# Guide for Reproducibility of A Scalable Model for Frequency Distribution of Low Occurrence Multi-words Towards Handling Very Large Spectrum of Text Corpora Sizes

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#### 1 Introduction

This is a guide to reproduce the algorithms and the results related to the content of the paper titled A Scalable Model for Frequency Distribution of Low Occurrence Multi-words Towards Handling Very Large Spectrum of Text Corpora Sizes. File ArchivedNgrams.zip must be uncompressed and all files in it must be placed in the same folder. These are the files used to generate the results (except figures) reported in the paper:

Def\_ValidationGHofC\_Phase.py
Def\_TestingGHofC\_Phase.py
Def\_ConstantsAndGlobalVars.py
Def\_DistByLangNandK.py
Def\_KThresolds.py
Def\_TestingCorporaSizes.py
Def\_ValidationCorporaSizes.py
Def\_Vocabulary.py
Def\_monotony.py
BaseLineModel.py
BaseLineModelDistKBD.py
BaseLineModelTrainingCorpora.py
BaseLineModelVocabulary.py
Def\_smooth\_spline\_results.pkl
Def\_smooth\_spline\_results\_const.pkl

And the following are the files related to the generation of the figures in the paper:

 $\label{lem:continuous} GerFig1k1.txt, powerlawVD1\_k1.txt, powerlawVD1\_k2.txt, strait\_k1.txt, strait\_k2.txt, GerFigDk.txt, EmmD1\_2grams.txt, EmmD1\_3grams.txt, EmmD2\_2grams.txt, EmmD2\_3grams.txt, PredD1\_2grams.txt, PredD1\_3grams.txt, PredD2\_2grams.txt, PredD2\_3grams.txt, GerFigWkbyC.txt, EmpWk11Gw.txt, \\$ 

EmpWk172Gw.txt, EmpWk31Gw.txt, EmpWk366Mw.txt, EmpWk373Gw.txt, EmpWk82Gw.txt, PredWk11Gw.txt, PredWk172Gw.txt, PredWk31Gw.txt, PredWk31Gw.txt, PredWk366Mw.txt, PredWk373Gw.txt, PredWk82Gw.txt.

Next sections explain how to reproduce the results of the paper.

## 2 Reproducing the Results of the Proposed Model

The results in the paper can be reproduced alternatively by:

- a) Running the learning phase first, and then running the test phase.
- b) Running the test phase based on the learning phase, which is already available.

Regarding a), if we want to train the model for a language and an n-grams size, we must use python function FindParametersCrossVal(Lang, NgramSize, Vi, Vf, Vs, InpercG, EnpercG, NstpG, InpercH, EnpercH, NstpH) available in the file Def\_ValidationGHofC\_Phase.py. The meaning of the parameters are: the language; the n-gram size; the initial, the final and the step values for vocabulary size searching; the initial, the final and the number of steps values for parameter  $g_k$  (relative values around estimated points performed by the algorithm as explained in the paper); the initial, the final and the number of steps values for parameter  $h_k$  (relative values around estimated points performed by the algorithm as explained in the paper). These parameters are respectively for Lang, NgramSize, Vi, Vf, Vs, InpercG, EnpercG, NstpG, InpercH, EnpercH and NstpH. As an example, we can run

FindParametersCrossVal('en',2,1.985e10,2.00e10,1e8,0.4,1.7,400,0.4,1.7,200) to train the model for English 2-grams. Then we will see:

It may take some minutes

k thresold for training corpora: 17

Processing for the hypothetical vocabulary size: 19850000000 Processing for the hypothetical vocabulary size: 19950000000

Best Vocabulary Size: 19950000000

The model can also be trained for German ('de') (since there are empirical n-gram counts available also for German in file **Def\_DistByLangNandK.py**). Other n-gram sizes (1,3,4,5,6) are also considered.

This function performs: the estimation of the vocabulary size and saves it in file  $\mathbf{Def}$ \_Vocabulary.py; calculates the k-threshold value – file  $\mathbf{Def}$ \_monotony.py is used for that – and saves it in file  $\mathbf{Def}$ \_KThresolds.py; saves the  $g_k$  and  $h_k$  values resulting from splines and saves them in file  $\mathbf{Def}$ \_smooth\_spline\_results.pkl. After the training phase, results can be obtained as explained next.

Regarding b) or for the case when training has already been done, results can be obtained by using some functions of file **Def TestingGHofC Phase.py**.

#### 2.1 $D_k$ Predictions for each Corpus Size

In order to obtain the D(k, C; L, n) values (the number of distinct n-grams with frequency greater or equal to k, for a corpus of size C, for language L and n-gram size n), we must run  $\mathbf{AvgDkRelErrorForEachTestCorpusWithScale}(\mathbf{L,n})$ . For English 2-grams, for example, it will be AvgDkRelErrorForEachTestCorpusWithScale('en',2). The results will be:

Prediction of Relative Errors and Mean Square Root of Dk Relative Errors for C=366397190:0.008778096245756954~0.012501532643591398Prediction of Relative Errors and Mean Square Root of Dk Relative Errors for C=11344756226:0.005131186374030359~0.007755717180264833Prediction of Relative Errors and Mean Square Root of Dk Relative Errors for C=31487751849:0.001390248848040873~0.0016880610900871554Prediction of Relative Errors and Mean Square Root of Dk Relative Errors for C=82718285912:0.003753852597988539~0.004208358650843619

The *corpora* sizes shown belong to the test set. Other n-gram sizes (1,3,4,5,6) can also be used, as well as German ('de').

Global Averages for Relative Errors, for C: 0.004763346016454182

## 2.2 $D_k$ Predictions for each k Value

To obtain the D(k, C; L, n) values for each k value, we must run (example for German 1-grams) **EachDkRelErrorTroughTestCorporaWithScale('de',1)**. The result will be:

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 1: 0.00849053017418453 \ 0.01218650642833925$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k >= 2:0.0067200672086275585\;0.009679657957020624$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 3:0.007771722277930806~0.012673282797699581

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 4:0.009088328578489901~0.015655086012544086$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 5: 0.010384346674730975 0.018366500908390555

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 6:0.010539299564900937~0.01942034915756724$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 7:0.009905875239372726~0.018642073728824215

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k >= 8 : 0.01036276032303929 \ 0.019552127288343102$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 9: 0.01015909784412201 0.0190171907397087

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 10:0.010158350377593947~0.01915349820125297$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 11:0.00984579931560585~0.017586457472024166

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 12:0.009340617522387946~0.016450594745856965

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 13:0.008641381239376428~0.015496191975991542

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 14:0.008342976111642999~0.015271078536159057$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 15:0.007495007791075804~0.013901153749269816$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 16:0.007320675764995028$  0.013118185838602345

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 32:0.006133040842653309~0.011437557708962582

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 64 : 0.002435957309486738 \ 0.002946845752580633$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 128: 0.000815778653078257 0.0012061565512207256

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 256:0.0012151504942092381~0.0014729533324368922$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 512:0.004117180032897926:0.00552603492760623

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 1024:0.006165708488898545~0.008875552862705962

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 2048 : 0.0029237630907731653 0.004250603817119361$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 4096 : 0.002899521855983909 \ 0.0031710731719632405$ 

Global Averages for Relative Errors for Dk: 0.007136372365669076

#### 2.3 $W_k$ Predictions for each Corpus Size

To obtain the W(k, C; L, n) values (the number of distinct n-grams with frequency equal to k, for a corpus of size C, for language L and n-gram size n), we must run (example for English 3-grams) **AvgWkRelErrorForEachTestCorpusWithScale('en',3)**. Result will be:

Prediction of Relative Errors and Mean Square Root of wk Relative Errors for  $C=366397189:0.018642560759557277\;0.023954128241150348$ 

Prediction of Relative Errors and Mean Square Root of wk Relative Errors for C = 11344756219 : 0.011709805185349254 0.01648502262704967

Prediction of Relative Errors and Mean Square Root of wk Relative Errors for C=31487751831:0.0042812559756207685~0.006066230839741724

Prediction of Relative Errors and Mean Square Root of wk Relative Errors for C=82718285866:0.0034558206248180624~0.005236572183512354 Global Averages for Relative Errors, for C:0.00952236063633634

#### 2.4 $W_k$ Predictions for each k Value

To obtain the W(k, C; L, n) values for each k value, we must run (example for German 4-grams) **EachWkRelErrorTroughTestCorporaWithScale('de',4)**. The result will be:

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk,  $k = 1:0.006245231763732335\;0.009674409157842576$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk, k = 2:0.008660560169364485~0.015195507344587841

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk, k = 3:0.014151072049539562~0.02617537350252586

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk, k = 4:0.00797595253373079~0.009940075914550626

Prediction of Relative Errors and Mean Square Root of Relative Errors for  $Wk, k = 5:0.003644634169138551\ 0.0062387260556056$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk,  $k = 6: 0.03046110800193105 \ 0.05306780442537013$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk, k = 7:0.0047830080276595~0.0071817324271585365

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk,  $k = 8:0.009510117730532548\ 0.015004339364116112$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk, k = 9:0.00506753558088398 0.006639749181006347

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk,  $k = 10:0.0083602061303187\;0.012146890712848066$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk, k = 11:0.05336989963602594~0.10281428610760057

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk, k = 12:0.024837679346496666~0.04645613836591567

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk, k = 13:0.03768685628978186~0.06910938113726507

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk, k = 14:0.04719759931729035 0.08955720774117897

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk, k = 15:0.019518677205597842~0.027517275731672747

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk, k = 16:0.031106685796940845~0.06123668509236676

Prediction of Relative Errors and Mean Square Root of Relative Errors for Wk, k=17:0.011538907135197422~0.016462212374322608 Global Averages for Relative Errors, for Wk:0.01906563122848014

## 3 Reproducing the Results of the Baseline Model

To reproduce the results of the Baseline model, some functions of the file **Base-LineModel.py** must be used.

## 3.1 $D_k$ Predictions for each Corpus Size

To reproduce these results, we must run (example for English 2-grams) **AntModTestDkActualC('en', 2)**. The results will be:

 $\label{eq:decomposition} Dk\ average\ and\ square\ Errors\ for\ c, Lang,\ Ngram:\ 366397190\ en\ 2:0.3479429689129384\\ 0.35150103822720175$ 

 $Dk \ average \ and \ square \ Errors \ for \ c, Lang, \ Ngram: \ 11344756226 \ en \ 2:0.1845444617935846 \\ 0.1882913860994461$ 

 $\label{eq:def:Dk} Dk\ average\ and\ square\ Errors\ for\ c, Lang,\ Ngram:\ 31487751849\ en\ 2:0.12648210955272876\ 0.1298042564802903$ 

 $\label{eq:decomposition} Dk\ average\ and\ square\ Errors\ for\ c, Lang,\ Ngram:\ 82718285912\ en\ 2:0.06189720790184233\ 0.08666983670089452$ 

 $Global\ Error: 0.18021668704027355$ 

## 3.2 $D_k$ Predictions for each k Value

These results will be given by function AntModTestDkActualKbyK('en', 1) (example for English 1-grams):

Dk average and square Errors for k, Lang, Ngram: 1 en 1: 0.8743386248218455 0.9280331731987299

Dk average and square Errors for k, Lang, Ngram: 2 en 1:0.7197735260330508 0.7355192031933593

 $\label{eq:def:Dk} Dk \ average \ and \ square \ Errors \ for \ k, Lang, \ Ngram: \ 3 \ en \ 1:0.677761189943357 \\ 0.6922814990487511$ 

 $\label{eq:decomposition} Dk \ average \ and \ square \ Errors \ for \ k, Lang, \ Ngram: \ 4 \ en \ 1:0.6405244316106208 \\ 0.6560912454585672$ 

 $\label{eq:decomposition} Dk \ average \ and \ square \ Errors \ for \ k, Lang, \ Ngram: 5 \ en \ 1:0.6126767963340082\\ 0.628579381322395$ 

Dk average and square Errors for k, Lang, Ngram: 6 en 1:0.5789157863016462 0.5943161024109519

 $\begin{tabular}{ll} $Dk$ average and square Errors for $k$, $Lang$, $Ngram$: 7 en 1: 0.5660245284792901$\\ 0.5812936360351759 \end{tabular}$ 

 $\label{eq:def:Dk} Dk \ average \ and \ square \ Errors \ for \ k, Lang, \ Ngram: \ 8 \ en \ 1:0.541280214691845\\ 0.5561677491848851$ 

Dk average and square Errors for k, Lang, Ngram: 9 en 1:0.5298732155588947 0.5438293138686668

Dk average and square Errors for k, Lang, Ngram: 10 en 1:0.5214961551405822 0.5357427230197834

Dk average and square Errors for k, Lang, Ngram: 11 en 1 : 0.5135993440667627 0.5278455160189464

Dk average and square Errors for k, Lang, Ngram: 12 en 1: 0.5012997893199591 0.5154477738561042

 $\label{eq:decomposition} Dk \ average \ and \ square \ Errors \ for \ k, Lang, \ Ngram: \ 13 \ en \ 1:0.4946363988982175\\ 0.5087963524025215$ 

 $\begin{tabular}{ll} Dk \ average \ and \ square \ Errors \ for \ k, Lang, \ Ngram: 14 \ en \ 1:0.4830666633430951\\ 0.49712056229538365 \end{tabular}$ 

 $\begin{tabular}{ll} $Dk$ average and square Errors for $k$, $Lang$, $Ngram$: 15 en 1: 0.4753733659590578 \\ 0.4895635307797699 \end{tabular}$ 

 $\label{eq:decomposition} Dk\ average\ and\ square\ Errors\ for\ k, Lang,\ Ngram:\ 16\ en\ 1:0.4660570506715273\\ 0.48021497098449234$ 

Global Error 0.57479356757336

#### 3.3 $W_k$ Predictions for each Corpus Size

The results for these predictions are obtained by function **AntModTestWkActualC('en', 4)** (example for English 4-grams). We will see:

 $\label{eq:wk} Wk\ average\ and\ square\ Errors\ for\ c, Lang,\ Ngram:\ 366397188\ en\ 4:0.45457538064158510.4581471864432782$ 

Wk average and square Errors for c, Lang, Ngram: 11344756212 en 4:0.40459252213688470.40777075747775654

Wk average and square Errors for c, Lang, Ngram: 31487751813 en 4:0.3965766410106921 0.40271421459657564

 $Wk\ average\ and\ square\ Errors\ for\ c, Lang,\ Ngram:\ 82718285820\ en\ 4:0.37375428196749866\\0.38146687989443634$ 

Global Error: 0.4073747064391651

## 3.4 $W_k$ Predictions for each k Value

For this case, function **AntModTestWkActualKbyK('en', 3)** must be run (example for English 3-grams):

 $\label{eq:wk} Wk\ average\ and\ square\ Errors\ for\ k, Lang,\ Ngram:\ 1\ en\ 3:0.1382600224269316\\ 0.1902472636337762$ 

 $Wk\ average\ and\ square\ Errors\ for\ k, Lang,\ Ngram:\ 2\ en\ 3:0.34537791251963024\ 0.3508103865959475$ 

 $\label{eq:wk} Wk\ average\ and\ square\ Errors\ for\ k, Lang,\ Ngram:\ 3\ en\ 3:0.3418870326112877\ 0.3475676970818502$ 

 $Wk\ average\ and\ square\ Errors\ for\ k, Lang,\ Ngram:\ 4\ en\ 3:0.35771802276845643\ 0.3641902048361738$ 

Wk average and square Errors for k, Lang, Ngram: 5 en 3 : 0.34627529987485350.3572764485904226

Wk average and square Errors for k, Lang, Ngram: 6 en 3: 0.37926604292987576 0.39026615743448617

 $Wk\ average\ and\ square\ Errors\ for\ k, Lang,\ Ngram:\ 7\ en\ 3:0.3462688999028242\ 0.3597628116842053$ 

 $Wk\ average\ and\ square\ Errors\ for\ k, Lang,\ Ngram:\ 8\ en\ 3:0.37740542837408864\ 0.3908476225002481$ 

Wk average and square Errors for k,Lang, Ngram: 9 en 3: 0.3531057187931252 0.36728618720761624

 $Wk \ average \ and \ square \ Errors \ for \ k, Lang, \ Ngram: 10 \ en \ 3:0.35827945200475820.37359439938138944$ 

 $Wk \ average \ and \ square \ Errors \ for \ k, Lang, \ Ngram: 11 \ en \ 3:0.334741795689396270.3484183770127716$ 

Wk average and square Errors for k, Lang, Ngram: 12 en 3:0.353226474099714970.364475569232437

 $Wk\ average\ and\ square\ Errors\ for\ k, Lang,\ Ngram:\ 13\ en\ 3:0.33295945209532085\ 0.34348130203108995$ 

Wk average and square Errors for k, Lang, Ngram: 14 en 3:0.3362720252764011 0.3507500373652505

Wk average and square Errors for k, Lang, Ngram: 15 en 3:0.327158749053925060.3388055126595562

Global Error 0.33521348856137273

## 4 Reproducing the Results for the Model Using Constant Parameters and Cross-Validation

As for the case of the proposed model in Section 2, results can be obtained by using the training phase, which is already available, or by training this model before the test phase.

For training the model, function FindParametersCrossVal('en', 2, 1.985e10, 2.00e10, 1e8, 0.4, 1.7, 400, 0.4, 1.7, 200, "ConstParms") (example for English 2-grams) from file Def\_ValidationGHofC\_Phase.py must be run. This is the same function as the one used in Section 2, so the meaning of the parameters are the same. Although, an additional parameter ("ConstParms") indicates a different model. The following results will be seen:

It may take some minutes

k\_thresold for training corpora: 17

Processing for the hypothetical vocabulary size: 19850000000 Processing for the hypothetical vocabulary size: 19950000000

Best Vocabulary Size: 19950000000

#### 4.1 $D_k$ Predictions for each Corpus Size

To obtain results for these predictions, function AvgDkRelErrorForEachT-estCorpusWithScale('en', 6, "ConstParms") (example for English 6-grams) from file Def\_TestingGHofC\_Phase.py is used. The results will be:

Prediction of Relative Errors and Mean Square Root of Dk Relative Errors for C=366397186:0.5470557736344195~0.7215757455003925

Prediction of Relative Errors and Mean Square Root of Dk Relative Errors for C=11344756198:0.06257949916140662~0.07249043107739996

Prediction of Relative Errors and Mean Square Root of Dk Relative Errors for C=31487751777:0.04731543884581135~0.0581867394836642

Prediction of Relative Errors and Mean Square Root of Dk Relative Errors for C=82718285728:0.013735208140885436~0.018957521096334837 Global Averages for Relative Errors, for C:0.16767147994563072

#### 4.2 $D_k$ Predictions for each k Value

For this case, function **EachDkRelErrorTroughTestCorporaWithScale('de', 5, "ConstParms")** (example for German 5-grams) must be called. Results will be:

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 1:0.06319805793682381~0.10405157721696671

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 2:0.005648711233361225\ 0.009821252460839189$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 3: 0.03674619459448325 0.06447605713897969

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 4 : 0.06282833195922559 0.1087022226471926$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 5: 0.0699172847664049 0.12178375077375486$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 6: 0.0780673370477064 0.13690189952688064$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 7: 0.0801316005338459 0.14159154554663253$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 8: 0.08539460377031968 0.15099329265283332

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 9: 0.08790645374002296 0.15661610612417642

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 10:0.09226138990755157~0.1645535068245945

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 11:0.09332739273300093~0.16744458121640882

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 12:0.0911111564703024~0.16294845156391152$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 13:0.09115990143855714~0.16436931356674866

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 14:0.089102233819684~0.16172553411509744$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 15:0.08838401794705207~0.15973252999412205

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 16:0.0854504108395549~0.1564568380716739

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 32:0.06808877653474203~0.13216717167706035$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 64 : 0.10139610651165959 \ 0.18827009953977752$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 128: 0.12182394304210789 0.19675882812783702

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 256:0.12857949130845694~0.19010648621965315$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk, k >= 512:0.2390661680930702~0.4271998592496243

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 1024 : 0.16096777666416442 0.2948498827365511$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for Dk,  $k \ge 2048 : 0.4429876756216375 \ 0.8774280399110012$ 

Prediction of Relative Errors and Mean Square Root of Relative Errors for  $Dk,\ k>=4096:0.7152260237005563\ 1.3498023552781868$ 

Global Averages for Relative Errors for Dk: 0.13244879334226214

## 5 Reproducing the Figures

Figure 1 of the paper, can be generated by running load "GerFig1k1.txt" in gnuplot context. File GerFig1k1.txt uses files: powerlawVD1\_k1.txt, powerlawVD1\_k2.txt, strait k1.txt and strait k2.txt for that.

Figure 2a can be generated by running load "GerFigDk.txt" in gnuplot context. File GerFigDk.txt uses files:  $EmmD1\_2grams.txt$ ,  $EmmD1\_3grams.txt$ ,  $EmmD2\_2grams.txt$ ,  $EmmD2\_3grams.txt$ ,  $PredD1\_2grams.txt$ ,  $PredD1\_3grams.txt$ ,  $PredD2\_2grams.txt$ ,  $PredD2\_2grams.txt$ ,  $PredD2\_2grams.txt$  for that.

Figure 2b can be obtained by running load "GerFigWkbyC.txt" in gnuplot context. File GerFigWkbyC.txt uses files: GerFigWkbyC.txt, EmpWk11Gw.txt, EmpWk172Gw.txt, EmpWk31Gw.txt, EmpWk