# Introduction to Natural Language Processing Assignment 4

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Quote (1)	Code	en-word	Quote (2)	Code	en-word	Quote (3)	Code	en-word
do	Make %2:41:00	Make %2:41:00	how	-	-	it	Information_ technology %1:09:00	-
you	-	-	it	Information _technology %1:09:00	-	is	Be %2:42:03	Be %2:42:03
train	Train %2:31:01	Train %2:31:01	is	Be %2:42:03	Be %2:42:03	the	-	-
for	-	-	we	-	-	responsibility	Duty %1:04:00	Duty %1:04:00
passing	Pass %2:38:00	Pass %2:38:00	have	Have %2:40:00	Have %2:40:00	of	-	-
tests	Trial %1:09:00	Trial %1:09:00	so	-	So %4:02:02	intellectuals	Intellectual %1:18:00	Intellectual %1:18:00
or	-	-	much	Much %3:00:00	Much %3:00:00	to	-	-
do	Make %2:41:00	Make %2:41:00	information	Information %1:10:00	Information %1:10:00	speak	Talk %2:32:00	Talk %2:32:00
you	-	-	,	-	-	the	-	-
train	Train %2:31:01	Train %2:31:01	but	-	-	truth	-	Truth %1:09:00
for	-	-	know	-	Know %2:31:01	and	-	-
creative	Creative %3:00:00	Creative %5:00:00	so	-	So %4:02:02	expose	Expose %2:39:02	Expose %2:39:02
inquiry	Inquiry %1:09:01	Inquiry %1:09:01	little	Small I%3:00:00	Little %5:00:00	lies	-	Lie %1:10:00

We are considering like relevant only the words that retrieve a Sense key for the recall calculation

Quote	Total	Correct	Relevant	Correct	Accuracy	Recall	F1
1	13	7	8	7	7/13=0.5385	%=0.875	0.6667
2	13	4	9	4	4/13=0.3077	4/9=0.4444	0.3636
3	13	5	8	5	5/130.3846	%=0.625	0.4762

```
[Synset('information_technology.n.01'), Synset('be.v.01'), Synset('duty.n.01'), Synset('intellectual.n.01'), Synset('talk.v.02'), Synset('expose.v.01')]
[Lemma('information_technology.n.01.information_technology'), Lemma('be.v.01.be'), Lemma('duty.n.01.duty'), Lemma('intellectual.n.01.intellectual'), Lemma('talk.v.02.talk'), Lemma('expose.v.01.expose')]
['information_technology%1:09:00::', 'be%2:42:03::', 'duty%1:04:00::', 'intellectual%1:18:00::', 'talk%2:32:00::', 'expose%2:39:02::']
```

[Synset('information\_technology.n.01'), Synset('be.v.01'), Synset('have.v.01'), Synset('much.a.01'), Synset('information.n.01'), Synset('small.a.01')] [Lemma('information\_technology.n.01.information\_technology'), Lemma('be.v.01.be'), Lemma('have.v.01.have'), Lemma('much.a.01.much'), Lemma('information.n.01.information'), Lemma('small.a.01.small')] ['information\_technology%1:09:00::', 'be%2:42:03::', 'have%2:40:00::', 'much%3:00:00::', 'information%1:10:00::', 'small%3:00:00::']

[Synset('make.v.01'), Synset('train.v.01'), Synset('pass.v.01'), Synset('trial.n.02'), Synset('creative.a.01'), Synset('inquiry.n.01')] [Lemma('make.v.01.make'), Lemma('train.v.01.train'), Lemma('pass.v.01.pass'), Lemma('trial.n.02.trial'), Lemma('creative.a.01.creative'), Lemma('inquiry.n.01.inquiry')] ['make%2:41:00::', 'train%2:31:01::', 'pass%2:38:00::', 'trial%1:09:00::', 'creative%3:00:00::', 'inquiry%1:09:01::']

### Question # 2

#### Part a

	Does	Racing	Zelda	Fighting	Nintendo	Is	Red
Does	0	0	9	1	1	16	0
Racing	0	2	0	1	1	17	0
Zelda	9	0	32	0	10	112	0
Fighting	1	1	0	0	1	23	0
Nintendo	1	1	10	1	6	45	1
Is	16	17	112	23	45	430	11
Red	0	0	0	0	1	11	4

#### Part c

## (c1) For each given word, determine the three words which are closest in the embedding space, i.e. the three most similar words:

Word	PCA model	Wiki-50d
Playstation	'n64', 'tomb', 'raider'	'xbox', 'nintendo', 'gamecube'
Walk	'travel', 'collect', 'explore'	'walking', 'walks', 'walked'
Love	'enjoy', 'playing', 'got'	'dream', 'life', 'dreams'

#### (c2) For each of the given sets of words, let the models determine the outlier:

Set of words	Truth	PCA model	Wiki-50d	
(mario, zelda, kirby, microsoft)	Microsoft	mario	microsoft	
(january, july, december, year)	year	year	year	

#### (c3) Determine the word that completes the following analogies.

Set of words	Truth	PCA model	Wiki-50d	
son:daughter :: boy:?	girl	nearly	girl	
n64:nintendo :: playstation:?	sony	n64	xbox	
acceptable:unacceptable:: reasonable:?	unreasonable	mesmerized	acceptable	
good:better :: tough:?	tougher	anyone	hard	
go:going :: play:?	playing	own	playing	

### Task(d)

We have tried to improve our vector embeddings by using larger dimensions (100d) and changing some parameters in PCA like what we read in documentation of PCA that setting value of random\_state either 0 or 42 have optimally resulted in good results so this time we ran our code with value 0. We have repeated the evaluation and here are the results:

1. 3-words which are close to (playstation) in the embedding space word2vec10d are : ['n64', 'tomb', 'raider']

- 2. 3-words which are close to (walk) in the embedding space word2vec10d are : ['travel', 'collect', 'explore']
- 3. 3-words which are close to (love) in the embedding space word2vec10d are : ['enjoy', 'playing', 'got']
- 4. Outliner found with using w2v50d is: mario
- 5. Outliner found with using w2v50d is: year
- 6. With using w2v50d :- boy : nearly
- 7. With using w2v50d :- playstation : n64
- 8. With using w2v50d :- reasonable : mesmerized
- 9. With using w2v50d :- tough : anyone
- 10. With using w2v50d :- play : own