

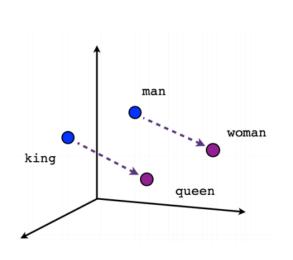
# Contextual word embeddings for recommender systems

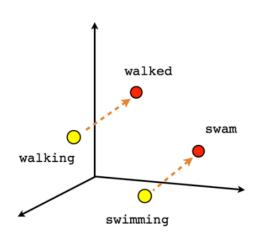
goal

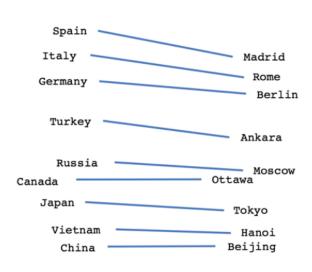
**GOAL:** To showcase a methodology to create recommendations using contextual word embeddings

**SECRET GOAL**: To show that hypothesis/experimental tinkering can lead to real business impact + open source experiments!

# word2vec- famous outcomes







Male-Female

Verb tense

Country-Capital

	100 10 100 0 0 10	4-	4!
word	represen	Itai	uons

word2vec is *not* deep learning;

The whole idea behind word2vec is to demonstrate that **you can get better word representations if you trade the model's complexity for efficiency**, i.e. the ability to learn from much bigger datasets.

"A word is known by the company it keeps"

## intuition for word2vec

word2vec is a shallow word embedding model.

#### STEPS:

- 1. Create word IDs 0 to length of the vocab in the corpus
- 2. IDs are mapped into vector space, taking their distributional properties into account
- 3. Training is online one example at a time
- 4. Target for CBOW and SKIP respectively predict the target word (w), given the context (c) i.e. P(w|c) OR predict the context, given the word i.e. P(c|w)

Thumb Rule: More data, use SKIP. Less data, use CBOW

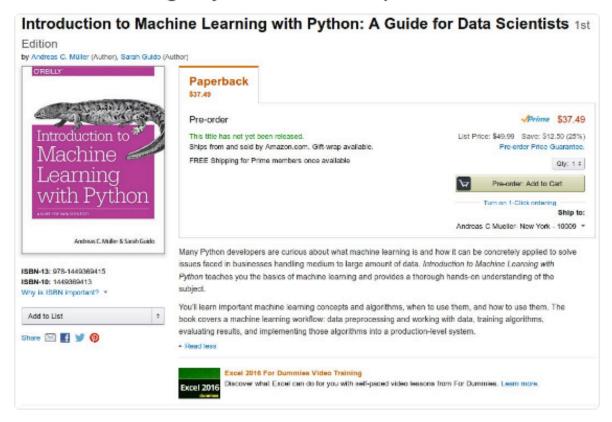
Or not, decide depending on your experiments. Conflicting views do exist

## recommendations can be broken, well sometimes!





# Amazon recommends "Excel for Dummies" when viewing my book... :weep:



### motivations

Word representations took the world by storm when it was introduced by Google. Numerous papers have been written after that in the domain of topic modelling, sentiment analysis and social media user profiling.

This is our attempt to propose and demonstrate a framework that's more rounded and preserves context while generating recommendations.

data

Amazon Reviews and user browsing history for more than 18 years, contains product reviews and metadata from Amazon, including 142.8 million reviews spanning May 1996 - July 2014.

Digital Music Reviews-800K+

**User Browsing History**: 9.4 M records

### models

#### **META MODEL:**

Created with the user browsing history. This contains more than 9.4 million product histories. The attempt is to mimic and improve upon existing product-product systems.

#### **USER REVIEW MODEL:**

Currently many food reviews or travel sites don't allow us to search on "context". This model intakes reviews and attempts to create a framework for a "contextual search engine".

#### **USER REVIEWS**

```
"reviewerID": "A2SUAM1J3GNN3B",
  "asin": "0000013714",
  "reviewerName": "J. McDonald",
  "helpful": [2, 3],
  "reviewText": "I bought this for my
husband who plays the piano. He is
having a wonderful time with Shankar
Mahadevan.
  "overall": 5.0,
  "summary": "Heavenly Highway Hymns",
  "unixReviewTime": 1252800000,
  "reviewTime": "09 13, 2009"
```

#### **METADATA**

```
"asin": "0000031852",
  "title": "Girls Ballet Tutu Zebra
Hot Pink",
  "price": 3.17,
  "imUrl": "http://ecx.images-
amazon.com/images/I/
51fAmVkTbyL. SY300 .jpg",
  "related":
    "also bought": ["B00JHONN1S",
"B002BZX8Z6", "B00D2K1M30", ....],
    "also viewed": ["B002BZX8Z6",
"B00JHONN1S", "B008F0SU0Y"...],
    "bought together": ["B002BZX8Z6"]
  "salesRank": {"Toys & Games":
211836}.
  "brand": "Coxlures",
  "categories": [["Sports & Outdoors",
"Other Sports", "Dance"]]
```

## transformations

#### **USER REVIEWS**

I bought this for my husband who plays the piano. He is having a wonderful time with Shankar Mahadevan.

5> <0000012714> ik

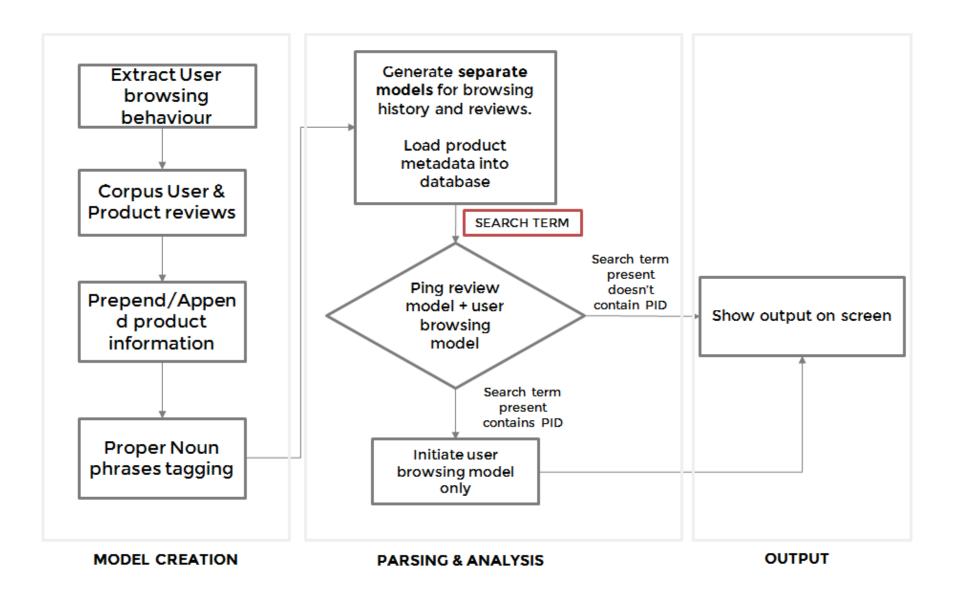
<p\_5> <0000013714> i bought this for my husband who plays the piano. he is having a wonderful\_time with shankar\_mahadevan. <0000013714> <p\_3.17>

#### **METADATA**

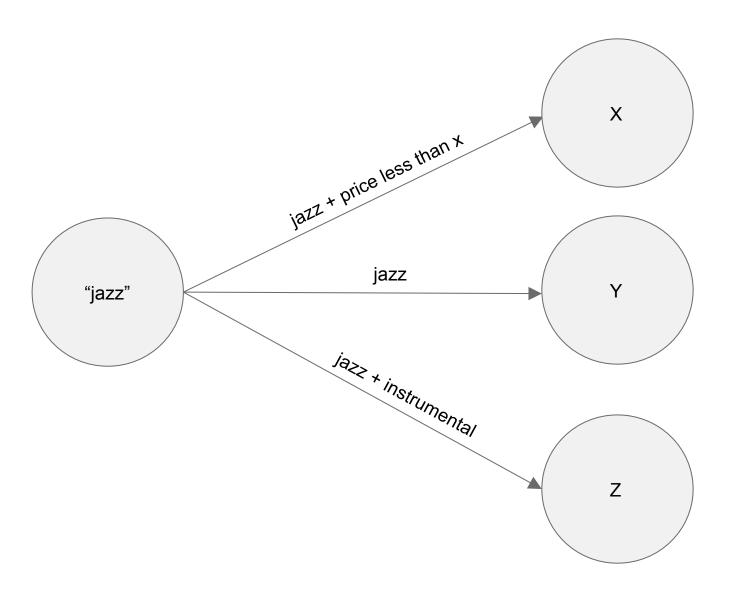
"0000031852", "B002BZX8Z6", "B00JHONN1S", "B008F0SU0Y"

"0000031852", "**b\_t\_**B002BZX8Z6", "**a\_b\_**B00JHONN1S", "**a\_v\_**B008F0SU0Y"

## pipeline



# end goal



## in action

Article/product titles re-writes for better SEO

Compact text features during ML - WIP

Validate and refine ontologies for domains

• SKU reorganization and genre validation

Sentiment Analysis, with Sentiwordnet

Recommendation engine for Indian restaurants

# DEMO

## what could have done better

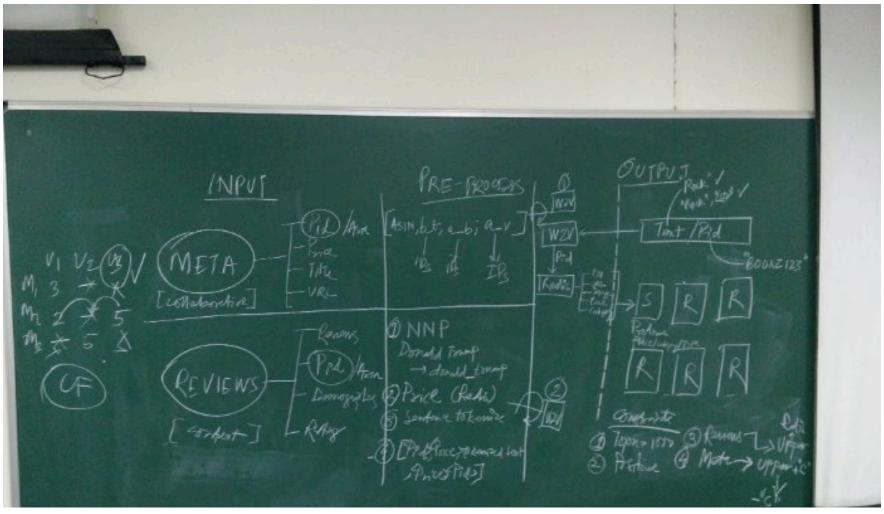
Better pre-processing using DBpedia – currently relies on a NNP parser only

• Investigating 'rogue' results and experimenting with more hyper-parameters

Mixing metadata and user reviews into one

Developing metrics and automated techniques for measuring performance

# **QUESTIONS?**



# https://github.com/manasRK/word2vec-recommender

Author	Working As	contact @
Manas Ranjan Kar	Practice Lead @ Juxt Smart Mandate	manasrkar91@gmail.com
Akhil Gupta	Intern, Amazon India	akhilgupta.official@gmail.com