

word2vec Demo

Efficient Estimation of Word Representations in Vector Space (Mikolov et al, 2013).

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Agenda

1 Installation

2 Demos

- Word vectors
- Word analogies
- From words to phrases

3 Additional notes

- Pre-trained models
- Other implementations

4 Optional

- Word and phrase accuracy
- Word classification

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word2vec source code

- <https://code.google.com/p/word2vec/>.
- Provides an efficient implementation of the continuous bag-of-words and skip-gram.
- Clean and well documented code in C.

word2vec source code

```

and@and-dblab: /opt/word2vec
and@and-dblab: /opt/word2vec 96x21
and@and-dblab: /opt/word2vec$ svn checkout http://word2vec.googlecode.com/svn/trunk/
A trunk/word2phrase.c
A trunk/LICENSE
A trunk/word-analogy.c
A trunk/compute-accuracy.c
A trunk/demo-analogy.sh
A trunk/demo-classes.sh
A trunk/demo-train-big-model-v1.sh
A trunk/demo-word-accuracy.sh
A trunk/demo-phrases.sh
A trunk/questions-words.txt
A trunk/demo-phrase-accuracy.sh
A trunk/demo-word.sh
A trunk/distance.c
A trunk/README.txt
A trunk/questions-phrases.txt
A trunk/word2vec.c
A trunk/makefile
Checked out revision 42.
and@and-dblab: /opt/word2vec$

```

word2vec source code

```
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk 132x24
and@and-dblab:~/Documents/Projects/C++/word2vec/trunk$ more makefile
CC = gcc
#Using -Ofast instead of -O3 might result in faster code, but is supported only by newer GCC versions
CFLAGS = -lm -pthread -O3 -march=native -Wall -funroll-loops -Wno-unused-result

all: word2vec word2phrase distance word-analogy compute-accuracy

word2vec: word2vec.c
$(CC) word2vec.c -o word2vec $(CFLAGS)
word2phrase: word2phrase.c
$(CC) word2phrase.c -o word2phrase $(CFLAGS)
distance: distance.c
$(CC) distance.c -o distance $(CFLAGS)
word-analogy: word-analogy.c
$(CC) word-analogy.c -o word-analogy $(CFLAGS)
compute-accuracy: compute-accuracy.c
$(CC) compute-accuracy.c -o compute-accuracy $(CFLAGS)
chmod +x *.sh

clean:
rm -rf word2vec word2phrase distance word-analogy compute-accuracy
and@and-dblab:~/Documents/Projects/C++/word2vec/trunk$
```

word2vec source code

```

and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk 123x46
and@and-dblab:~/Documents/Projects/C++/word2vec/trunk$ ./word2vec
WORD VECTOR estimation toolkit v 0.1c

Options:
Parameters for training:
-train <file>
    Use text data from <file> to train the model
-output <file>
    Use <file> to save the resulting word vectors / word clusters
-word-size <int>
    Set size of word vectors; default is 100
-window <int>
    Set max skip length between words; default is 5
-word-sample <float>
    Set threshold for occurrence of words. Those that appear with higher frequency in the training data
    will be randomly down-sampled; default is 1e-3, useful range is (0, 1e-5)
-negative <int>
    Use Hierarchical Softmax; default is 0 (not used)
    Number of negative examples; default is 5, common values are 3 - 10 (0 = not used)
-threads <int>
    Use <int> threads (default 12)
-iter <int>
    Run more training iterations (default 5)
-min-count <int>
    This will discard words that appear less than <int> times; default is 5
-alpha <float>
    Set the starting learning rate; default is 0.025 for skip-gram and 0.05 for CBOW
-classes <int>
    Output word classes rather than word vectors; default number of classes is 0 (vectors are written)
-debug <int>
    Set the debug mode (default = 2 = more info during training)
-binary <int>
    Save the resulting vectors in binary moded; default is 0 (off)
-save-vocab <file>
    The vocabulary will be saved to <file>
-read-vocab <file>
    The vocabulary will be read from <file>, not constructed from the training data
-cbow <int>
    Use the continuous bag of words model; default is 1 (use 0 for skip-gram model)

Examples:
./word2vec -train data.txt -output vec.txt -size 200 -window 5 -sample 1e-4 -negative 5 -hs 0 -binary 0 -cbow 1 -iter 3
and@and-dblab:~/Documents/Projects/C++/word2vec/trunk$

```

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demo-word.sh

```
## Compile the code...
make
## Download and unzip the training file...
if [ ! -e text8 ]; then
    wget http://mattmahoney.net/dc/text8.zip -O text8.gz
    gzip -d text8.gz -f
fi
## Run the model (taking time)...
time ./word2vec -train text8 -output vectors.bin -cbow 1 -size 200 -window 8 -negative 25 -hs 0
    ↪ -sample 1e-4 -threads 20 -binary 1 -iter 15
## Query word distances...
./distance vectors.bin
```

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demo-word.sh output

```
and@and-dblab:~/Documents/Projects/C++/word2vec/trunk$ ./demo-word.sh
make: Nothing to be done for 'all'.
--2015-11-12 18:18:10-- http://mattmahoney.net/dc/text8.zip
Resolving mattmahoney.net (mattmahoney.net)... 98.139.135.129
Connecting to mattmahoney.net (mattmahoney.net)|98.139.135.129|:80... connected.
HTTP request sent, awaiting response... 200 OK
Length: 31344016 (30M) [application/zip]
Saving to: text8.gz

text8.gz
  ↪ 100%[=====>] 29.89M 1.74MB/s in 18s

2015-11-12 18:18:28 (1.70 MB/s) - text8.gz saved [31344016/31344016]

Starting training using file text8
Vocab size: 71291
Words in train file: 16718843
Alpha: 0.000005 Progress: 100.10% Words/thread/sec: 113.47k
real 10m15.450s
user 36m52.552s
sys 0m4.388s
Enter word or sentence (EXIT to break):
```

demo-00.sh

```
## Get a small file...
head -c 5000000 text8 > text8_small
## Build the model...
./word2vec -train text8_small -output vectors_small.bin -cbow 1 -size 100 -window 5 -negative 0 -hs
↪ 25 -threads 1 -iter 4 -min-count 2 -binary 1
## Query word distances...
./distance vectors_small.bin
```

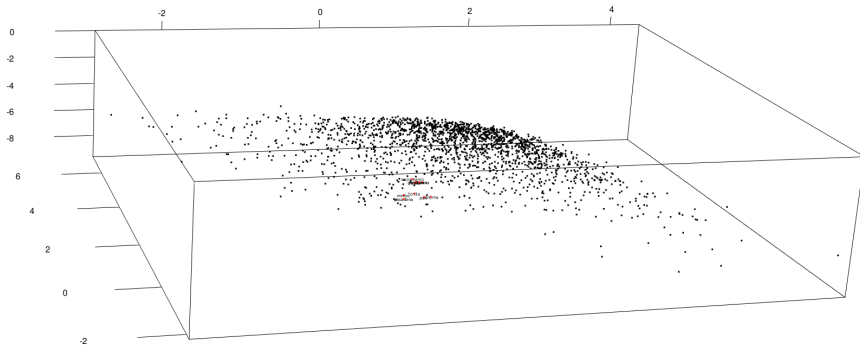
demo-01.sh

```
## Text model saving vocabulary
./word2vec -train text8_small -output vectors_small_50.txt -cbow 1 -size 50 -window 5 -negative 0
↳ -hs 25 -threads 1 -iter 4 -binary 0 -save-vocab vocab.txt
## Text model with just 3 dimensions
./word2vec -train text8_small -output vectors_small_3.txt -cbow 1 -size 3 -window 5 -negative 0 -hs
↳ 25 -threads 1 -iter 4 -binary 0
## See the results...
echo "Text model size 50..."
head -n 5 vectors_small_50.txt
echo "Vocabulary..."
head -n 5 vocab.txt
echo "Text model size 3..."
head -n 5 vectors_small_3.txt
```

demo-word.sh revisited

- **distance** can load a pre-trained model...
- Let's try some examples...
 - 1 california
 - 2 sciences
 - 3 happiness
 - 4 man
 - 5 ...

— + ×



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- **Word analogies**
- From words to phrases

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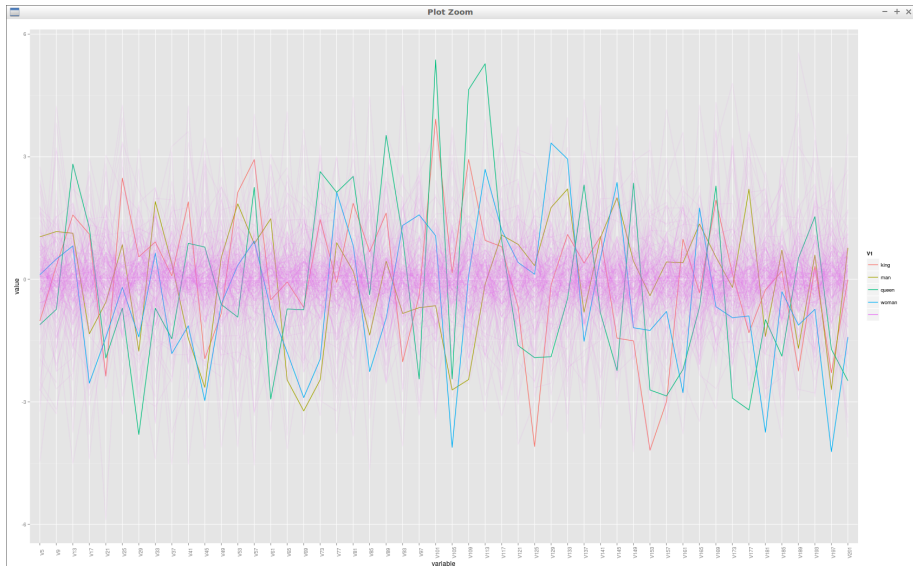
4 Optional

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Interesting properties of the word vectors

- $\overrightarrow{\text{paris}} - \overrightarrow{\text{france}} + \overrightarrow{\text{italy}} \cong \overrightarrow{\text{rome}}$
- $\overrightarrow{\text{king}} - \overrightarrow{\text{man}} + \overrightarrow{\text{women}} \cong \overrightarrow{\text{queen}}$

Parcoord plot



demo-analogy.sh

```
## Same that before...
make
if [ ! -e text8 ]; then
    wget http://mattmahoney.net/dc/text8.zip -O text8.gz
    gzip -d text8.gz -f
fi
echo -----
echo Note that for the word analogy to perform well, the model
echo should be trained on much larger data set
echo Example input: paris france berlin
echo -----
time ./word2vec -train text8 -output vectors.bin -cbow 1 -size 200 -window 8 -negative 25 -hs 0
    ↪ -sample 1e-4 -threads 20 -binary 1 -iter 15
## Call word-analogy script...
./word-analogy vectors.bin
```

demo-analogy.sh

- Some examples...

- 1 paris france bogota ...
- 2 king man queen ...
- 3 boy girl brother ...
- 4 chicago illinois memphis ...
- 5 poland zloty sweden ...
- 6 bad worst good ...
- 7 child children mouse ...
- 8 going went selling ...
- 9 mexico mexican peru ...
- 10 berlin germany riyadh¹ ...
- 11 woman angel man ...
- 12 heaven hell man ...

¹word2phrase will address the problem...

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From words to phrases and beyond

- It is desirable to have only one vector for representing 'los_angeles'.
- How to get vector representation of larger pieces of text no just words?
- **word2phrase**
- Pre-processing the training data set to form phrases.

demo-phrases.sh

```

and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk 132x24
and@and-dblab:~/Documents/Projects/C++/word2vec/trunk$ ./word2phrase
WORD2PHRASE tool v0.1a

Options:
Parameters for training:
  -train <file>          Use text data from <file> to train the model
  -output <file>         Use <file> to save the resulting word vectors / word clusters / phrases
  -min-count <int>       This will discard words that appear less than <int> times; default is 5
  -threshold <float>     The <float> value represents threshold for forming the phrases (higher means less phrases); default 100
  -debug <int>           Set the debug mode (default = 2 = more info during training)
Examples:
./word2phrase -train text.txt -output phrases.txt -threshold 100 -debug 2
and@and-dblab:~/Documents/Projects/C++/word2vec/trunk$

```


demo-phrases.sh

```
## Compile...
make
## Download...
if [ ! -e news.2012.en.shuffled ]; then
    wget http://www.statmt.org/wmt14/training-monolingual-news-crawl/news.2012.en.shuffled.gz
    gzip -d news.2012.en.shuffled.gz
fi
## Pre-process...
sed -e "s/'/'/g" -e "s/'/'/g" -e "s/'/'/g" < news.2012.en.shuffled | tr -c "A-Za-z'_ \n" " " >
    ↪ news.2012.en.shuffled-norm0
time ./word2phrase -train news.2012.en.shuffled-norm0 -output news.2012.en.shuffled-norm0-phrase0
    ↪ -threshold 200 -debug 2
time ./word2phrase -train news.2012.en.shuffled-norm0-phrase0 -output
    ↪ news.2012.en.shuffled-norm0-phrase1 -threshold 100 -debug 2
tr A-Z a-z < news.2012.en.shuffled-norm0-phrase1 > news.2012.en.shuffled-norm1-phrase1
## Model...
time ./word2vec -train news.2012.en.shuffled-norm1-phrase1 -output vectors-phrase.bin -cbow 1 -size
    ↪ 200 -window 10 -negative 25 -hs 0 -sample 1e-5 -threads 20 -binary 1 -iter 15
## Deploy...
./distance vectors-phrase.bin
```

demo-phrases.sh

```

and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk 120x22
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk$ head -n 12 news.2012.en.shuffled-norm0
Prang's initial success came from small prints and cards including the first Christmas cards but with the arrival of
the war he began to issue maps.
This single mining project is one of the main reasons for the amazing economic growth in the country said Dale Ch
oi an analyst at Origo Partners a private equity company that advises investors on China and Mongolia
On Dec six days before Christmas Jocelyn Earnest's close friend Marcy Shepherd who had been texting with h
er all day became concerned when Jocelyn never responded to messages she sent that evening
It is a phenomenon affecting the whole industry
British male solo artist Ed Sheeran
On Sunday morning the clean up crew jumped at the sight of a scurrying rodent shouting rat as it scrambled under t
he roof.
If you accidentally slide past the base you don't get called out of the baseline he said adding that qualified in th
is case
Damage to both ships is being evaluated with both ships currently operating under their own power
Next year it will also replace its Douro River ship in Portugal which stops in Porto and the wine growing town of Pinh
o with the motor passenger Queen Isabel
He tried to cross the street
The Labour MP Clive Betts will captain the other side
They are very stable they have stuck with the same starting lineup and don't expect many changes except for the injured
striker Helder Postiga
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk$
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk$

```

demo-phrases.sh

```

and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk 120x22
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk$ head -n 12 news.2012.en.shuffled-norm0-phrase0
Prang's initial success came from small prints and cards including the first Christmas cards but with the arrival of th
e war he began to issue maps Calendar R Web Google Translate Ph.D. Misc Riverside Apartm Riverside Studen
This single mining project is one of the main reasons for the amazing economic growth in the country said Dale Choi an
analyst at Origo Partners a private equity company that advises investors on China and Mongolia
On Dec six days before Christmas Jocelyn Earnest's close friend Marcy Shepherd who had been texting with her all day be
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British male solo artist Ed Sheeran
On Sunday morning the clean up crew jumped at the sight of a scurrying rodent shouting rat as it scrambled under their
feet
If you accidentally slide past the base you don't get called out of the baseline he said adding that qualified in this
case
Damage to both ships is being evaluated with both ships currently operating under their own power
Next year it will also replace its Douro River ship in Portugal which stops in Porto and the wine growing town of Pinh
o with the passenger Queen Isabel
He tried to cross the street
The Labour MP Clive Betts will captain the other side
They are very stable they have stuck with the same starting lineup and don't except any change except for the injured s
triker Helder Postiga
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk$
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk$

```

demo-phrases.sh

```

and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk 120x22
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk$ head -n 12 news.2012.en.shuffled-norm0-phrase1
Prang's initial success came from small prints and cards including the first Christmas cards but with the arrival of th
e war he began to issue maps Calendar R Web Google Translate Ph.D. Misc Riverside Apartm Riverside Studen
This single mining project is one of the main reasons for the amazing economic growth in the country said Dale Choi an
analyst at Origo Partners a private equity company that advises investors on China and Mongolia
On Dec six days before Christmas Jocelyn Earnest's close friend Marcy Shepherd who had been texting with her all day be
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It is a phenomenon affecting the whole industry
British male solo artist's Sheeran
On Sunday morning the clean up crew jumped at the sight of a scurrying rodent shouting rat as it scrambled under their
feet
If you accidentally slide past the base you don't get called out of the baseline he said adding that qualified in this
case
Damage to both ships is being evaluated with both ships currently operating under their own power
Next year it will also replace its Douro River ship in Portugal which stops in Porto and the wine growing town of Pinh
o with the passenger Queen Isabel
He tried to cross the street
The Labour MP Clive Betts will captain the other side
They are very stable they have stuck with the same starting lineup and don't except any change except for the injured s
triker Helder Postiga
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk$
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk$

```

demo-phrases.sh output

```
...
Starting training using file news.2012.en.shuffled-norm1-phrase1
Vocab size: 681320
Words in train file: 283545447
Alpha: 0.000005 Progress: 100.00% Words/thread/sec: 162.97k
real 115m6.531s
user 434m57.904s
sys 1m4.464s
Enter word or sentence (EXIT to break):
```

demo-phrases.sh

```

and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk 87x31
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk$ ./distance_vectors-phase.bin
Enter word or sentence (EXIT to break): restaurants los angeles
Word: restaurants Position in vocabulary: 2444
Word: los_angeles Position in vocabulary: 878
-----
• the training algorithm: hierarchical softmax (better for infrequent words) vs
  dim: Words vector Cosine distance
  -----
  • eateries 0.651376
  • downtown 0.645015
  • santa_monica 0.641607
  • upscale_restaurants 0.633591
  • san_francisco 0.632140
  • southern_california 0.624869
  • upscale 0.613263
  • restaurant 0.609788
  • downtown_los_angeles 0.609320
  • bay_area 0.608609
  • retail_shops 0.607322
  • new_york 0.604622
  • beverly_hills 0.599124
  • seafood_restaurants 0.592477
  • new_york_city 0.590868
  • osteria_mozza 0.589431
  • hotels 0.576938
  • pizzeria_mozza 0.574118
  • pizzerias 0.572582
Enter word or sentence (EXIT to break):

```

demo-phrases.sh

```

and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk 93x22
and@and-dblab:~/Documents/Projects/C++/word2vec/trunks$ ./word-analogy_vectors-phrase.bin
Enter three words (EXIT to break): berlin germany riyadh

Word: berlin Position in vocabulary: 3207
Word: germany Position in vocabulary: 932
Word: riyadh Position in vocabulary: 16906

----- Word Distance -----
Additional notes 31      saudi_arabia 0.761608
Pre-trained mod... 32      saudi      0.612114
Other Implemen... 33      saudi_arabia's 0.605735
Optional 34      united_arab_emirates 0.595843
Word and phras... 35      uae 0.594940
Word classificati... 36      gulf_states 0.583258
                        saudis 0.581073
                        qatar 0.549350
                        oman 0.548390
                        saudi_arabia_qatar 0.528623
Enter three words (EXIT to break):

```

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demo-train-big-model-v1.sh

```
#####  
#  
# Script for training good word and phrase vector model using public corpora, version 1.0.  
# The training time will be from several hours to about a day.  
#  
# Downloads about 8 billion words, makes phrases using two runs of word2phrase, trains  
# a 500-dimensional vector model and evaluates it on word and phrase analogy tasks.  
#  
#####
```

- GoogleNews-vectors-negative300.bin.gz
 - 100 billion words
 - 300 dimensional vectors
 - 1.6 GB
- freebase-vectors-skipgram1000.bin.gz
 - 100 billion words
 - 1000 dimensional vectors
 - 2.5 GB
- Some tips about performance and where obtain more training data can be found in <https://code.google.com/p/word2vec/>.

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Ports of the framework...

- in Python: gensim²
- in Java: deeplearning4j³
- in R: tmcn⁴
- in Scala: Apache Spark⁵

²<http://rare-technologies.com/deep-learning-with-word2vec-and-gensim/>

³<http://deeplearning4j.org/word2vec.html>

⁴<http://rpackages.ianhowson.com/rforge/tmcn.word2vec/man/word2vec.html>

⁵<https://spark.apache.org/docs/latest/mllib-feature-extraction.html#word2vec>

Thank you!!!

Download this presentation at <https://goo.gl/pE8Mjw>

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demo-word-accuracy.sh

```
## Same than before...
make
if [ ! -e text8 ]; then
  wget http://mattmahoney.net/dc/text8.zip -O text8.gz
  gzip -d text8.gz -f
fi
time ./word2vec -train text8 -output vectors.bin -cbow 1 -size 200 -window 8 -negative 25 -hs 0
  ↪ -sample 1e-4 -threads 20 -binary 1 -iter 15
## Test accuracy...
./compute-accuracy vectors.bin 30000 < questions-words.txt
## threshold is used to reduce vocabulary of the model for fast approximate evaluation
## to compute accuracy with the full vocabulary, use:
## ./compute-accuracy vectors.bin < questions-words.txt
```

questions-words.txt

• <https://word2vec.googlecode.com/svn/trunk/questions-words.txt>

```

questions-words.txt (-/Documents/Projects/C++/word2vec/trunk) - gedit
File Edit View Search Tools Documents Help
Open Save Undo Redo Find
questions-words.txt x
capital-common countries
Athens Greece Baghdad Iraq
Athens Greece Bangkok Thailand
Athens Greece Beijing China
Athens Greece Berlin Germany
Athens Greece Bern Switzerland
Athens Greece Cairo Egypt
Athens Greece Canberra Australia
Athens Greece Hanoi Vietnam
Athens Greece Havana Cuba
Athens Greece Helsinki Finland
Athens Greece Islamabad Pakistan
Athens Greece Kabul Afghanistan
Athens Greece London England
Athens Greece Madrid Spain
Athens Greece Moscow Russia
Athens Greece Oslo Norway
Athens Greece Ottawa Canada
Athens Greece Paris France
Athens Greece Rome Italy
Athens Greece Stockholm Sweden
Athens Greece Tehran Iran
Athens Greece Tokyo Japan
Baghdad Iraq Bangkok Thailand
Baghdad Iraq Beijing China
Baghdad Iraq Berlin Germany
Baghdad Iraq Bern Switzerland
Baghdad Iraq Cairo Egypt
Baghdad Iraq Canberra Australia
Baghdad Iraq Hanoi Vietnam
Baghdad Iraq Havana Cuba
Baghdad Iraq Helsinki Finland
Baghdad Iraq Islamabad Pakistan
Baghdad Iraq Kabul Afghanistan
Baghdad Iraq London England
Baghdad Iraq Madrid Spain
Baghdad Iraq Moscow Russia
Baghdad Iraq Oslo Norway
Baghdad Iraq Ottawa Canada
Baghdad Iraq Paris France
Baghdad Iraq Rome Italy
Baghdad Iraq Stockholm Sweden
Baghdad Iraq Tehran Iran
Baghdad Iraq Tokyo Japan
Baghdad Iraq Athens Greece
Bangkok Thailand Beijing China
Bangkok Thailand Berlin Germany
Bangkok Thailand Bern Switzerland
Plain Text Tab Width: 8 Ln 1, Col 1 INS

```

questions-phrases.txt

https://word2vec.googlecode.com/svn/trunk/questions-phrases.txt

```

questions-phrases.txt (-/Documents/Projects/C++/word2vec/trunk) - gedit
File Edit View Search Tools Documents Help
[Icons] Open Save Undo Redo Find
questions-phrases.txt *
:
newspapers
Albuquerque Albuquerque Journal Baltimore Baltimore Sun
Albuquerque Albuquerque Journal Boston Boston Globe
Albuquerque Albuquerque Journal Cincinnati Cincinnati Enquirer
Albuquerque Albuquerque Journal Cleveland Cleveland Plain Dealer
Albuquerque Albuquerque Journal Charleston Charleston Gazette
Albuquerque Albuquerque Journal Chicago Chicago Tribune
Albuquerque Albuquerque Journal Columbus Columbus Dispatch
Albuquerque Albuquerque Journal Dallas Dallas Morning News
Albuquerque Albuquerque Journal Dayton Dayton Daily News
Albuquerque Albuquerque Journal Denver Denver Post
Albuquerque Albuquerque Journal Dothan Dothan Eagle
Albuquerque Albuquerque Journal Fort Collins Fort Collins Coloradoan
Albuquerque Albuquerque Journal Fresno Fresno Bee
Albuquerque Albuquerque Journal Houston Houston Chronicle
Albuquerque Albuquerque Journal Indianapolis Indianapolis Star
Albuquerque Albuquerque Journal Knoxville Knoxville News Sentinel
Albuquerque Albuquerque Journal Los Angeles Los Angeles Times
Albuquerque Albuquerque Journal Miami Miami Herald
Albuquerque Albuquerque Journal Milwaukee Milwaukee Journal Sentinel
Albuquerque Albuquerque Journal Minneapolis Minneapolis Star Tribune
Albuquerque Albuquerque Journal New Haven New Haven Register
Albuquerque Albuquerque Journal New York New York Times
Albuquerque Albuquerque Journal Oakland Oakland Tribune
Albuquerque Albuquerque Journal Philadelphia Philadelphia Inquirer
Albuquerque Albuquerque Journal Portland Portland Oregonian
Albuquerque Albuquerque Journal Sacramento Sacramento Bee
Albuquerque Albuquerque Journal Salt Lake Salt Lake Tribune
Albuquerque Albuquerque Journal San Antonio San Antonio Express News
Albuquerque Albuquerque Journal San Francisco San Francisco Chronicle
Albuquerque Albuquerque Journal San Jose San Jose Mercury News
Albuquerque Albuquerque Journal Seattle Seattle Times
Albuquerque Albuquerque Journal Tallahassee Tallahassee Democrat
Albuquerque Albuquerque Journal Waco Waco Tribune Herald
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Baltimore Baltimore Sun Cleveland Cleveland Plain Dealer
Baltimore Baltimore Sun Charleston Charleston Gazette
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Baltimore Baltimore Sun Columbus Columbus Dispatch
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Plain Text Tab Width: 8 Ln 1, Col 1 INS

```

demo-word-accuracy.sh output

```

and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk$ ./demo-word-accuracy.sh
capital-common-countries:
ACCURACY TOP1: 83.00 % (420 / 506)
Total accuracy: 83.00 % Semantic accuracy: 83.00 % Syntactic accuracy: -nan %
capital-world:
ACCURACY TOP1: 62.67 % (910 / 1452)
Total accuracy: 67.93 % Semantic accuracy: 67.93 % Syntactic accuracy: -nan %
currency:
ACCURACY TOP1: 20.90 % (56 / 268)
Total accuracy: 62.26 % Semantic accuracy: 62.26 % Syntactic accuracy: -nan %
city-in-state:
ACCURACY TOP1: 49.01 % (770 / 1571)
Total accuracy: 56.78 % Semantic accuracy: 56.78 % Syntactic accuracy: -nan %
family:
ACCURACY TOP1: 77.78 % (238 / 306)
Total accuracy: 58.35 % Semantic accuracy: 58.35 % Syntactic accuracy: -nan %
gram1-adjective-to-adverb:
ACCURACY TOP1: 18.25 % (138 / 756)
Total accuracy: 52.11 % Semantic accuracy: 58.35 % Syntactic accuracy: 18.25 %
gram2-opposite:
ACCURACY TOP1: 23.53 % (72 / 306)
Total accuracy: 50.42 % Semantic accuracy: 58.35 % Syntactic accuracy: 19.77 %
gram3-comparative:
ACCURACY TOP1: 62.22 % (784 / 1260)
Total accuracy: 52.73 % Semantic accuracy: 58.35 % Syntactic accuracy: 42.81 %
gram4-superlative:
ACCURACY TOP1: 38.93 % (197 / 506)
Total accuracy: 51.72 % Semantic accuracy: 58.35 % Syntactic accuracy: 42.11 %
gram5-present-participle:
ACCURACY TOP1: 39.31 % (390 / 992)
Total accuracy: 50.17 % Semantic accuracy: 58.35 % Syntactic accuracy: 41.39 %
gram6-nationality-adjective:
ACCURACY TOP1: 86.29 % (1183 / 1371)
Total accuracy: 55.50 % Semantic accuracy: 58.35 % Syntactic accuracy: 53.25 %
gram7-past-tense:
ACCURACY TOP1: 38.21 % (509 / 1332)
Total accuracy: 53.33 % Semantic accuracy: 58.35 % Syntactic accuracy: 50.18 %
gram8-plural:
ACCURACY TOP1: 63.71 % (632 / 992)
Total accuracy: 54.22 % Semantic accuracy: 58.35 % Syntactic accuracy: 51.96 %
gram9-plural-verbs:
ACCURACY TOP1: 34.77 % (226 / 650)
Total accuracy: 53.19 % Semantic accuracy: 58.35 % Syntactic accuracy: 50.59 %
Questions seen / total: 12268 19544 62.77 %
and@and-dblab: ~/Documents/Projects/C++/word2vec/trunk$

```

Agenda

1 Installation

2 Demos

- Word vectors
- Word analogies
- From words to phrases

3 Additional notes

- Pre-trained models
- Other implementations

4 Optional

- Word and phrase accuracy
- Word classification

demo-classes.sh

```
## Same than before...
make
if [ ! -e text8 ]; then
    wget http://mattmahoney.net/dc/text8.zip -O text8.gz
    gzip -d text8.gz -f
fi
## Train the model with classes rather than vectors...
time ./word2vec -train text8 -output classes.txt -cbow 1 -size 200 -window 8 -negative 25 -hs 0
    ↪ -sample 1e-4 -threads 20 -iter 15 -classes 500
## Sort the result by the second column...
sort classes.txt -k 2 -n > classes.sorted.txt
echo The word classes were saved to file classes.sorted.txt
```

demo-classes.sh

```
## Let's build a small model...  
./word2vec -train text8_small -output classes.txt -cbow 1 -size 50 -window 5 -negative 0 -hs  
  ↪ 12 -sample 1e-4 -threads 20 -iter 3 -classes 10  
sort classes.txt -k 2 -n > classes.sorted.txt
```