



## **Lesson2:**

# **Modelling the Web with Simple Statistical Descriptive Text Models**

## **Unit4:**

# **Test if lesser words are required on Simple English Wikipedia to understand a larger fraction than on English Wikipedia**

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Introduction to Web Science Part 2

Emerging Web Properties





## Completing this unit you should

- Understand what a log-log plot is
- Improve your skills in reading and interpreting diagrams
- Know about the word rank / frequency plot
- Should be able to transfer a histogram or curve into a cumulative distribution function



## Strategy to fulfil our test

- count the frequency of words in both corpora
- Sort the words descending to their frequency
  - This creates a ranking
- Create a plot displaying the frequency depending on the rank
- Transform this to the cumulative plot in order to test our prediction

# Counting words is really simple in python

```
In [39]: def readWordsFromWiki(filename):
        """
        opens a file which has one sentence per line (without punctuation marks)
        returns a list with all words
        """
        f = open(filename)
        allWords=[]
        for line in f:
            line = line[:-1]
            words = line.split()
            allWords.extend(words)
        return allWords

allSimpleWords = readWordsFromWiki("../datasets/simpleWikiAbstractsOneSentencePerLine")
allEnWords = readWordsFromWiki("../datasets/enWikiAbstractsOneSentencePerLine")
```

```
In [40]: from collections import Counter
c=Counter(allSimpleWords)
words,frequencies = zip(*c.most_common())
print words[0:10], frequencies[0:10]

('the', 'is', 'a', 'of', 'in', 'and', 'it', 'was', 'to', 'an') (134415, 89447, 81349, 80376, 80309, 39475, 27820, 25554, 18726, 15620)
```

```
In [23]: cEn=Counter(allEnWords)
enWords,enFrequencies = zip(*cEn.most_common())
print enWords[0:10], enFrequencies[0:10]

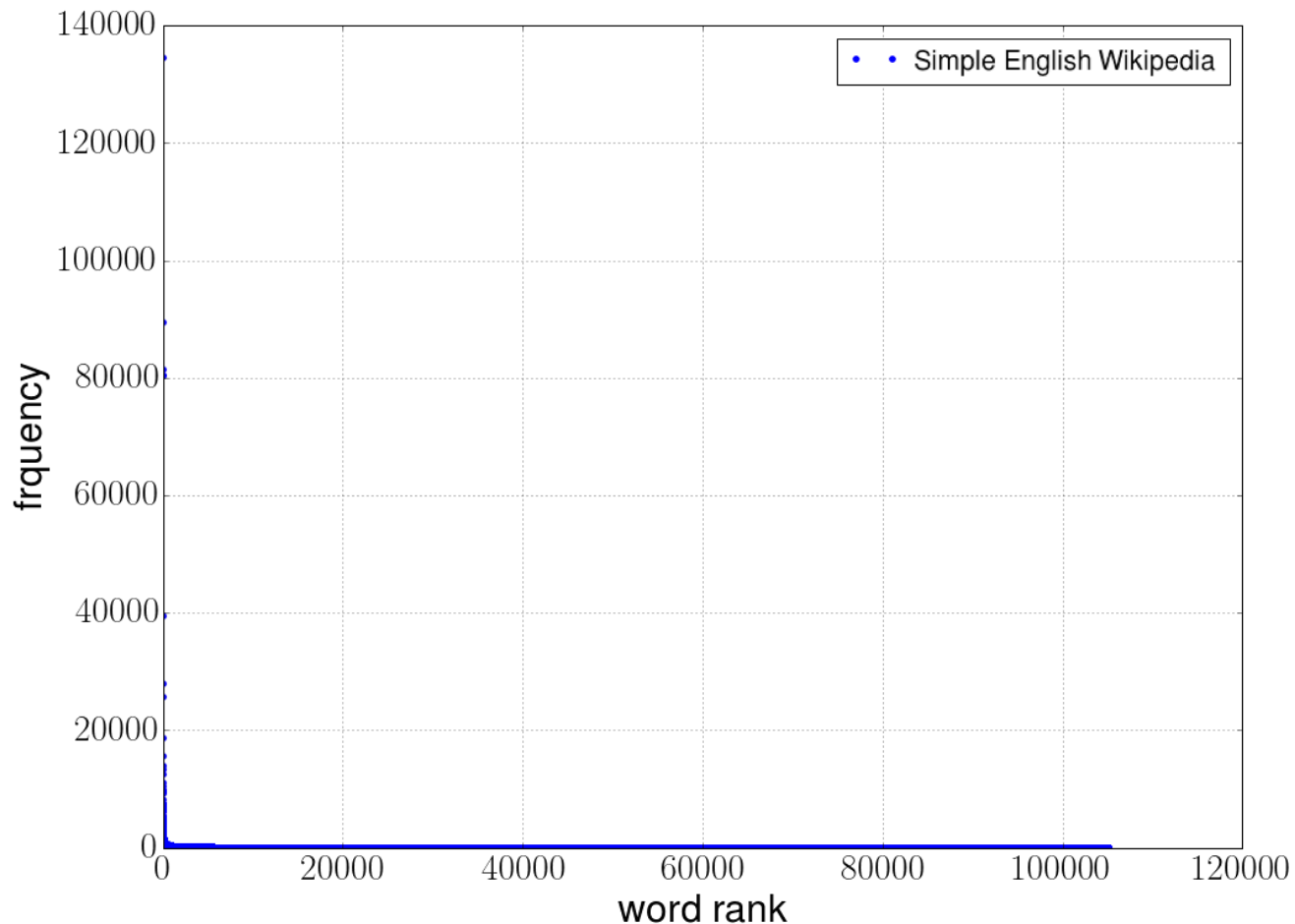
('the', 'of', 'in', 'a', 'is', 'and', 'was', 'to', 'by', 'it') (5307042, 3247413, 2810037, 2594795, 2331626, 1983945, 1128009, 1085090, 748863, 591726)
```



# Lets look at the rank frequency diagram

- As you can see, you see nothing (:

Wordrank frequency diagram on Wikipedia data sets

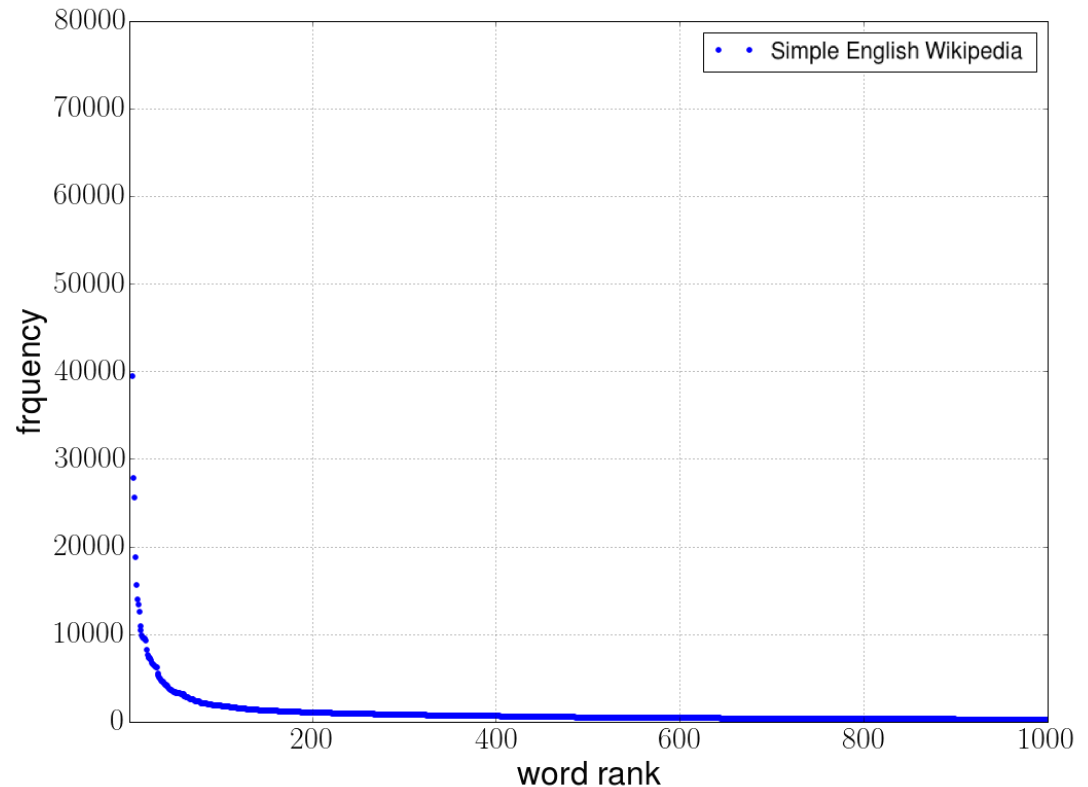


# Maybe zooming the axis helps a little bit?

Problems with this plot:

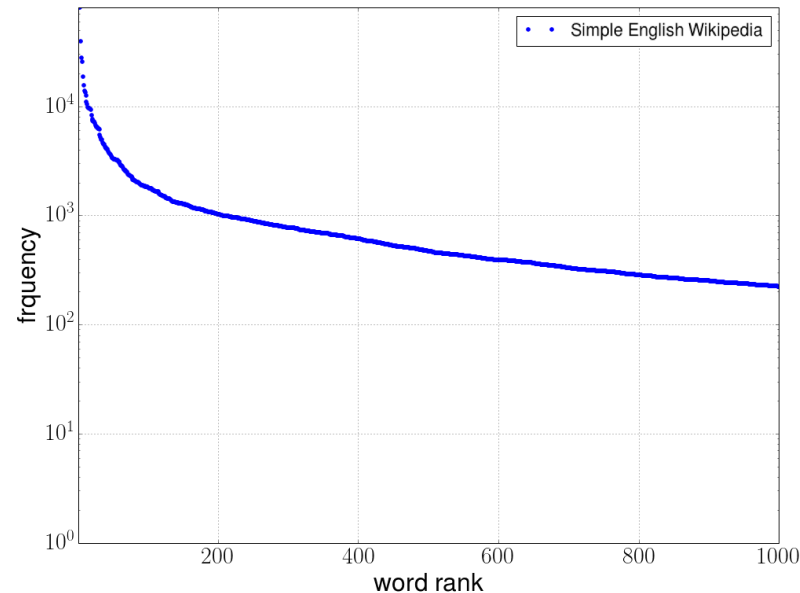
- The frequency of words with a rank bigger than 200 can not be distinguished
- What if all ranks should be displayed?

Wordrank frequency diagram on Wikipedia data sets (Zoomed)

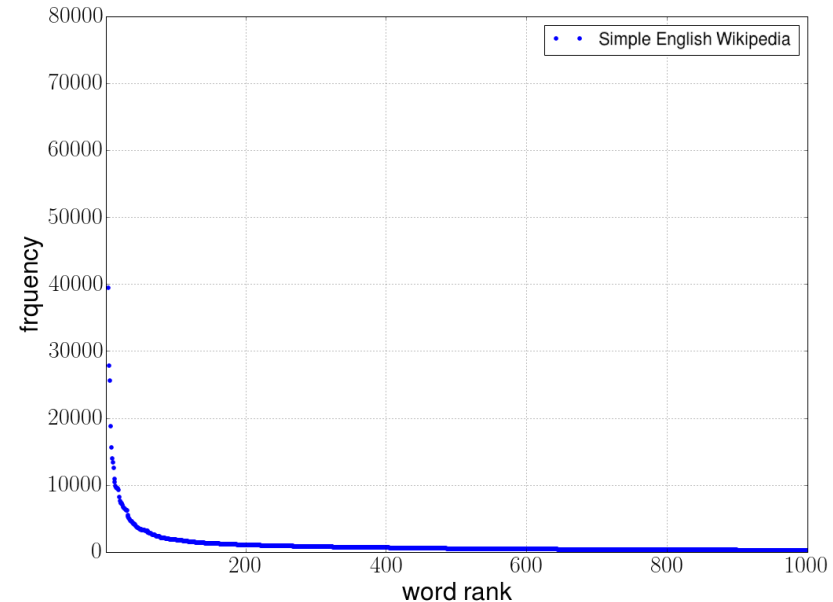


# Changing the y-axis to a logarithmic scale

Wordrank frequency diagram on Wikipedia data sets (Zoomed)



Wordrank frequency diagram on Wikipedia data sets (Zoomed)

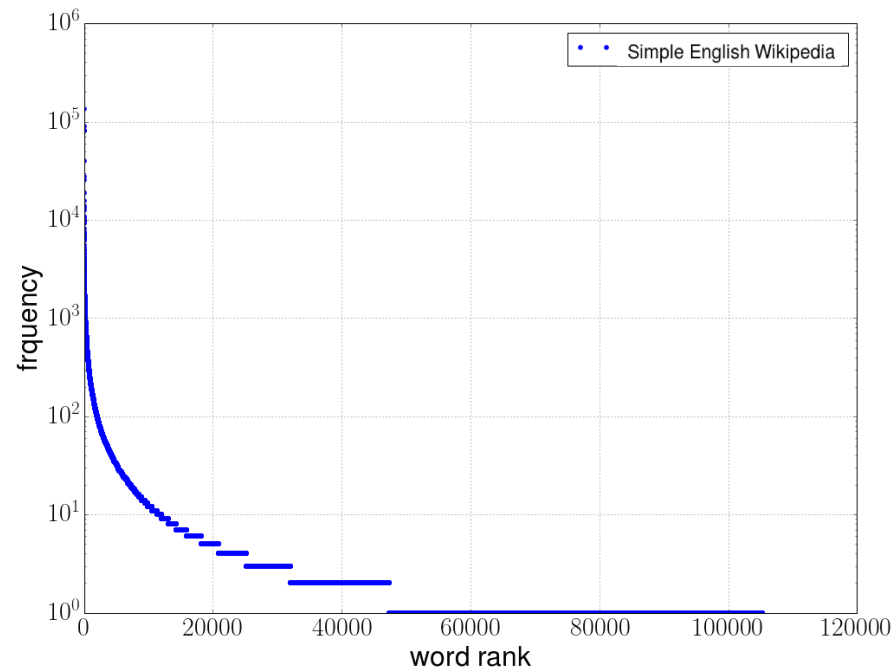


- Same data being used
- Very different visualization
- What happens if we include all ranks again?



# Displaying the full x-axis

Wordrank frequency diagram on Wikipedia data sets (Zoomed)

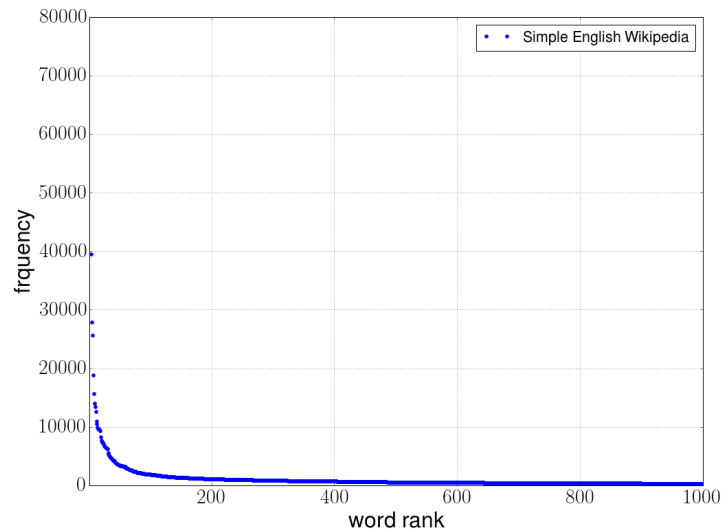


- Similar problems as before:
  - Top ranks can almost not be distinguished
- Do the same trick as before

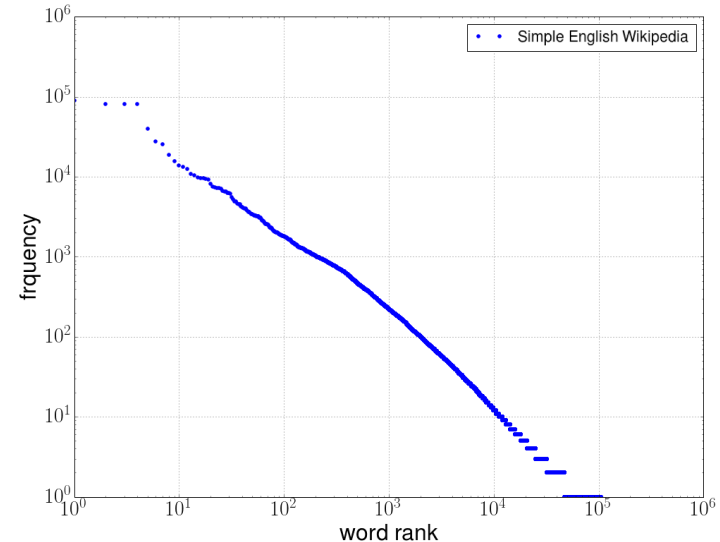


# Compare linear scale plot with log-log plot

Wordrank frequency diagram on Wikipedia data sets (Zoomed)



Wordrank frequency diagram on Wikipedia data sets (log-log scale)



## Linear

Every interval displays a **fixed range** of numbers

**Adding** a constant number (10 k) to go from one scale unit to the next one

Can visualize best what is happening **in a certain interval** - Usually the highest order of magnitude

## Logarithmic

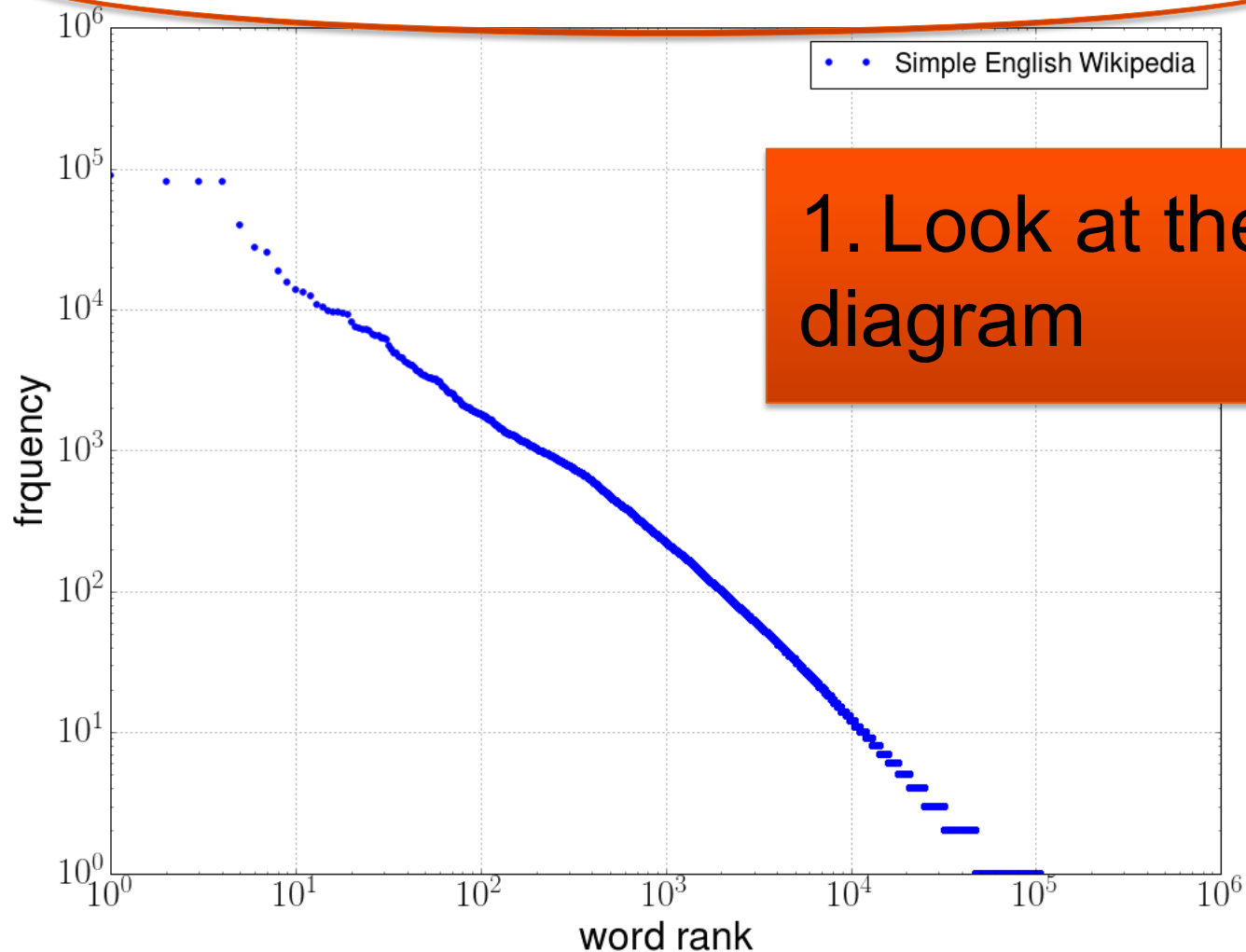
Every interval displays one **order of magnitude**

**Multiplying** with a constant number (in our case 10) to go from one scale to the next

Can visualize best what happens in **each** order of magnitude

# 6 Steps of mastering reading (log-log) plots

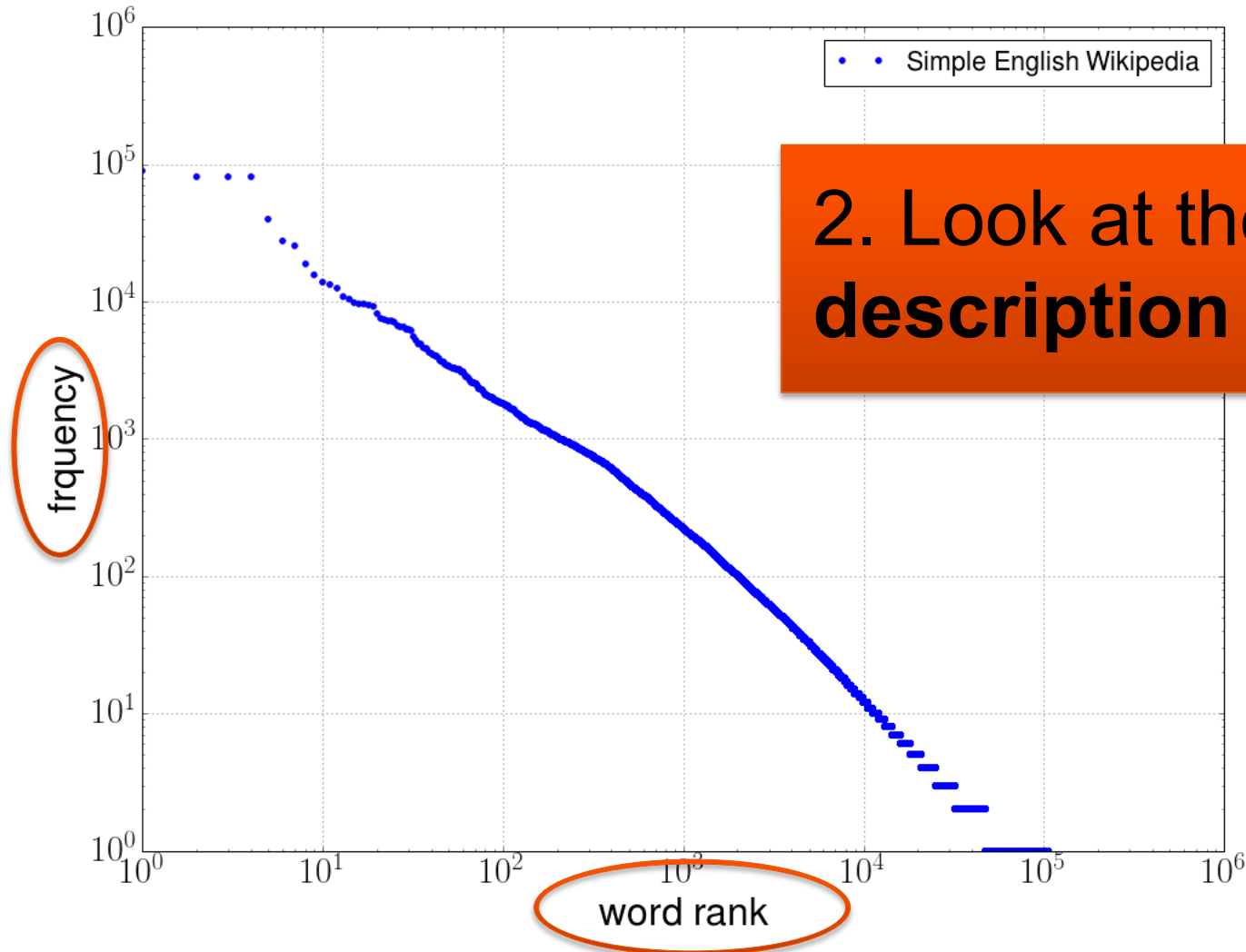
Wordrank frequency diagram on Wikipedia data sets



1. Look at the **title** of the diagram

# 6 Steps of mastering reading (log-log) plots

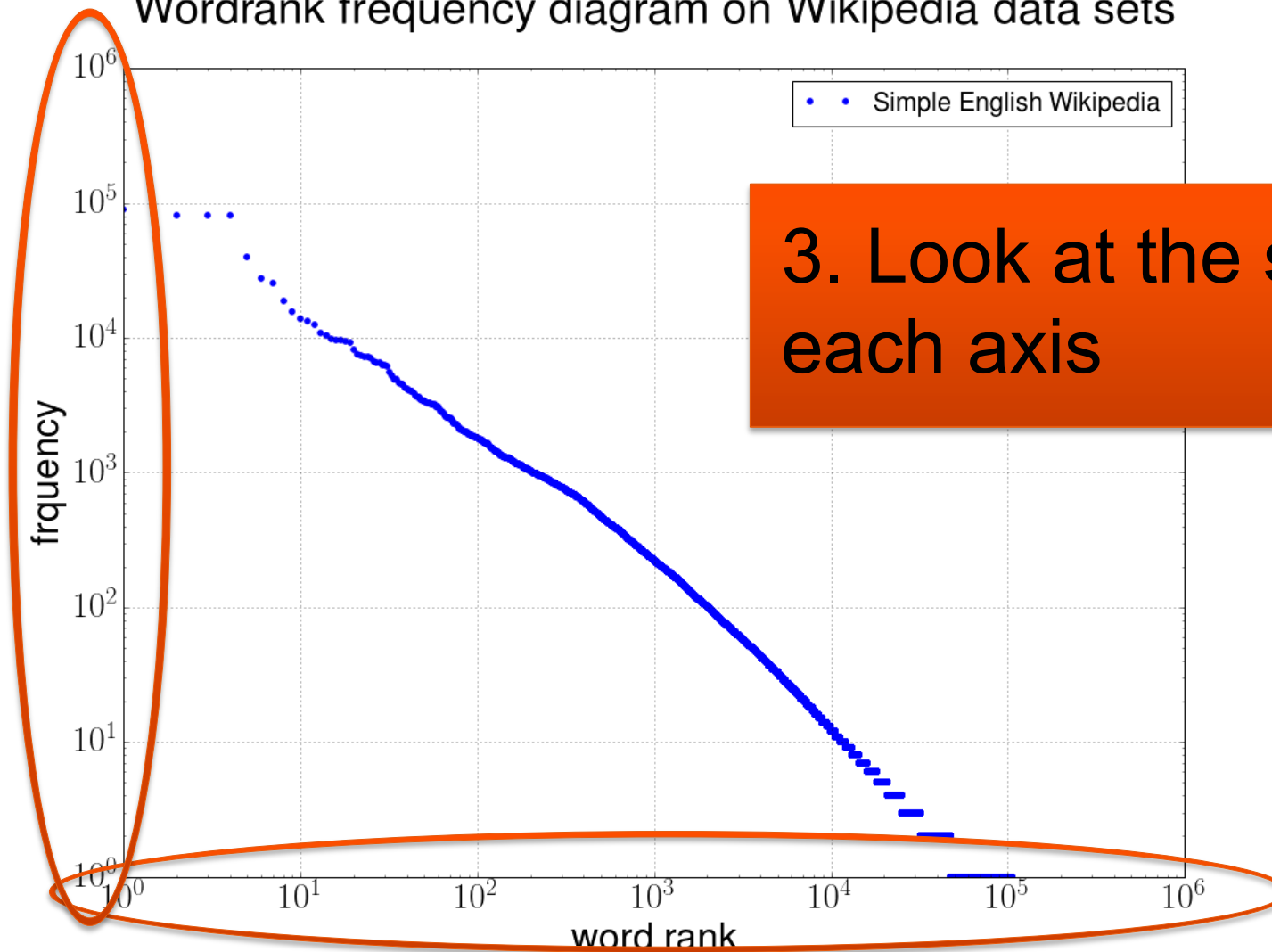
Wordrank frequency diagram on Wikipedia data sets



2. Look at the **description** of both axis

# 6 Steps of mastering reading (log-log) plots

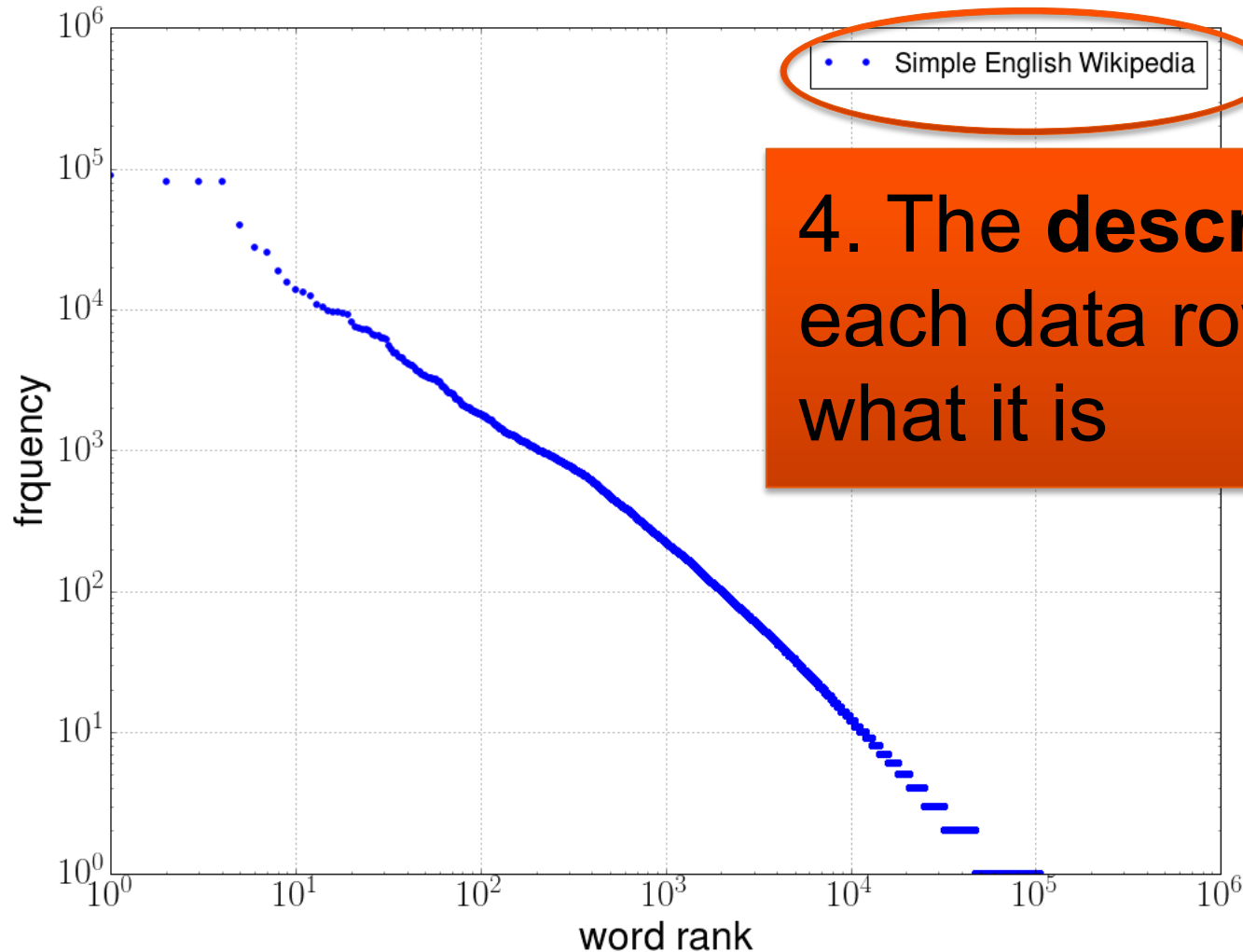
Wordrank frequency diagram on Wikipedia data sets



3. Look at the **scale** of each axis

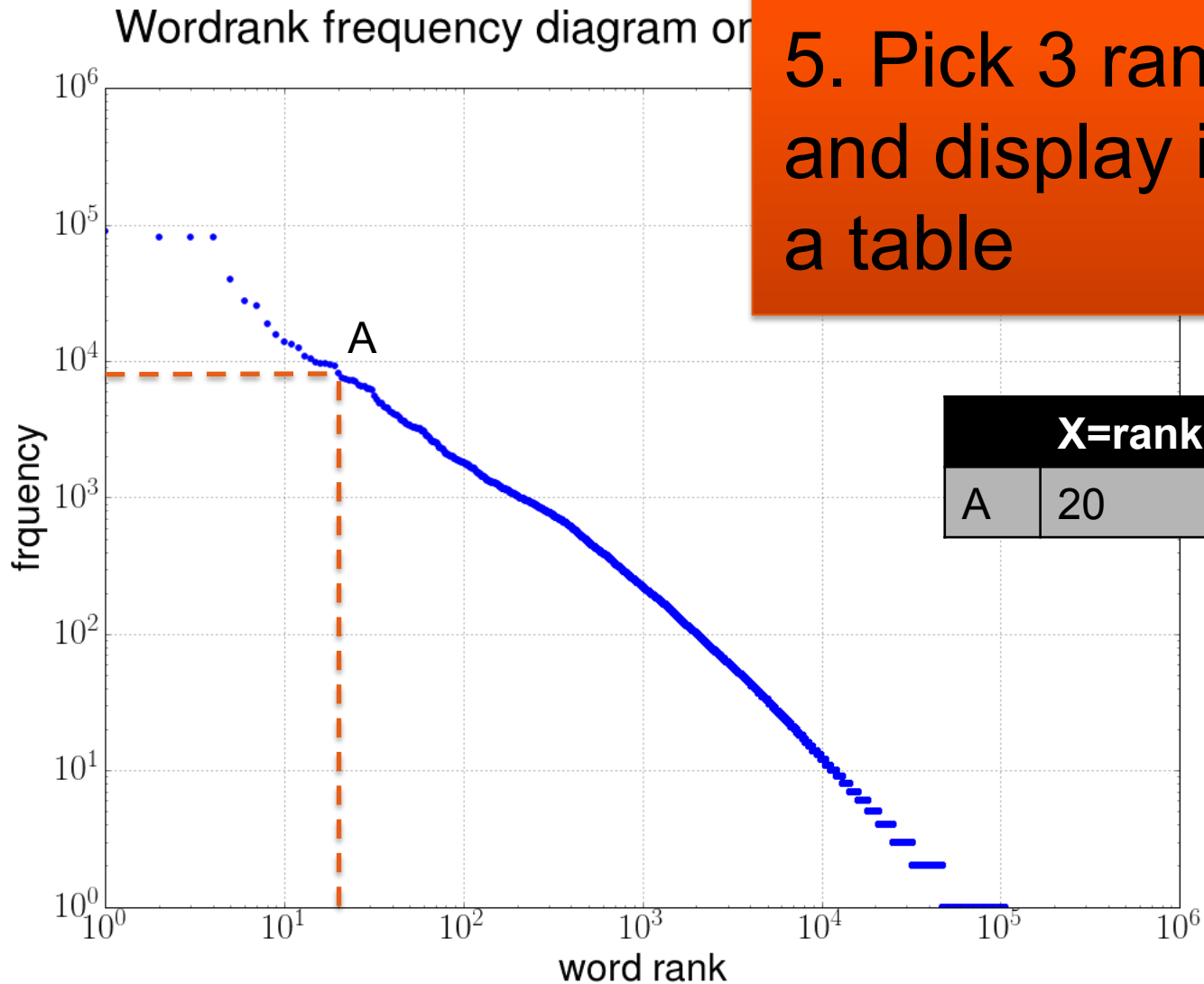
# 6 Steps of mastering reading (log-log) plots

Wordrank frequency diagram on Wikipedia data sets



4. The **description** of each data row tells you what it is

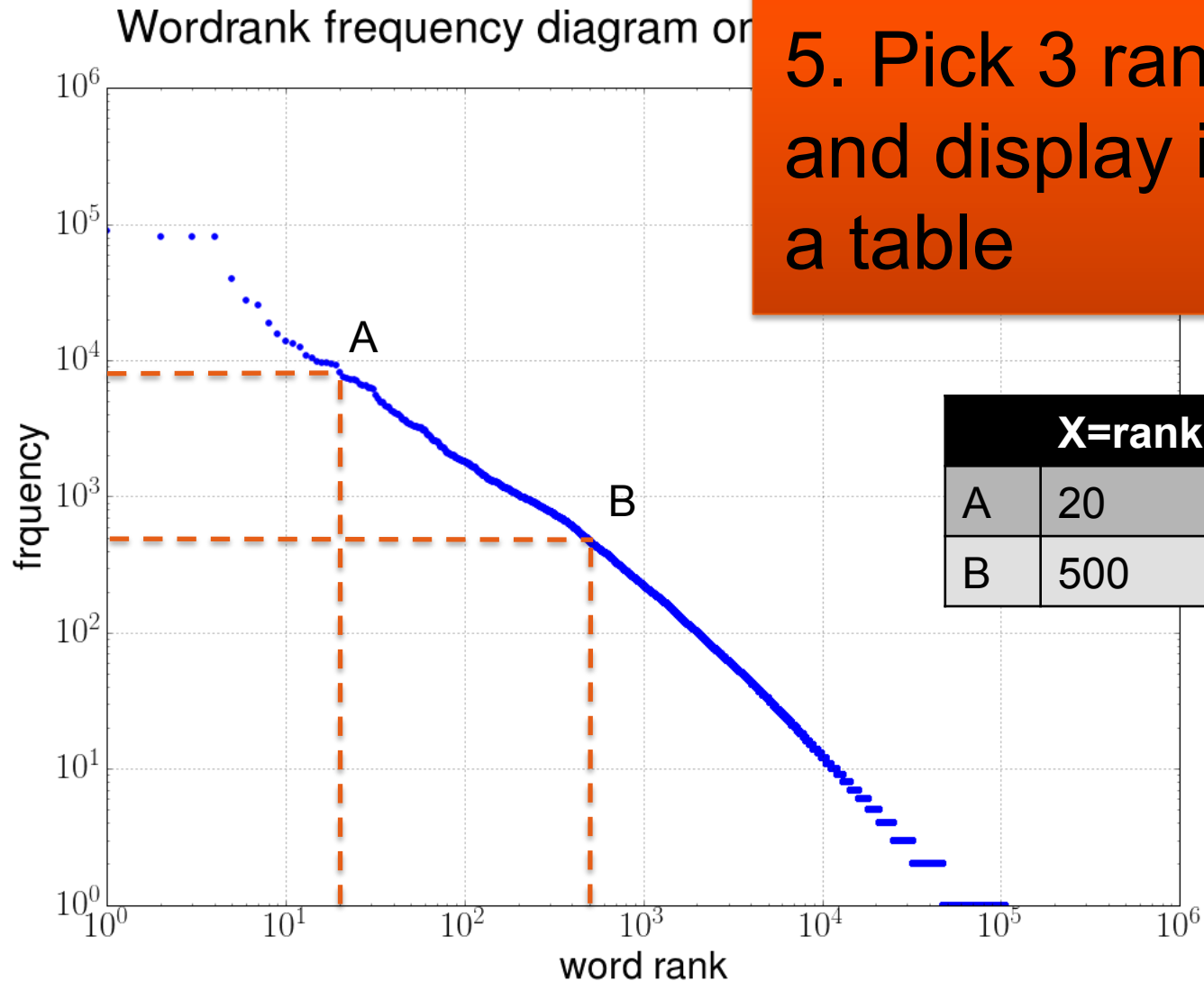
# 6 Steps of mastering reading (log-log) plots



5. Pick 3 random points and display its values in a table

	X=rank	Y=frequency
A	20	~8000

# 6 Steps of mastering reading (log-log) plots

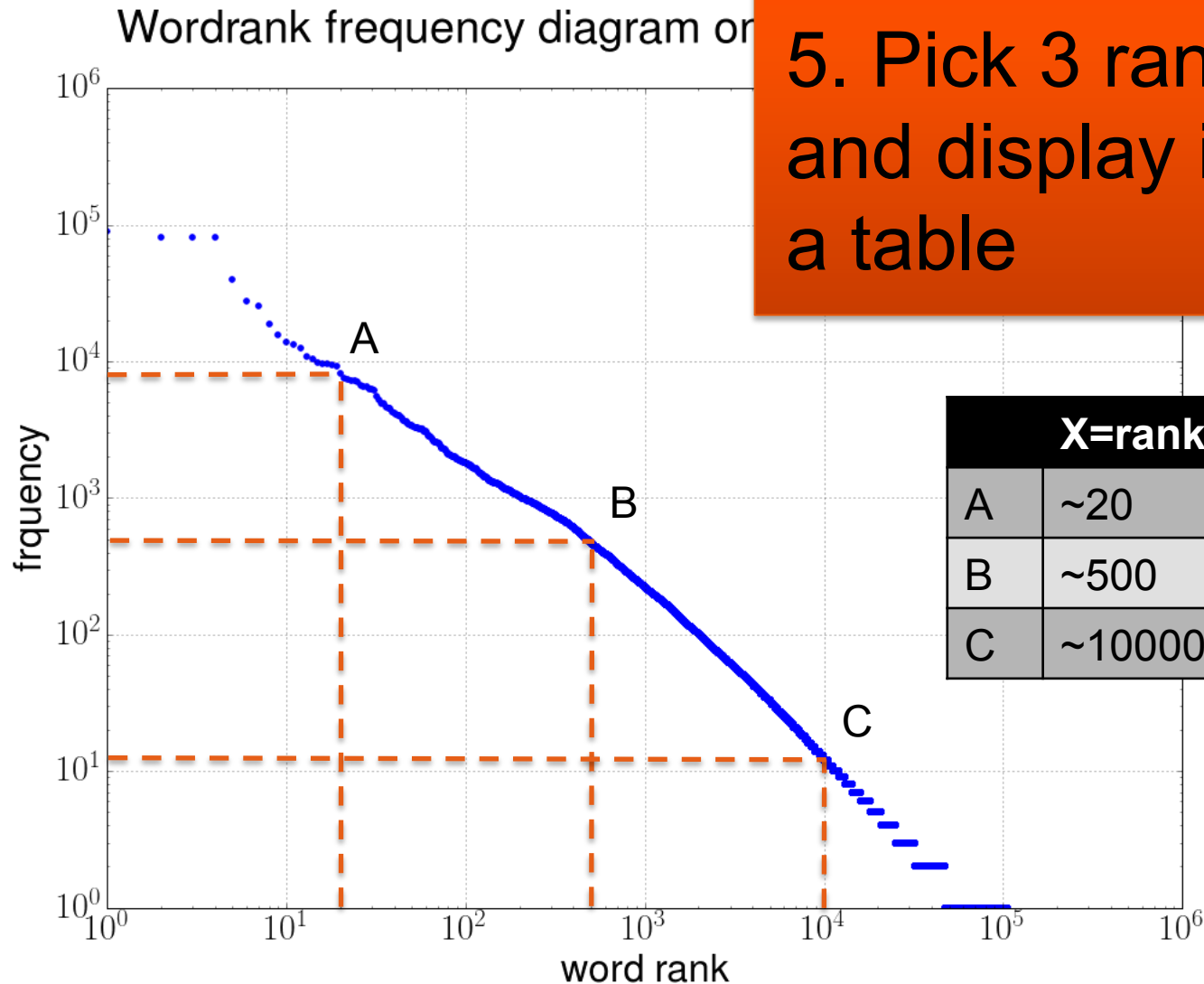


5. Pick 3 random points and display its values in a table

	X=rank	Y=frequency
A	20	~8000
B	500	~500



# 6 Steps of mastering reading (log-log) plots



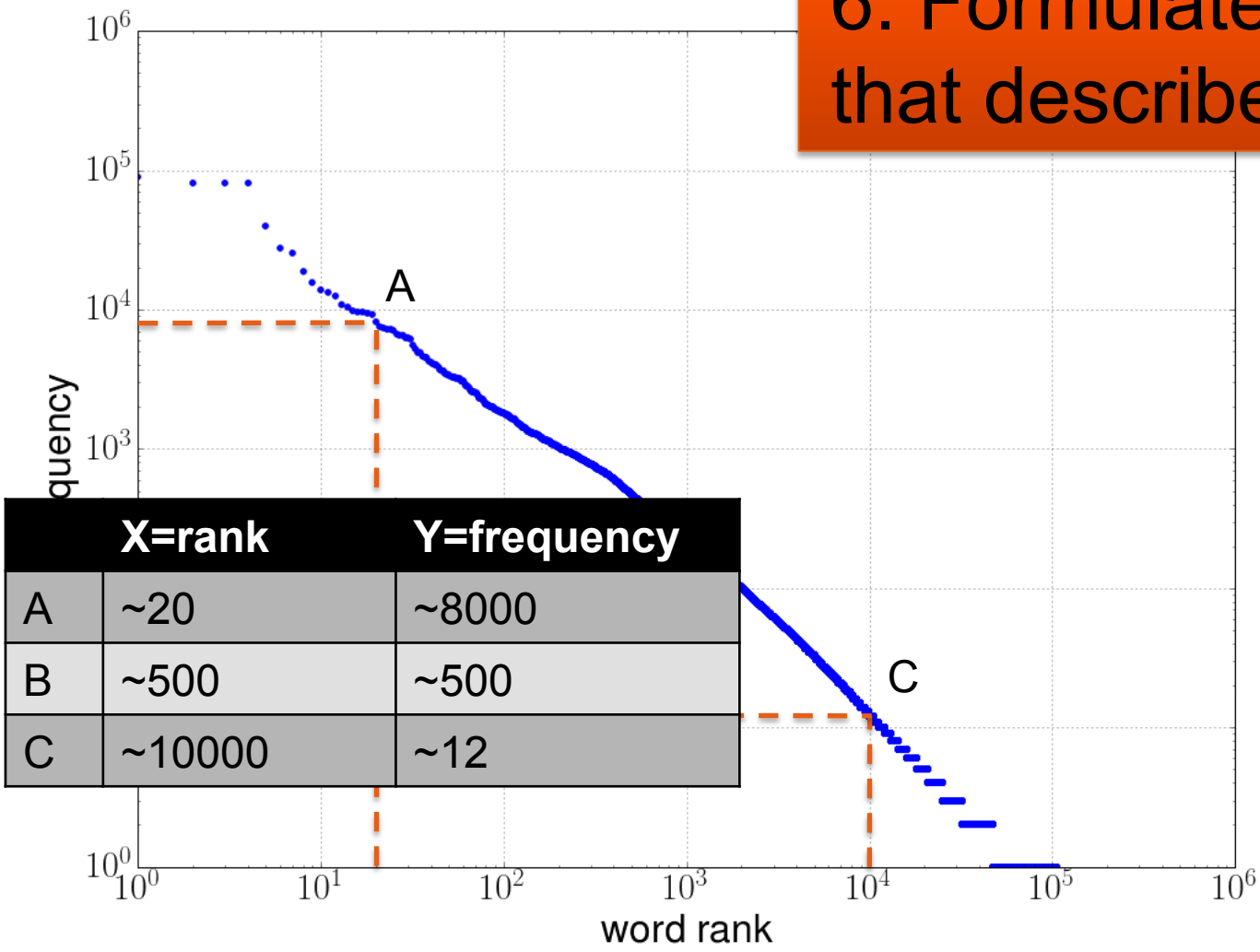
5. Pick 3 random points and display its values in a table

	X=rank	Y=frequency
A	~20	~8000
B	~500	~500
C	~10000	~12

# 6 Steps of mastering reading (log-log) plots

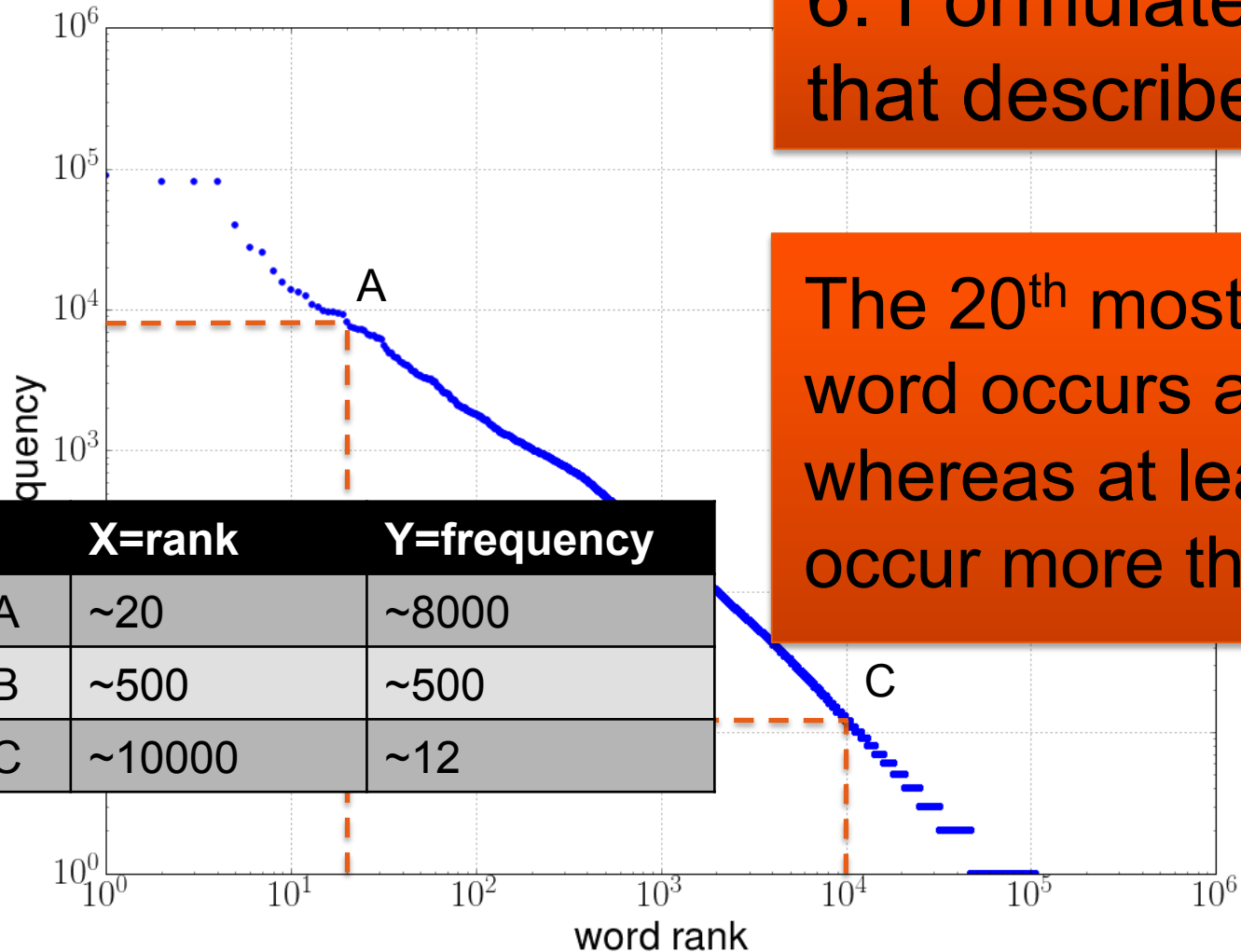
6. Formulate a sentence that describes the plot

Wordrank frequency diagram or



# 6 Steps of mastering reading (log-log) plots

Wordrank frequency diagram or



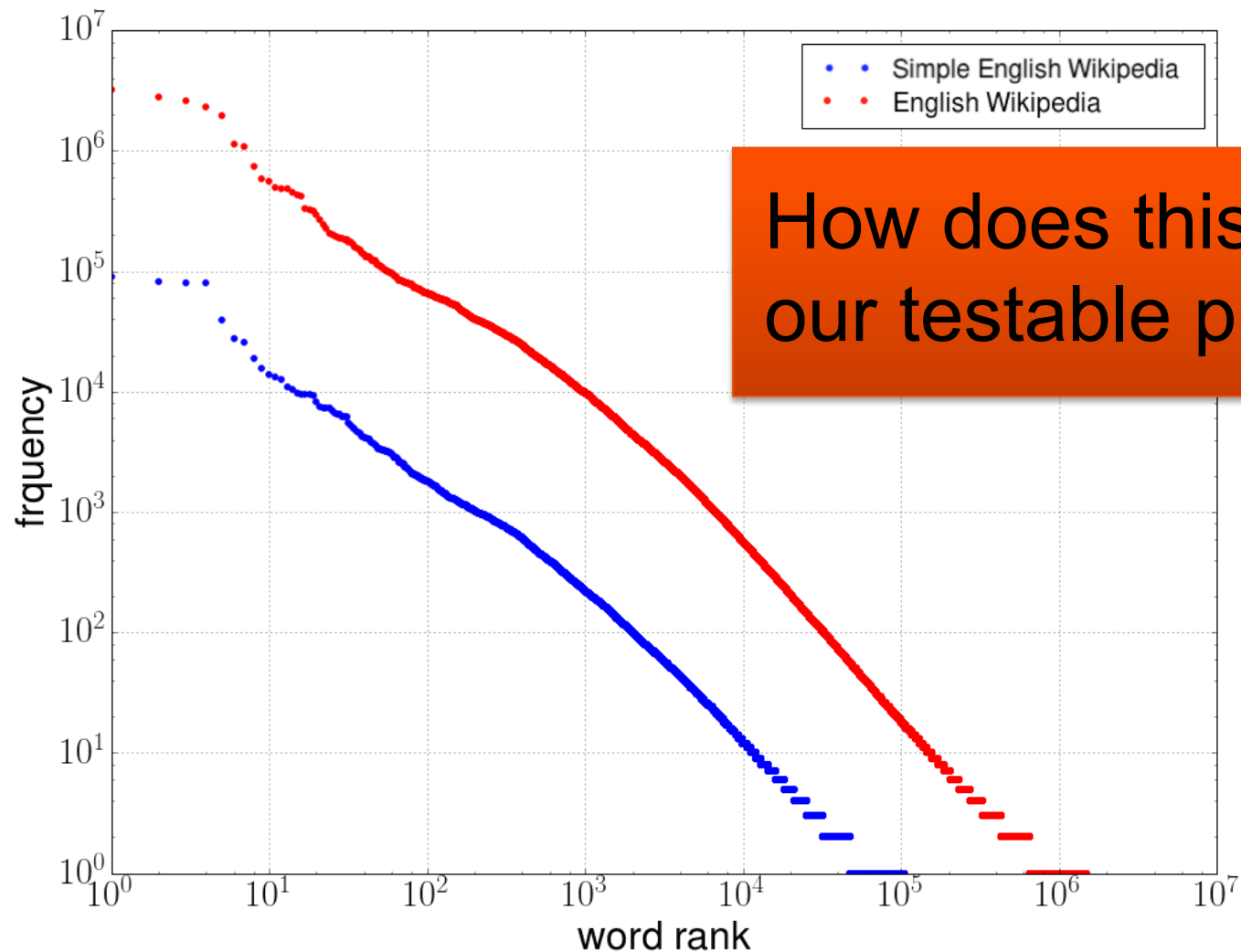
6. Formulate a sentence that describes the plot

The 20<sup>th</sup> most frequent word occurs about 8k times whereas at least 10k words occur more than ten times.



# Visualizing both data sets

Wordrank frequency diagram on Wikipedia data sets (log-log scale)

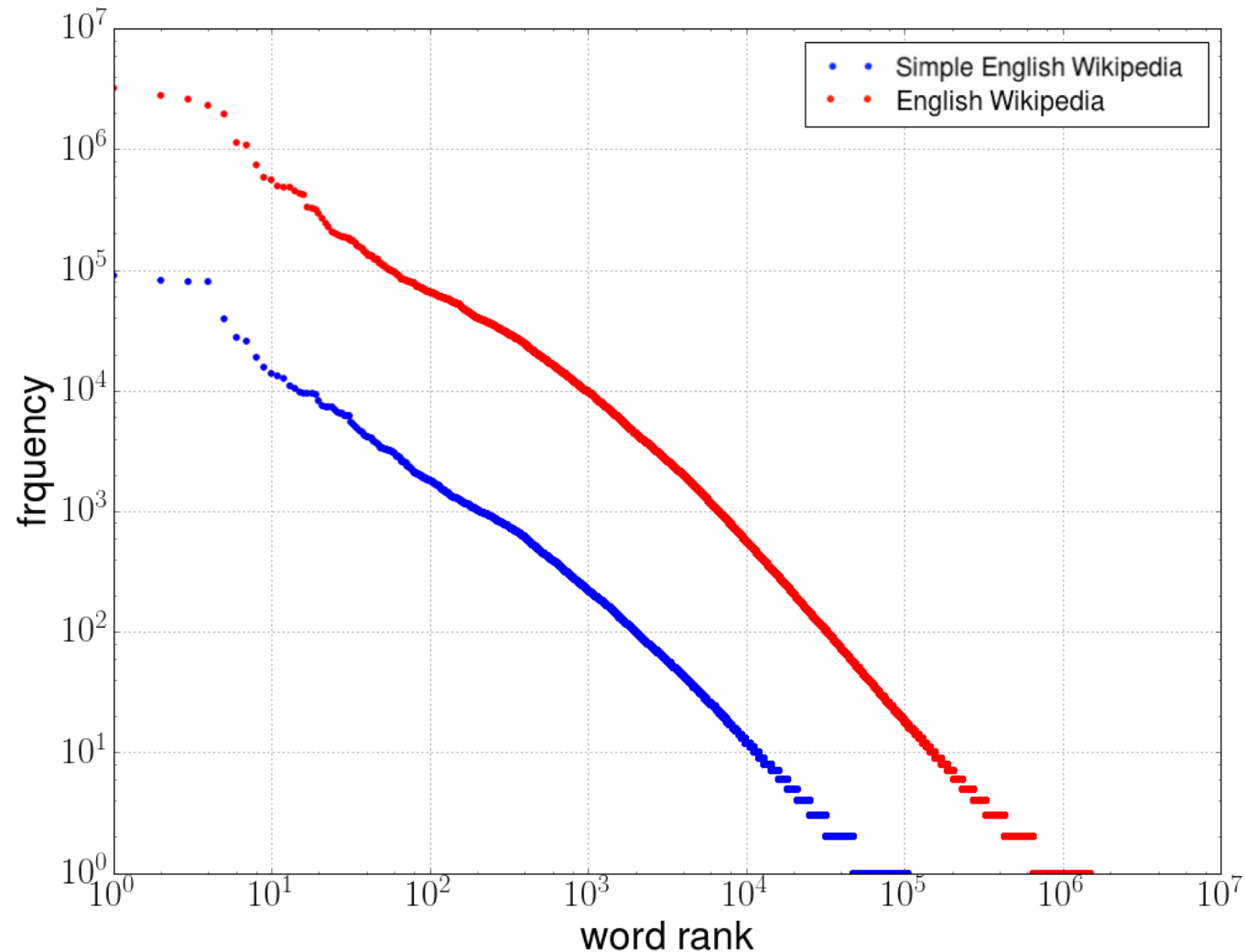


How does this support our testable prediction?



# Beware word order not the same!

Wordrank frequency diagram on Wikipedia data sets (log-log scale)





## Comparing the top 10 words

	Simple English Wiki	English Wiki
1 <sup>st</sup>	the	the
2 <sup>nd</sup>	is	of
3 <sup>rd</sup>	a	in
4 <sup>th</sup>	of	a
5 <sup>th</sup>	in	is
6 <sup>th</sup>	and	and
7 <sup>th</sup>	it	was
8 <sup>th</sup>	was	to
9 <sup>th</sup>	to	by
10 <sup>th</sup>	an	it
<b>Average frequency</b>	<b>20.04</b>	<b>57.74</b>
<b>Median frequency</b>	<b>1</b>	<b>1</b>



# Creating the Cumulative Distribution Function

```
In [ ]: from collections import Counter

def getWordCDF(f):
    allWords=readWordsFromWiki(f)
    c=Counter(allWords)
    words,frequencies = zip(*c.most_common())
    cumsum = np.cumsum(frequencies)
    normedcumsum = [x/float(cumsum[-1]) for x in cumsum]
    wrank = {words[i]:i+1 for i in range(0,len(words))}
    return wrank,normedcumsum

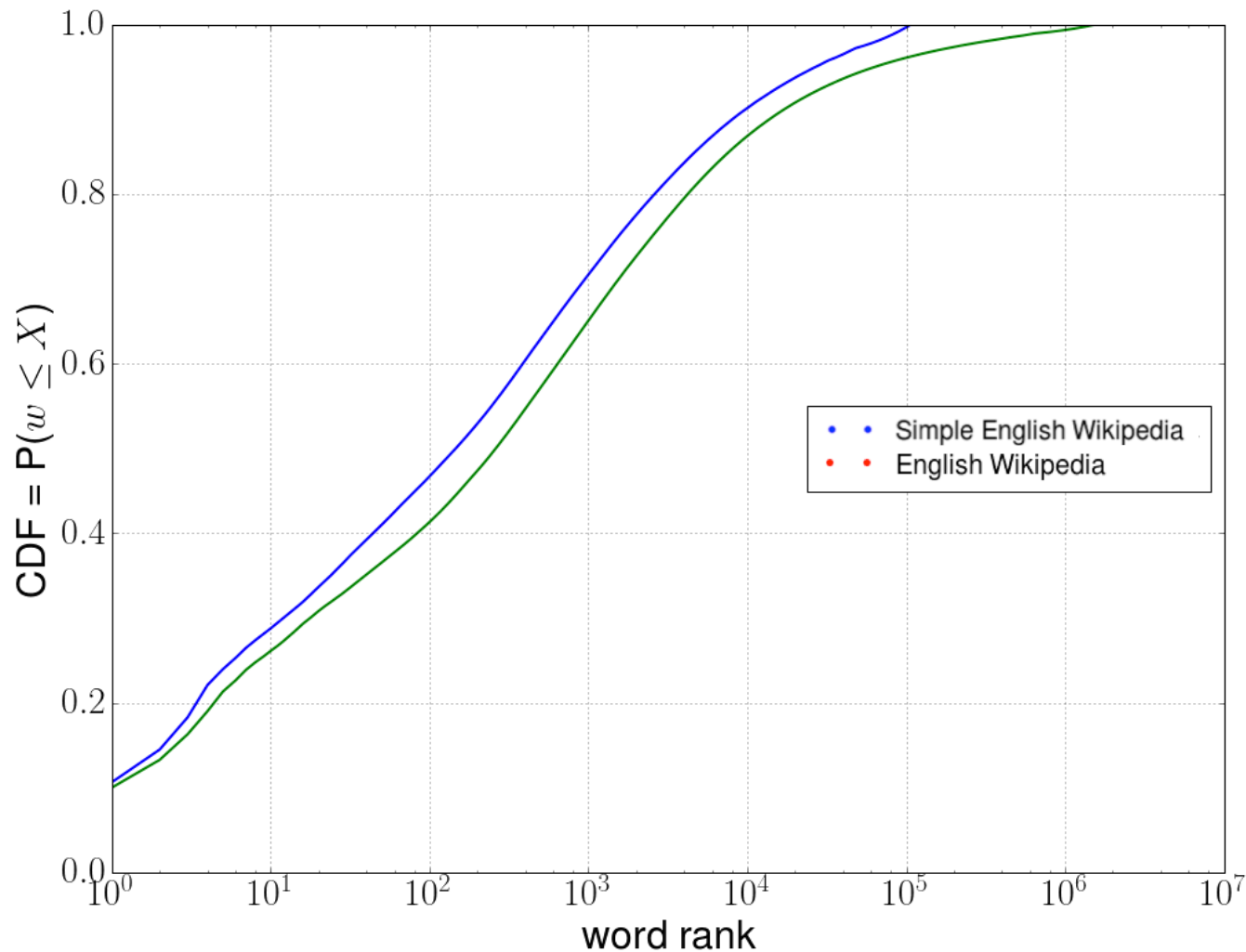
f = open("../datasets/simpleWikiAbstractsOneSentencePerLine")
simpleWordRanks, simpleNormedCumsum = getWordCDF(f)

f = open("../datasets/enWikiAbstractsOneSentencePerLine")
enWordRanks, enNormedCumsum = getWordCDF(f)
```



# Visualizing the CDF!

CDF wordrank frequency diagram on Wikipedia data sets





## Now lets be critical!

- Understanding 80% of all words does not necessarily mean that one understands 80% of the text
- Or do you understand the meaning of:
  - But it is her Schadenfreude
- English Wikipedia Corpus much bigger / more articles than Simple English
  - Comparing apples and peaches?
- Counting is ambiguous: Various forms for the “same” word like:
  - word, words
  - be, was, were, am, is
  - have, has, had

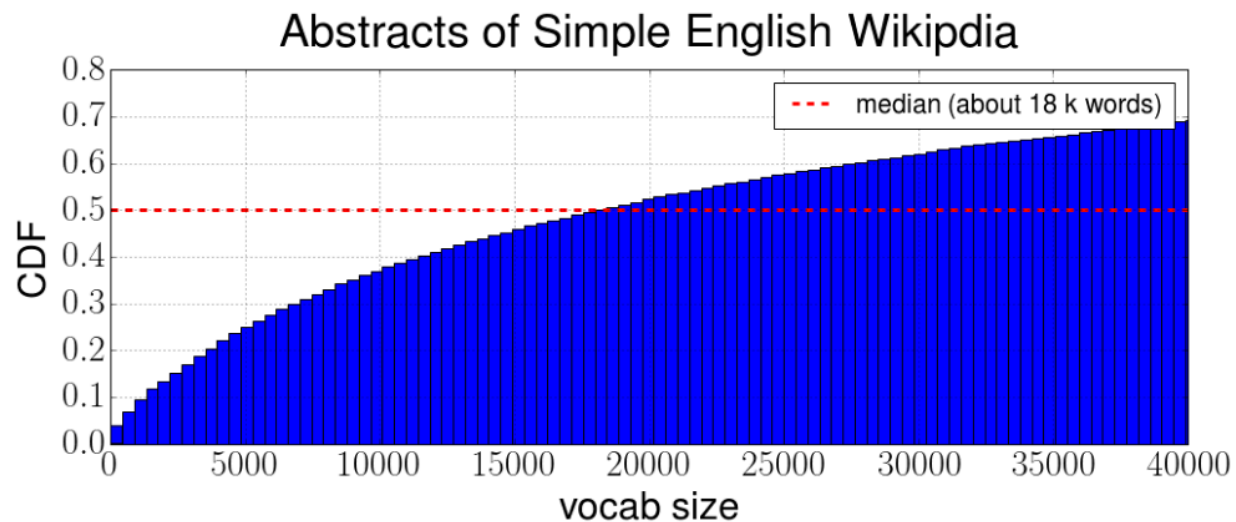
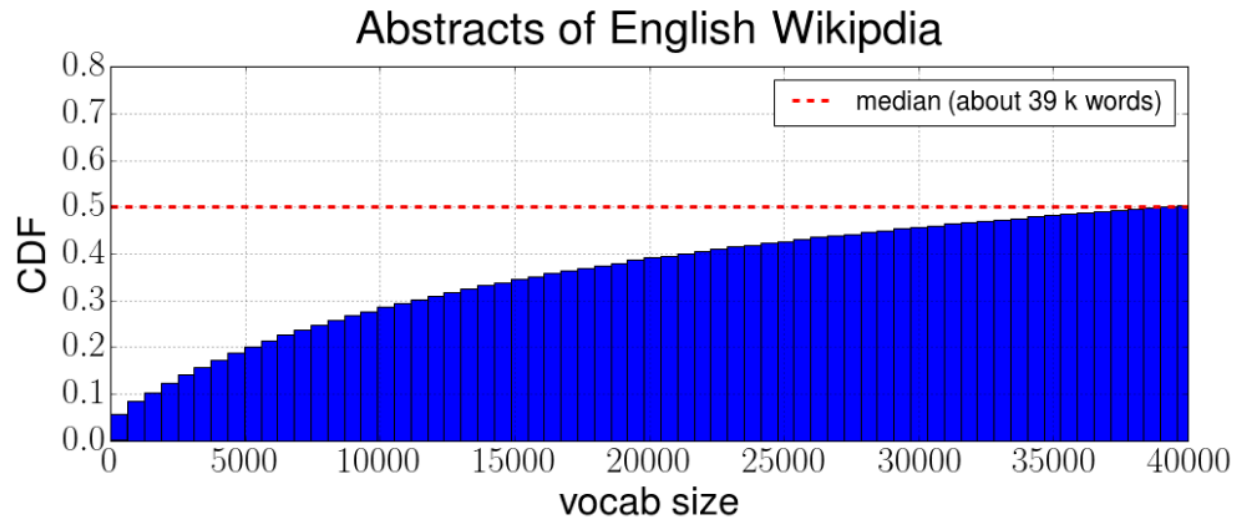


## **We could change the question a little bit**

- How many words does one need to know all words in a given sentence?
- Can be done with the same tools and techniques
- Lets dig directly into the results

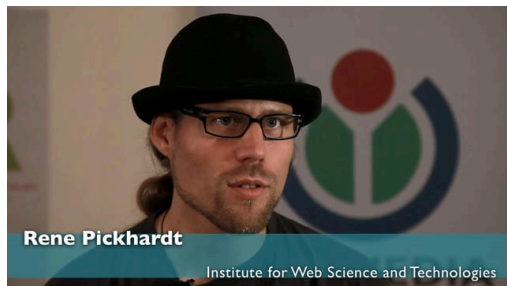
# Repeat on sentences instead of words

CDF for understanding all words in a sentence given a vocab of top popular words of a certain size





# Thank you for your attention!



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