



Lecture 1: Introduction to Natural Language Processing

- **Course Introduction**
- Overview of Natural Language Processing (NLP)
 Assignment Project Exam Help
 Word Meaning and Representation
- Count-base the Repute ont gtion
 - One-hot Encoding
 - Bag of Wow eChat: cstutorcs
 - Term Frequency-Inverse Document Frequency
- Next Week Preview
 - Prediction-based Word Representation



Dr Caren Han

Education

- B CompScience (1st Class Honours)
- PhD Ansity Similariti Profeste Exam Help

Australian Young httipver teathing Excellence Award

Teacher of the Year 2020 Award

CSTUTOTCS

Machine Learning, Natural Language Processing, etc.

NLP Research Experience (8 years experience)

- U.S. Air Force: Robotics Natural Language Processing Interface
- U.S. Navy: Deep Learning with NLP
- **Hyundai Co.**: Car Navigation Dialog System
- NLP Top-tier Conferences Best Paper Award/Best Area Paper Award



COMP5046 Natural Language Processing

This unit introduces computational linguistics and the statistical techniques and algorithms used to automatically process natural languages. It will review the core statistics and information theory, and the basic linguistics, required to understand natural language processing (NLP).

NLP is used in Awide range of applitations, including information legisland extraction; question answering; machine translation; and classifying and clustering of documents. This unit will explore the key challenges of natural language to computational modelling, and the state-of-the-art approaches to the key NLP sub-tasks, including tokenisation, morphological analysis, word sense representation, part-of-speech tagging, named entity recognition and other information extraction.

Students will implement many of these sub-tasks in labs and assignments, that can be used in the real-world cases. The wit picture tigote the process that is central to creating training data for interesting application. With this unit, students can develop the innovative application that can be used in the real world.

Where to find the course information?

Unit Outline - COMP5046

https://www.sydney.edu.au/units/COMP5046

Canvas - COMP5046

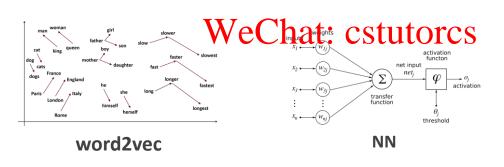
https://canvas.sydney.edu.au/courses/30912

What will you learn in this course?

The focus of this course is on the review and comparison of <u>models and methods that have achieved state-of-the-art results on various NLP tasks</u> such as question answering (QA) and machine translation.

In this comprehensive review to property of the past, present, and future of NLP. In addition, students will learn some of the current best practices for applying deep learning in NLP

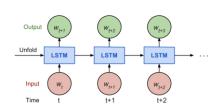
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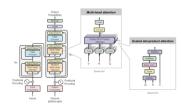


Dependency Parsing

Part-of-Speech Tagging



Language Modelling



Transformer

INTRODUCTION



What will you learn in this course?

Week 1: Introduction to Natural Language Processing (NLP)

Week 2: Word Embeddings (Word Vector for Meaning)

Week 3: Work Spijestin Week 3: Work Week 3:

Week 4: Word Classification with Machine Learning II

Week 5: Language Fundamental //tutorcs.com

Week 6: Part of Speech Tagging

Week 7: Dependency Parsing Chat: cstutorcs

Week 8: Language Model and Natural Language Generation

NLP Techniques

NLP and

Week 9: Information Extraction: Named Entity Recognition

Week 10: Advanced NLP: Attention and Reading Comprehension

Week 11: Advanced NLP: Transformer and Machine Translation

Week 12: Advanced NLP: Pretrained Model in NLP

Advanced Topic

Week 13: Future of NLP and Exam Review



I DO assume you can program

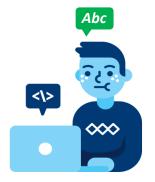
- By that, I mean you are a confident programmer
- · Labs will Any sing in represent The ject Exam Help
- Assessment will involve programming https://tutorcs.com
- Python recommended; other popular languages accepted
- There will be NO NON-programming option for assignments
- · But it's more than just programming:
 - algorithms, mathematics and (esp.) statistics
 - linguistics and intuition about language
 - analytical thinking





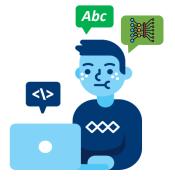
I DO <u>NOT</u> assume you are a linguist

- But you do need to know roughly how to identify a noun/verb/etc.
- · We will think critically about prove us et anguage Help



I DO NOT assume you are a deep learning researcher

- But you do need to know (really) roughly how machine learning works.
- · We will think critically how to use text deta and empeddings
- and about how deep learning models capture aspects of language (context)
 https://tutorcs.com



The NLP Big Picture



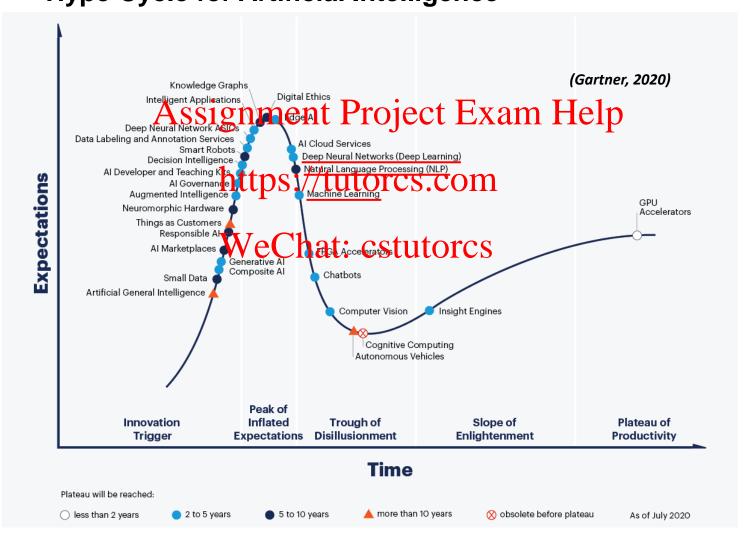
The purpose of Natural Language Processing: Overview

Understanding Searching Application **Generation** Translation Help Sentiment. Search Analysis Classification Dialog **Topic** https<u>://tut</u>orcs.com Modelling When Sebastian thrun Person started at Google ORG IN 2007 DATE **Entity Extraction Parsing** Claudia sat on a stool **NLP Stack PoS Tagging** She sells seashells [she/PRP] [sells/VBZ] [seashells/NNS] **Stemming** Drinking, Drank, Drunk Drink **Tokenisation** How is the weather today [How] [is] [the] [weather] [today]

The NLP ERA



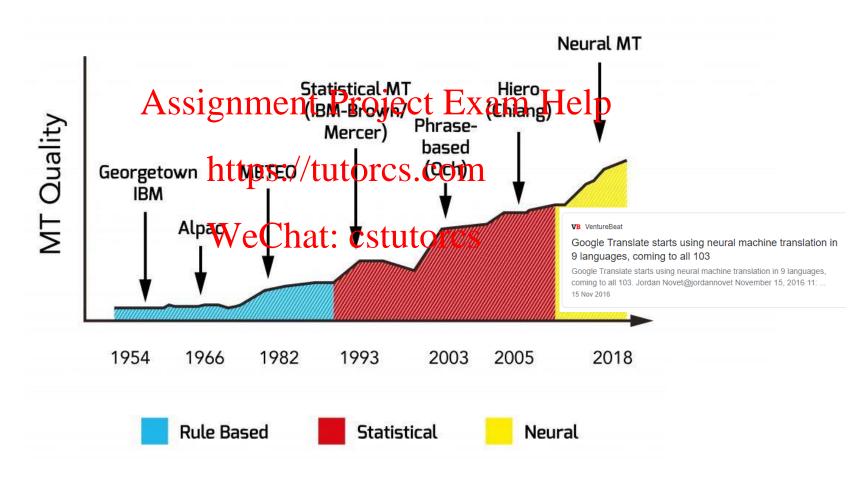
Hype Cycle for Artificial Intelligence



The NLP ERA



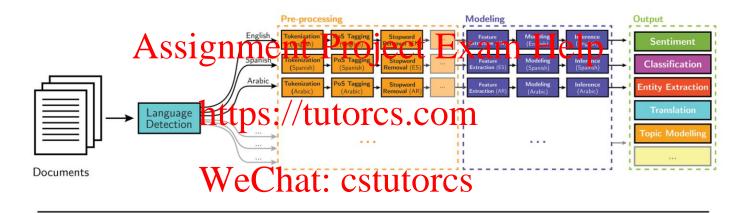
NLP Techniques – with the Trend of Machine Translation



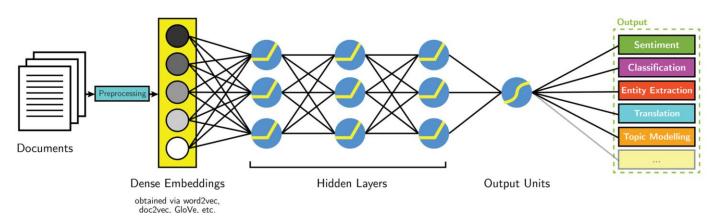


Language Modelling using Deep Learning-based NLP Techniques

Classical NLP



Deep Learning-based NLP





Assessment Overview

Assessment	Weight	Due
Lab Exercise	10%	Multiple Weeks
Assignment 1	20%	Week 8 (Friday 11.59pm – AU time)
Assignment 2 ASSIGN	ngent Proje	Week 14 (Friday 11.59pm – AU time)
Final Exam	50%	Exam Period

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Lab Exercises

• Programming tasks done in weekly computer tasks

Assignments

- Take place through the teaching period
- Implementation and Documentation

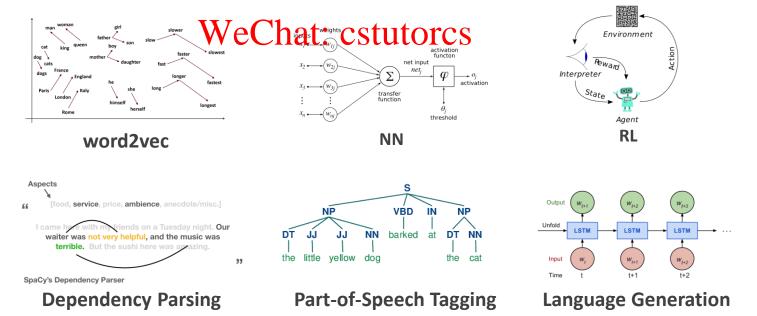
Penalties for lateness

• 10% of the available marks per day late; maximum 7 days late



Lab Exercise – 10%

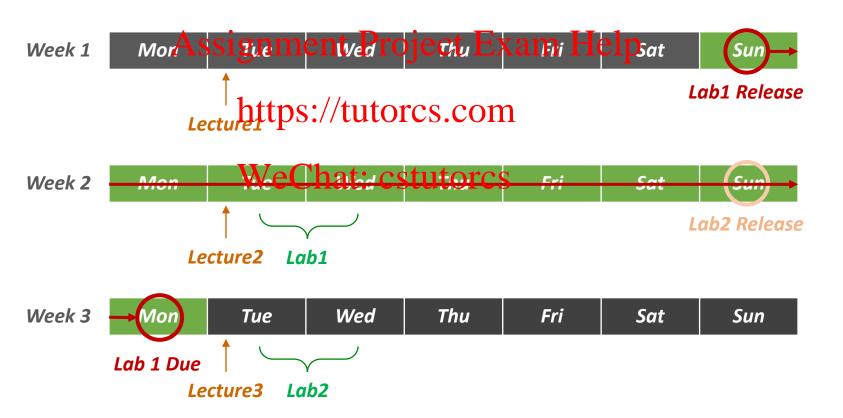
- In the Lab, students need to do the small tasks (1% for each week).
- · 2-3 tasks Ares gizen the ead to Pwhat you learned in the previous lecture.
- You must have been assessed as having completed 10 out of 11 in order to get the 10% for weekly lab exercise.





Lab Exercise – 10%

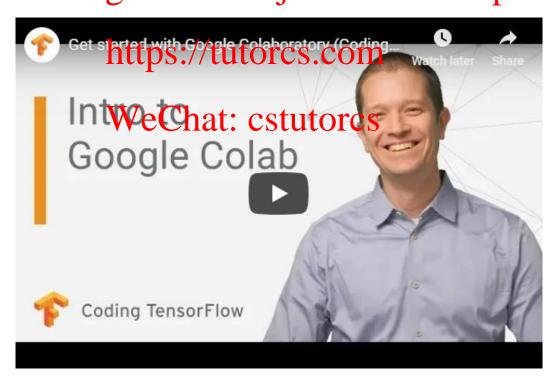
When to submit the Weekly Lab Exercise (e.g. Lab1 Release and Submission)



What do we do during Labs?

In Labs, Students will use Google CoLab

Colaboratory is a free Jupyter notebook environment that requires no setup and runs entirely in the cloud. With Colaboratory you can write and execute code, save and share your analyses, and coesis powerful porning resources, all for free from your browser.





Assignment 1 (20%)

The focus of this assignment 1 is implementing the text/word embeddings/language models, which 1)understands the language and 2)produces the detection/prediction/generation decision.

NOTE: Assignanting it be with the project of the man Help

Assignment 21(20%)//tutorcs.com

The focus of this assignment 2 is proposing a natural language model to produce high performance (such a paccurage, consistency plausibility, validity, and distribution) in different NLP tasks. The different NLP tasks can be Question Answering, Named Entity Recognition, Text Classification, or etc.

NOTE: Assignment 2 would be a group assignment (2 people in group). However, you can do individually only if you want.

Final Exam (50%)

The final exam will be a short take-home exam hosted on Canvas (3 hours duration). You will be asked to answer variety of theoretical questions.

The sample exam questions will be shared in the week 13 lecture.

ASSESSMENT



ASSESSMENT Due

Week 1: Introduction to Natural Language Processing (NLP)

Week 2: Word Embeddings (Word Vector for Meaning)

Week 3: Work Spijeginnite Inthi Project Exam Help

Week 4: Word Classification with Machine Learning II

Week 5: Language Fundamental //tutorcs.com

Week 6: Part of Speech Tagging Chat: cstutorcs

Week 8: Language Model and Natural Language Generation

Assignment 1 Due

Week 9: Information Extraction: Named Entity Recognition

Week 10: Advanced NLP: Attention and Reading Comprehension

Week 11: Advanced NLP: Transformer and Machine Translation

Week 12: Advanced NLP: Pretrained Model in NLP

Week 13: Future of NLP and Exam Review



Start assignments early!

- All assignments involve coding and report writing
- Reports Are the primary deliverable ct Exam Help
- Assignments will Be very different from last year's
- Reports will be submitted to Turnitin through Canvas
- Code is also submitted (for assignments 1 and 2) and retained
- We will use code plagiarism detection tools
- Clearly reference any copied/adapted code portions and cite their origins



Start assignments early!

- Starting early means you will work while you sleep
- · Don't wastesall your time Project Exam Help
- If you're stuck, ask early
- We might be able to offer you alternatives



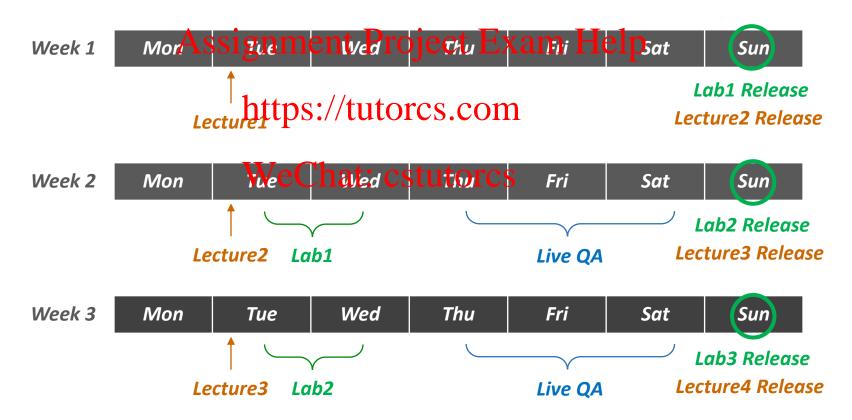
Working Hours

- Work 12 hours per week for this course (including 3 contact hours);
- · Attend 2 Acsosi granteen to Project Exam Help
 - Tuesday 3 5pm
 - https://tutorcs.com Lectures are recorded, but don't depend on it!
- Attend 1 hour of the tal transplaces turnercs
- Participate respectfully in discussions in lectures and labs;
- · Complete all assessment tasks on time.



Classes and Release Date

When the Class and Release Dates are





Full Course Timetable at course website:

- Lecture: Tue 3-5pm
- Tutorial ATSISTS frim Entry Projecto Exam Help
- LiveQA: Thu, Fri, Sat 9-10pm

https://tutorcs.com



Who are we?

Unit Coordinator & Lecturer

- Caren.Han@sydney.edu.au
- · No fixed consultation hour, please arrange a time to see me.

https://tutorcs.com

Teaching Assistants & Tutors WeChat: cstutorcs

• They will introduce themselves really soon!

For Qs related to course content, please post in Ed.

For Qs related to admin, please contact to the unit coordinator

Please put [COMP5046] in the title of the email

READINGS (OPTIONAL)



No Textbook Recommended, but if you really want to read some

Deng, L., & Liu, Y. (Eds.). (2018). Deep Learning in Natural Language Processing. Springer.

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Rao, D., & McMahan, B. (2019). Natural Language Processing with PyTorch: Build Intelligent Language Applications using Deep Learning. O'Reilly Media, Inc.".

Manning, C. D., Manning, C. D., & Schutze, H. (1998). Foundations of statistical natural language processing. MIT press.



Ask yourself...

- How much work will you be devoting to this unit, each week?
- · Who should so ignorialitic Piero jeter Exam Help
- When is the first assessment due? https://tutorcs.com
- What do you do if you get sick during semester?
- What is Turnitin? We Chat: cstutorcs
- What programming language do you need to know?

1

What do we learn? (NLP and Deep Learning)



WeChat: cstutorcs

"We have an incompetent administration, and if I am not elected President, that will not change over the next four years—but it must change, and it must change now."

(June 13th, 201

"We wanted border security. We wanted very, very strong military. We wanted all of the things that we're going to get..."

(CPAC Feb. 24th, 201,

We fed 270,000 words spoken by Trump into a computer program that studies language patterns.

"Now, they do charge you tariff on trucks, when we send trucks and other things over there." (Announcement June 16th, 201s

"I only want to admit people who share our values and love our people."

(June 22nd, 2016

https://www.youtube.com/watch?v=EFHyzuqjaok



Lecture 1: Introduction to Natural Language Processing

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- 4. Count-based Representation
 - One-hot Encoding

 - Bag of WweChat: cstutorcs
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- Next Week Preview



- language stores knowledge
- · language Agggigning the new Projecte Exam Help
- language is a key to culture and human experience https://tutorcs.com
- language is a natural interface for humans



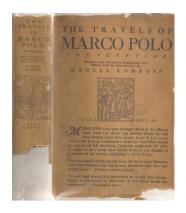


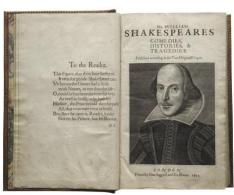
- language stores knowledge
- · language Aconing unique to the language aconing unique to th
- language is a key to culture and human experience https://tutorcs.com
- language is a natural interface for humans





- language stores knowledge
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- language stores knowledge
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Natural Language Processing (NLP)



What is Natural Language Processing?

Natural Language Processing (NLP) is the branch of artificial intelligence focused on developing systems that allow computers to communicate with people using everyday language

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Computational Linguistics

It concerns how computational methods can and the understanding of language

Communication WeChat: cstutorcs

The goal in the production / comprehension of language is communication.

Natural Language Processing (NLP)



Communication for the speaker:

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I'm speakingttps://tutorcs.com

- Intention: Decide when and what information should be transmitted (a.k.a. strategic generator) Cilian require planning and reasoning about agents' goals and beliefs.
- Generation: Translate the information to be communicated (in internal logical representation or "language of thought") into string of words in desired natural language (a.k.a. tactical generation).
- Synthesis: Output the string in desired modality, text or speech.

Natural Language Processing (NLP)



Communication for the hearer:



- Perception: Map input modality to a string of words, e.g. optical character recognition (OCR) or epoch recognition (CCR) or epo
- Analysis: Determine the information content of the string.
 - Syntactic interpretation (parsing): Find the correct parse tree showing the phrase structure of the string.
 - Semantic Interpretation: Extract the (literal) meaning of the string.
 - Pragmatic Interpretation: Consider effect of the overall context on altering the literal meaning of a sentence.
- Incorporation: Decide whether or not to believe the content of the string and add it to the Knowledge Base.



What is special about NLP?

Human language is a system specifically constructed to convey meaning and is **not produced by a physical manifestation of any kind**. In that way, it is very different from vision or any other machine learning task.

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 Most words are just symbols for an extra-linguistic entity: the word is a signifier that mabit to significant that mabit to significant the significant control of the significa

"Computer" We Chat: cstutores

"Whaaaaaaa"







Ambiguity

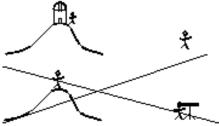
I saw the man on the hill with a telescope.

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https://tutorcs.com

WeChatt cstutores







Ambiguity is Explosive

Ambiguities compound to generate enormous numbers of possible interpretations.

In English, a sentence ending in n prepositional phrases has over 2n interpretations.

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"I saw the man with the telescope": 2 parses

"I saw the man on https://thtetores.com parses

"I saw the man on the hill in Texas with the telescope": 14 parses

"I saw the man on the hill in Texas with the telescope at noon.": 42 parses

"I saw the man on the hill in Texas with the telescope at noon on Monday" 132 parses



Ambiguity is Ubiquitous

Speech Recognition

- "recognize speech" vs. "wreck a nice beach"
- "youth in Assignaments Project Exam Help

Syntactic Analysiattps://tutorcs.com

• "I ate spaghetti with chopsticks" vs. "I ate spaghetti with meatballs." WeChat: cstutorcs

Semantic Analysis

- "The dog is in the pen." vs. "The ink is in the pen."
- "I put the plant in the window" vs. "Ford put the plant in Mexico"



Even human struggle with understanding

```
Assigmiss ryouexam Help doesp://tuequesh
"Let's get back together".
```

333333



The difficulty level in various NLP tasks

Easy

- Spell Checking
- · Keyword According nment Project Exam Help
- Finding Synonyms

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Medium

• Extracting Information from documents (including websites)

Difficult

- Semantic Analysis (What is the meaning of query statement?)
- Machine Translation
- Coreference Resolution
- Question Answering



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WORD REPRESENTATION



How to represent the meaning of the word?

Definition: meaning (Collins dictionary).

- the idea that it is presents and which can be explained using other words.
- the thoughts or ideas that are intended to be expressed by it.

https://tutorcs.com

signifier (symbol) signified (idea or thing)

"Apple"





WORD REPRESENTATION



How do we have usable meaning in a computer?

- Common solution: Use e.g. WordNet, a thesaurus containing lists of synonym sets and hypernyms ("is a" relationships).

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e.g. synonym sets containing "good":

e.g. hypernyms of "panda":

```
from nltk.corpus import when some poses = { 'n':'noun', 'v':'verb', 's':'adj (s)', from nltk.corpus import wordnet as wn
'a':'adj', 'r':'adv'}
                                                                   panda = wn.synset("panda.n.01")
for synset in wn.synsets("good");
                                                                   hyper = lambda s: s.hypernyms()
                                                                  Ois Canda. closure (hyper))
print("{}: {}".format(poses y y tet.pat], C
", ".join([l.name() for l in synset.lemmas()]
```

```
noun: good
noun: good, goodness
noun: good, goodness
noun: commodity, trade good, good
adj: good
adj (sat): full, good
adj: good
adj (sat): estimable, good, honorable, respectable
adj (sat): beneficial, good
adj (sat): good
adj (sat): good, just, upright
adverb: well, good
adverb: thoroughly, soundly, good
```

```
[Synset('procyonid.n.01'),
Synset('carnivore.n.01'),
Synset('placental.n.01'),
Synset('mammal.n.01'),
Synset('vertebrate.n.01'),
Synset('chordate.n.01'),
Synset('animal.n.01'),
Synset('organism.n.01'),
Synset('living thing.n.01'),
Synset('whole.n.02'),
Synset('object.n.01'),
Synset('physical entity.n.01'),
Synset('entity.n.01')]
```

WORD REPRESENTATION



Problems with resources like WordNet

- Great as a resource but missing nuance
 - e.g. "proficient" is listed as a synonym for "good". This is only correct in some contexts.
 - e.g. "glad" can be synonym for "fXXX off"???

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- Missing new meanings of words

 - e.g., wicked, badass, nifty, wizard, genius, ninia, bombast
 Impossible to keep up-to-date!
- WeChat: cstutorcs Subjective
- Requires human labor to create and adapt
- Can't compute accurate word similarity



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One-hot Encoding

In traditional NLP, we regard words as discrete symbols.

Hot (True) Cold (False)

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Words can be represented by **one-hot vectors**:

- 1. The categorical haltips be intemports intemporal values (index)
- 2. each integer value is represented as a binary vector that is all zero values except the indexwithe integer, which is marked with a 1.

$$\begin{array}{c} & & & & & & & & \\ & & & & & & \\ & & & & & \\ & & & & \\ & & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ &$$

Vector dimension = number of words in vocabulary



Sun, Mar 10 4

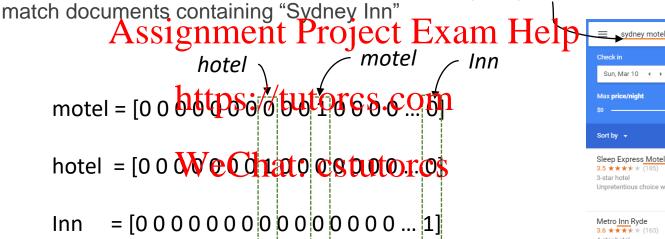
Mon, Mar 11

2 ▼

Problem with one-hot vectors

Problem #1. No word similarity representation

Example: in web search, if user searches for "Sydney motel", we would like to



There is no natural notion of similarity for one-hot vectors!

Problem #2. Inefficiency

Vector dimension = number of words in vocabulary

Each representation has only a single '1' with all remaining 0s.



Bag of Words (BOW)

- A bag-of-words model (BoW) is a representation of text that describes the occurrence of words within a document. It involves two things:
 - A vocabulary of known words.
 - · Amasignmente Project Ewam Help
- It is called a "happ" of words in the document is discarded. The model is only concerned with whether known words occur in the document, not where in the document.

 WeChat: cstutorcs





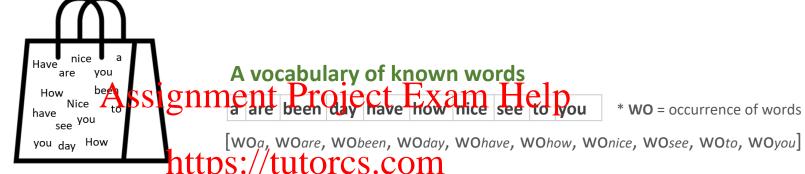
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Bag of Words (BOW)



How are you = [0, 1, 0, 0, 0, 1, 0, 0, 0, 1]

How have you been ve e [0,0,1; c, \$1,1,0,5,0,1]

Nice to see you = [0, 0, 0, 0, 0, 0, 1, 1, 1, 1]

Have a nice day = [1, 0, 0, 1, 1, 0, 1, 0, 0, 0]

Total Frequency = [1, 1, 1, 1, 2, 2, 2, 1, 1, 3]

a	are	been	day	have	how	nice	see	to	you
1	1	1	1	2	2	2	1	1	3



Why use BoW?

 The intuition is that documents are similar if they have similar content. Further, that from the content alone we can learn something about the meaning of the document.

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Problem with BoW

• Discarding word stap signotest the context, and in turn meaning of words in the document (semantics). Context and meaning can offer a lot to the model, that if modeled could tell the difference between the same words differently arranged ("this is interesting" vs "is interesting").

S1= I love you but you hate me
S2= I hate you but you love me
WORDS



Term Frequency-Inverse Document Frequency

• Term Frequency-Inverse Document Frequency (TF-IDF) is a way of representing *how important a word* is to a document in a collection or corpus.

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$$W_{i,j} = W_{i,j} \times \log \left(\frac{\partial M}{\partial f_i}\right)$$

https://tutorcs.com
 $W_{i,j} = W_{i,j} = W_$

- The Term Frequency is a count of how many times a word occurs in a given document (synonymous with bag of words)
- The Document Frequency is the number of times a word occurs in a corpus of documents



Term Frequency

$$w_{i,j} = tf_{i,j} \times \log(\frac{N}{df})$$
Assignment Project Exami Help Like Bow

the type of accurrences of term i in document j

Document #1: I like apple Chat: cstutorcs

Document #3: Sweet and yellow banana banana

Document #4: Sweet fruit

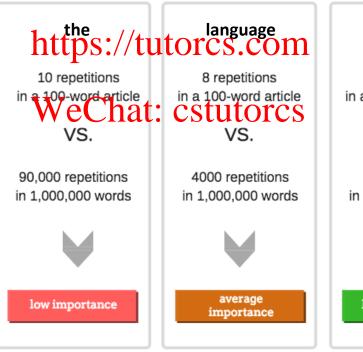
	and	apple	banana	fruit	1	like	sweet	yellow
D#1	0	1	0	0	1	1	0	0
D#2	0	0	1	0	1	1	0	0
D#3	1	0	2	0	0	0	1	1
D#4	0	0	0	1	0	0	1	0



What if we just use Term Frequency Only?

• It is known that certain terms, such as "is", "of", and "that", may appear a lot of times but have little importance. Thus we need to weigh down the frequent terms while scale up the rare ones.

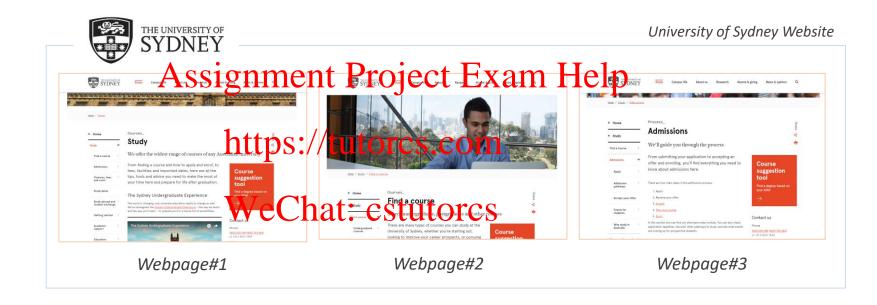
Assignment Project Exam Help







Can we use Term Frequency Only?





Inverse Document Frequency

Why do we need log? $w_{i,j} = tf_{i,j} \times \log \left(\frac{N}{df}\right)$ Assignment Project Exam Help

> *N* = total number of documents httpsimbattores open taining term i

Document #1: I like apple WeChat: cstutorcs

Document #3: Sweet and yellow banana banana

Document #4: Sweet fruit

N	=	4
---	---	---

	and	apple	banana	fruit	1	like	sweet	yellow
df	1	1	2	1	2	2	2	1



Inverse Document Frequency

Why do we need log? $w_{i,j} = tf_{i,j} \times \log(\frac{N}{df})$ Assignment Project Exam Help

> *N* = total number of documents httpsimbattores open taining term i

with logWeChat:nestutorcs

idf(d,t) = log(n/df(t))

df(t)idf(d,t)word1

100

word2

word3	1,000	3
word4	10,000	2
word5	100,000	1
word6	1,000,000	0

Without log

$$idf(d,t)=n/df(t)$$

	df(t)	idf(d,t)
word1	1	1,000,000
word2	100	10,000
word3	1,000	1,000
word4	10,000	100
word5	100,000	10
word6	1,000,000	1



Inverse Document Frequency

$$w_{i,j} = tf_{i,j} \times \log(\frac{N}{df})$$

Assignment Project Exam Help^{1+dfi sometimes, why?}

N = 4

N = total number of documents https://datorcoroom/taining term i

Document #1: I like apple Chat: cstutorcs

Document #3: Sweet and yellow banana banana

Document #4: Sweet fruit

	and	apple	banana	fruit	1	like	sweet	yellow
df	1	1	2	1	2	2	2	1



Inverse Document Frequency

$$w_{i,j} = tf_{i,j} \times \log(\frac{N}{df})$$
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N = total number of documents httpsimbattoresnoomntaining term i

Document #1: I like apple Chat: cstutorcs

Document #3: Sweet and yellow banana banana

Document #4: Sweet fruit

N = 4

	and	apple	banana	fruit	1	like	sweet	yellow
df	1	1	2	1	2	2	2	1
idf (with 1+ <i>dfi</i>)	Inv(4/(1+1)) =0.693147	Inv(4/(1+1)) =0.693147	Inv(4/(2+1)) =0.287682	Inv(4/(1+1)) =0.693147	Inv(4/(2+1)) =0.287682	Inv(4/(2+1)) =0.287682	Inv(4/(2+1)) =0.287682	Inv(4/(1+1)) =0.693147



Term Frequency Inverse Document Frequency

$$w_{i,j} = tf_{i,j} \times \log\left(\frac{N}{df}\right)$$
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https://tutorcs.com

Document #1: I like apple Chat: cstutorcs

Document #3: Sweet and yellow banana banana

Document #4: Sweet fruit

	and	apple	banana	fruit	1	like	sweet	yellow
D#1	0	0.693147	0	0	0.287682	0.287682	0	0
D#2	0	0	0.287682	0	0.287682	0.287682	0	0
D#3	0.693147	0	0.575364	0	0	0	0.287682	0.693147
D#4	0	0	0	0.693147	0	0	0.287682	0



Sparse Representation

With COUNT based word representation (especially, one-hot vector), linguistic information was represented with sparse representations (high-dimensional features)

However, with the recent popularity and success of word embeddings (low dimensional, distributed representations), neural-based models have achieved superior results on various language-related tasks as compared to traditional machine learning models with high-dimensional features.



Lecture 1: Introduction to Natural Language Processing

- Course Introduction
- Overview of Natural Language Processing (NLP)
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 Word Meaning and Representation
- Count-basted Workst Representation
 - One-hot Encoding

 - Bag of WweChat: cstutorcs
 Term Frequency-Inverse Document Frequency
- Next Week Preview



How to Represent the Word Similarity!

How to represent the word similarity with dense vector



Try this with word2vec

Word Algebra

Enter all three words, the first two, or the last two and see the words that result.





Reference for this lecture

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