

# INFORMATION RETRIEVAL

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Lecture 2

# LECTURE OUTLINE

\*Now embellished with diagrams

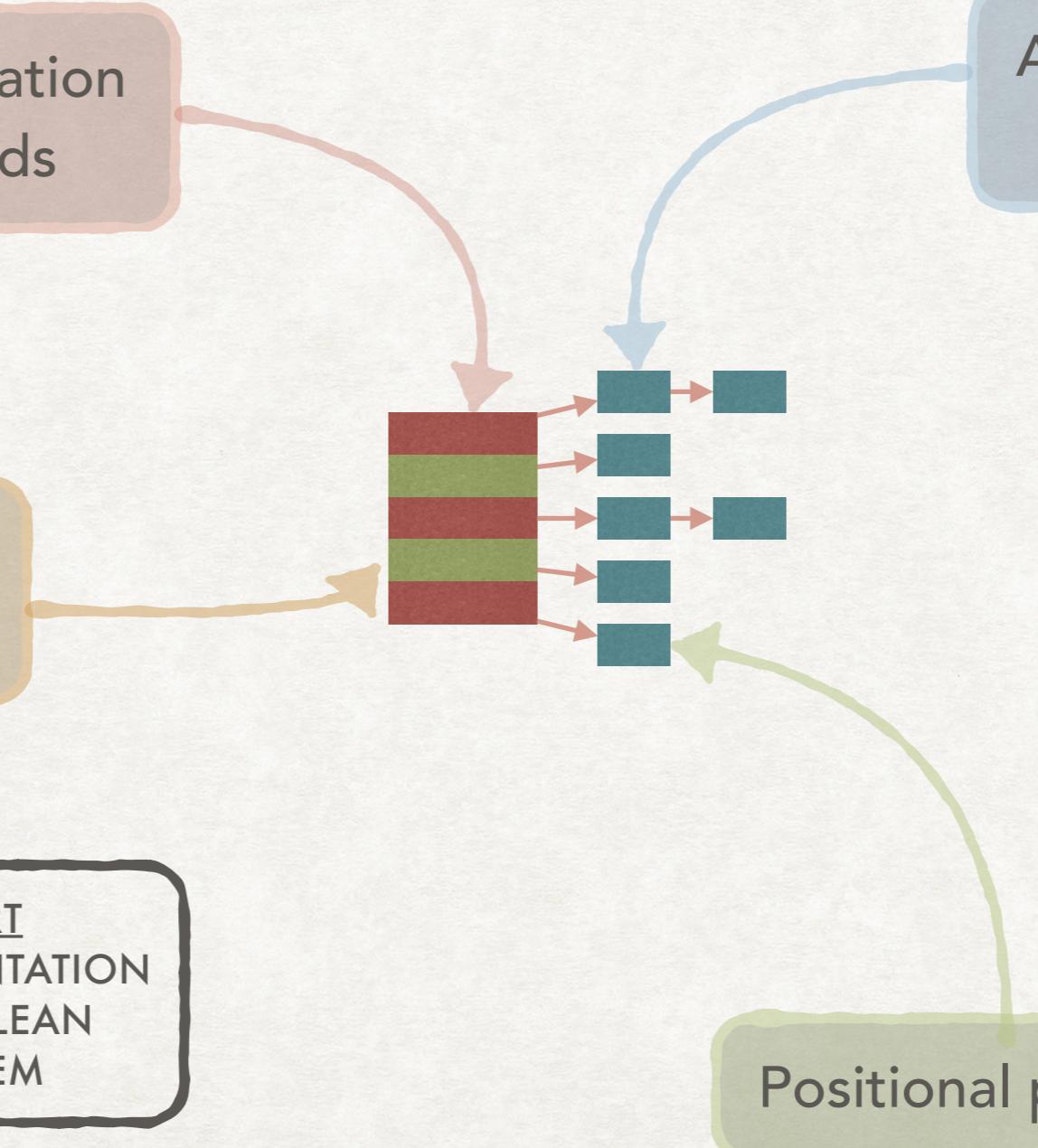
Stemming & Lemmatization  
Removing Stop Words

Basic operations on  
inverted indices

PRACTICAL PART  
A PYTHON IMPLEMENTATION  
OF A SIMPLE BOOLEAN  
RETRIEVAL SYSTEM

Arrays, linked lists,  
and skip lists

Positional postings



# **INVERTED INDEX: UNION AND INTERSECTION**

# HOW TO IMPLEMENT AN INVERTED INDEX

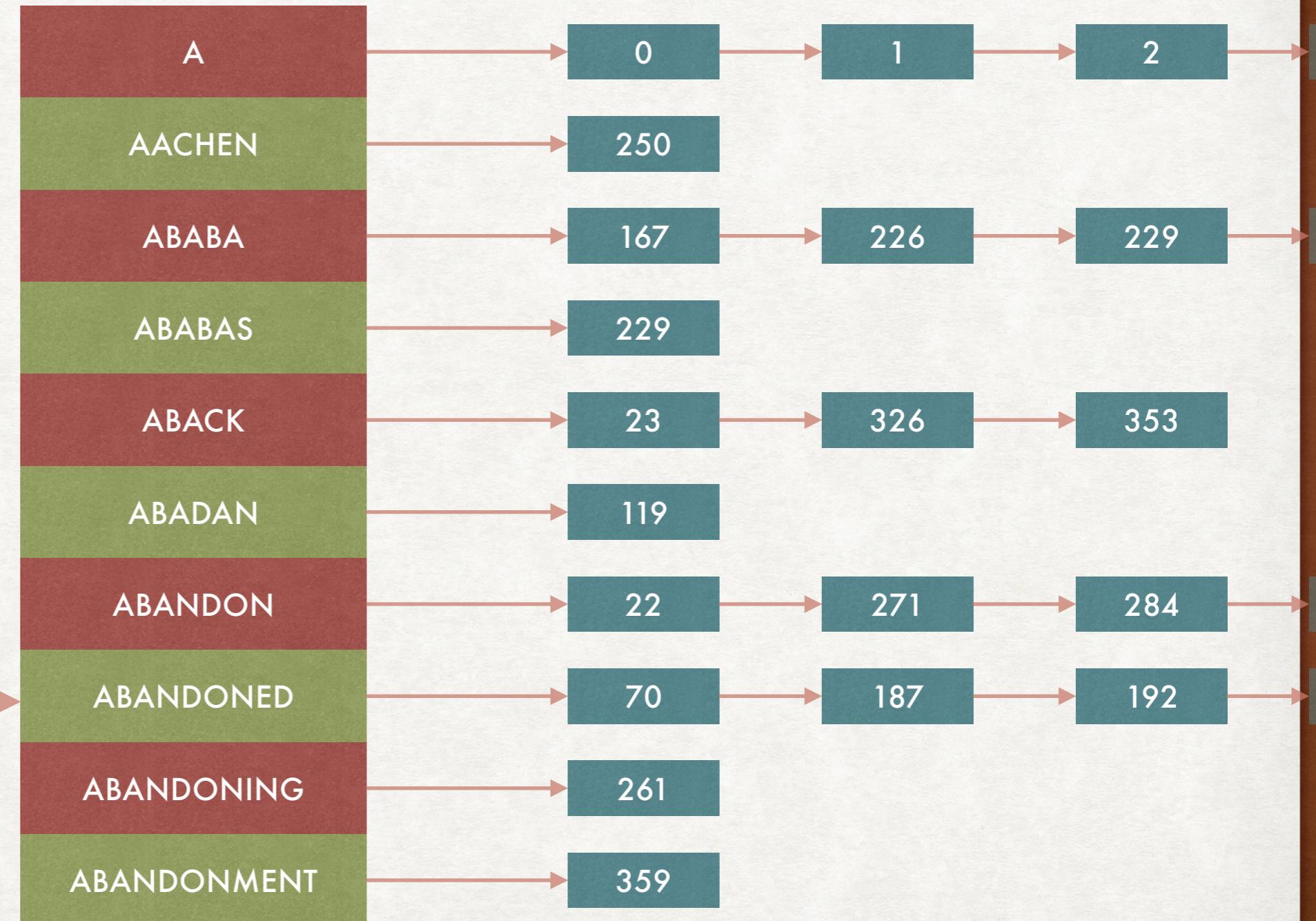
## BASIC IMPLEMENTATION AND OTHER IMPROVEMENTS

- We will spend some time in discussing how to implement and improve the inverted index
- Basic functionality: answer queries of the form
  - term1 AND term2
  - term1 OR term2
- Additional functionalities:  
term1 NEAR term2, "term1 term2", term1\* (wildcards), etc.
- How to compress the index, how to update it, etc.

# ANSWERING A SIMPLE QUERY

## A SINGLE WORD QUERY

QUERY  
ABANDONED



We find the term  
in the list of terms

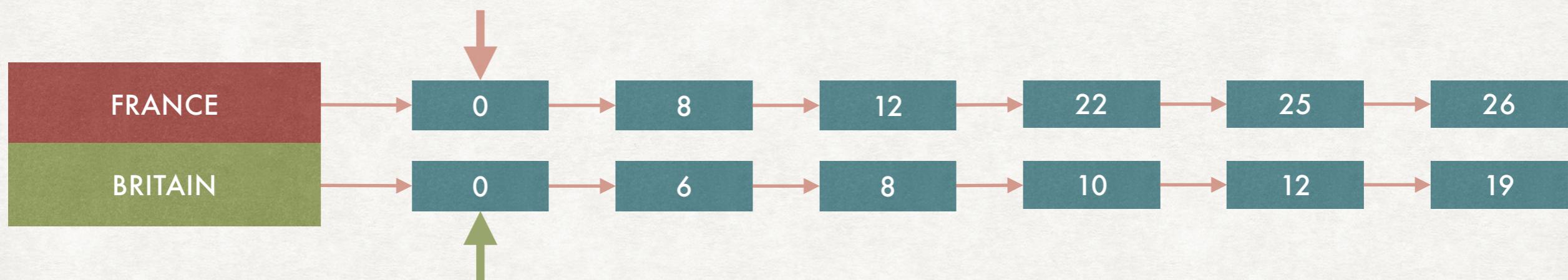
We return the  
associated list of terms

# ANSWERING AN “AND” QUERY

## NOW WITH TWO WORDS

QUERY	FRANCE	AND	BRITAIN
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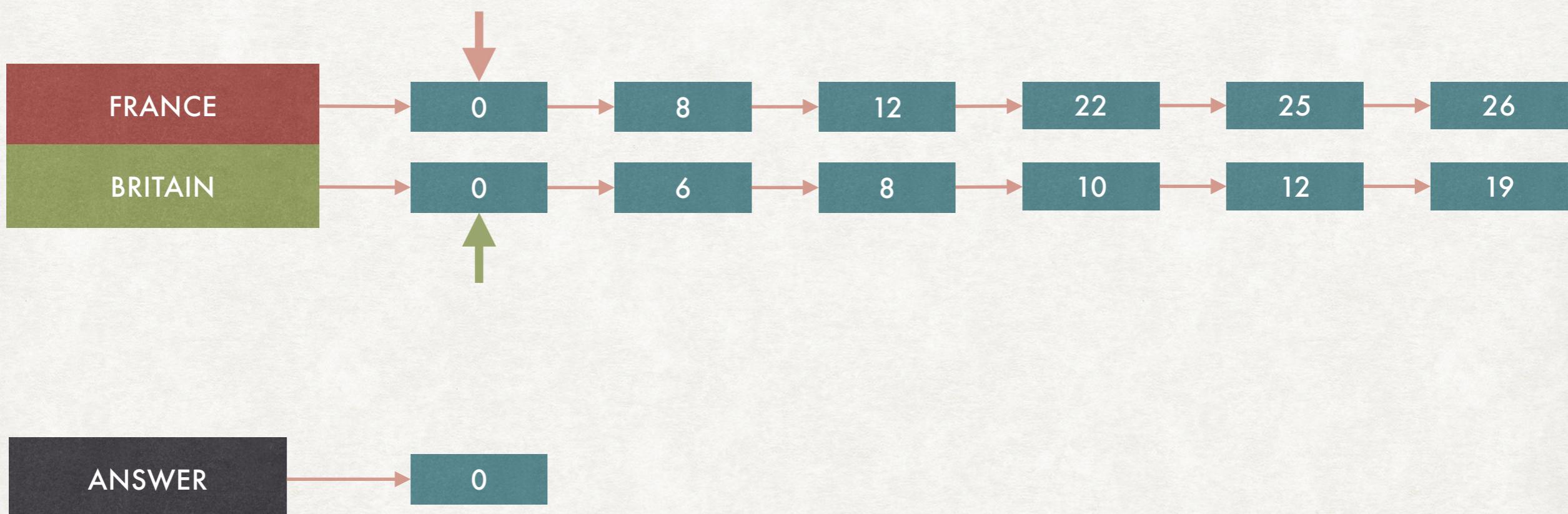
Now we need to compare the two lists of documents



ANSWER

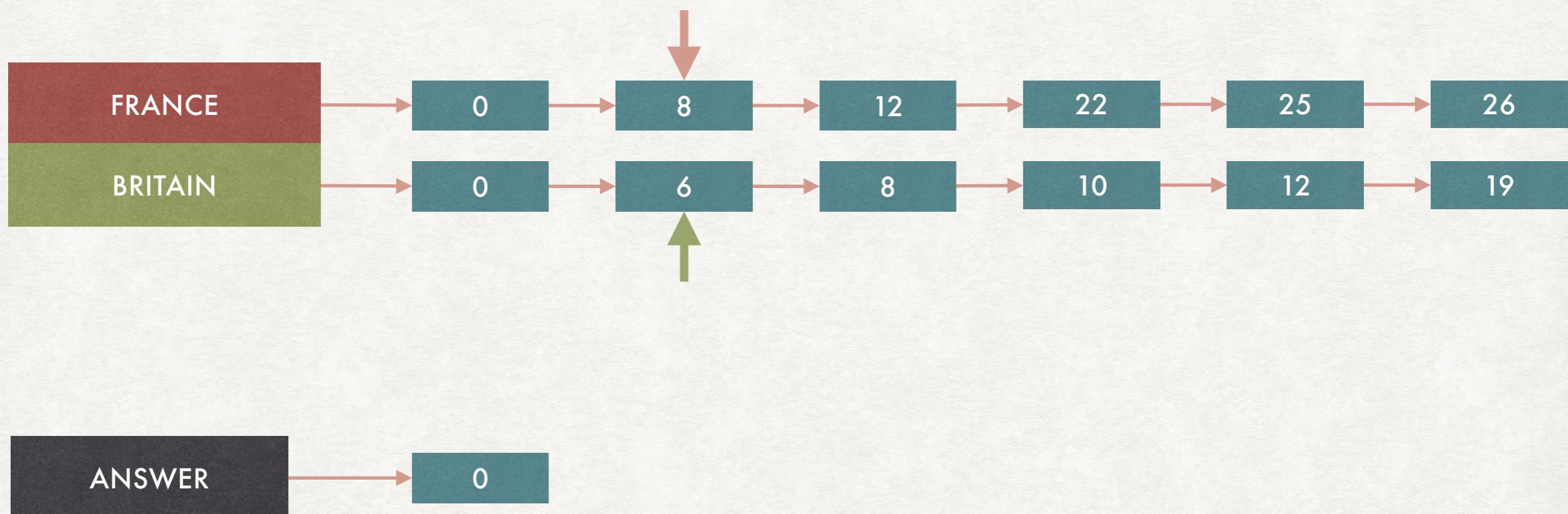
# ANSWERING AN “AND” QUERY

## NOW WITH TWO WORDS



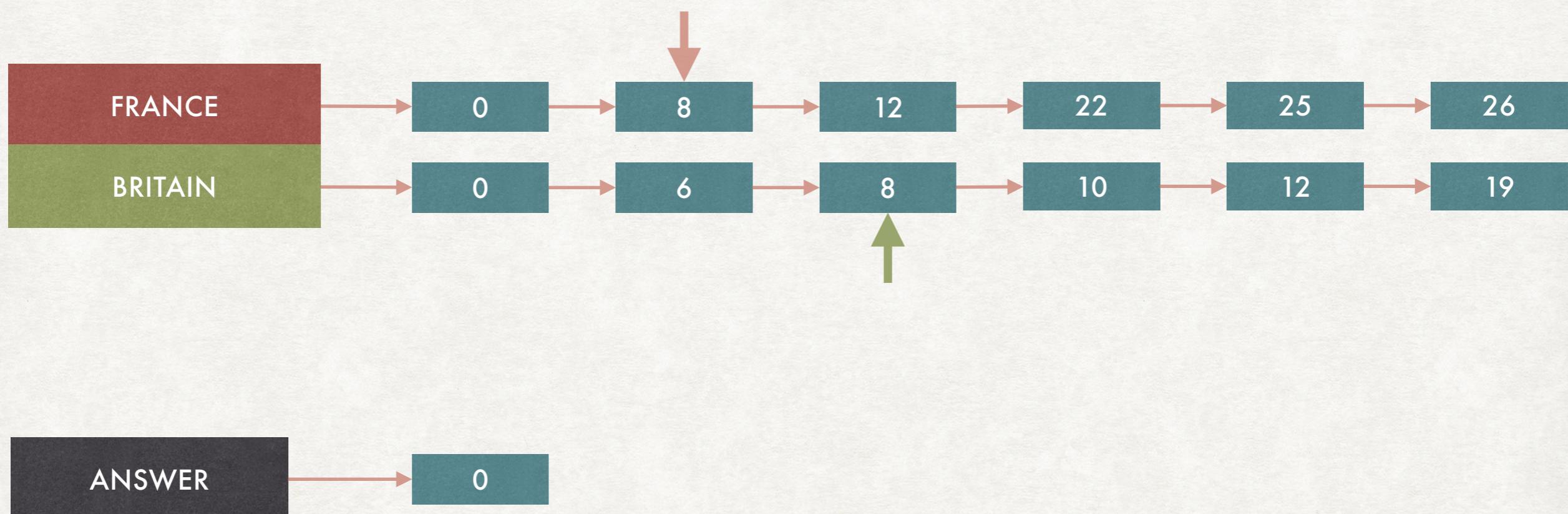
# ANSWERING AN “AND” QUERY

## NOW WITH TWO WORDS



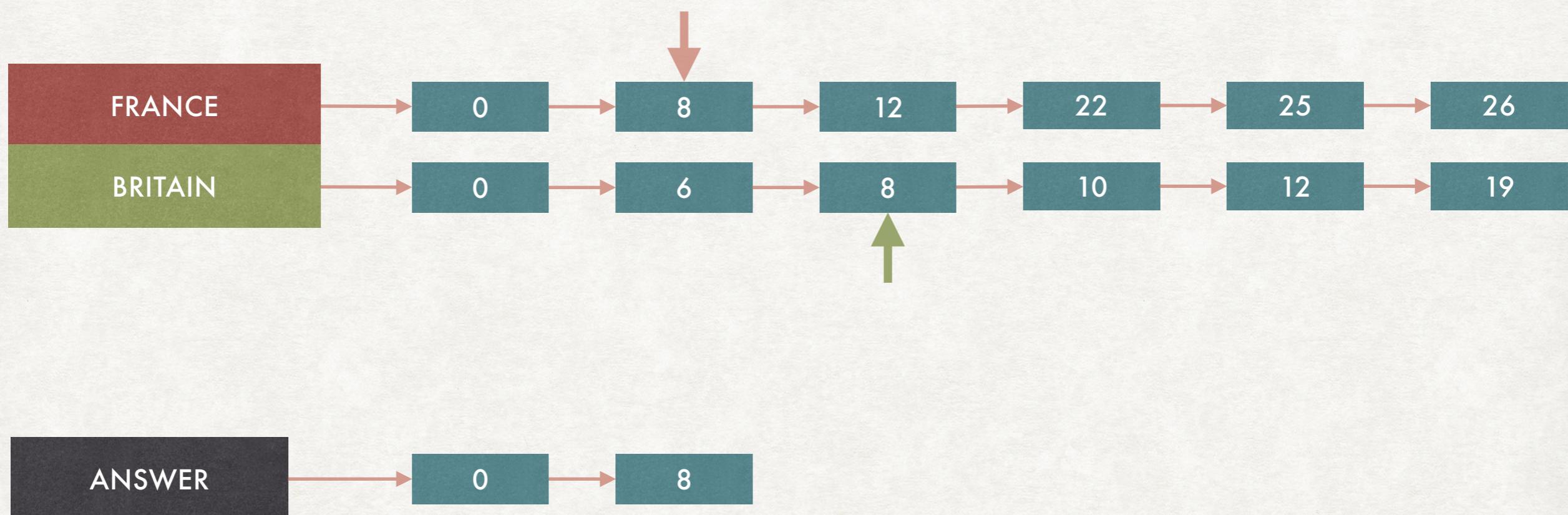
# ANSWERING AN “AND” QUERY

## NOW WITH TWO WORDS



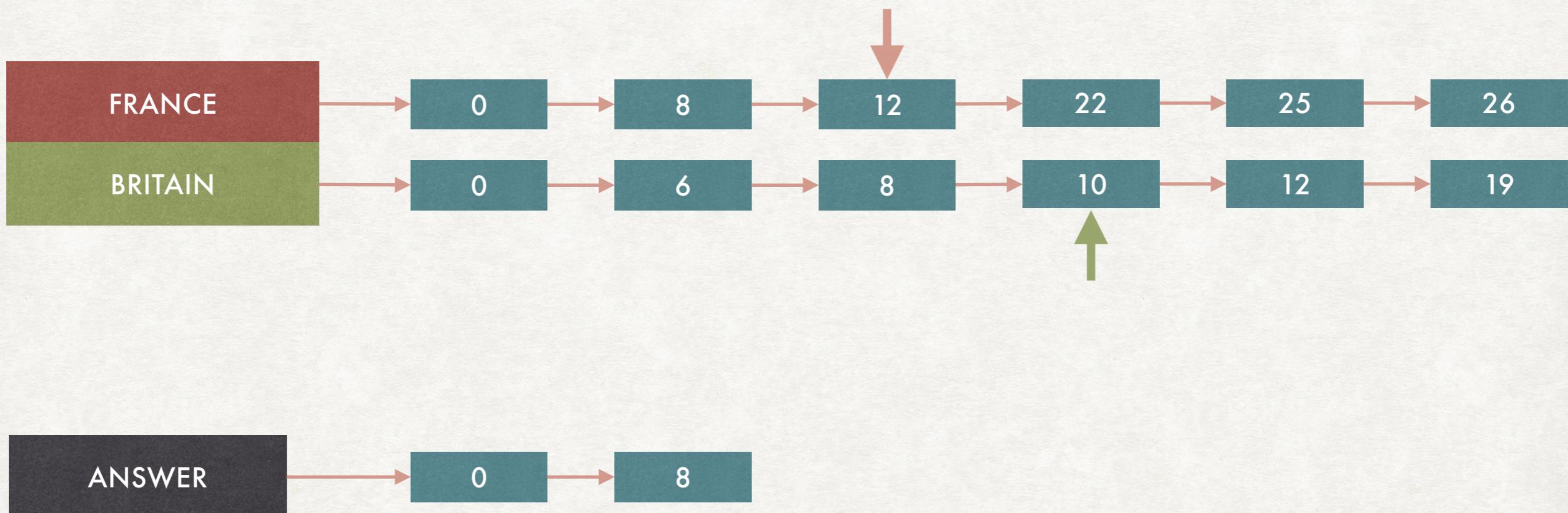
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## NOW WITH TWO WORDS



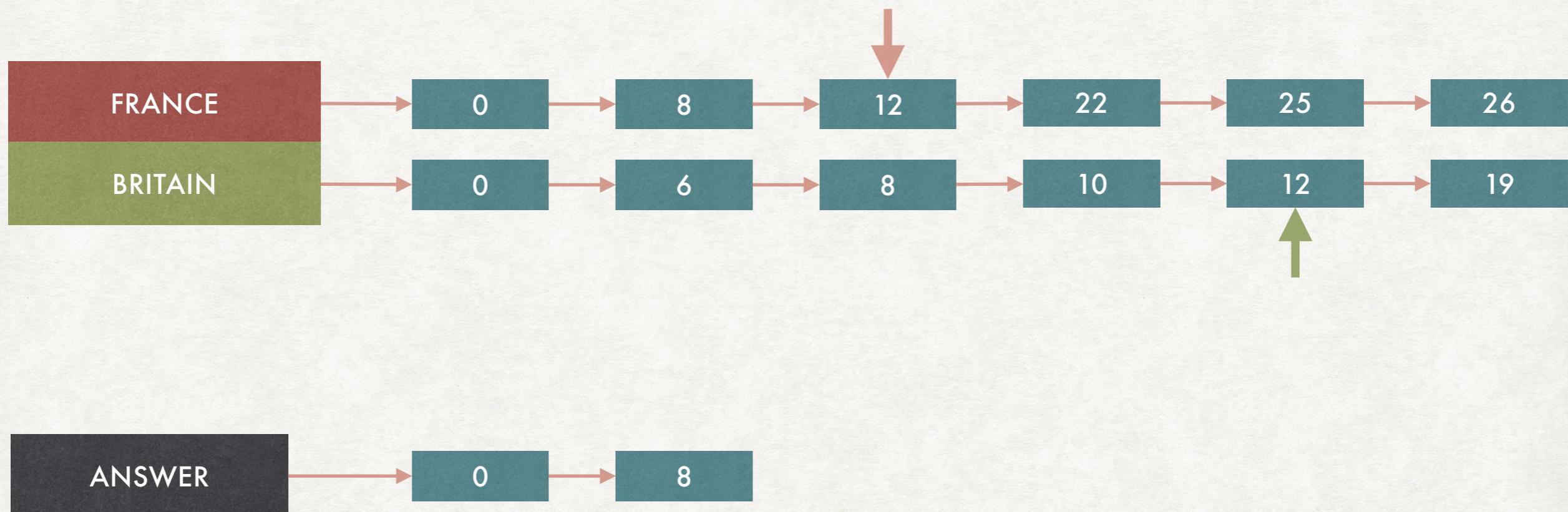
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## NOW WITH TWO WORDS



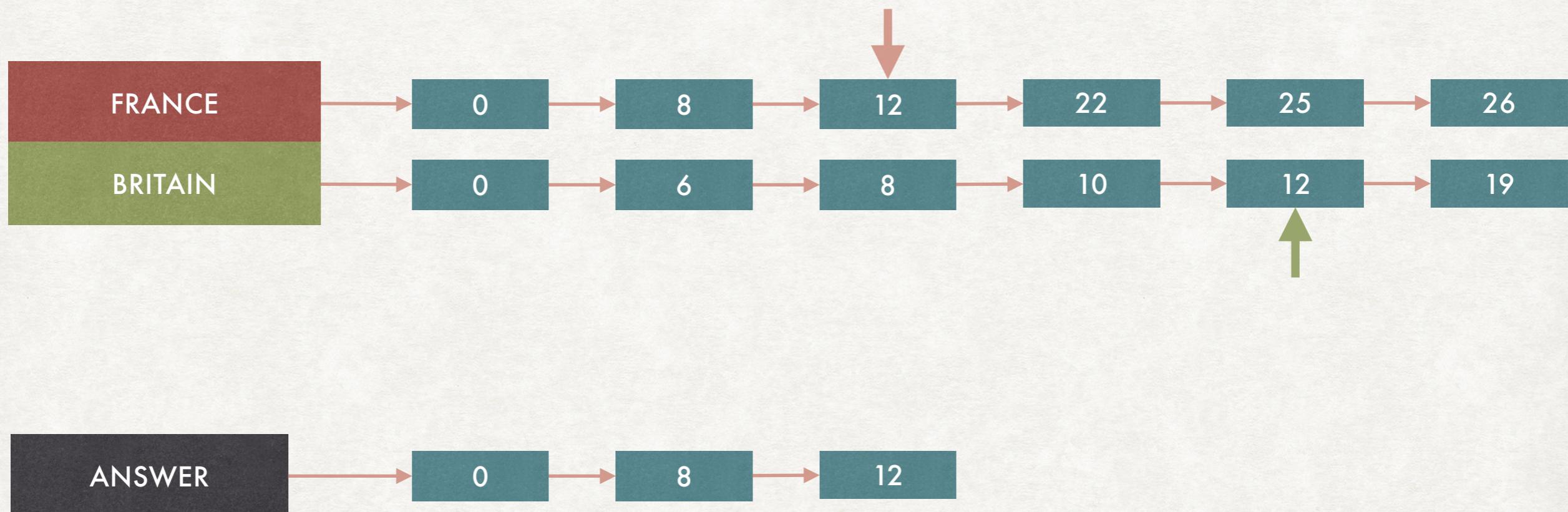
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## NOW WITH TWO WORDS



# ANSWERING AN “AND” QUERY

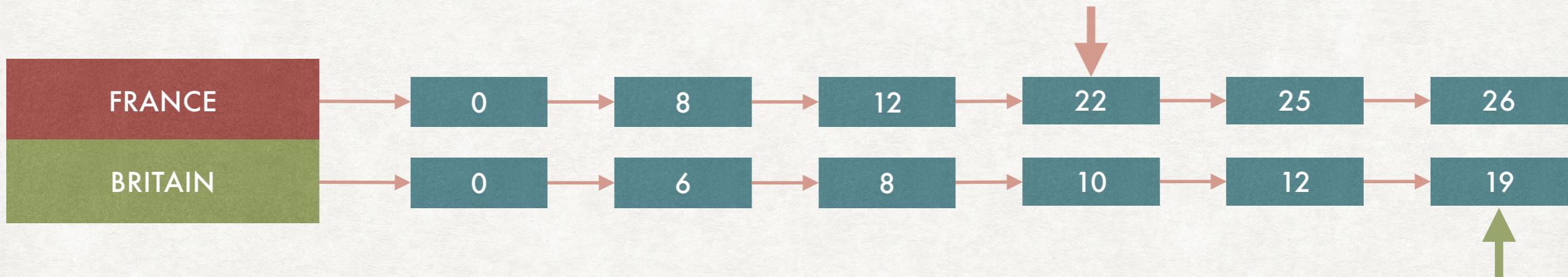
## NOW WITH TWO WORDS



# ANSWERING AN “AND” QUERY

## NOW WITH TWO WORDS

QUERY	FRANCE	AND	BRITAIN
-------	--------	-----	---------



Complexity: linear in the lengths of the lists

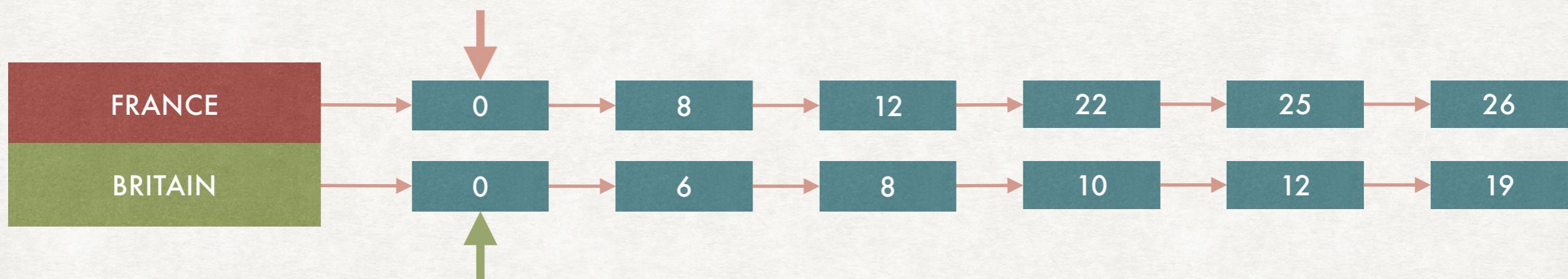


Size of the answer  $\leq$  minimum of the lengths of the lists

# ANSWERING A “OR” QUERY WITHOUT DUPLICATES!

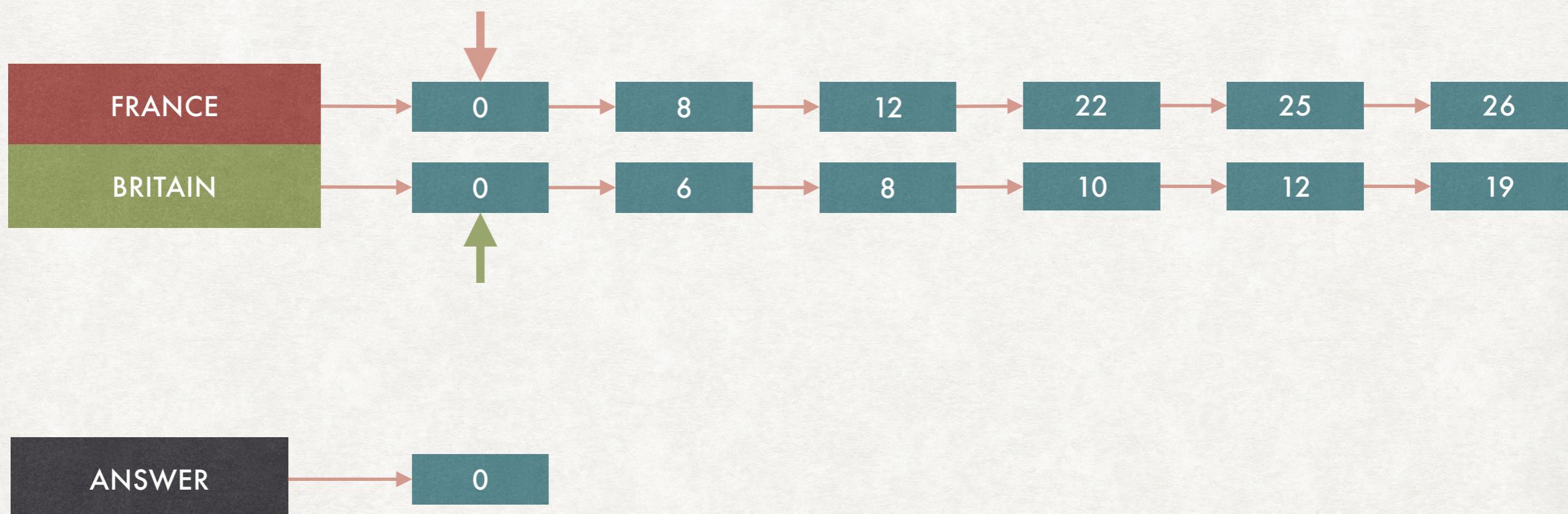
QUERY	FRANCE	OR	BRITAIN
-------	--------	----	---------

We still need to compare the two lists of documents

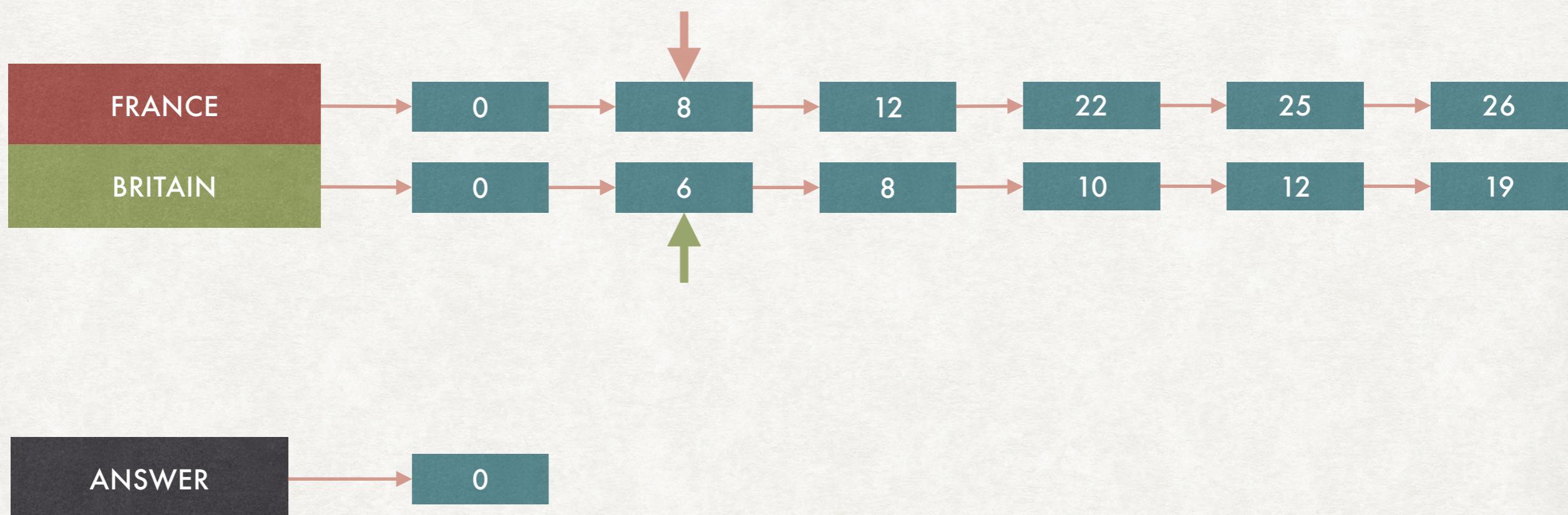


ANSWER

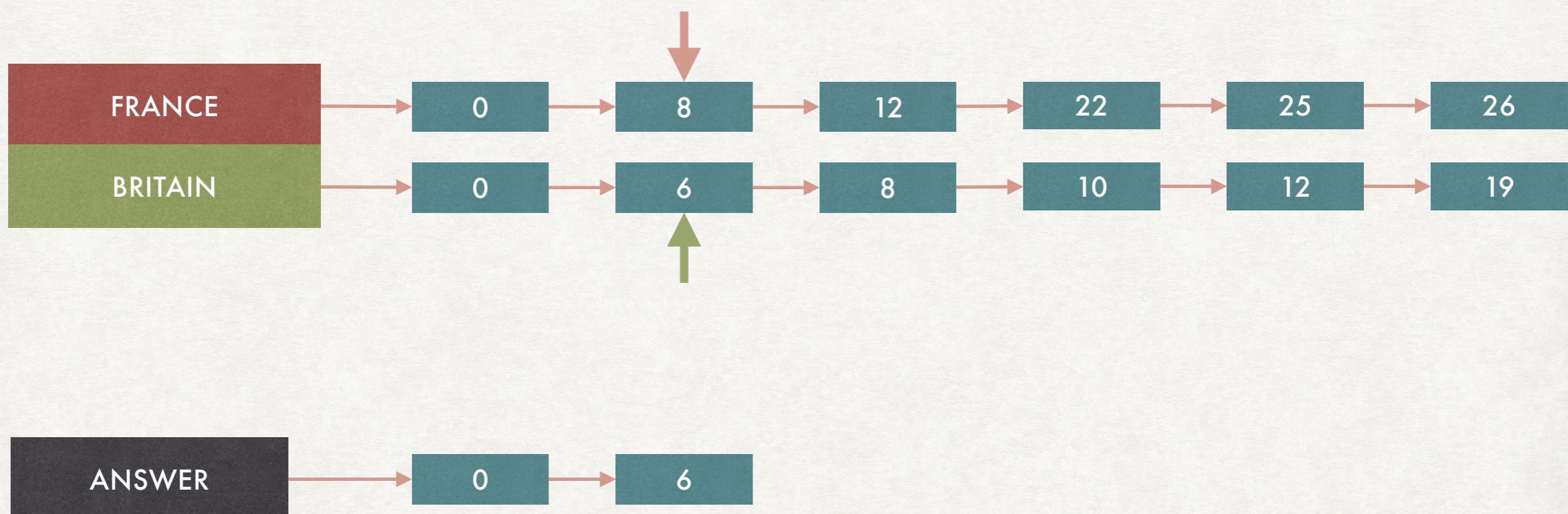
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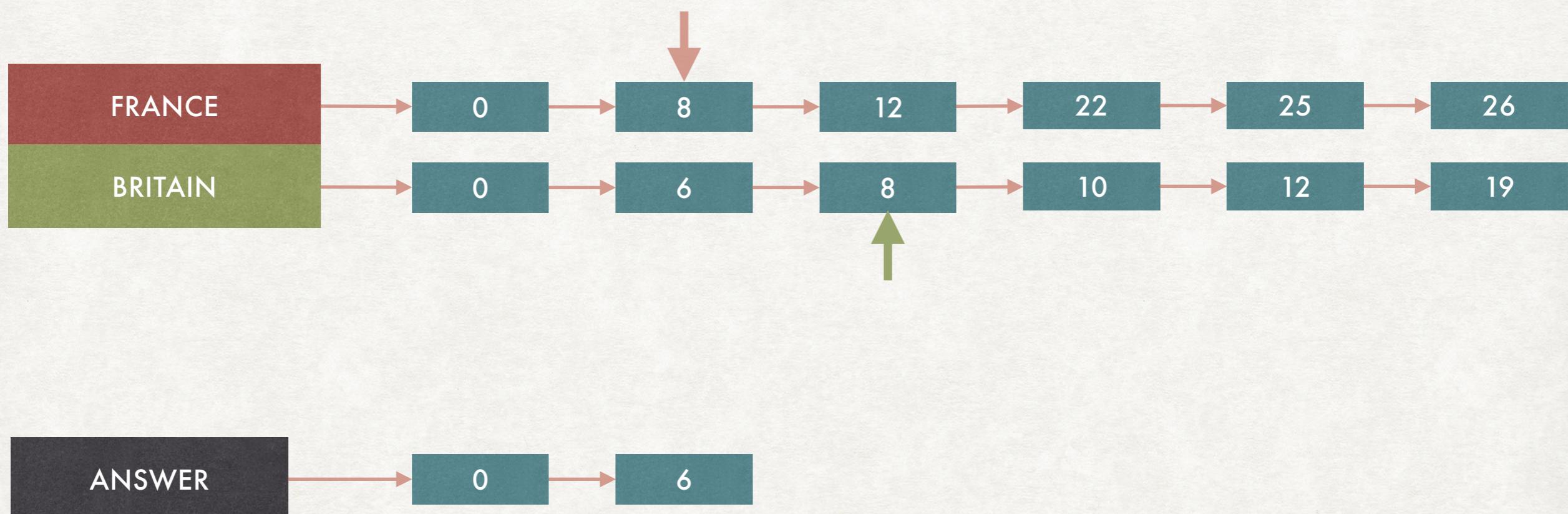
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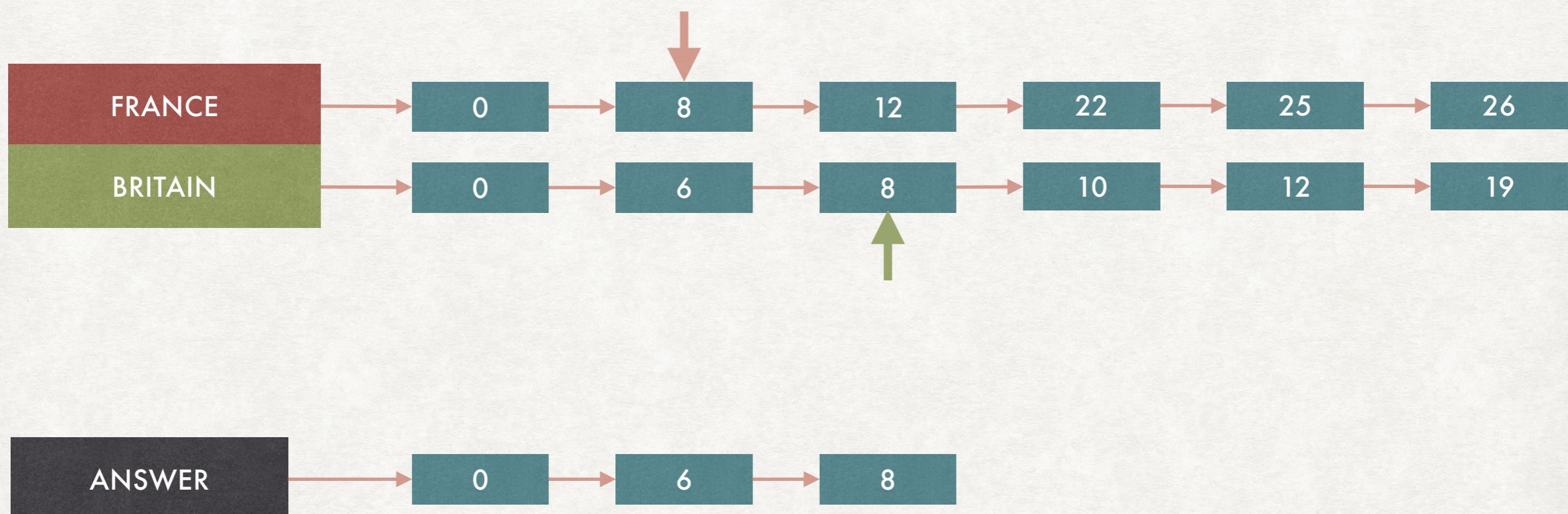
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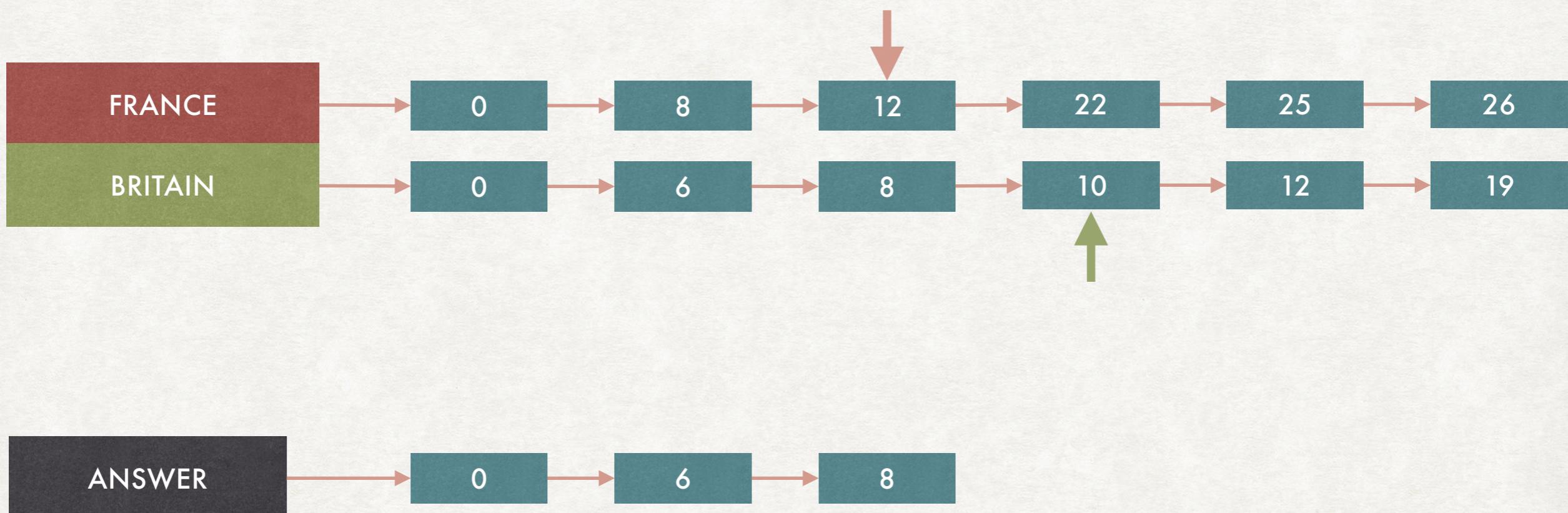
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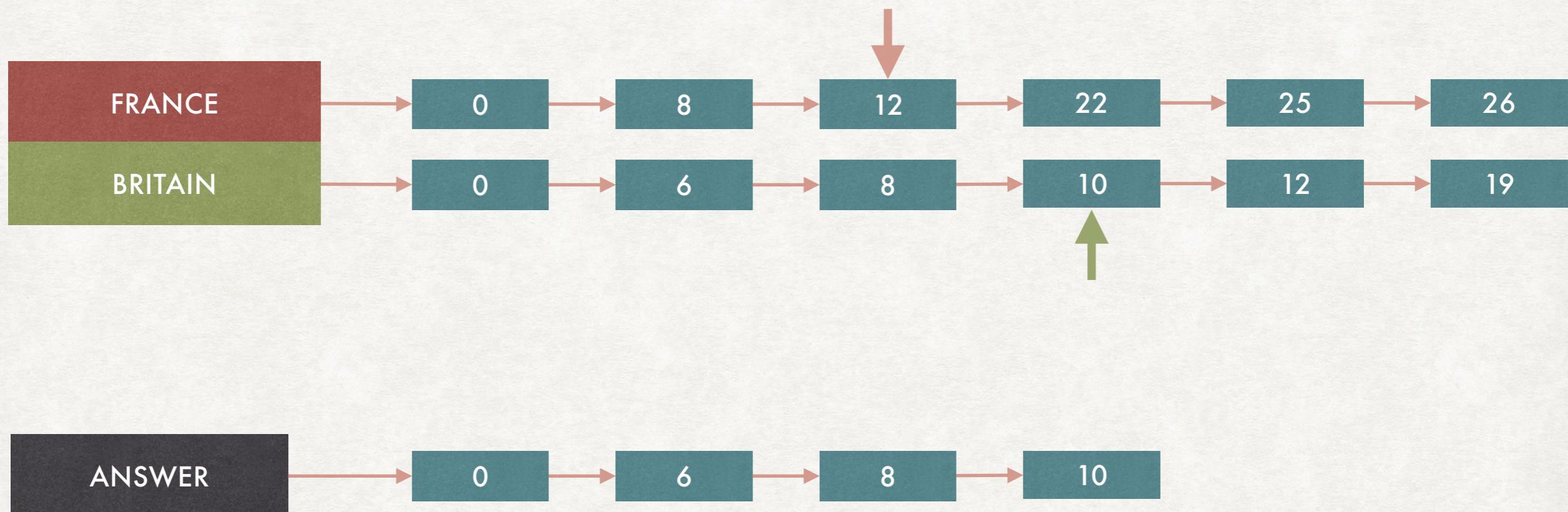
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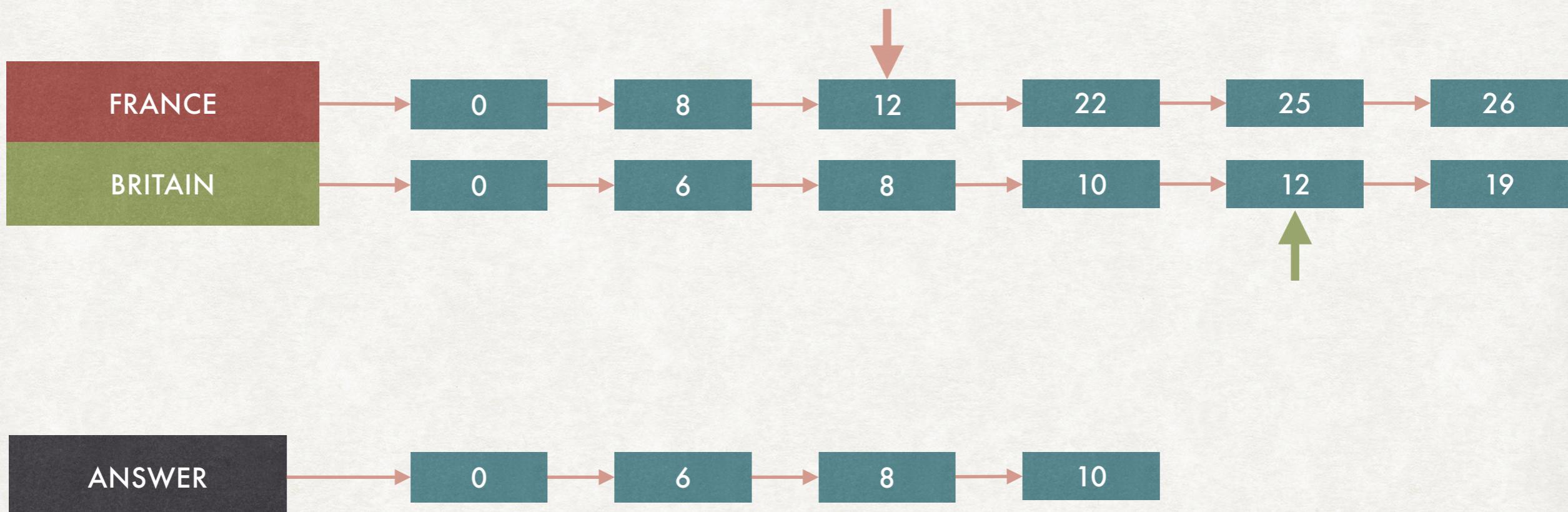
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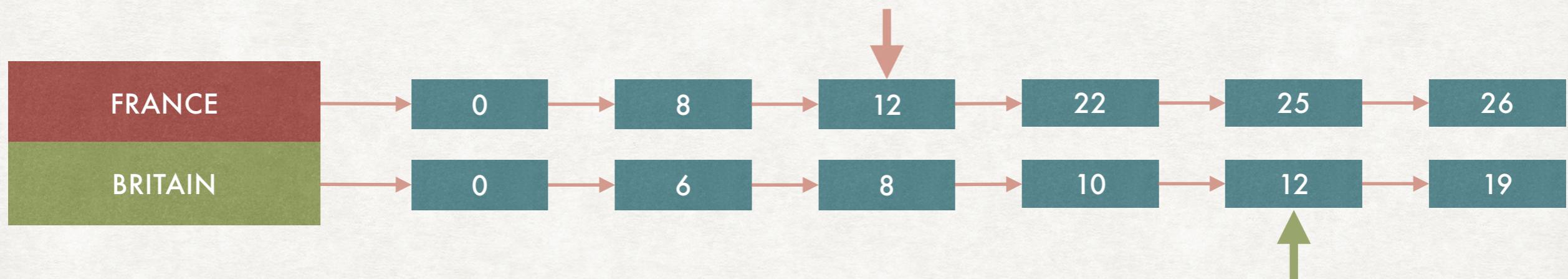
# ANSWERING A “OR” QUERY WITHOUT DUPLICATES!



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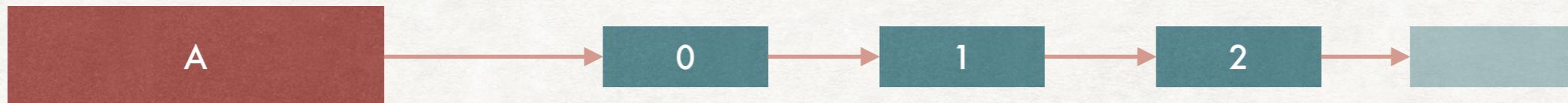
Complexity: linear in the lengths of the lists



Size of the answer  $\leq$  sum of the lengths of the lists

# IS THAT ALL?

HINT: NO



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



Some terms are very similar semantically.

Example

Do we really want to keep “CAR” and “CARS” separated?

# **IMPROVING THE QUALITY OF RETRIEVAL**

# TERMINOLOGY (4)

## THIS TIME FOR TOKENIZATION

- **Token:** instance of a sequence of characters
- **Type:** collection of all tokens with the same character sequence
- **Term:** a type that is inserted into the dictionary

THE CAT IS INSIDE THE BOX      Text

THE CAT IS INSIDE THE BOX      Tokens

THE CAT IS INSIDE THE BOX      Types (notice only one instance of "the")

CAT      INSIDE      BOX      Terms (after removal of common words)

# TOKENIZATION

## SPLITTING THE TEXT IN WORDS

- First step in the indexing process is to decide what is the granularity of the indexing (i.e., return chapters or paragraphs instead of entire books).
- 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 other 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:

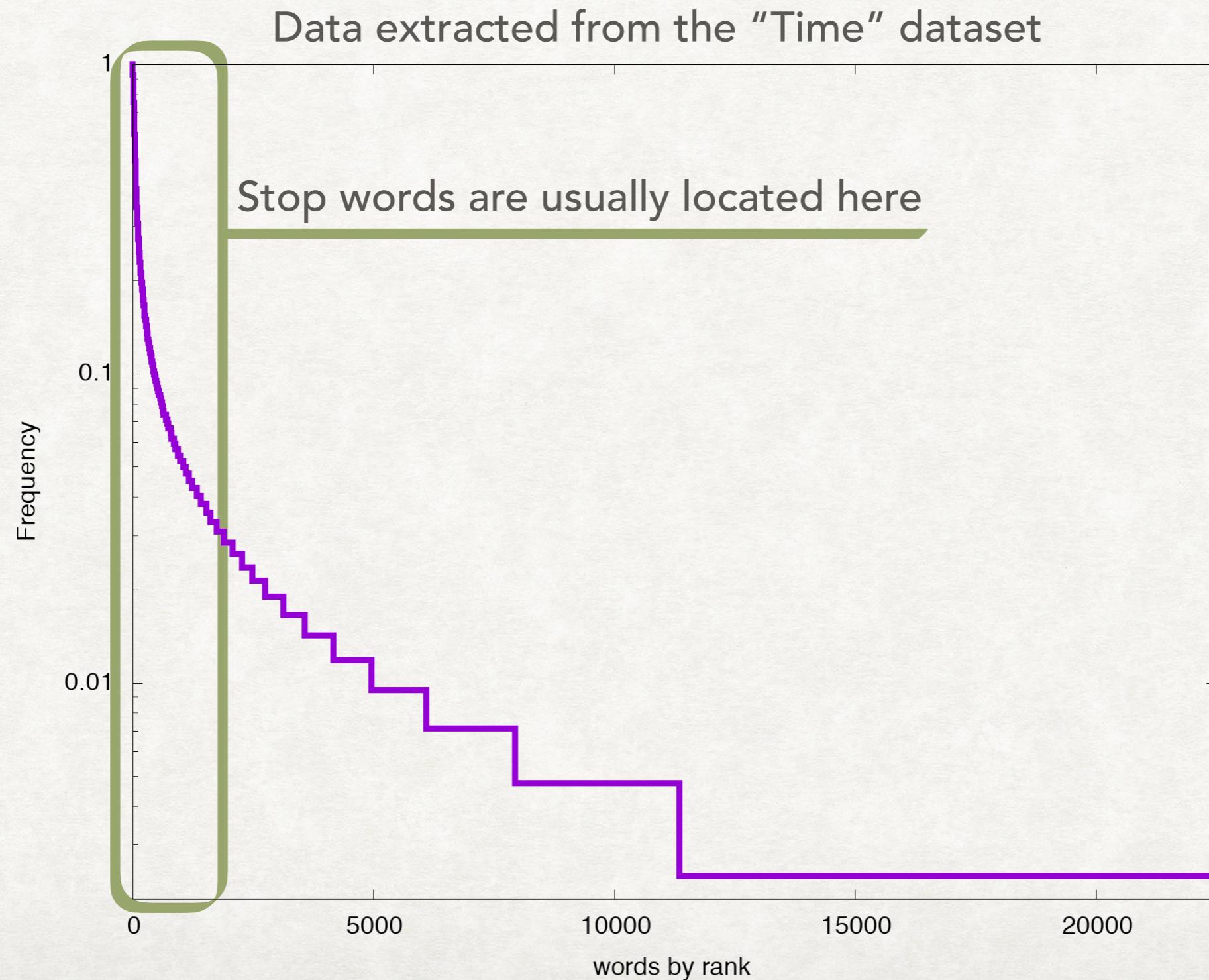


Some terms are not useful: “A” is in all the documents!

- **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.

# 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 is to have small (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").
  - 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

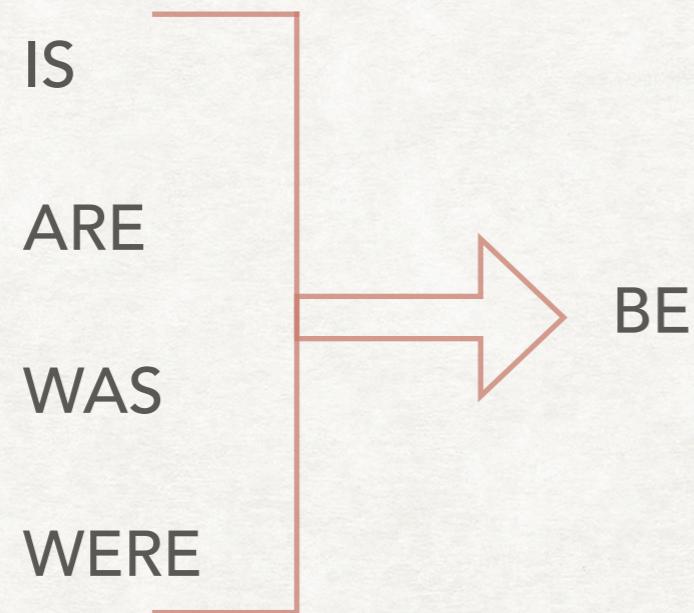
windows (can mean both the object and the OS)  Windows (the OS)

This can be solved by saving (possibly asymmetric) relations between token

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

# STEMMING AND LEMMATIZATION

## REDUCE WORDS TO A COMMON BASE FORM



**Idea**  
reduce all variants of a word  
to a “common root”

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

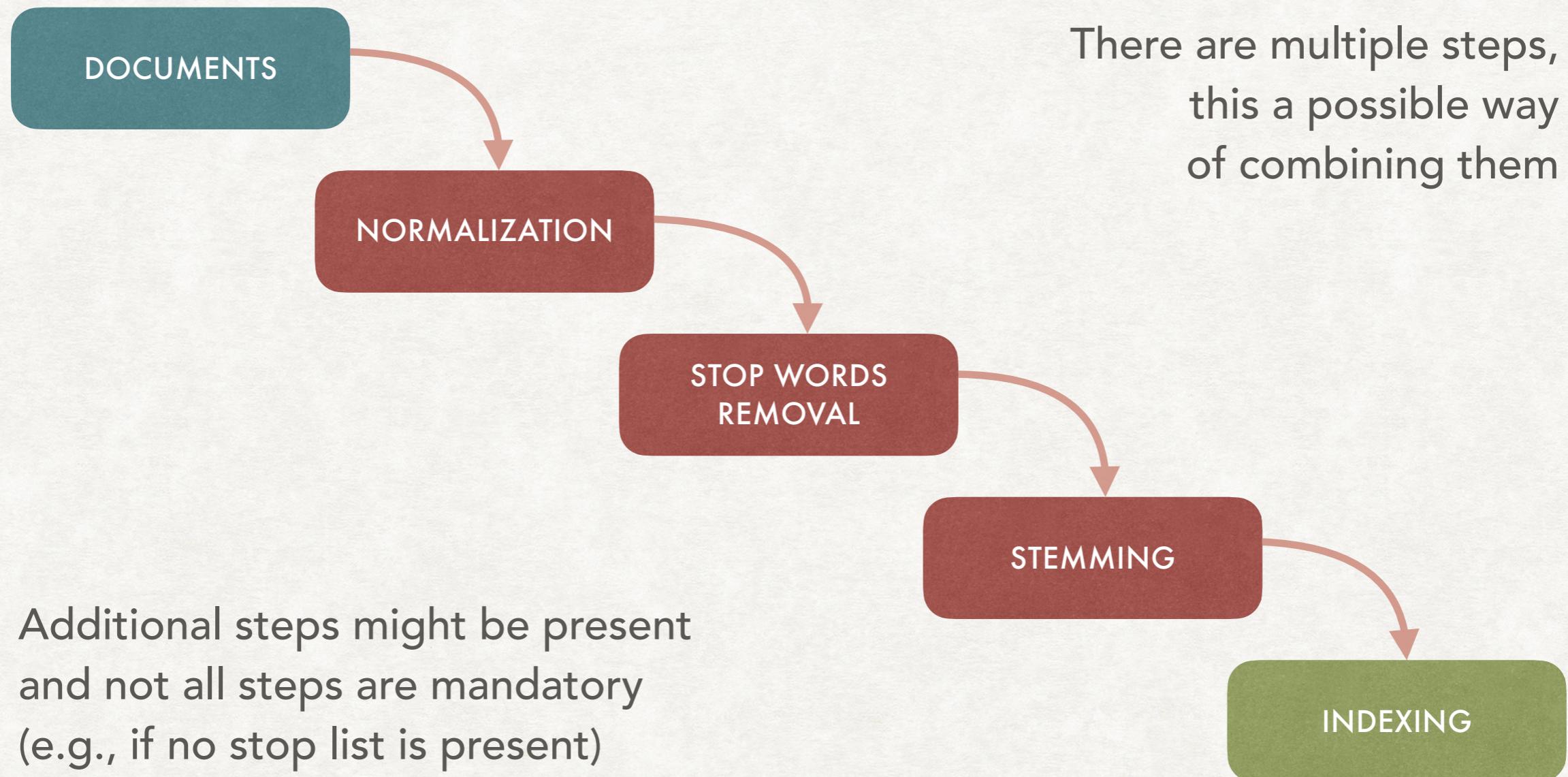
Five stages applied sequentially.

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

Rule	
SSES → SS	caressess → caress
IES → S	poinies → poni
SS → SS	caress → caress
S →	cats → cat

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

# THE “PREPROCESSING” PIPELINE



# **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

THE CAT

CAT IS

IS INSIDE

INSIDE THE

THE BOX

Terms

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

“inside the box”



“inside the” AND “the box”

# BIWORD INDEXES

## POSSIBLE PROBLEMS

Text:                   **INSIDE THE HOUSE THERE IS THE BOX**

Original Query:     “inside the box”                           **No Match**

Rewritten Query:    “inside the” AND “the box”                           **Match**

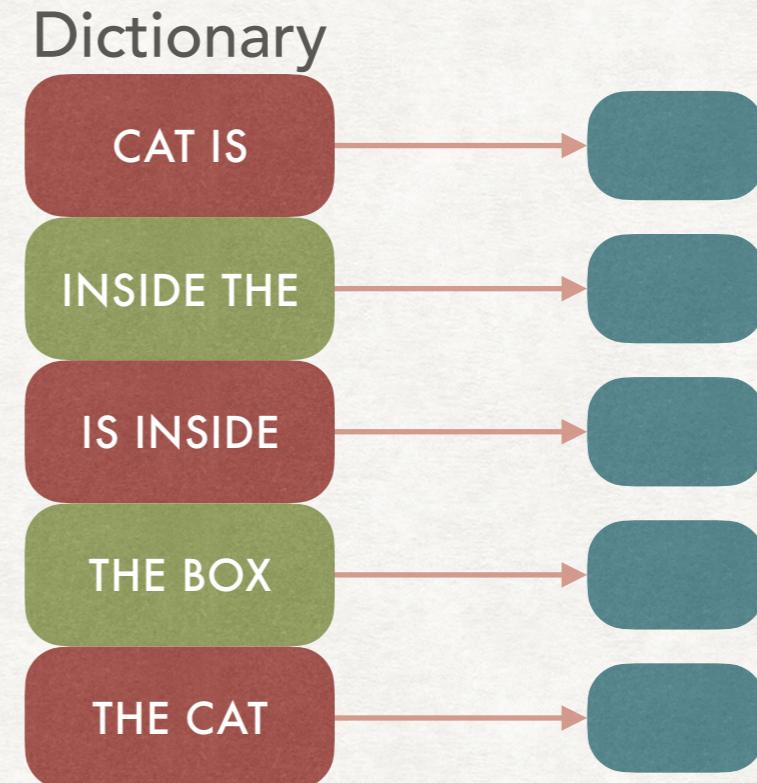
Rewriting the query might generate false positives  
(but it works quite well in practice)

# 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

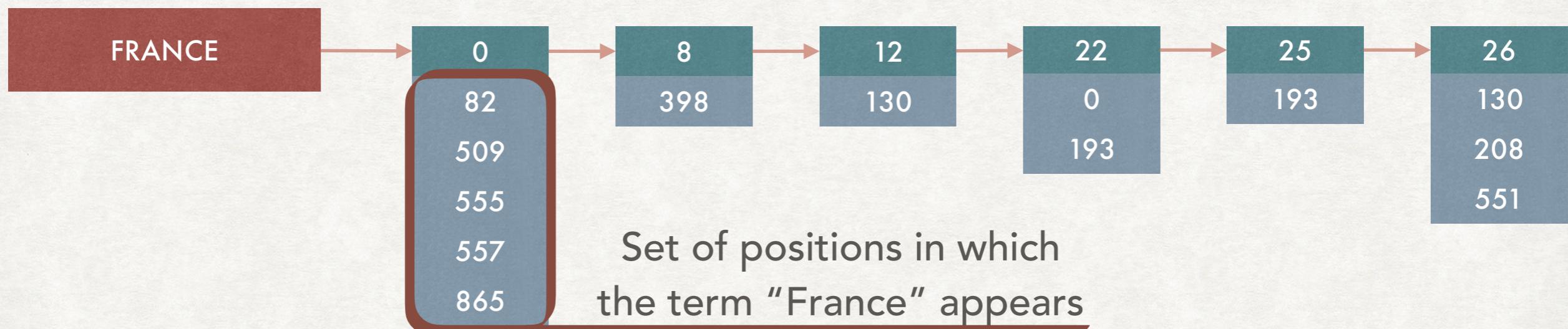
## EXTENSIONS AND FURTHER OBSTACLES

- 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

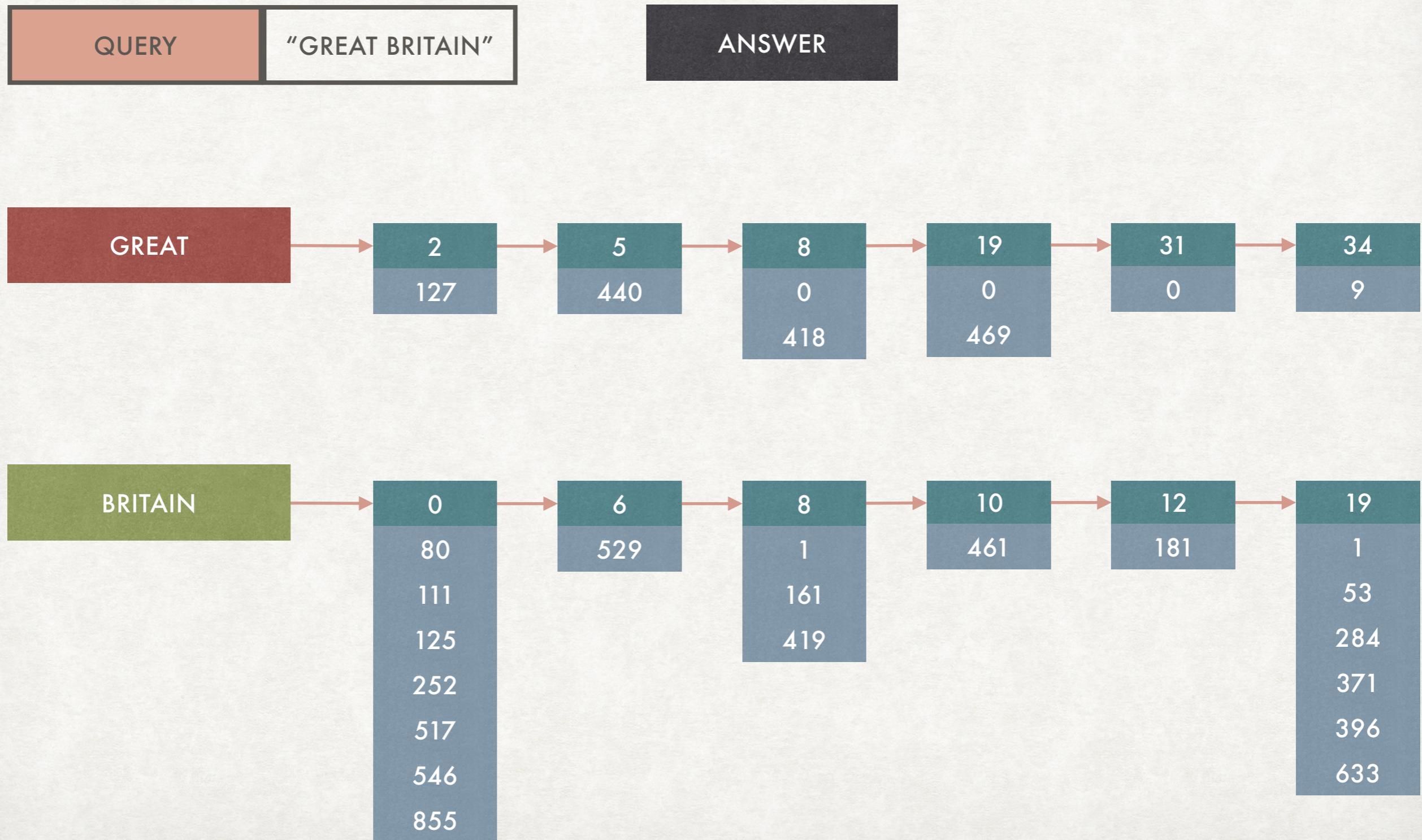
# POSITIONAL INDEXES

## 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.



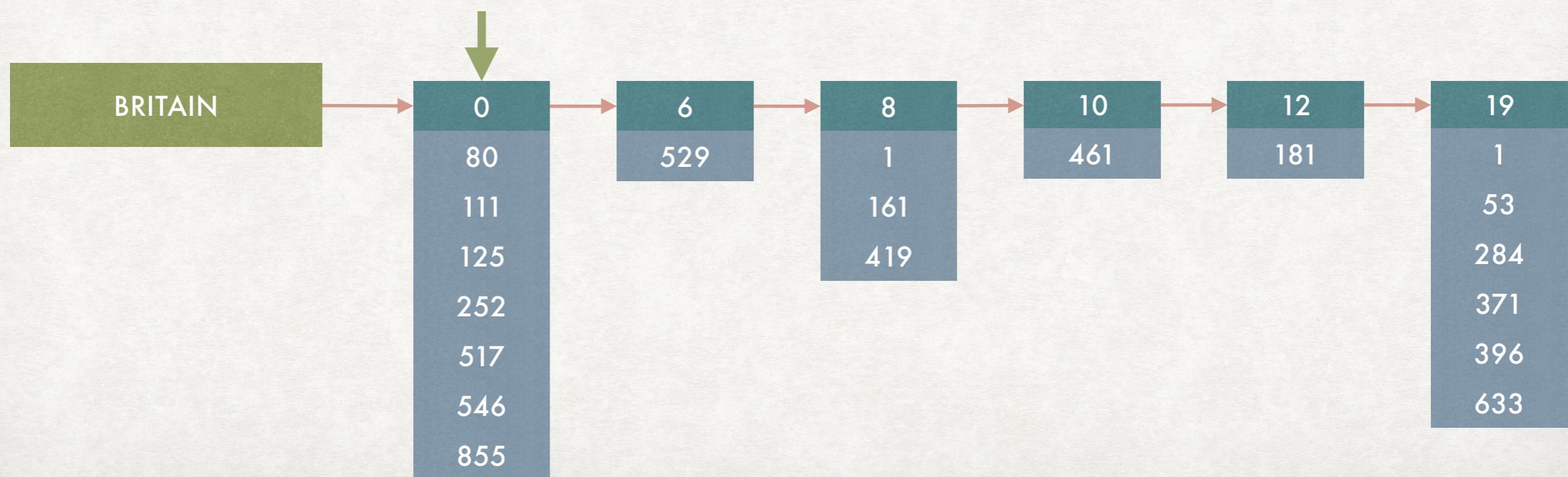
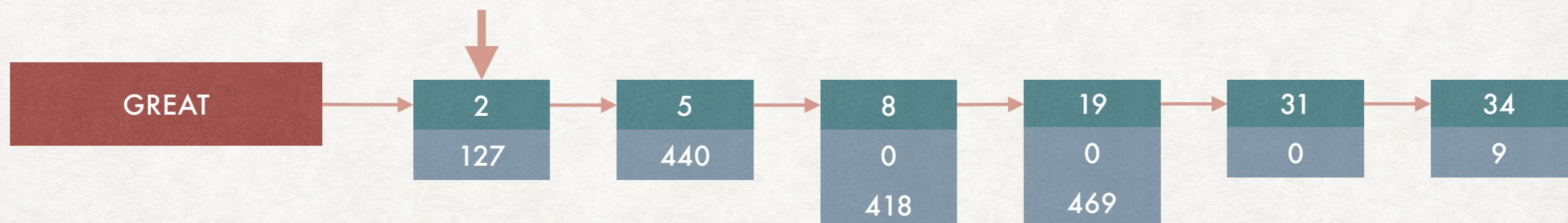
# ANSWERING A PHRASE QUERY WITH POSITIONAL INDEXING



# ANSWERING A PHRASE QUERY WITH POSITIONAL INDEXING

QUERY      "GREAT BRITAIN"

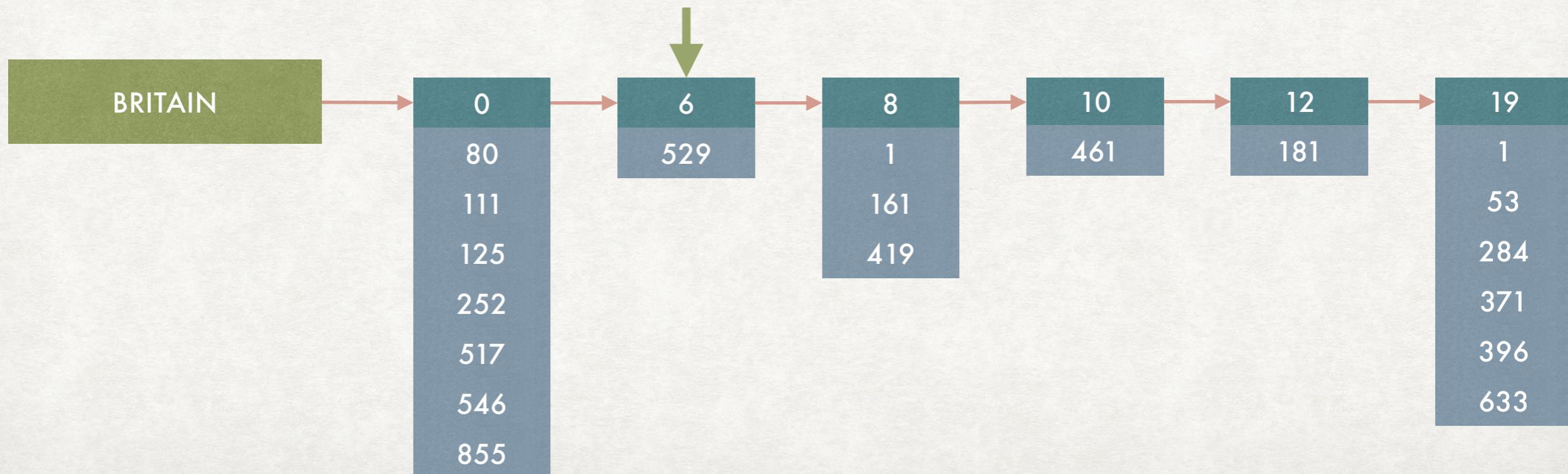
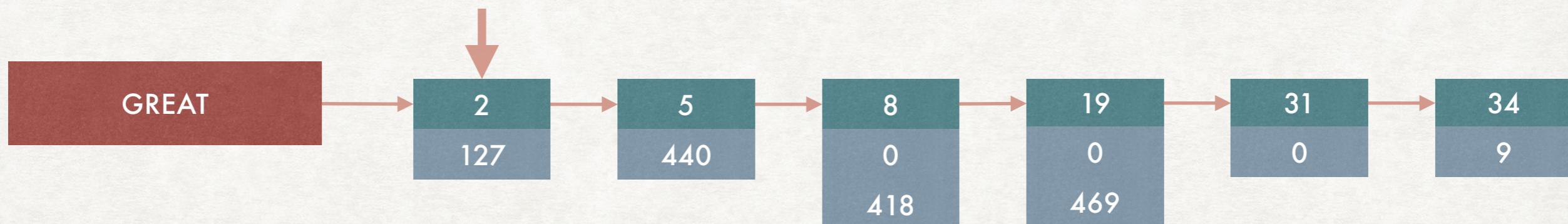
ANSWER



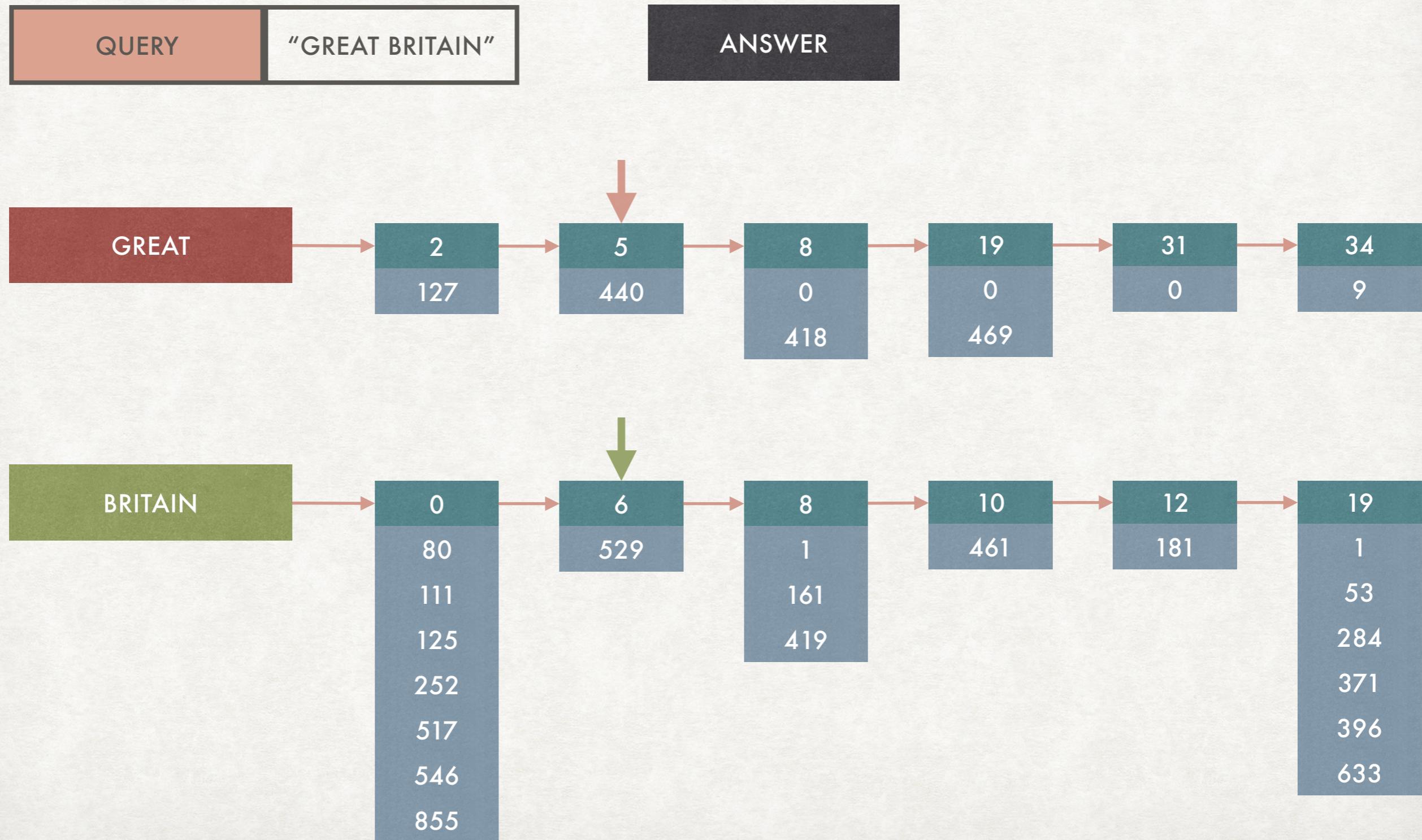
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ANSWER



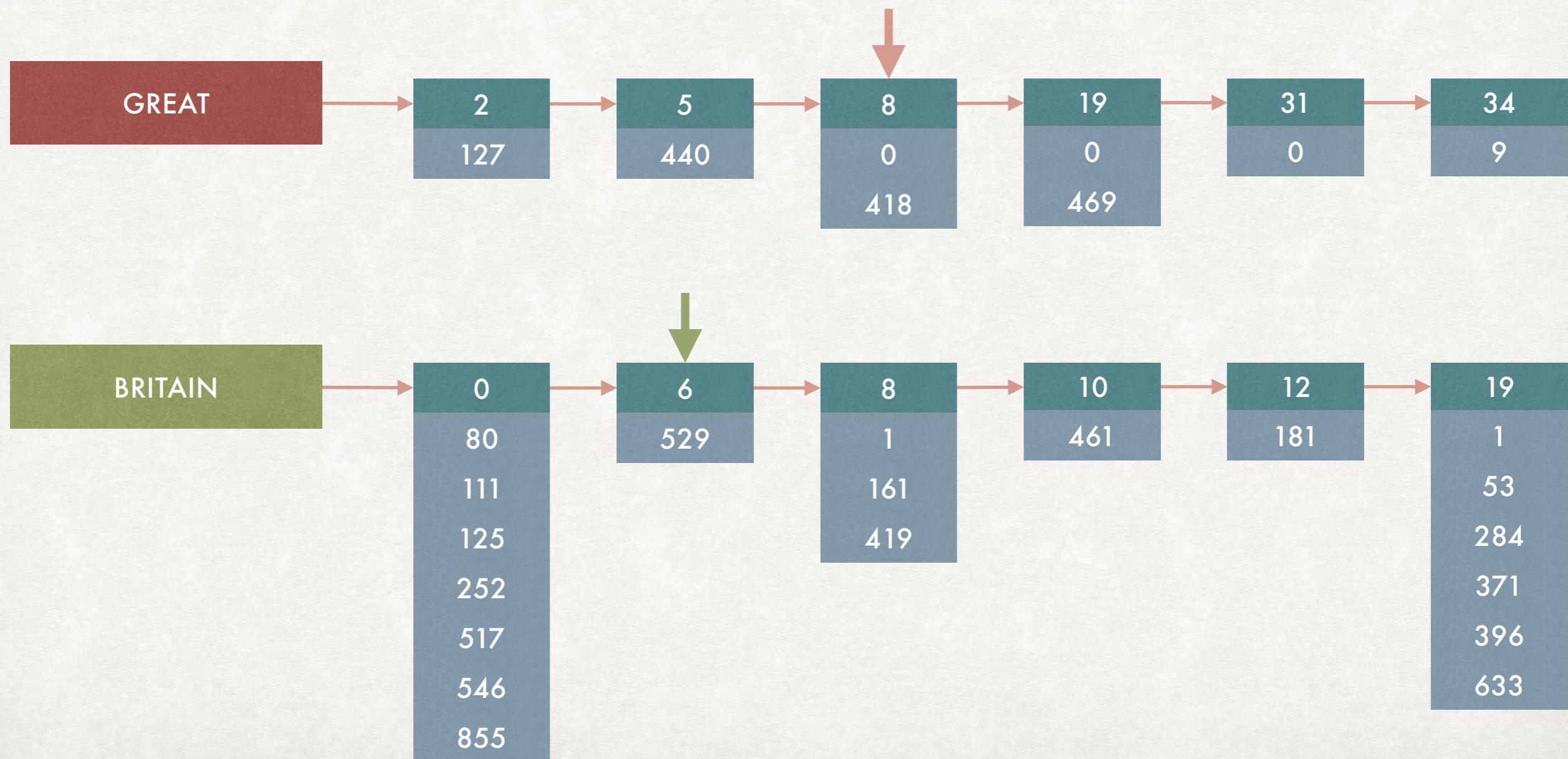
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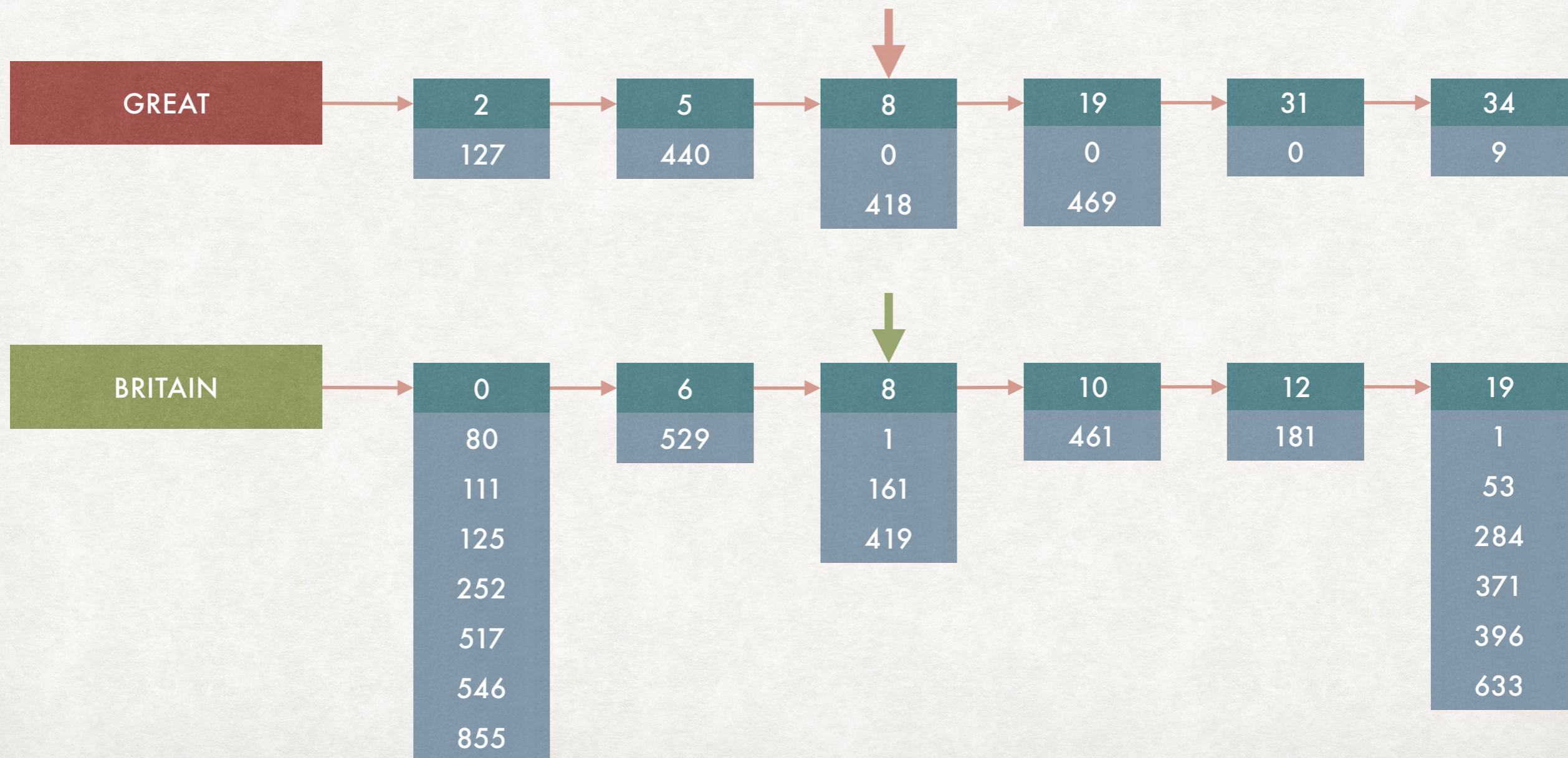
ANSWER



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QUERY      "GREAT BRITAIN"

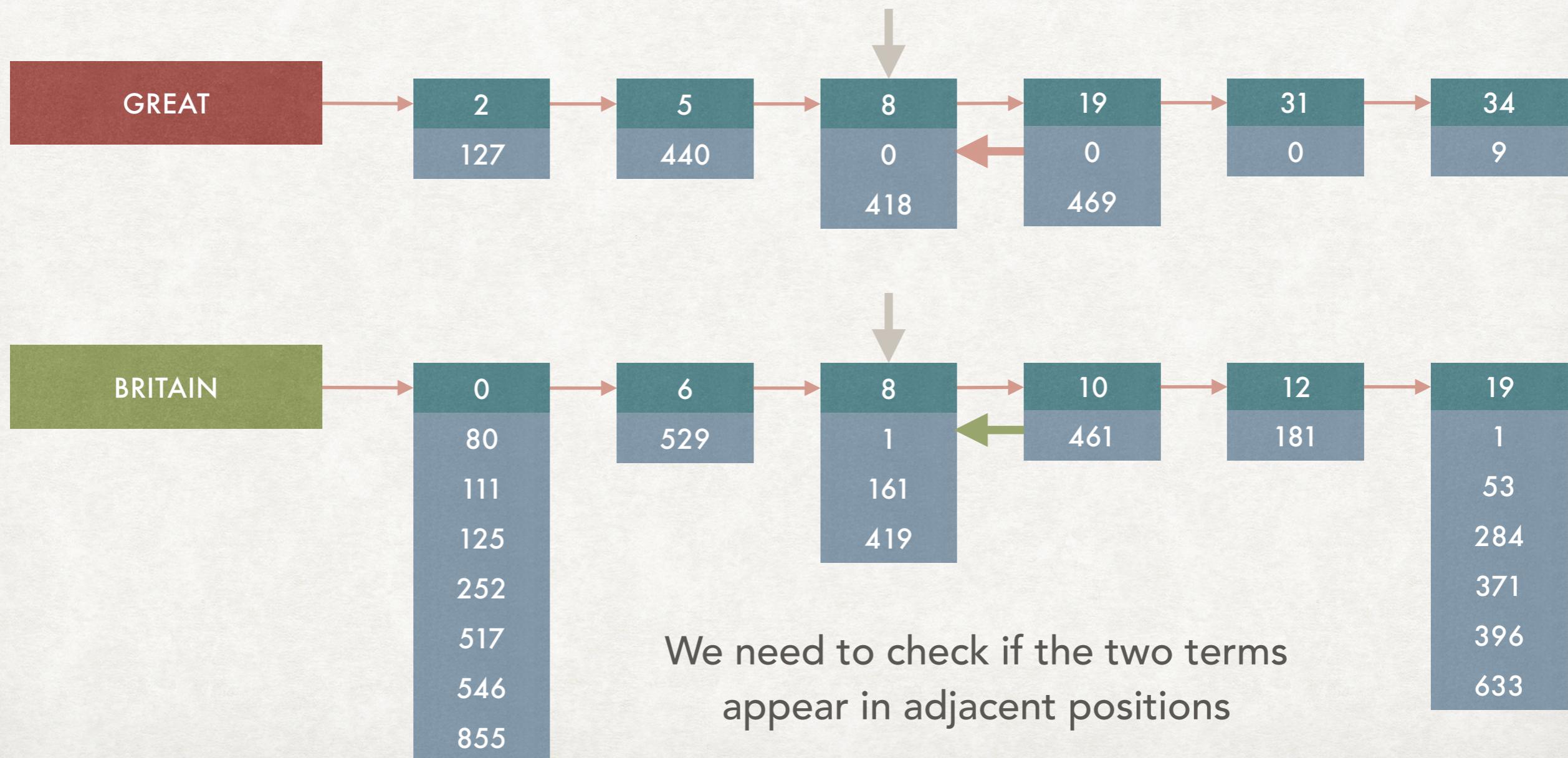
ANSWER



# ANSWERING A PHRASE QUERY WITH POSITIONAL INDEXING

QUERY      "GREAT BRITAIN"

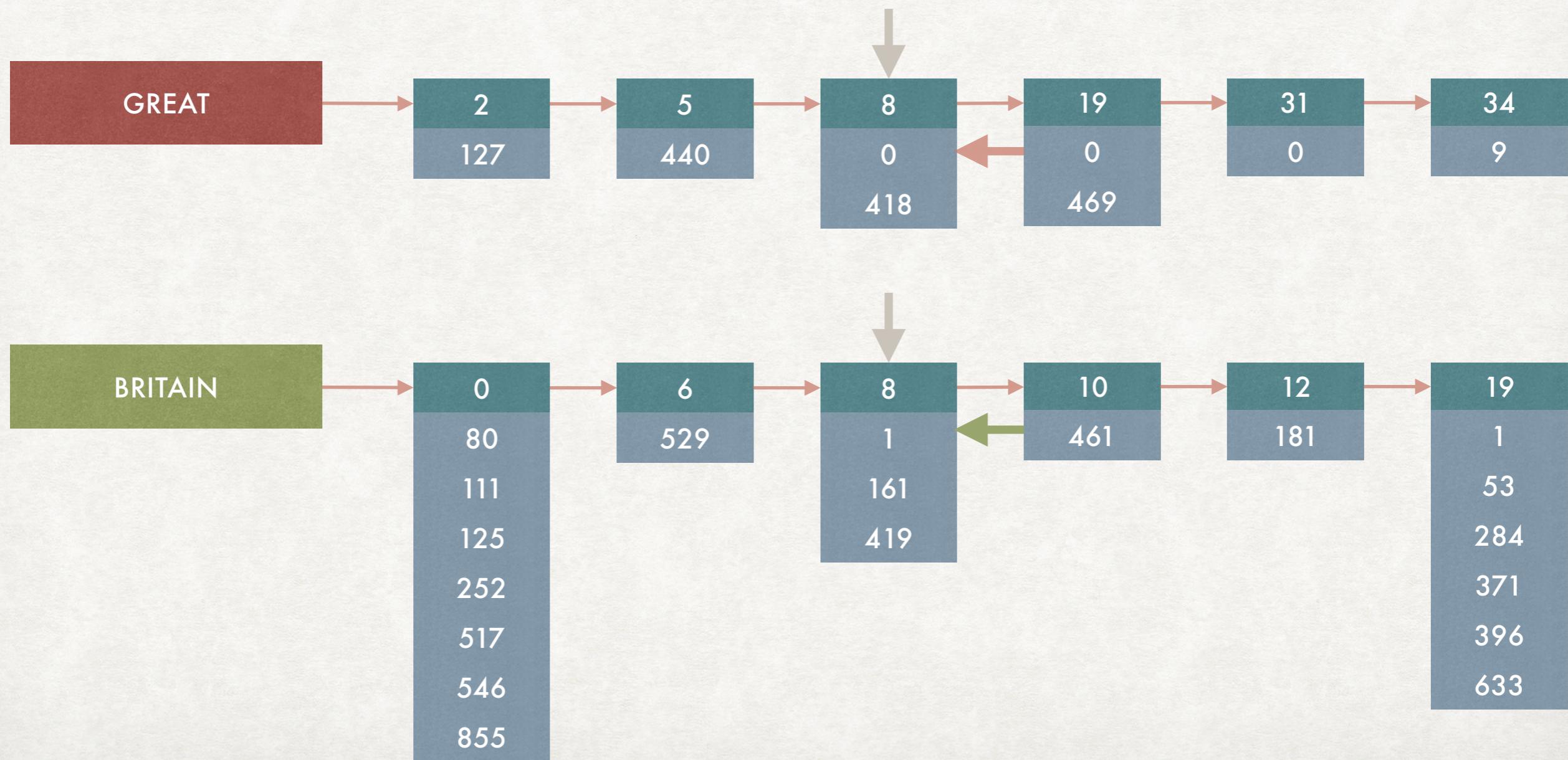
ANSWER



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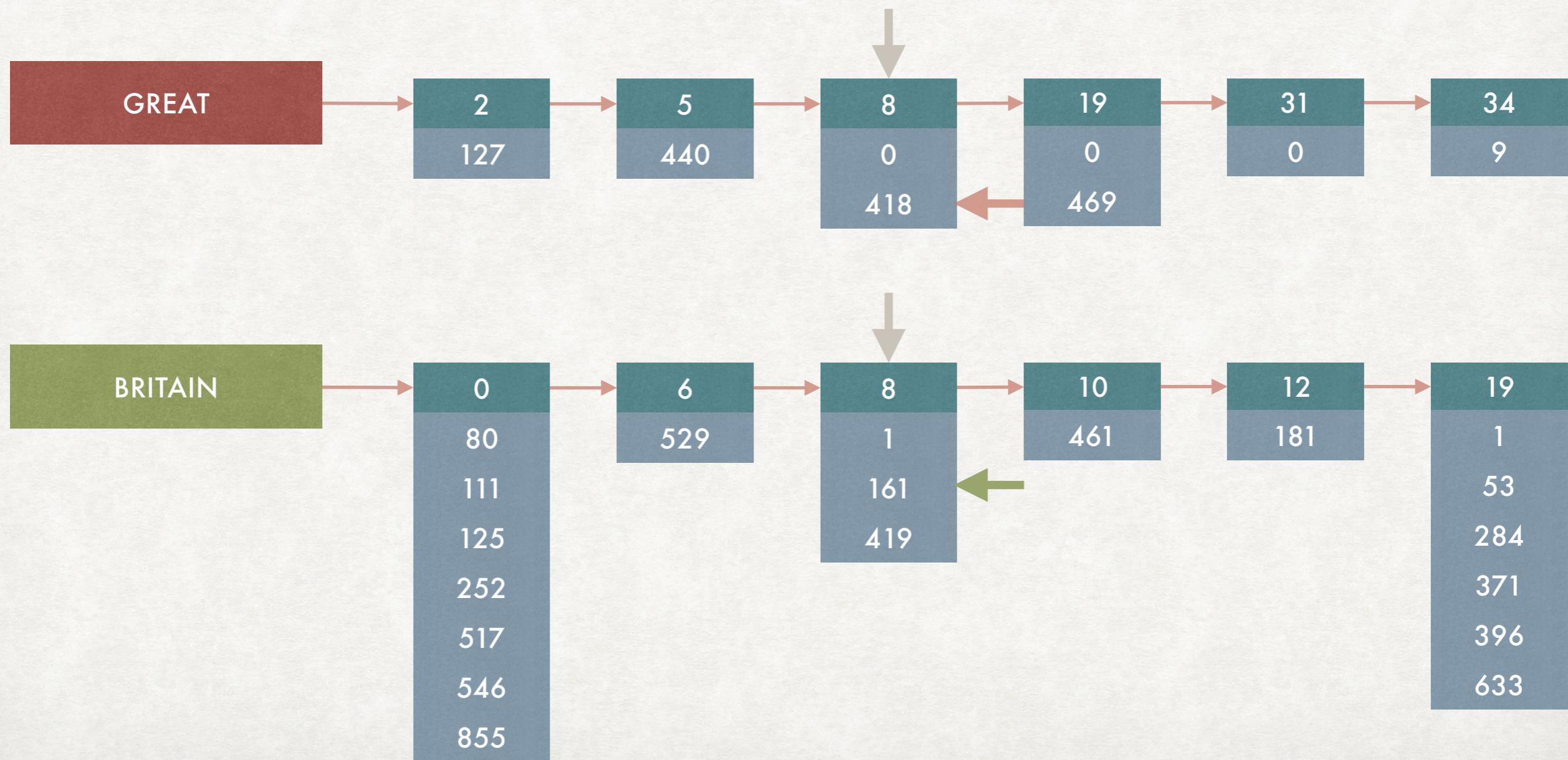
ANSWER → 8  
0



# ANSWERING A PHRASE QUERY WITH POSITIONAL INDEXING

QUERY      "GREAT BRITAIN"

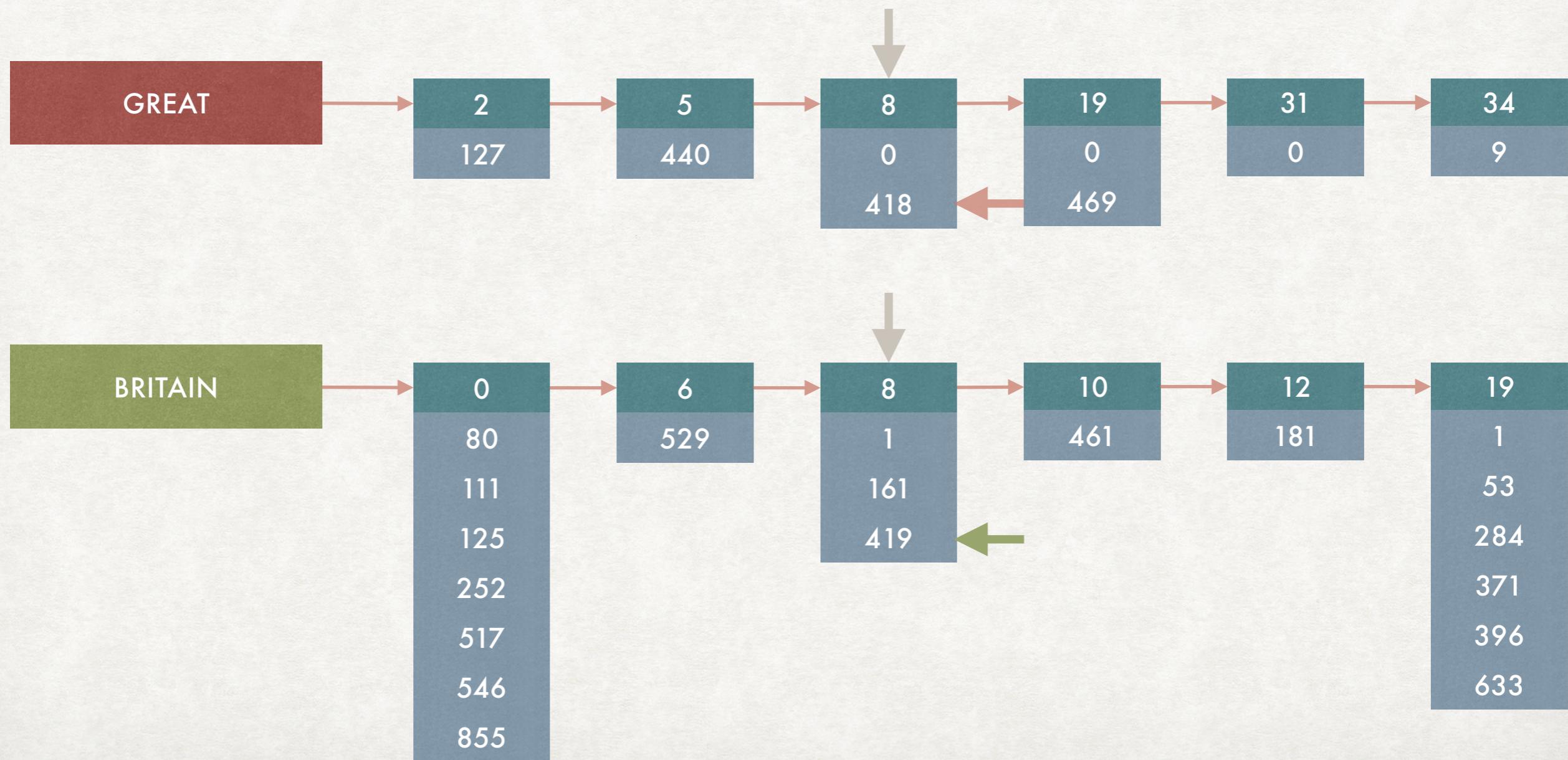
ANSWER → 8  
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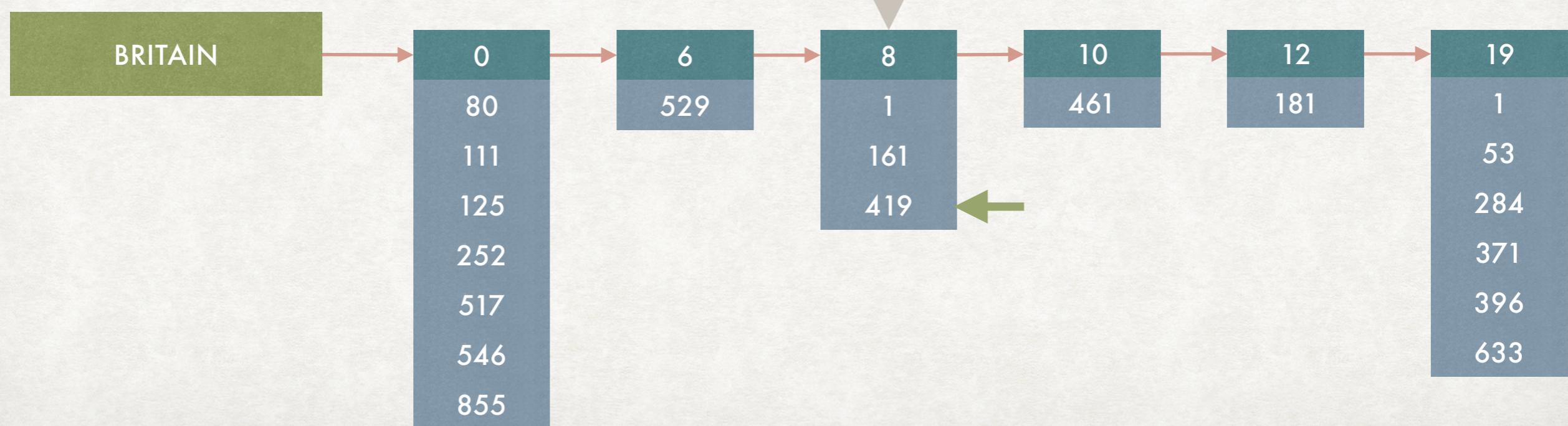
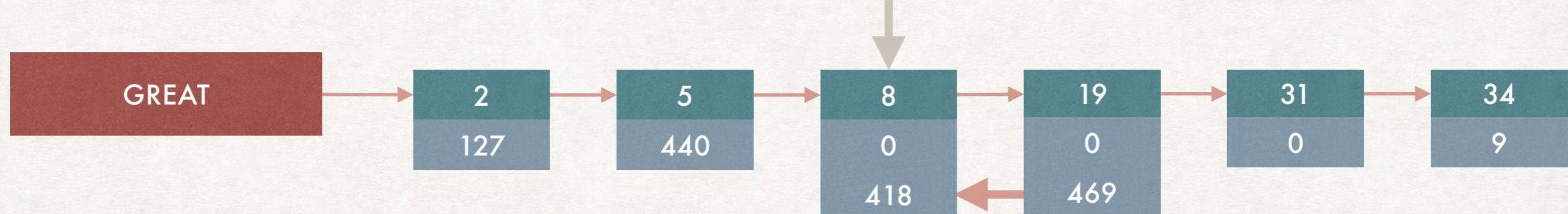
ANSWER → 8  
0



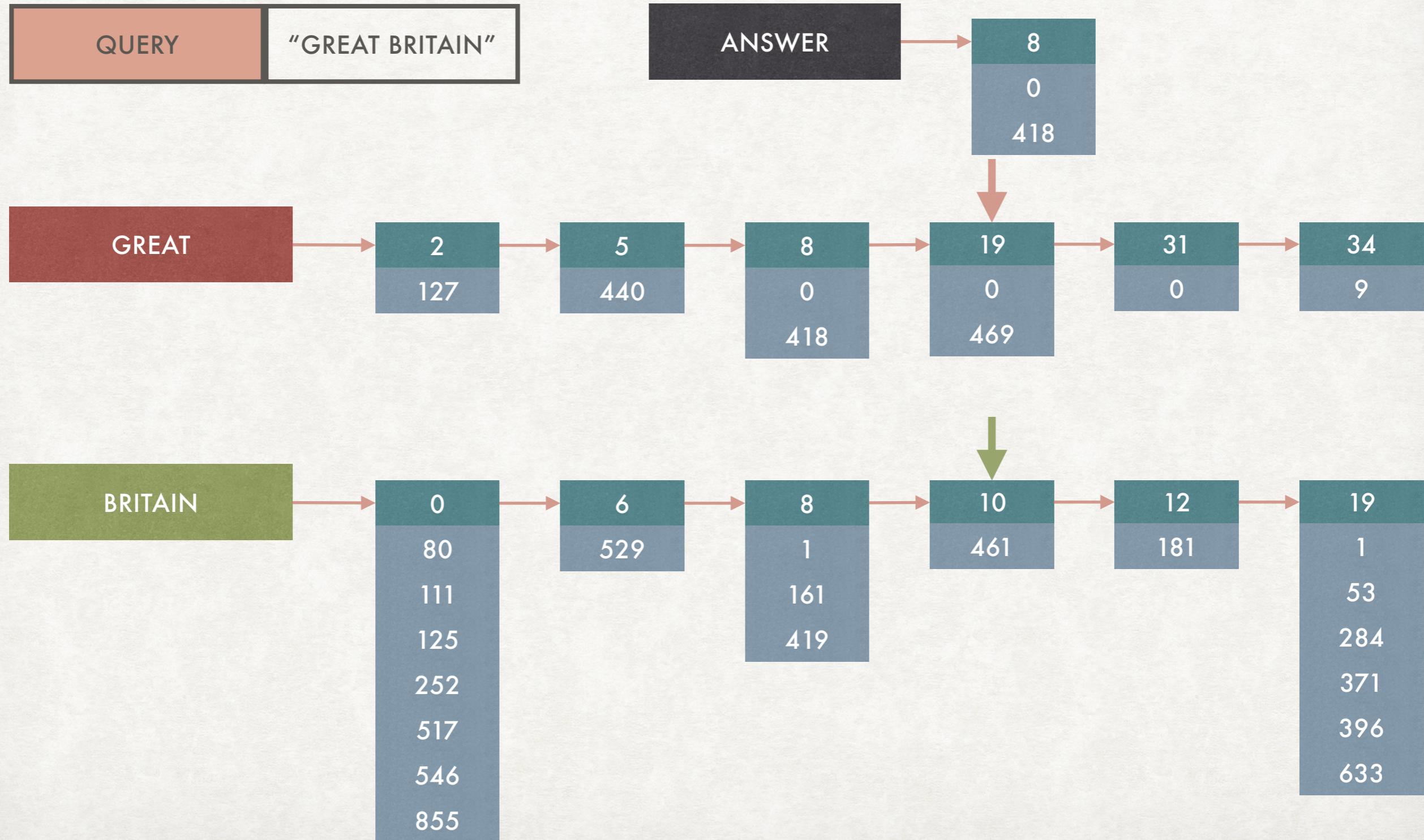
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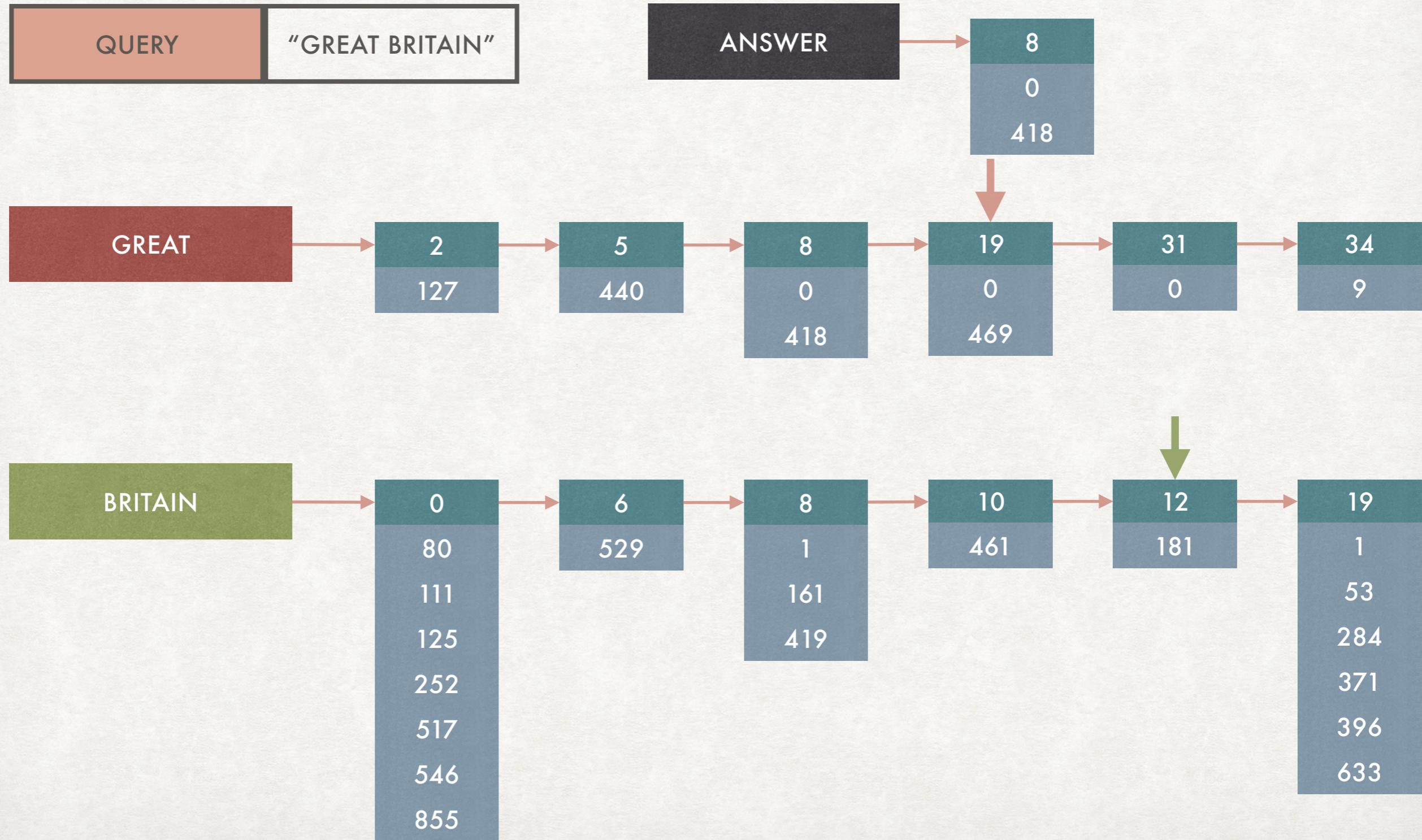
ANSWER → 8  
0  
418



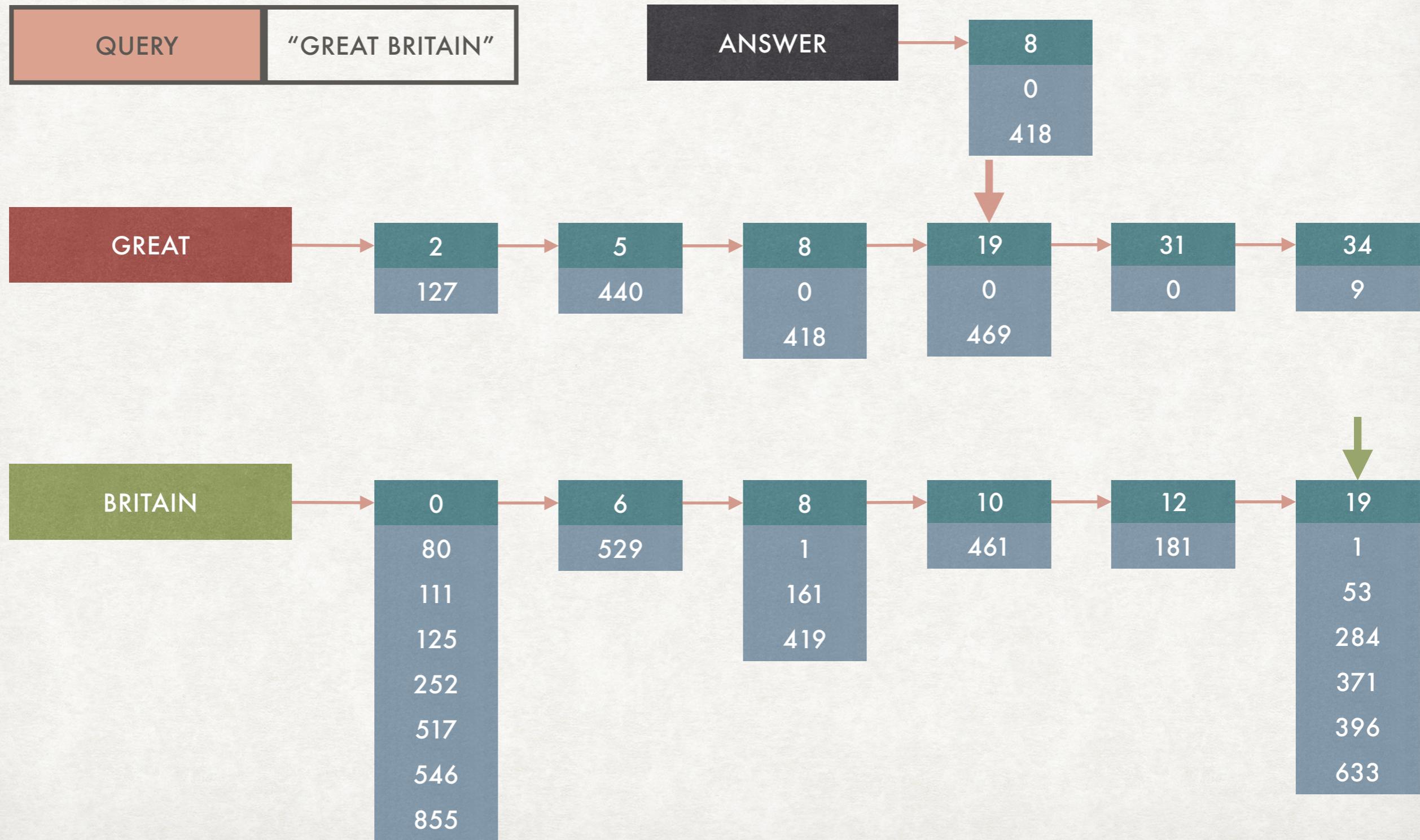
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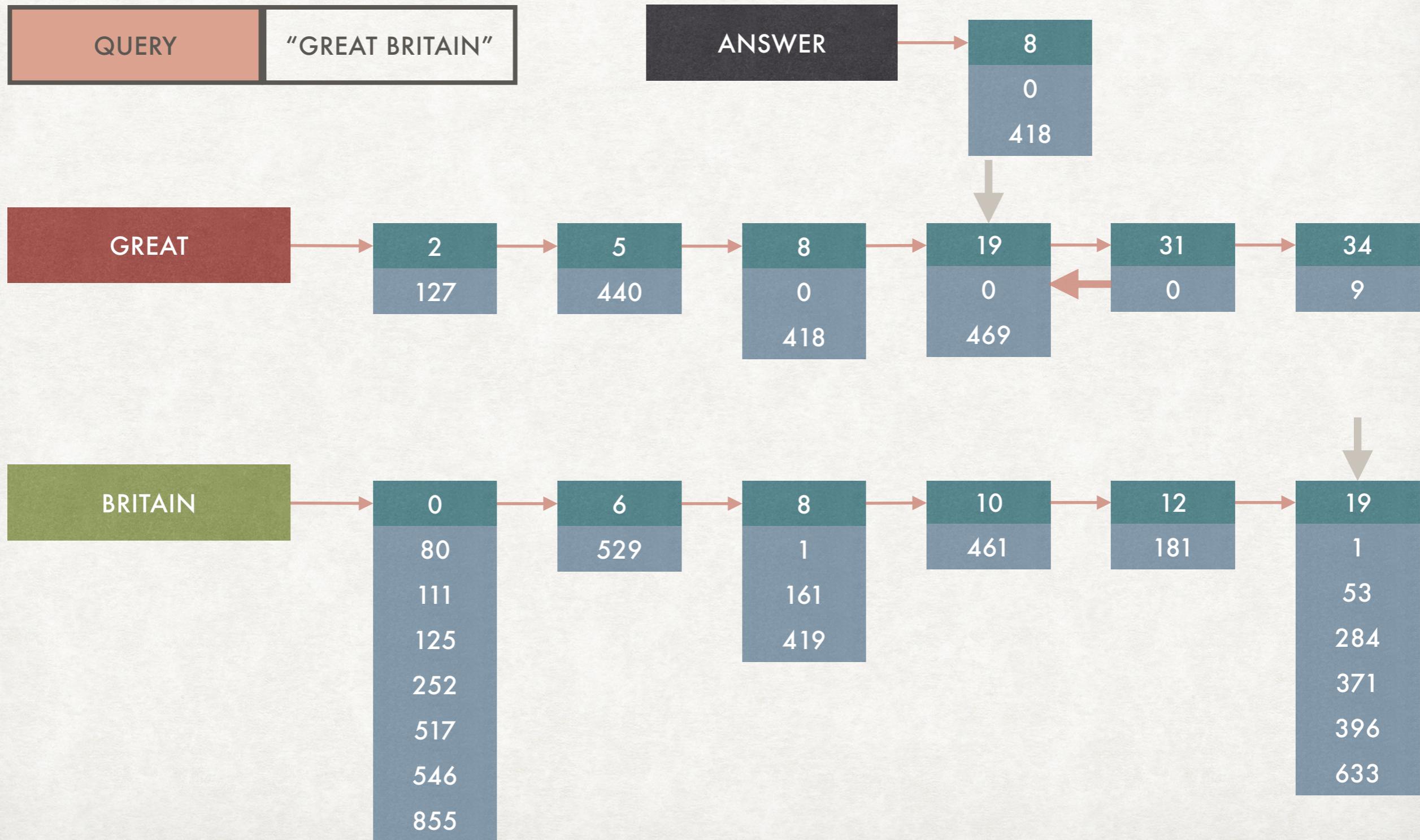
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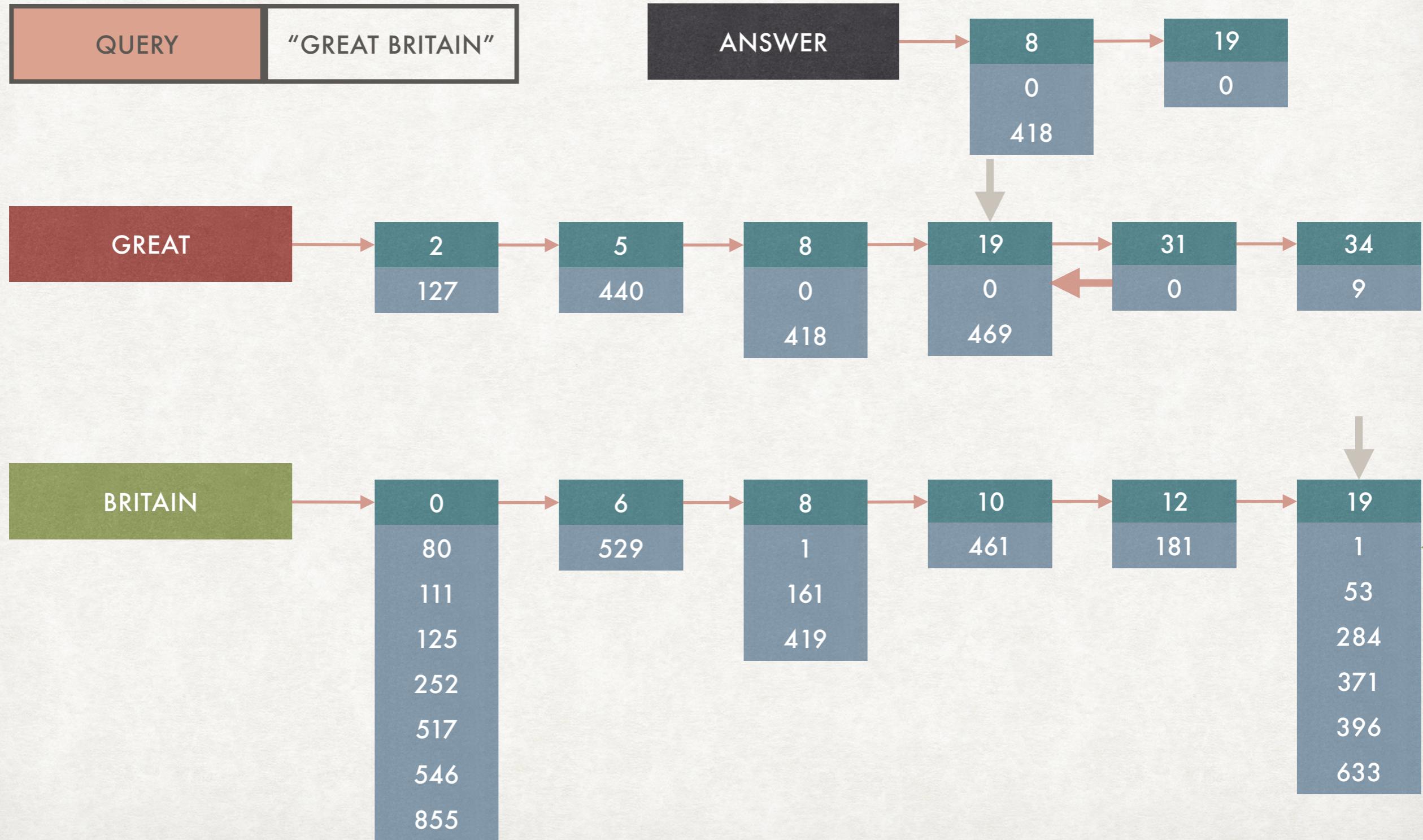
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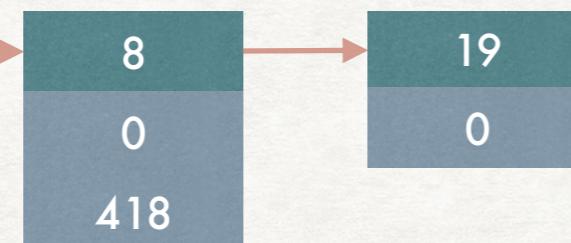
# POSITIONAL INDEXING: SUMMARY

## THE GOOD, THE BAD, AND THE UGLY

- The positional index can be used to support the operators of the form “ $\text{term}_1 /k \text{ term}_2$ ” with  $k$  an integer indicating the maximum number of words that can be between  $\text{term}_1$  and  $\text{term}_2$ .
- 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

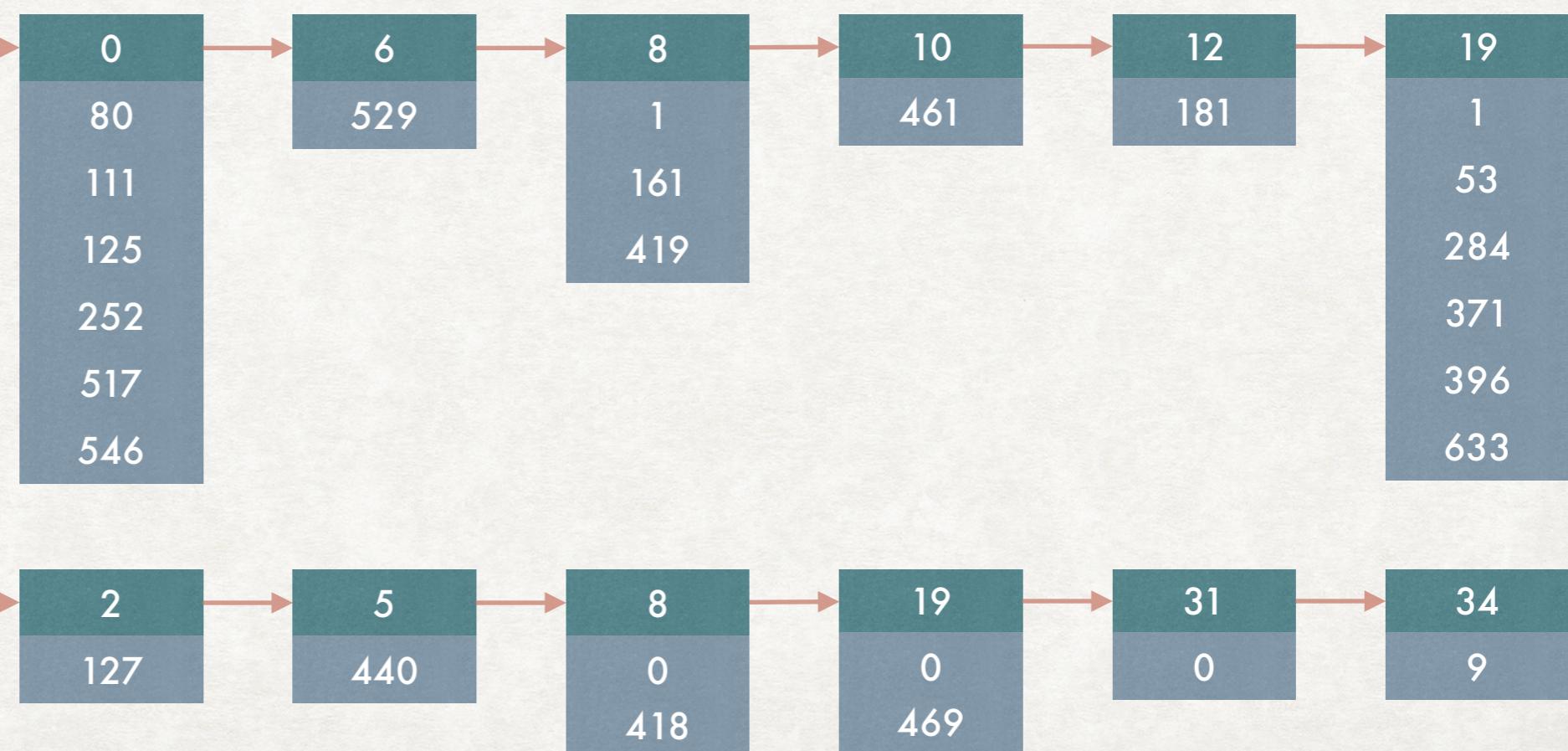
GREAT BRITAIN  
...  
...



Phrase index for frequently asked queries

Positional index for all other queries

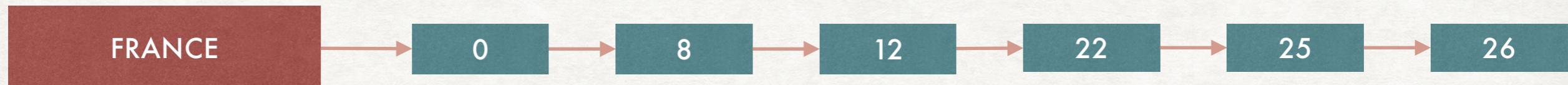
BRITAIN  
...  
...  
...  
GREAT



# IMPROVING THE INVERTED INDEX

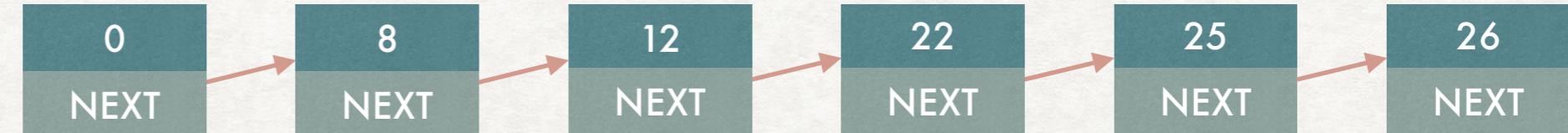
# ARRAYS OR LINKED LISTS?

## WHAT TO USE FOR THE POSTING LISTS?



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

Singly linked lists



cheap insertion and updates  
pointer overhead, poor memory locality (pointers chasing)

Variable length arrays

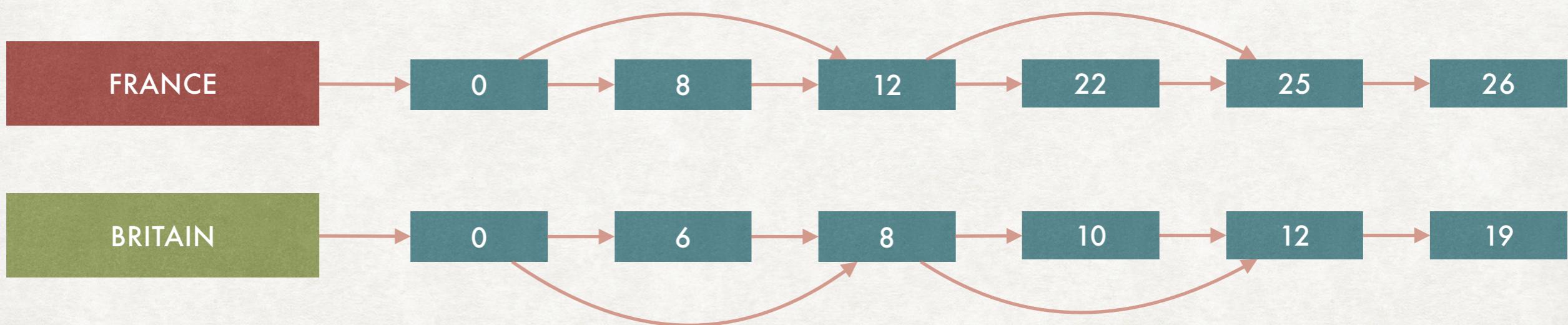


no pointers overhead, contiguous memory  
difficult to update

# SKIP LISTS

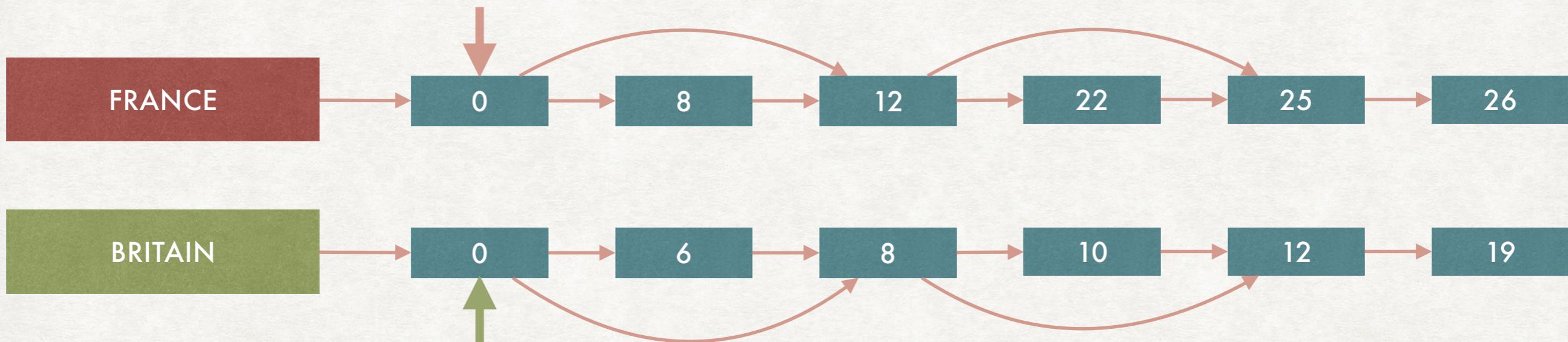
## FASTER INTERSECTION

- 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



# AN EXAMPLE QUERY

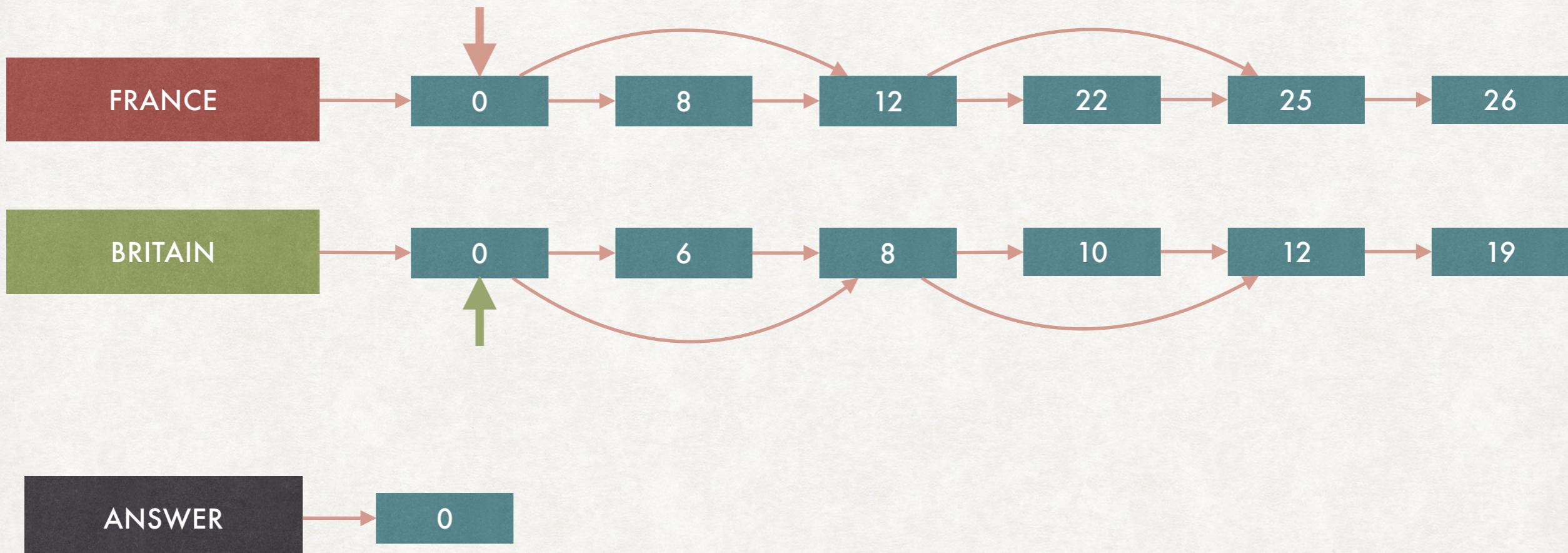
## SAVING TIME WITH SKIP LISTS



ANSWER

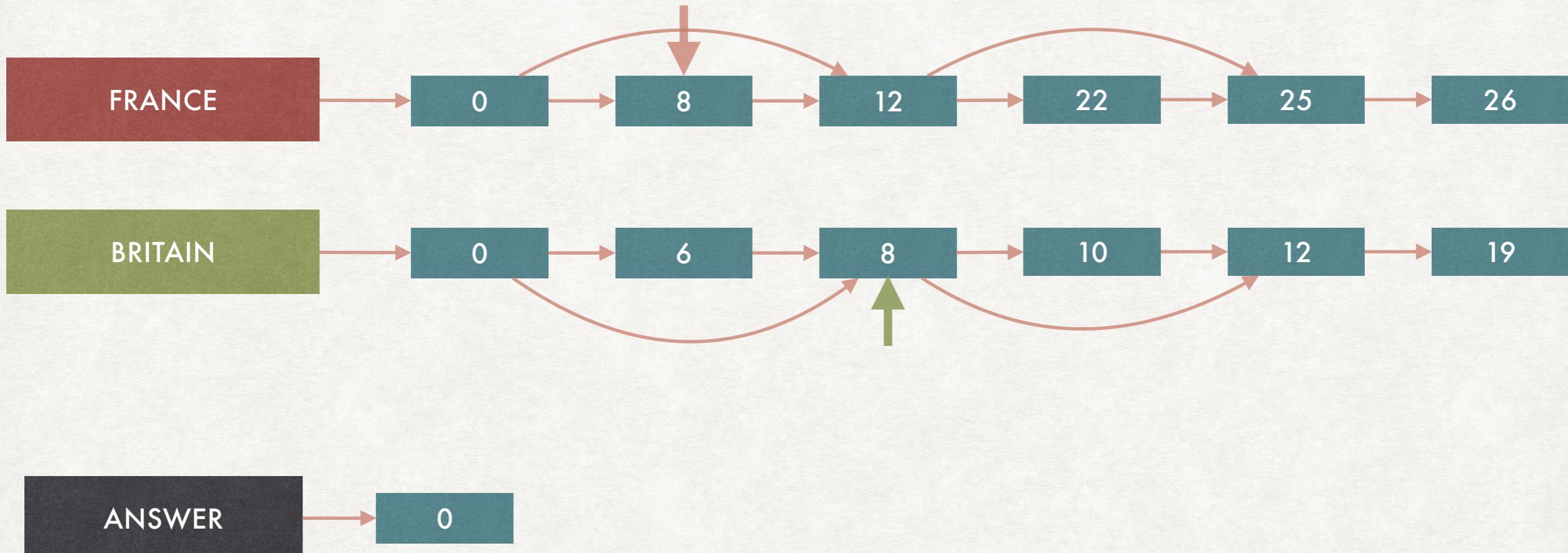
# AN EXAMPLE QUERY

## SAVING TIME WITH SKIP LISTS



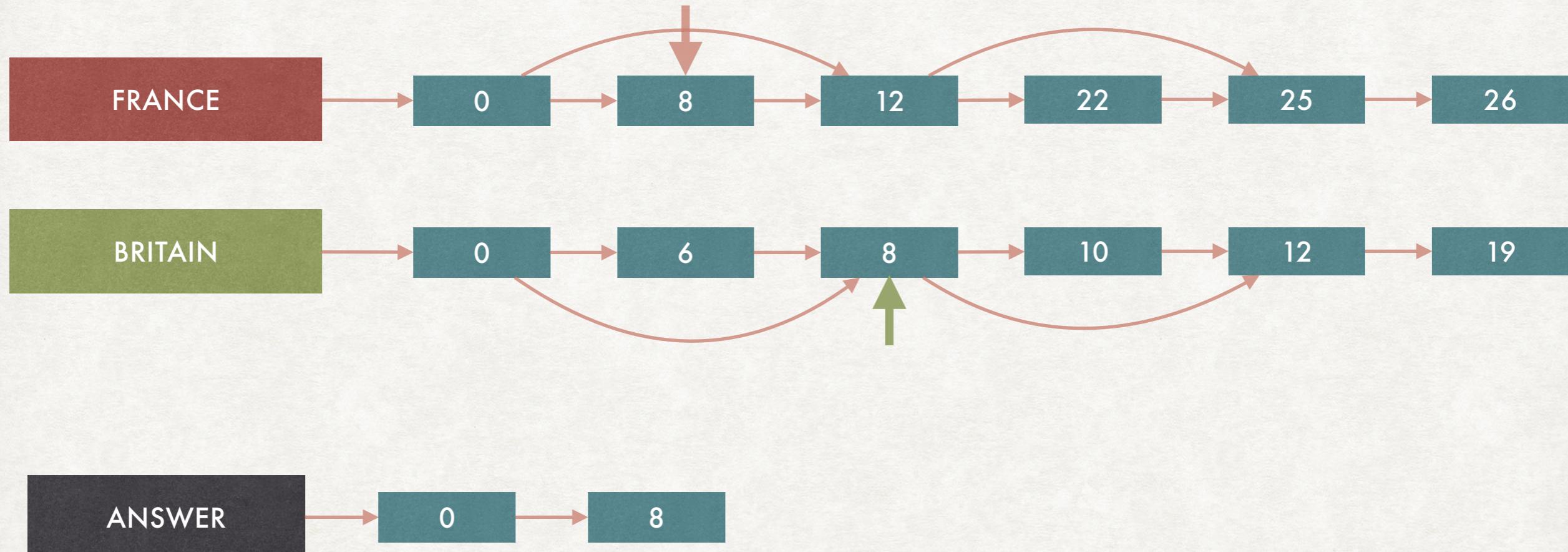
# AN EXAMPLE QUERY

## SAVING TIME WITH SKIP LISTS



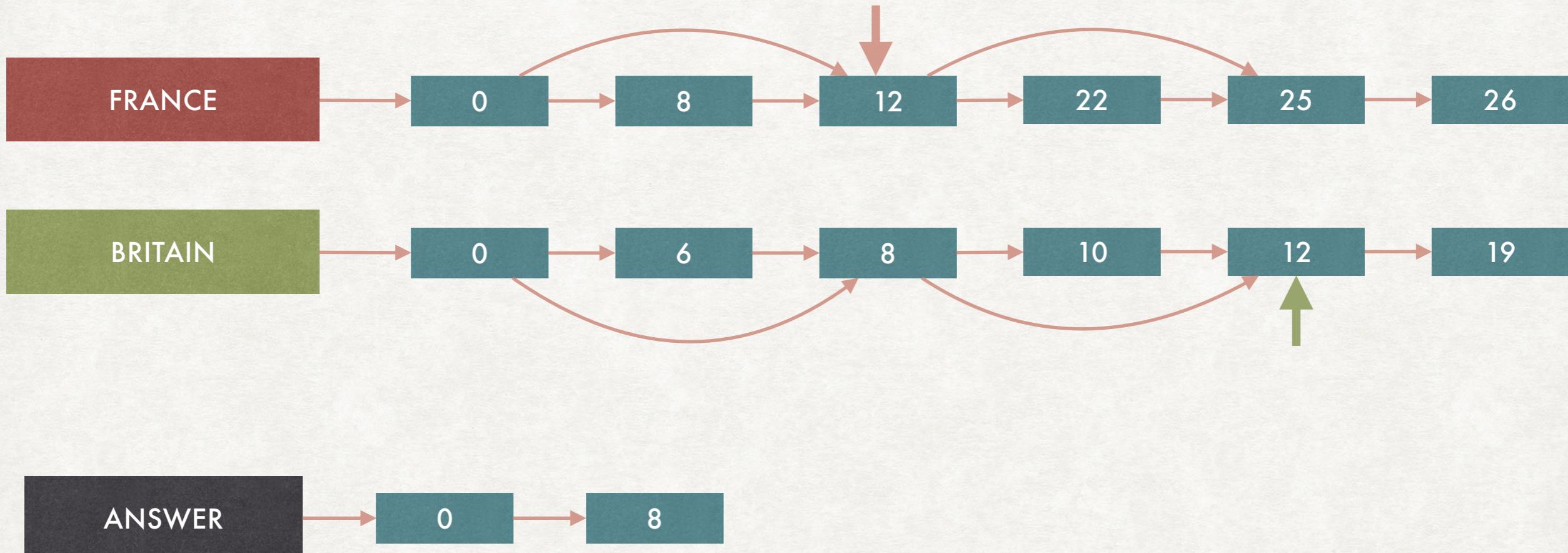
# AN EXAMPLE QUERY

## SAVING TIME WITH SKIP LISTS



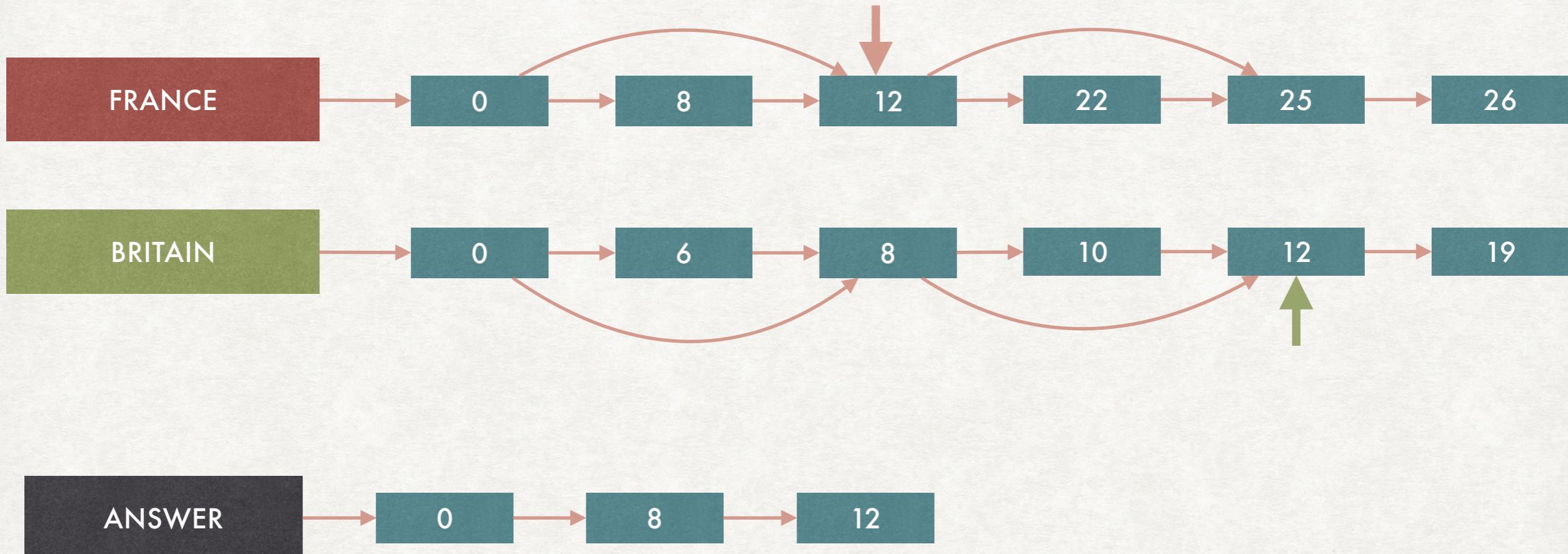
# AN EXAMPLE QUERY

## SAVING TIME WITH SKIP LISTS



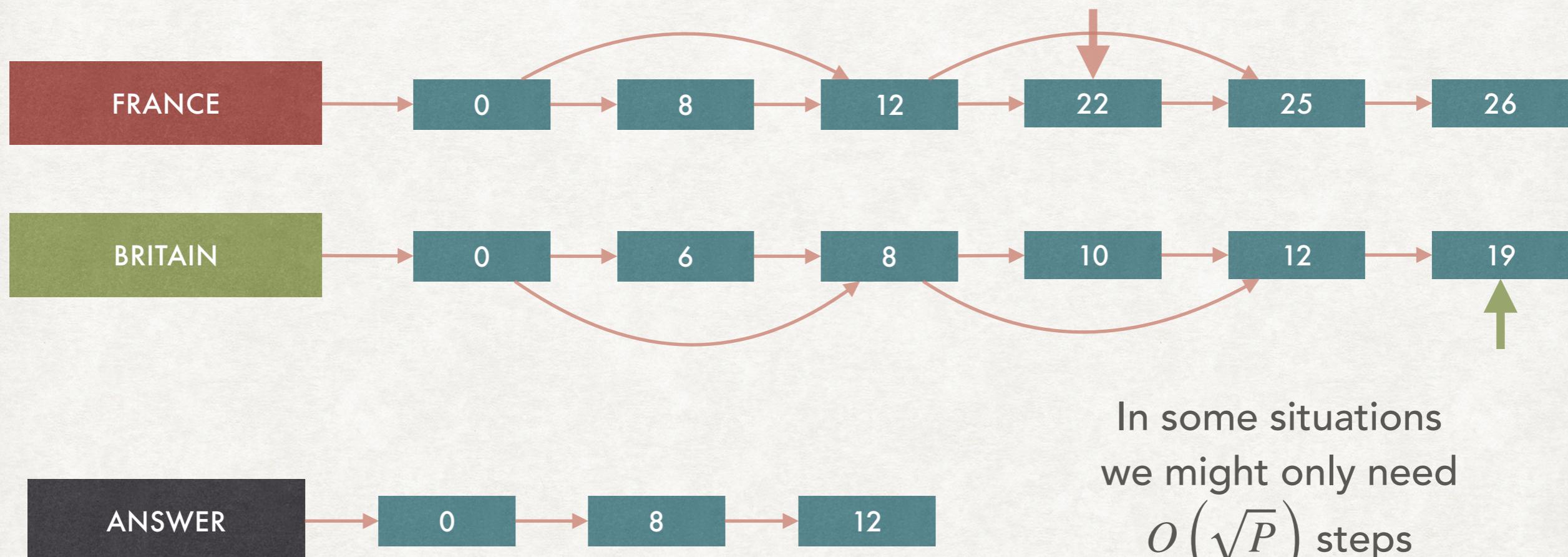
# AN EXAMPLE QUERY

## SAVING TIME WITH SKIP LISTS



# AN EXAMPLE QUERY

## SAVING TIME WITH SKIP LISTS



In some situations  
we might only need  
 $O(\sqrt{P})$  steps  
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: <http://www.cs.cmu.edu/~ark/personas/>, more than 42k movie descriptions