German wings challenge

Kai Chen

Resume

- Oct 2018 present, Data Science Manager, Unitymedia, Germany
- May 2017 Oct 2018, Senior Data Scientist, AGT International, Germany
- Sep 2012 Jun 2017, PhD in CS, University of Fribourg, Switzerland
- Jun 2015 Sep 2015, Visiting PhD, Chinese Academy of Sciences, China
- Sep 2009 Sep 2012, Master in CS, University of Fribourg, Switzerland
- Oct 2005 Feb 2008, Bachelor in CS, University of Applied Science,
 Switzerland

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Use Cases

- Case 1: Sentiment Analysis
 - Predict a review is positive or negative

- Case 2: Topic Modeling
 - Find topics in the reviews

Methodology

- Data Loading
 - Load data from a txt file and save the data into a pandas dataframe
- EDA (Exploratory Data Analysis)
 - Plot distribution of variables
 - Show relationship between the variables
 - Text analysis: Plot word frequencies, Topic modelling with LDA

Feature Engineering

- TFIDF (term frequency-inverse document frequency)
 - Count Features
- dimensionality reduction using truncated SVD
- Word Embedding: GloVe

Evaluation Metrics

- Cross-Entropy Loss
- Precision Recall
- ROC (Receiver Operating Characteristics)

Modelling

Logistic regression, Gradient boost machine, Deep Learning

Error Analysis

Notebook

Word Embedding

The slides are taken from, Sequence Models, Deep Learning Specialization, Coursera by Andrew Ng.

Word representation

```
V = [a, aaron, ..., zulu, \langle UNK \rangle]
```

1-hot representation

Man Woman King Queen Apple Orange (5391) (9853) (4914) (7157) (456) (6257)

N = 10,000

I want a glass of prange ______.

I want a glass of apple______.

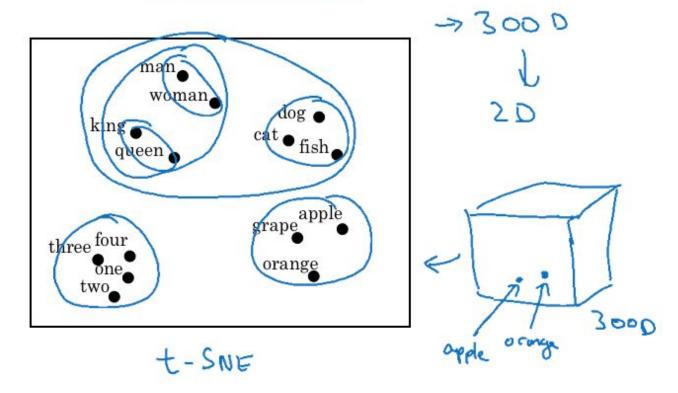




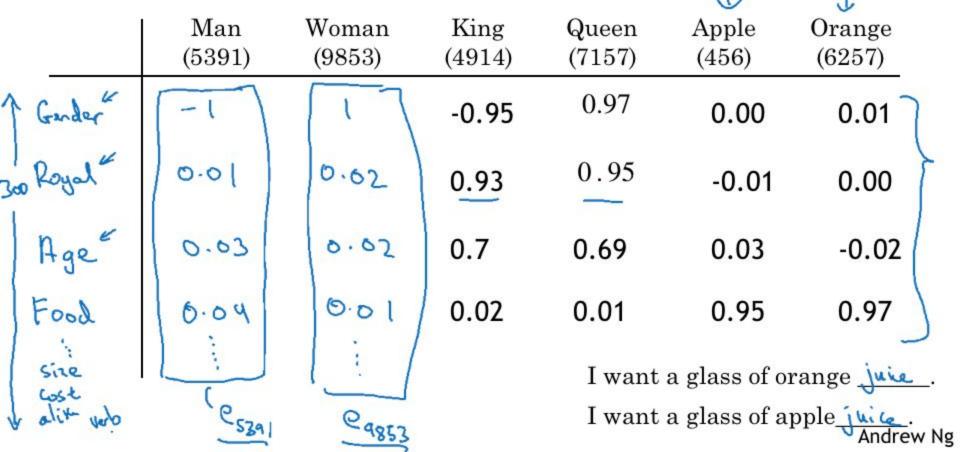


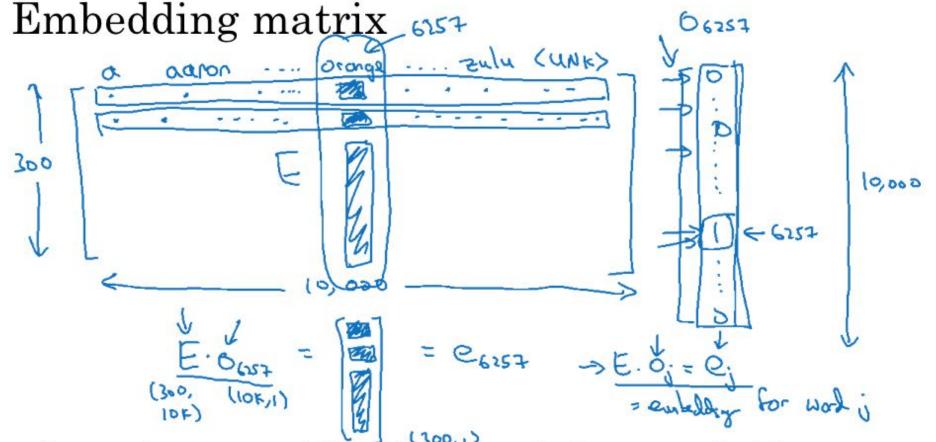
Andrew Ng

Visualizing word embeddings



Featurized representation: word embedding





In practice, use specialized function to look up an embedding.

Andrew Ng

Analogies

	Man (5391)	Woman (9853)	King (4914)	Queen (7157)	Apple (456)	Orange (6257)
Gender	$\left(-1\right)$	1	-0.95	0.97	0.00	0.01
Royal	0.01	0.02	0.93	0.95	-0.01	0.00
Age	0.03	0.02	0.70	0.69	0.03	-0.02
Food	0.09	0.01	0.02	0.01	0.95	0.97
	25391 2 man	2 woman		eman - em	eman ≈ [0]	
Mon -> Woman as King ->! Queen & [-2]						
Ç	2 may - Quamar	& Cking -	23 tames		V Lo.	1

Analogies using word vectors man king woman cat fish queen Theor 3000->20 Momon apple three four grape orange• t-SAE $e_{man} - e_{woman} \approx e_{king} - e_{?}$ Man 300 D Sim Qw Find word wi arg max

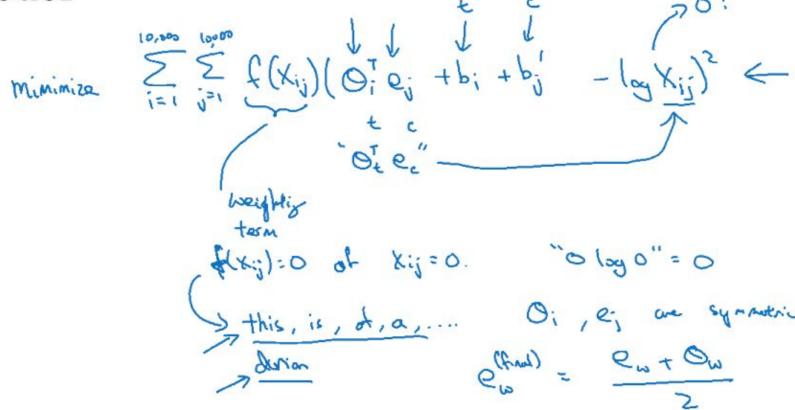
Andrew Ng

30 - 75%

GloVe (global vectors for word representation)

I want a glass of orange juice to go along with my cereal.

Model



Work Experience

Work at Unitymedia

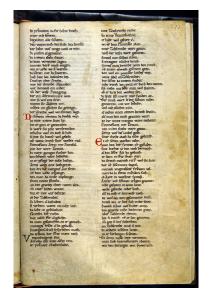
- Recommendation System
- Churn Prevention
- ETL in Hadoop

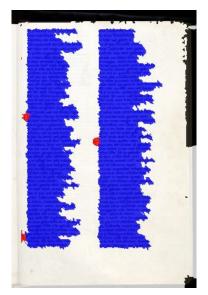
Work at AGT International

- Punch Recognition with Deep Learning
- A/B Testing for Punch Recognition Evaluation
- Anomaly Detection

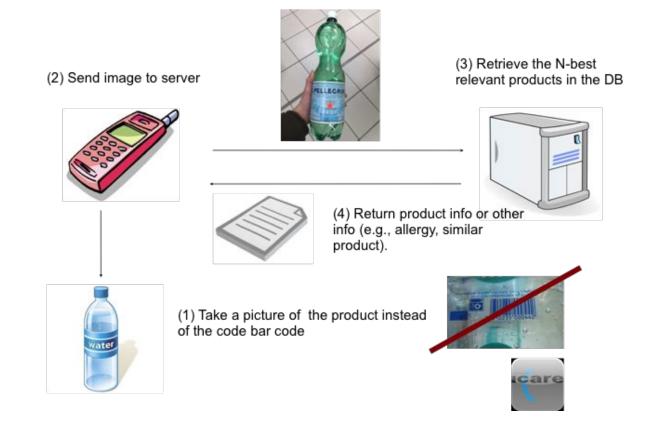
PhD thesis: Historical Document Layout Analysis with Machine Learning

- Goal
 - Developing a general page segmentation method with minimal prior knowledge.
- Basic Idea
 - Page Segmentation —— Pixel Labeling

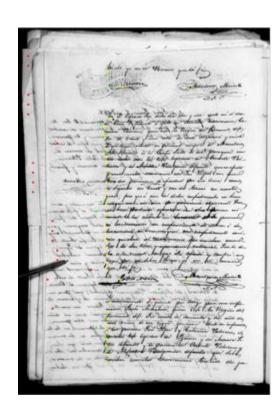


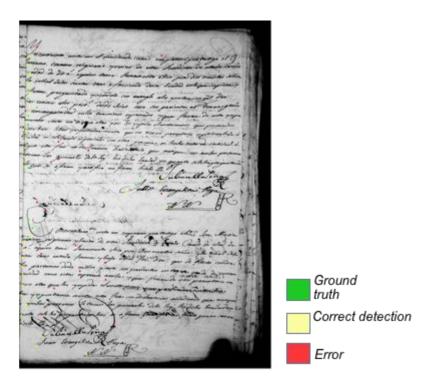


Camera-based image retrieval (master project)

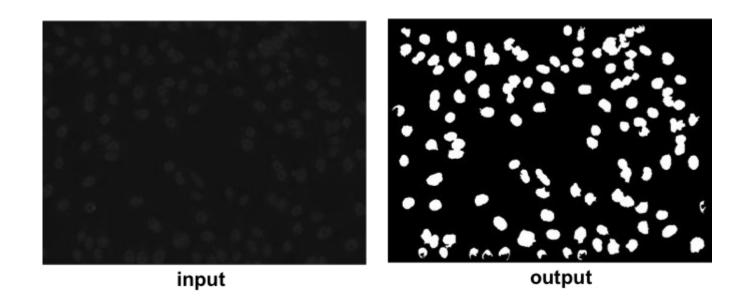


Text line detection in historical document





Cell image segmentation



Other Projects

- Structured Data
 - Booking prediction
 - Revenue per click prediction
 - Consumer shopping prediction
 - Car price prediction
- Computer Vision (CV)
 - Product category classification
- Natural Language Processing (NLP)
 - Toxic Comment Classification (Kaggle challenge, top 16%)
- CV + NLP
 - Product description generation

Why I want to join zeroG?