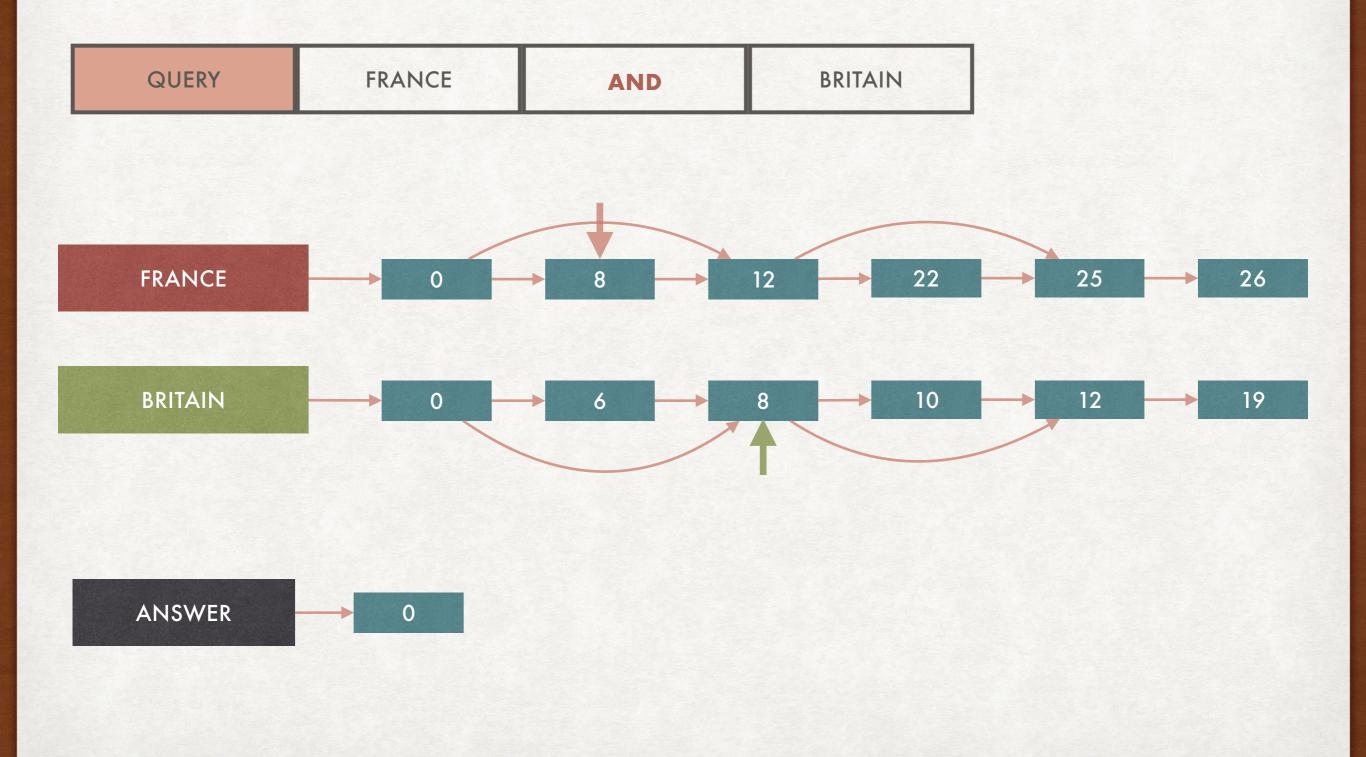
## SKIP LISTS FASTER INTERSECTION

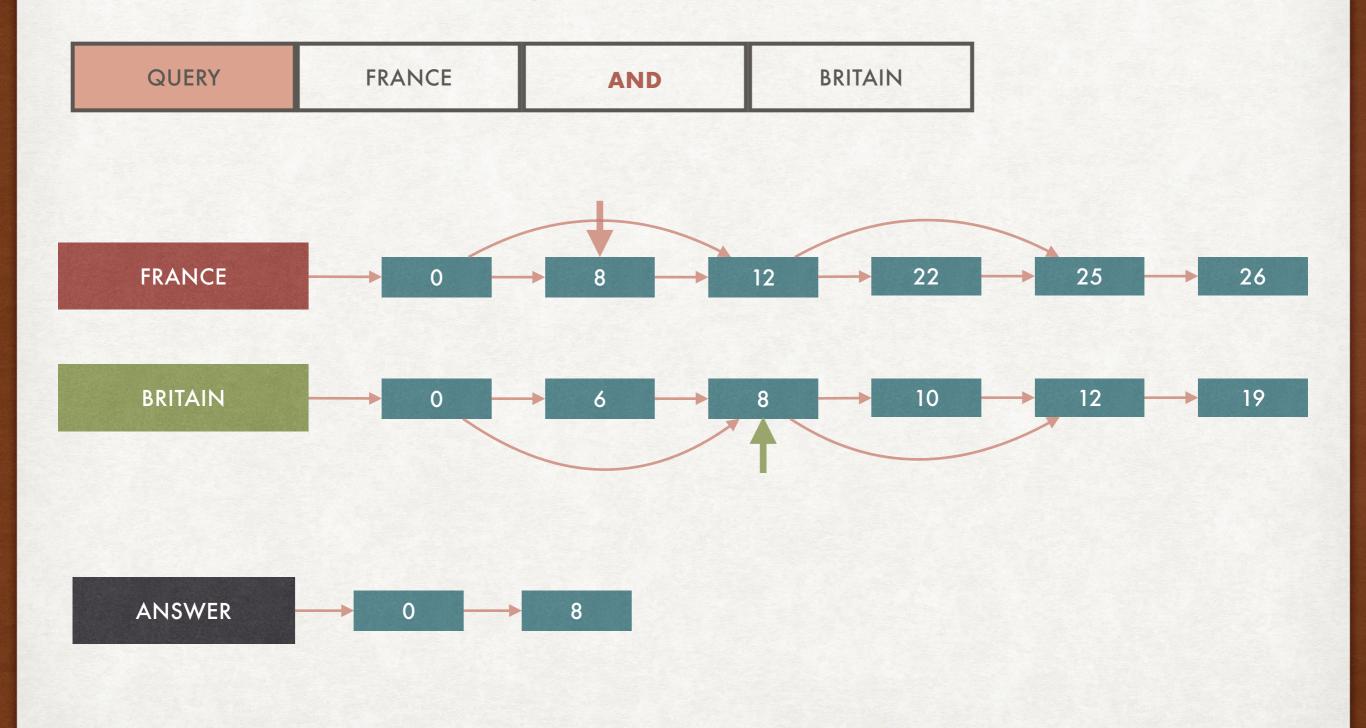
- We add additional forward pointers every k postings inside a list. The forward pointer "skips" a certain number of postings.
- A rule of thumb is, for a postings list of P postings to use  $\sqrt{P}$  evenly spaced skip pointers

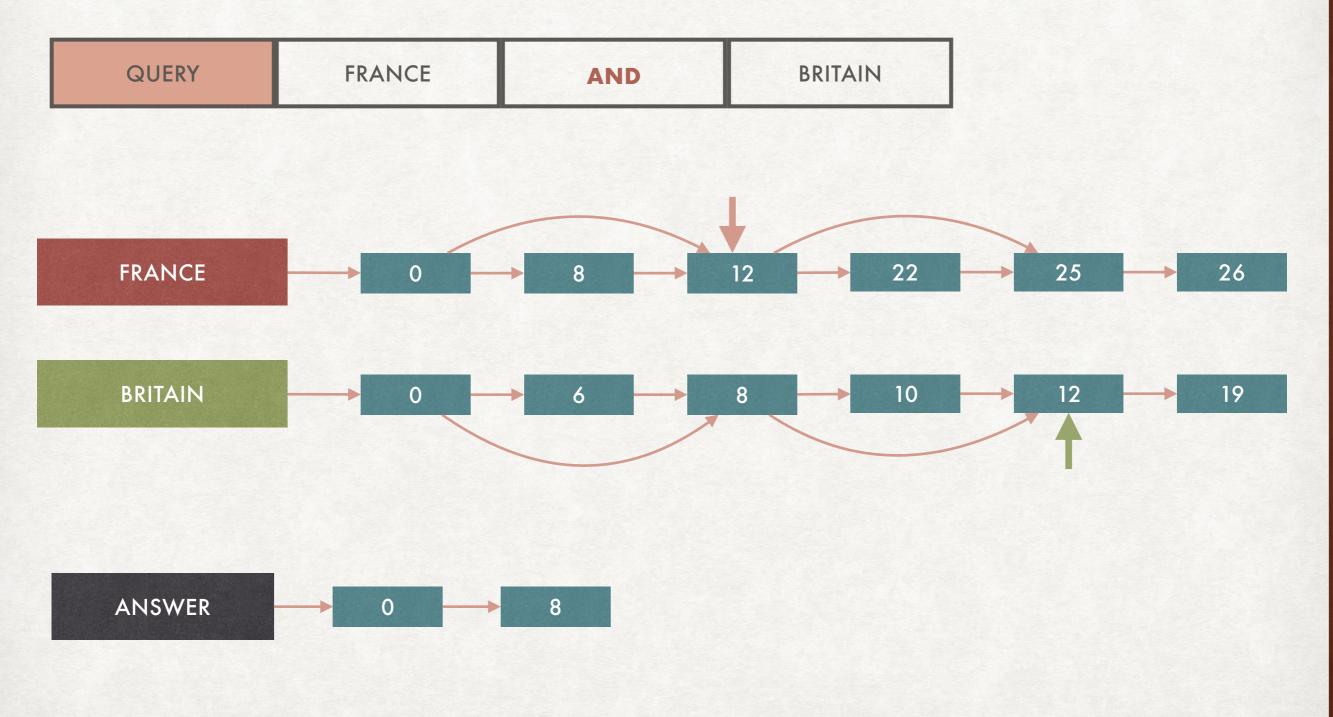


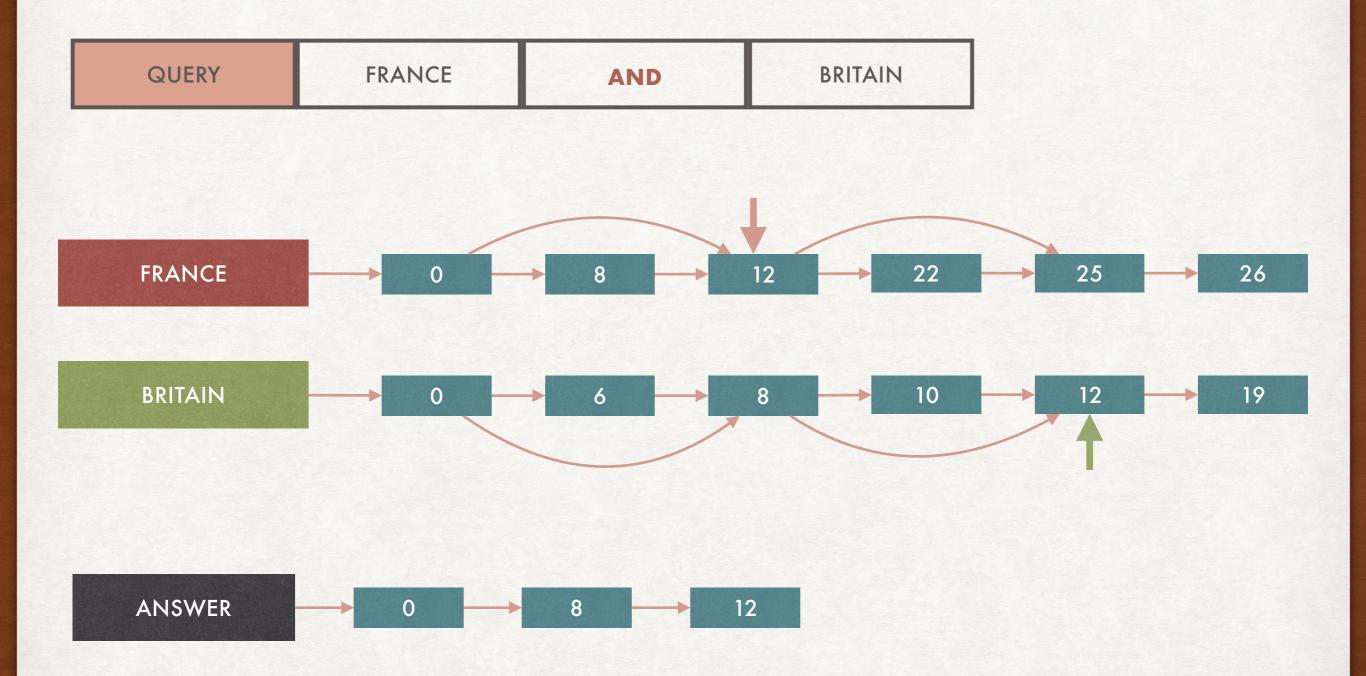


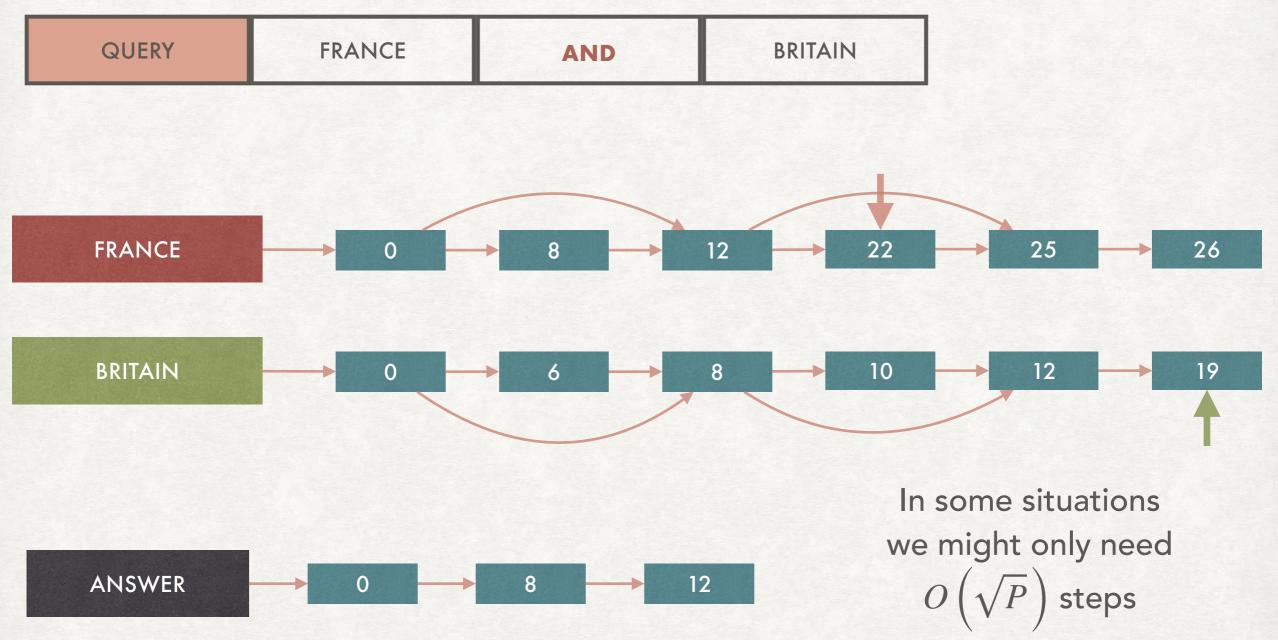












to traverse a list

## EXERCISES THE PRACTICAL PART

- We are going to implement some of the algorithms and data structure described in this course
- We use Python 3, but you can follow along with any other programming language
- While IR systems must be efficient, we will sometimes allow for inefficiencies for the sake of more readable code
- Dataset that we use: <u>http://www.cs.cmu.edu/~ark/personas/</u>, more than 42k movie descriptions