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# Approximate Couting

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Abstract –The approximate counting algorithms allow counting a very large number of events using a small amount of memory, in other others, it is possible to use a small counter to keep approximate counts of large numbers. In this article, it is explored the fixed and the decreasing probability counters.

Keywords -Approximate, Counting, Probability, Algorithm, Decreased, Fixed

#### I. Introduction

The algorithms of approximate counters are very important since they allow estimating exact values, with an expected error that can be controlled [1].

They can be useful in examining large data streams for patterns, such as, in applications of data compression, sight and sound recognition, and other artificial intelligence applications [2].

## II. PROBLEM DESCRIPTION

The goal is to count the number of occurrences of letters in text files and also to identify the most common ones. This will be done in three different ways: with an exact counter, a fixed probability counter (1 / 8) and with a decreasing probability counter  $(1 / 2\hat{k})$ .

A fixed probability counter will use the same probability for every event that will decide to either count that event or not.

A decreased probability counter will use a probability different for each letter, this probability will get smaller every time an event is counted for the letter in analysis.

A series of tests will be performed to analyse the computational efficiency and limitations of the developed counters.

#### III. APPROACH AND IMPLEMENTATION

## A. Parameters Management

The program allows the user to pass some arguments that have different actions. The user can set a few parameters that allow for an easier use of the program. It can be passed the name of the file to perform the letter counting, the number of most occurrent letters to display and also the amount of repetitions/trials to perform the counting.

```
Usage: python3 main.py
-f <File Name for Counting Letters: str>
-k <Top k Most Occurrent Letters: int>
-r <Repetitions for Testing: int>
```

Fig. 1
PROGRAM ARGUMENTS

# B. Exact Counter

This is a counter with a very simple implementation. All it does is read the given file in chunks, count the letters and store in a dictionary with the exact number of occurrences of each letter. It also removes all non-alphabetical characters and transforms every letter to upper case.

Reading the file chunk by chunk could be crucial since it might be necessary to handle very large files, therefore, it may not have memory to read and store the whole file.

#### C. Fixed Probability Counter

This counter has a similar implementation to the previous one, except it only counts the letter occurrence if a random generated number by python is smaller or equal to 1/8.

To estimate the events after the counting, it can be simply done by multiplying the occurrences of the letter by the inverse of the probability, in this case, 8.

# D. Decreasing Probability Counter

For the decreasing probability counter, each letter will have different probabilities of being counted, as the more occurrences it has, the less probability it will have. The probability used was 1 / 2  $\hat{k}$ , where k is the number of occurrences of a letter.

In order to speed this process, it was kept another dictionary with the probability of each letter already calculated, this increased the performance by a big margin, because after a while the probability of counting a letter becomes so small it is rarely even counted, and to count each time the same probability takes way more time than having it only counted once every time it's value changes, that is when it is effectively counted. Since this problem is mainly focused in decreasing memory usage, this optimization might not be viable, but was used to speed up the testing process.

After the counting, the following formula was used to estimate the events of each letter:

```
int(a ** k - 1)
```

In this case, the a represents the base used, that was 2, and k the value of the counter of a given letter.

#### E. Results

For the testing part, it was used the text file of the Bible, for these firsts tests, it was always used the English version.

As it was mentioned previously, the program allows setting the number of trials to execute and also choosing the number of the most frequent letters to display.

# F. Results for 1 Trial

For a first analysis, it was performed a test with a single repetition and chosen to display the top 10 most frequent letters.

Total Elapse	r 1 repetition: d Time: 1.043 s Counted: 3218643
Average Valu Measure	es for a Repetition: Value
Counting Tim	e (s) 1.043
Alphabet Siz	
Events	3.21864e+06
Mean	123794
Minimum	1234
Maximum	400817
Top 10 Most Letter  E T O A H N I S R	Frequent Letters: Exact Events

Fig. 2
EXACT COUNTER RESULTS FOR 1 REPETITION

As it can be seen, it took slightly over a second to perform the test, and counted more than 3 million events. It also displays the alphabet size, which was the full English alphabet, 26, the mean, minimum, and maximum number of events for all letters, and then the top 10 most frequent letter.

These results are stored to be used for comparison with the next counters.

For the next two approximate counters, other statistics are calculated.

As it can be seen from the figures 3 and 4, the execution time of the fixed probability counter was a bit smaller than the decreasing one, this can be explained due to the fact that the decreasing probability counter has to calculated new probabilities for each letter. However, both of them take less time than the exact counter because of less memory accesses are done to increment the counters. As it can be seen, the fixed probability counted less a million events, and the decreasing one, only counted 402 events. The difference is huge when compared to the 3 million of the exact counter.

The alphabet size was the same in all counters, which means, every counter managed to count every letter at least once. The total and mean number of events has a smaller error in the fixed probability one, this is a bit obvious since it counted a lot more events.

Fixed Prob	pability	Counter	with 1 / 8	3				
Results for Total Elap Total Ever	osed Time	e: 0.649						
Average Va	alues for	a Repet	tition:					
Measure			Value	Absolute Erro	r	Relative	Error (%)	
Counting 1	 Гіте (s)	0.	.649					
Alphabet 9						0.0		J
Events			.22946e+06	10813.0 415.89		0.34		J
		124210		415.89		0.34		
						7.29		
Maximum		402584		1767.0		0.44		
Top 10 Mos	t Freque	ent Lette	erc.					
Letter		Max	Mean	Mean Absolute	Error	Mean	Relative Error (	(%)
	402584	402584	402584		1767		0.	.44
	286784	286784	286784		2206		0.	.78
	279688	279688	279688		198		0.	.07
A N I	266584	266584	266584		674		0.	. 25
N	216912	216912	216912		841		0.	. 39
		207368			818		0.	. 39
S	186656				1124		0.	.61
R D	176280				1707			.98
		143440	143440		2407		1.	.71
Accuracy:								
Precision								
Average Pi	recision	(relativ	/e order):	100.00 %				

 $\label{eq:Fig. 3} {\it Fixed Probability Counter Results for 1 Repetition}$ 

Average V Measure		r a Repe	tition: Value	Absolute Error	Rel	ative	Error (%	
Counting	Time (s)	0	.793					
lphabet	Size	26		0.0	0.0			
vents			.30493e+06	86291.0	2.6	8		
1ean		127113		86291.0 3318.89	2.6	8		
Minimum		511			58.			
1aximum		524287		123470.0	30.	8		
Top 10 Mo		ent Lett	ers:					
_etter	Min	Max		Mean Absolute E	rror	Mean	Relative	Error (%
etter 	Min			Mean Absolute E		Mean	Relative	Error (%
	524287	524287	Mean  524287			Mean	Relative	Error (% 84.2
	524287 524287	524287 524287	Mean  524287 524287		 397 <b>0</b> 9	Mean	Relative	
 [ [	524287 524287 262143	524287 524287 262143	Mean  524287 524287 262143	23 12 2	 39709 23470 20433	Mean	Relative	84.2 30.8 8.4
 [ ] 	524287 524287 524287 262143 262143	524287 524287 524287 262143 262143	Mean 524287 524287 524287 262143	23 12 2 8	9709 23470 20433 37570	Mean	Relative	84.2 30.8 8.4 50.1
 [ ] 	524287 524287 524287 262143 262143 262143	524287 524287 262143 262143 262143	Mean 524287 524287 262143 262143	23 12 2 8 7	23470 23470 20433 37570 76611	Mean	Relative	84.2 30.8 8.4 50.1 41.2
	524287 524287 262143 262143 262143 262143	524287 524287 262143 262143 262143 262143	Mean 524287 524287 262143 262143 262143		39709 23470 20433 37570 76611	Mean	Relative	84.2 30.8 8.4 50.1 41.2 6.3
	524287 524287 262143 262143 262143 262143 262143	524287 524287 262143 262143 262143 262143 262143	Mean	23 12 2 8 7 1	39709 23470 20433 37570 76611 17743	Mean	Relative	84.2 30.8 8.4 50.1 41.2 6.3
 [ : : : : : : :	524287 524287 524287 262143 262143 262143 262143 131071	524287 524287 524287 262143 262143 262143 262143 131071	Mean 	23 12 2 8 7 1 14	39709 23470 20433 37570 76611 17743 16243	Mean	Relative	84.2 30.8 8.4 50.1 41.2 6.2 126.1
 [ : : : : : : :	524287 524287 262143 262143 262143 262143 262143 131071 131071	524287 524287 262143 262143 262143 262143 262143	Mean 524287 524287 262143 262143 262143 262143 131071 131071	23 12 2 8 7 1 1 13	39709 23470 20433 37570 76611 17743	Mean	Relative	84.2 30.8 8.4 50.2 41.2 6.3

Fig. 4
Decreasing Probability Counter Results for 1
Repetition

From the top 10 most frequent letters, it is also shown the error values compared to the exact counter. Overall, all errors are much smaller on the fixed probability one than on the decreasing one, which is to be expected since the total events counted is much smaller. It also shows the accuracy, precision, and the average precision considering relative order of the top letters displayed. The fixed probability counter had the same order as the exact counter, but the decreasing one, although it had high accuracy and precision, it didn't have any letter in the correct order.

# G. Results for 100 Trials

In order to better visualize the differences, these algorithms will be executed multiple times to better understand the results, in this case, 100 trials.

For the exact counter, it can be seen the total elapsed time, the number of counted events, and the average time for 1 repetition, all the other results are obviously the same as with 1 repetition, so they are not displayed here, but can be consulted on the figure 2.

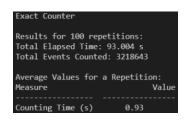


Fig. 5
EXACT COUNTER RESULTS FOR 100 REPETITIONS

Visualizing the figures 6 and 7, the same conclusions regarding time elapsed, events counted, and alphabet size can be made from these results as the previous one with only 1 repetition.

Edward David	L-L2724	. C		0				
rixed Pro	poadility	Counter	with 1 /	8				
Results f	or 100 r	epetitio	ns:					
Total Ela								
Total Eve	nts Coun	ted: 402	230.96					
Average \	raiues to	га кере		Absolute Error	Relative	(W)		
neasure			varue	Appointe ELLOI.	Ketative	ELLOI. (%)		
Counting								
Alphabet				0.0	0.0			
Events			.21785e+06		0.02			
Mean					0.02			
Minimum					0.6			
Maximum		400748		68.84	0.02			
Top 10 Mc	ok Engan	ant Late						
Letter				Mean Absolute Erro	r Mean	Relative Error (%)	Variance	Standard Deviation
E	396368	403936	400748	68.8		0.02	2.83761e+06	1684.52
T	281808	287576	284653	75.2		0.03		1276.88
О	275656			310.3		0.11		1544.59
Α	263928	269816		24.9		0.01		
Н	236112		241454	255.9		0.11		1340.75
N	212904			118.6		0.05		1237.17
I	203984		208120	66.3		0.03		1215.47
S	183280	187576		47.2			956088	977.798
R	171136		174455	117.8		0.07		
D		143440	140882	151.6	18	0.11	1.11304e+06	1055.01
Accuracy: Precision								
			ve order):	100 00 %				
Average F	Lectaton	(Leigit	ve order):	100.00 A				

Fig. 6
Fixed Probability Counter Results for 100 Repetitions

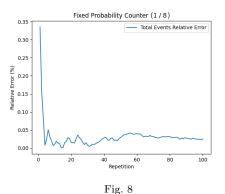
Measure		r a Repet	Value	Absolute Error	Relative Error (%			
Counting	Time (s)		847					
				0.0	0.0			
Events			19325e+06	25394.12	0.79			
				976.7	0.79			
Minimum Maximum				372.28	30.17			
		574094		173277.36	43.23			
		ent Lette Max	rs:			e Error (%)	Variance	Standard Deviatio
Letter 	ost Frequ Min	ent Lette Max	rs:		ror Mean Relativ		Variance 4.43241e+10	
Letter  E	ost Frequ Min  131071	ent Lette Max  1048575	rs: Mean	Mean Absolute Er	ror Mean Relativ	8.44		
Letter  E A	ost Frequ Min  131071 65535	ent Lette Max  1048575 1048575	rs: Mean  367001	Mean Absolute Er	ror Mean Relativ	8.44 10.35	4.43241e+10	210533
Letter E A O	ost Frequ Min  131071 65535 65535	ent Lette Max  1048575 1048575 1048575	ms: Mean  367001 294911	Mean Absolute Er 33816 27653	ror Mean Relativ	8.44 10.35 1.86	4.43241e+10 4.66863e+10	210533 216070
Letter E A O T H	ost Frequ Min  131071 65535 65535 65535	ent Lette Max 1048575 1048575 1048575 1048575 524287	Mean 367001 294911 285081 280493 227409	Mean Absolute Er 33816 27653 5194 4084 14301	ror Mean Relativ 	8.44 10.35 1.86 1.44	4.43241e+10 4.66863e+10 5.27744e+10 3.28204e+10 1.73908e+10	210533 216070 229727 181164 131874
Letter E A O T H	ost Frequ Min 131071 65535 65535 65535 65535	ent Lette Max 1048575 1048575 1048575 1048575 524287	Mean 	Mean Absolute Er 33816 27653 5194 4084	ror Mean Relativ 	8.44 10.35 1.86 1.44	4.43241e+10 4.66863e+10 5.27744e+10 3.28204e+10	210533 216070 229727 181164
Letter E A O T H N R	ost Frequ Min 131071 65535 65535 65535 65535 65535 32767	1048575 1048575 1048575 1048575 1048575 524287 1048575 1048575	Mean 367001 294911 285081 280493 227409 218889 199884	Mean Absolute Er 33816 27633 5194 4084 14301 2818 25310	.4 .6 .92 .1 .24	8.44 10.35 1.86 1.44 5.92 1.3	4.43241e+10 4.66863e+10 5.27744e+10 3.28204e+10 1.73908e+10 2.26104e+10 2.50933e+10	210533 216070 229727 181164 131874 150368 158409
Letter E A O T H N R	ost Frequ Min  131071 65535 65535 65535 65535 65535 65535 32767 65535	ent Lette Max  1048575 1048575 1048575 524287 1048575 524287 524287	Mean	Mean Absolute Er 33816 27653 5194 4084 14301 2818 25310 12234	nor Mean Relativ .4 .6 .92 .1 .24 .8	8.44 10.35 1.86 1.44 5.92 1.3 14.5	4.43241e+10 4.66863e+10 5.27744e+10 3.28204e+10 1.73908e+10 2.26104e+10 2.50933e+10 1.7566e+10	210533 216670 229727 181164 131874 15988 158409
Top 10 M Letter E A O T H N R I I S	ost Frequ Min 131071 65535 65535 65535 65535 65535 32767	ent Lette Max  1048575 1048575 1048575 524287 1048575 524287 524287	Mean	Mean Absolute Er 33816 27633 5194 4084 14301 2818 25310	ror Mean Relativ .4 .6 .92 .1 .24 .8 .4	8.44 10.35 1.86 1.44 5.92 1.3 14.5 5.88	4.43241e+10 4.66863e+10 5.27744e+10 3.28204e+10 1.73908e+10 2.26104e+10 2.50933e+10	210533 216070 229727 181164 131874 159368 158409 13257 98083.

Fig. 7
Decreasing Probability Counter Results for 100
Repetitions

However, averaging all repetitions and getting the of number of estimated events, the mean, minimum and maximum, much smaller errors are obtained. The minimum and maximum values have much higher relative errors on the decreasing probability counter comparing to the fixed one. This can be easily explained, letters with little occurrences will be harder to estimate using a decreased probability because the probability grows smaller than the fixed one really fast, so it might end up skipping a lot of values. A similar explanation for the maximum can also be made, if by chance the letter that occurs many times is counted even with a minimal probability, it will estimate a much larger number of events than what it really happens.

Considering the top most frequent letters, it also displays the minimum, maximum, the mean, the mean absolute and relative errors, the variance and the standard deviation of the registered values. These values help visualize the amount of variation or dispersion of a set of values. A low standard deviation indicates that the values tend to be close to the mean of the all the estimated counters for that letter, while a high standard deviation indicates that the values are spread out over a wider range [3].

It can be observed that the minimum and maximum values do not deviate too much from the mean on the fixed probability counter, but in the decreasing one, the deviations are much higher, for example on the letter E, has the minimum value of 130 071 and the maximum of over a million, but the mean is 367 001, which shows that it can have a big variance in some repetitions, having a standard deviation of 210 533. On the fixed probability counter, they are around 1 000, which is significantly smaller. The accuracy and precision of the top letters was also higher than with just 1 repetition for the decreasing probability counter, and also the average precision considering the relative order increased from 0% to 34.33%.



Fixed Probability Counter Relative Error of the Average Total Estimated Events for 100 Repetitions

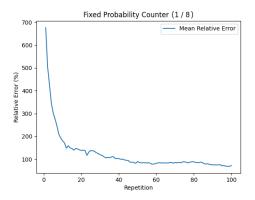


Fig. 9
Fixed Probability Counter Mean Relative Error of all
the Estimated Events for 100 Repetitions

Regarding the fixed probability counters, the graphics above, 8 and 9, show, respectively, the estimated total events relative error, and the average of the approximations relative errors, this is, the mean error of the estimated occurrences of each letter comparing to the exact counter value. Both of these two graphics, tend to stabilize after approximately 60 repetitions, this demonstrates that if the problem is only memory related, it can be obtained approximations with minimal errors by averaging the results of 60 repetitions, however, the elapsed time would be much higher.

Comparing the previous graphs with the ones for the decreased probability counter, 10 and 11, it can be observed they also tend to stabilize after 60 repetitions, however, they stabilize on much higher error values and have much higher fluctuations, which means the graph doesn't look as linear as the previous.

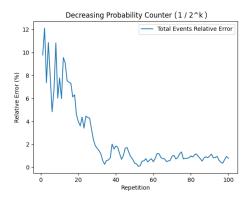


Fig. 10

Decreased Probability Counter Relative Error of the Average Total Estimated Events for 100 Repetitions

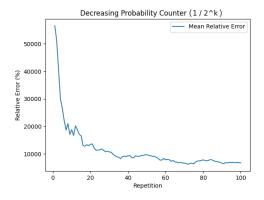


Fig. 11
Decreased Probability Counter Mean Relative Error of all the Estimated Events for 100 Repetitions

# H. Different Languages Analysis

For the previous tests, the English version of the Bible was the one used, but it is also interesting to test the same book but in different languages. So it was also used in the German and Portuguese versions. These two languages have different alphabets, as it can be seen from the figures 2 and 12, the alphabet size of the English one has 26 letters, the German has 30 letters and the Portuguese 42, a big reason for that difference is the accents, since English does not have any, and also, German alphabet has other letters, for example, 'ß', and the Portuguese as well, such as 'Ç'.



Exact Counter	
Results for 1	renetition:
	Time: 0.796 s
	Counted: 2985343
TOTAL EVELLES	Counted: 2985343
Average Value	es for a Repetition:
Measure	Value
Counting Time	e (s) 0.796
Alphabet Size	42
Events	2.98534e+06
Mean	71079.6
Minimum	1
Maximum	406759
Top 10 Most F	requent Letters:
Letter E	xact Events
E	406759
Α	348622
0	332322
S	282939
R	198395
I	161611
D	153589
U	136263
М	132984
N	132549

Fig. 12
EXACT COUNTERS FOR GERMAN BIBLE (LEFT) AND
PORTUGUESE BIBLE (RIGHT)

However, they still have similar letters, and the most frequent ones tend to be the vowels 'A', 'E', 'I', but 'O' in English and the 'U' in German do not show in the top 10. Also, consonants like 'N', 'S' and 'R' are very frequent in these 3 languages.

# IV. CONCLUSION

Overall, it was tested two different approximate counters, in order to analyse their computational efficiency and evaluate the approximations' errors.

From the results, it can be concluded that it is possible to count very large number of events while having an amount of memory and estimations' errors that can be controlled by adjusting the probabilities used or/and the algorithms used.

#### References

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