

CIS530 HW5 Submission Example

2 SimLex-999 Dataset Revisited (10 points)

Note **5 Extra points** will be awarded for creativity and a more thorough qualitative analysis.)

2.1 What is the least similar 2 pairs of words based on human judgement scores and vector similarity? Do the pairs match? [3 points]

Results

- least similar 2 pairs based on human judgement scores
 1. (new, ancient)
 2. (shrink, grow)
- least similar 2 pairs based on vector similarity
 1. (house, key)
 2. (flower, endurance)
- Do the pairs match?
 - No

2.2 What is the most similar 2 pairs of words based on human judgement scores and vector similarity? Do the pairs match? [3 points]

Results

- most similar 2 pairs based on human judgement scores
 3. (vanish, disappear)
 4. (quick, rapid)
 - 5.
- most similar 2 pairs based on vector similarity
 3. (south, north)
 4. (north, west)
- Do the pairs match?
 - no

2.3 Provide correlation scores and p values for the following models:

How do those correlation value compare to each other? [4 points]

Results

- glove.6B.50d.magnitude
 - Correlation = 0.18100126067449063,
 - P Value = 1.2242211264976856e-17
- glove.6B.100d.magnitude
 - Correlation = 0.20506409092608713,
 - P Value = 3.41228663395174e-22
- glove.6B.200d.magnitude
 - Correlation = 0.23670323199262908,
 - P Value = 4.9936324557833286e-29
- glove.6B.300d.magnitude
 - Correlation = 0.25894302181101986,
 - P Value = 2.080389068003349e-34
- glove.840B.300d.magnitude
 - Correlation = 0.2860664813618063,
 - P Value = 1.293335613361039e-41
- How do those correlation value compare to each other?
 - When the dimensions of the magnitude increases, the p-value goes down significantly and the correlation increases. Which means the similarity is getting closer to human's judgement when the dimensions increases.

3.1 Cluster Randomly (1 points)

3.1 Run clustering on dev data, report the f_scores from the dev data [1 point]

- Dev data

Target	k	Paired F-Score
smell.v	4	0.2857
image.n	9	0.1204
express.v	7	0.1562
talk.v	6	0.2428
play.v	34	0.0308
miss.v	8	0.1972
produce.v	7	0.1968
write.v	9	0.1507
provide.v	7	0.2204

party.n	5	0.2017
bank.n	9	0.0741
plan.n	3	0.3693
shelter.n	5	0.2500
difference.n	5	0.2564
eat.v	6	0.2186
mean.v	6	0.1772
treat.v	8	0.1610
use.v	6	0.2407
suspend.v	6	0.1304
judgment.n	7	0.1480
organization.n	7	0.1734
interest.n	5	0.1704
paper.n	7	0.2044
operate.v	7	0.1237
receive.v	13	0.0903
watch.v	5	0.2041
rule.v	7	0.1267
simple.a	5	0.2000
atmosphere.n	6	0.1812
expect.v	6	0.2105
different.a	1	1.0000
begin.v	8	0.0897
note.v	3	0.3684
win.v	4	0.2756
source.n	9	0.1260
performance.n	5	0.2347
wash.v	13	0.0849
hear.v	5	0.1835
climb.v	6	0.1963
degree.n	7	0.1985

=> Average Paired F-Score: 0.1592

3.2 Cluster with Sparse Representations (6 points)

3.2.1 Run clustering on dev data, report the f_scores from the dev data [1 point]

- Dev data

Target	k	Paired F-Score
judgment.n	7	0.1989
produce.v	7	0.3147
bank.n	9	0.3373
expect.v	6	0.3011
suspend.v	6	0.2222
receive.v	13	0.2028
paper.n	7	0.3892
organization.n	7	0.2775

degree.n	7	0.2947
atmosphere.n	6	0.2291
plan.n	3	0.5102
win.v	4	0.4304
simple.a	5	0.1818
performance.n	5	0.2713
mean.v	6	0.3375
different.a	1	1.0000
begin.v	8	0.2694
hear.v	5	0.3368
wash.v	13	0.1026
smell.v	4	0.3301
miss.v	8	0.2182
express.v	7	0.2504
eat.v	6	0.2938
treat.v	8	0.2101
play.v	34	0.0884
talk.v	6	0.3462
note.v	3	0.5333
shelter.n	5	0.3636
write.v	9	0.1742
source.n	9	0.2374
watch.v	5	0.3717
interest.n	5	0.1792
provide.v	7	0.3820
image.n	9	0.2037
use.v	6	0.3366
rule.v	7	0.1811
operate.v	7	0.2400
climb.v	6	0.3655
party.n	5	0.2419
difference.n	5	0.4000

=> Average Paired F-Score: 0.2529

3.2.2 Provide a brief description of your method in the report, making sure to describe the vector space model you chose, the clustering algorithm you used, and the results of any preliminary experiments you might have run on the dev set. [5 points]

- Brief Description of your method
 - [Description / Background]

I have used a different clustering algorithm which is Agglomerative Clustering. The method is the same as Cluster with Sparse Representation. The cooccurrence matrix is the same, but instead of using K-Means algorithm, I switched to use Agglomerative Clustering. Agglomerative Clustering starts the cluster from each point and K-Means assign k random centroids at the beginning. That's the difference. Agglomerative clustering is slower than K-means but it's more stable.

- [the vector space model you chose]

The vector space I used is the same as Sparse Representation which is “coocvec-500mostfreq-window-3.filter.magnitude”

- [the clustering algorithm you used]

I have used a different clustering algorithm which is Agglomerative Clustering, which is a type of hierarchical clustering algorithm. It supports 4 different linkage methods

Linkage	Explanation of distance between clusters
ward	Minimize total within-cluster variance
complete	Max of distances between points in each cluster
average	Average of distances between all pairs
single	Min of distances between points

- [the results of any preliminary experiments]

Linkage	Overall F-score	
ward	0.2389	
complete	0.2832	
average	0.3430	
single	0.3599	Using “single” achieved highest f-score

Using “single” achieves best performance

The F-Score from Single in Agglomerative Clustering:

Target	k	Paired F-Score
degree.n	7	0.4666
atmosphere.n	6	0.3595
play.v	34	0.1712
smell.v	4	0.4839
talk.v	6	0.6142
plan.n	3	0.6387
party.n	5	0.3572
express.v	7	0.4345
win.v	4	0.4813
paper.n	7	0.4570
rule.v	7	0.2880
produce.v	7	0.4385
note.v	3	0.6400
performance.n	5	0.4427
bank.n	9	0.2429
miss.v	8	0.3511
operate.v	7	0.2844
write.v	9	0.3516
source.n	9	0.3378
mean.v	6	0.4361
expect.v	6	0.4205
difference.n	5	0.4778
image.n	9	0.2885
shelter.n	5	0.4244
use.v	6	0.6223
wash.v	13	0.1963
interest.n	5	0.3121
organization.n	7	0.3958

treat.v	8	0.4033
begin.v	8	0.3688
climb.v	6	0.3375
different.a	1	1.0000
provide.v	7	0.7063
suspend.v	6	0.2830
simple.a	5	0.2564
hear.v	5	0.3234
eat.v	6	0.4495
judgment.n	7	0.3309
receive.v	13	0.2289
watch.v	5	0.4400

=> Average Paired F-Score: 0.3599

3.3 Cluster with Dense Representations (8 points)

3.3.1 Run clustering on dev data, report the f_scores from the dev data [1 point]

- Dev data

Target	k	Paired F-Score
judgment.n	7	0.2498
produce.v	7	0.3720
bank.n	9	0.5937
expect.v	6	0.4500
suspend.v	6	0.4151
receive.v	13	0.1486
paper.n	7	0.5574
organization.n	7	0.3622
degree.n	7	0.4295
atmosphere.n	6	0.3171
plan.n	3	0.6387
win.v	4	0.4936
simple.a	5	0.2857
performance.n	5	0.4221
mean.v	6	0.2952
different.a	1	1.0000
begin.v	8	0.3933
hear.v	5	0.2595
wash.v	13	0.1672
smell.v	4	0.3117
miss.v	8	0.1795
express.v	7	0.4044
eat.v	6	0.4226
treat.v	8	0.3188
play.v	34	0.1568
talk.v	6	0.6120
note.v	3	0.8400
shelter.n	5	0.4574
write.v	9	0.3152

source.n	9	0.2659
watch.v	5	0.3140
interest.n	5	0.3440
provide.v	7	0.6249
image.n	9	0.3051
use.v	6	0.4022
rule.v	7	0.2888
operate.v	7	0.2835
climb.v	6	0.2811
party.n	5	0.3479
difference.n	5	0.4794

=> Average Paired F-Score: 0.3401

3.3.2 Provide a brief description of your method in the report that includes the vectors you used, and any experimental results you have from running your model on the dev set. [5 points]

- Brief Description of your method
 - [Description / Background]

I used the Agglomerative Clustering Method, for each word's paraphrases, I create k lists. K is the number of clusters. For each paraphrase, I get the embedding from the vector space model I use. If the paraphrase doesn't exist in the vector space, I randomly select a vector. I generated a matrix X by doing this. I created the model with K as the number of clusters. Then I fit the X with the model. After getting the labels through fitting, I reconstruct K paraphrase lists based on the model labels.

- [the vector space model you chose]

Used GoogleNews-vectors-negative300.magnitude

- [the clustering algorithm you used]

Used Agglomerative Clustering cluster algorithm with specified k

I have used a different clustering algorithm which is Agglomerative Clustering, which is a type of hierarchical clustering algorithm. It supports 4 different linkage methods

Linkage	Explanation of distance between clusters
ward	Minimize total within-cluster variance
complete	Max of distances between points in each cluster
average	Average of distances between all pairs
single	Min of distances between points

- [the results of any preliminary experiments]

Linkage	Overall F-score	
ward	0.2915	
complete	0.3034	
average	0.3704	
single	0.3798	Using "single" achieved highest f-score

Target	k	Paired F-Score
judgment.n	7	0.2521
produce.v	7	0.4331
bank.n	9	0.6154
expect.v	6	0.4675
suspend.v	6	0.3023
receive.v	13	0.2016
paper.n	7	0.6070
organization.n	7	0.3906
degree.n	7	0.4019
atmosphere.n	6	0.4491
plan.n	3	0.6432
win.v	4	0.4892
simple.a	5	0.2759
performance.n	5	0.4370
mean.v	6	0.3646
different.a	1	1.0000
begin.v	8	0.3382
hear.v	5	0.3241
wash.v	13	0.2886
smell.v	4	0.4677
miss.v	8	0.3217
express.v	7	0.4253
eat.v	6	0.4584
treat.v	8	0.3906
play.v	34	0.1453
talk.v	6	0.6358
note.v	3	0.8400
shelter.n	5	0.5304
write.v	9	0.3484
source.n	9	0.3106
watch.v	5	0.4700
interest.n	5	0.3376
provide.v	7	0.7403
image.n	9	0.3347
use.v	6	0.6252
rule.v	7	0.3228
operate.v	7	0.2977
climb.v	6	0.3412
party.n	5	0.3807
difference.n	5	0.4788

=> Average Paired F-Score: 0.3798

The f-score is 0.3798. It's more than the f-score from the sparse embedding experiment. Which indicate more density helps to improve the correlation to human's results.

3.3.3 In addition, for Task 3.2 and 3.3, do an analysis of different errors made by each system – i.e. look at instances that the word-context matrix representation gets wrong and dense gets right, and vice versa, and see if there are any interesting patterns. There is no right for this. [2 points]

- To compare, make a dataframe **Target word | F score_ sparse | F score_ dense | Difference**

K-Means on both sparse and dense embeddings

Target Word	k	F Score Sparse	F Score Dense	Difference
write.v	9	0.2205	0.1633	0.0572
treat.v	8	0.2505	0.2658	-0.0153
note.v	3	0.5333	0.619	-0.0857
paper.n	7	0.3783	0.3784	-1E-04
talk.v	6	0.3507	0.2812	0.0695
wash.v	13	0.12	0.1789	-0.0589
source.n	9	0.1743	0.3111	-0.1368
degree.n	7	0.3279	0.3531	-0.0252
operate.v	7	0.231	0.2491	-0.0181
bank.n	9	0.3373	0.7333	-0.396
smell.v	4	0.2947	0.35	-0.0553
use.v	6	0.4184	0.2926	0.1258
provide.v	7	0.3312	0.2495	0.0817
mean.v	6	0.3804	0.3145	0.0659
hear.v	5	0.3368	0.2712	0.0656
suspend.v	6	0.1905	0.386	-0.1955
party.n	5	0.2762	0.4209	-0.1447
watch.v	5	0.3361	0.4964	-0.1603
climb.v	6	0.35	0.2798	0.0702
image.n	9	0.2062	0.3038	-0.0976

performance.n	5	0.2544	0.3798	-0.1254
organization.n	7	0.2523	0.2224	0.0299
eat.v	6	0.2574	0.3386	-0.0812
expect.v	6	0.3654	0.3512	0.0142
different.a	1	1	1	0
shelter.n	5	0.3129	0.3805	-0.0676
simple.a	5	0.1538	0.2857	-0.1319
play.v	34	0.0903	0.1218	-0.0315
receive.v	13	0.2018	0.1628	0.039
interest.n	5	0.2145	0.3395	-0.125
begin.v	8	0.2022	0.358	-0.1558
atmosphere.n	6	0.3489	0.3123	0.0366
difference.n	5	0.3088	0.4235	-0.1147
plan.n	3	0.572	0.4451	0.1269
rule.v	7	0.2077	0.2846	-0.0769
miss.v	8	0.25	0.2703	-0.0203
produce.v	7	0.2721	0.2705	0.0016
express.v	7	0.3106	0.2698	0.0408
win.v	4	0.4497	0.4737	-0.024
judgment.n	7	0.2245	0.3219	-0.0974

- Now sort based on Difference to identify words where one works better than other
 - [Target word | F score_ sparse | F score_ dense | Difference table]
 - [Error analysis]

When the word has K as 1, there's no difference.

When the word has a very large K, for example “play” has k=34, it’s f-score is very low.

Overall, the F-Score Dense performs better than F-Score Sparse.

Target Word	k	F Score Sparse	F Score Dense	Difference	Comments

				=F Score Dense – F Score Sparse	
bank.n	9	0.3373	0.7333	0.396	This work has highest f-score under densed embedding
suspend.v	6	0.1905	0.386	0.1955	
watch.v	5	0.3361	0.4964	0.1603	
begin.v	8	0.2022	0.358	0.1558	
party.n	5	0.2762	0.4209	0.1447	
source.n	9	0.1743	0.3111	0.1368	
simple.a	5	0.1538	0.2857	0.1319	
performance.n	5	0.2544	0.3798	0.1254	
interest.n	5	0.2145	0.3395	0.125	
difference.n	5	0.3088	0.4235	0.1147	
image.n	9	0.2062	0.3038	0.0976	
judgment.n	7	0.2245	0.3219	0.0974	
note.v	3	0.5333	0.619	0.0857	
eat.v	6	0.2574	0.3386	0.0812	
rule.v	7	0.2077	0.2846	0.0769	
shelter.n	5	0.3129	0.3805	0.0676	
wash.v	13	0.12	0.1789	0.0589	
smell.v	4	0.2947	0.35	0.0553	
play.v	34	0.0903	0.1218	0.0315	
degree.n	7	0.3279	0.3531	0.0252	
win.v	4	0.4497	0.4737	0.024	
miss.v	8	0.25	0.2703	0.0203	
operate.v	7	0.231	0.2491	0.0181	
treat.v	8	0.2505	0.2658	0.0153	
paper.n	7	0.3783	0.3784	1E-04	

different.a	1	1	1	0	
produce.v	7	0.2721	0.2705	-0.0016	
expect.v	6	0.3654	0.3512	-0.0142	
organization.n	7	0.2523	0.2224	-0.0299	
atmosphere.n	6	0.3489	0.3123	-0.0366	
receive.v	13	0.2018	0.1628	-0.039	
express.v	7	0.3106	0.2698	-0.0408	
write.v	9	0.2205	0.1633	-0.0572	
hear.v	5	0.3368	0.2712	-0.0656	
mean.v	6	0.3804	0.3145	-0.0659	
talk.v	6	0.3507	0.2812	-0.0695	
climb.v	6	0.35	0.2798	-0.0702	
provide.v	7	0.3312	0.2495	-0.0817	
use.v	6	0.4184	0.2926	-0.1258	
plan.n	3	0.572	0.4451	-0.1269	

3.4 Cluster without K (6 points)

3.4.1 Run clustering on dev data, report the f_scores from the dev data [1 point]

- Dev data

Target	k	Paired F-Score
degree.n	7	0.4460
atmosphere.n	6	0.3825
play.v	34	0.1753
smell.v	4	0.5033
talk.v	6	0.6414
plan.n	3	0.6745
party.n	5	0.3739
express.v	7	0.4256
win.v	4	0.5445
paper.n	7	0.6022
rule.v	7	0.3317
produce.v	7	0.4968
note.v	3	0.7719

performance.n	5	0.4761
bank.n	9	0.5688
miss.v	8	0.3058
operate.v	7	0.3060
write.v	9	0.3567
source.n	9	0.3109
mean.v	6	0.4011
expect.v	6	0.4642
difference.n	5	0.5078
image.n	9	0.3086
shelter.n	5	0.5194
use.v	6	0.6658
wash.v	13	0.2579
interest.n	5	0.3383
organization.n	7	0.3955
treat.v	8	0.4046
begin.v	8	0.3389
climb.v	6	0.3646
different.a	1	1.0000
provide.v	7	0.7470
suspend.v	6	0.4878
simple.a	5	0.3889
hear.v	5	0.3351
eat.v	6	0.4729
judgment.n	7	0.3298
receive.v	13	0.2256
watch.v	5	0.4646

=> Average Paired F-Score: 0.3949

3.4.2 Provide a brief description of your method in the report that includes the vectors you used, and any experimental results you have from running your model on the dev set. [5 points]

- Brief Description of your method
 - [Description / Background]

For each target word's paraphrases, I run the Agglomerative Clustering method with different k from 2 to the number of paraphrases. When the paraphrases contain less than 2 words, no need to do clustering. For each value for k, I record the silhouette score. I will pick K clusters which gives the highest silhouette score. Repeat this process for each word.

- [the vector space model you chose]

The vector I used is GoogleNews-vectors-negative300.magnitude

- [the clustering algorithm you used]

The clustering method used is Agglomerative Clustering. I used "single" in linkage because based on sparse and dense experiment, "single" always achieve the highest correlation.

- [the results of any preliminary experiments]

Target	k	Paired F-Score

degree.n	7	0.4460	
atmosphere.n	6	0.3825	
play.v	34	0.1753	
smell.v	4	0.5033	
talk.v	6	0.6414	
plan.n	3	0.6745	
party.n	5	0.3739	
express.v	7	0.4256	
win.v	4	0.5445	
paper.n	7	0.6022	
rule.v	7	0.3317	
produce.v	7	0.4968	
note.v	3	0.7719	
performance.n	5	0.4761	
bank.n	9	0.5688	
miss.v	8	0.3058	
operate.v	7	0.3060	
write.v	9	0.3567	
source.n	9	0.3109	
mean.v	6	0.4011	
expect.v	6	0.4642	
difference.n	5	0.5078	
image.n	9	0.3086	
shelter.n	5	0.5194	
use.v	6	0.6658	
wash.v	13	0.2579	
interest.n	5	0.3383	
organization.n	7	0.3955	
treat.v	8	0.4046	
begin.v	8	0.3389	
climb.v	6	0.3646	
different.a	1	1.0000	
provide.v	7	0.7470	
suspend.v	6	0.4878	
simple.a	5	0.3889	
hear.v	5	0.3351	
eat.v	6	0.4729	
judgment.n	7	0.3298	
receive.v	13	0.2256	
watch.v	5	0.4646	

=> Average Paired F-Score: 0.3949

The f-score is 0.3949. It's highest comparing to sparse representation and dense representation. Which means that each word might have a best k which can device the paraphrases.

Leaderboards [2 points + 3 bonus]

- No writing is required here, but be sure to submit a valid result (not -1) to Gradescope. [2 points]