

A Glimpse of NLP in Industry

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Outline

- My Journey & motivations (5 mins)
- Use Case: Geolocation Prediction (20 mins)
- Academia and Industry comparisons (5 mins)
- NLP landscape in industry applications (10 mins)
- Mindset for Industry (10 mins)
- Questions and Answers (10 mins)

My Journey with NLP

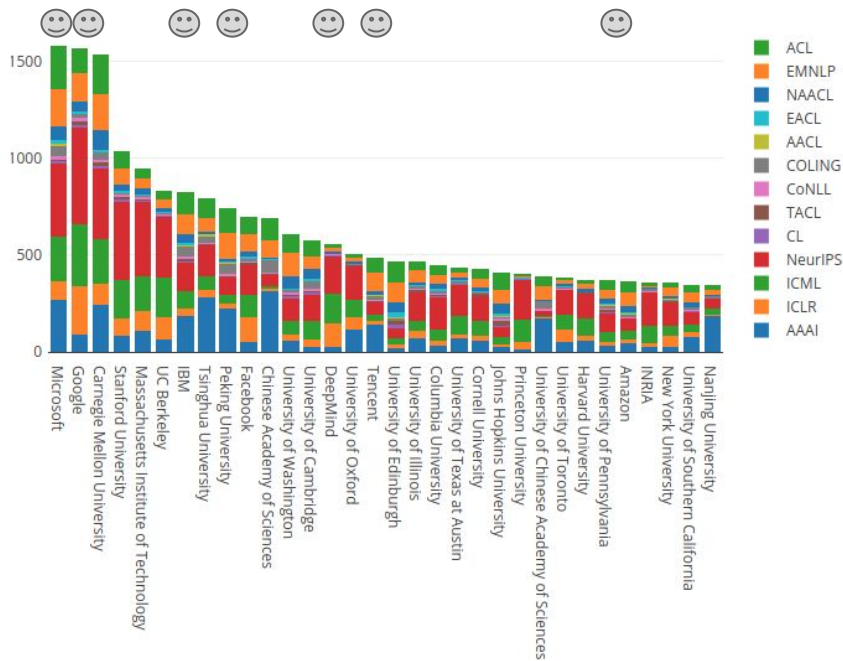
Industry Research Institutions: Microsoft Research Asia (2007-2009), IBM Research Australia (2014-2016)

Universities: University of Melbourne/NICTA (2010-2014)

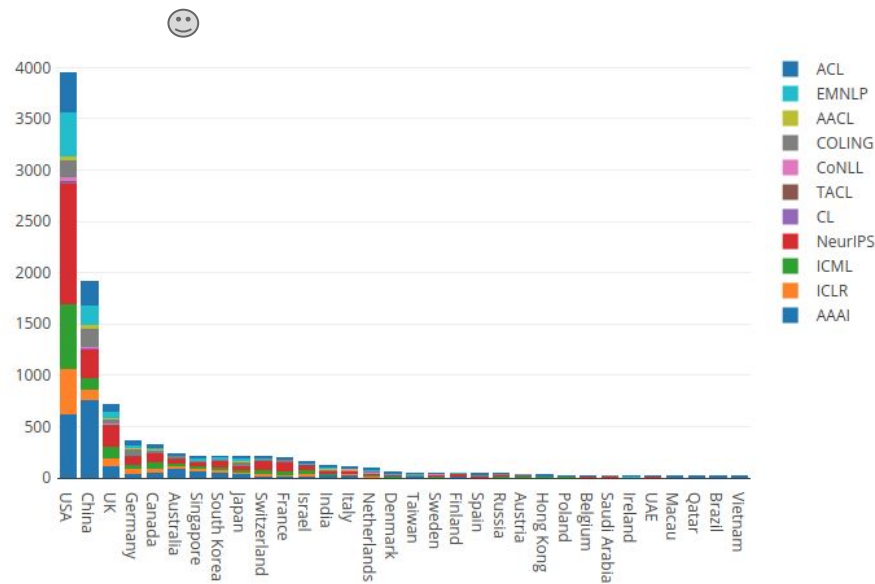
Professional Firms: Start-up (2016-2017), Kaplan (2017-2018), Accenture (2018-now)

Why should I care NLP/ML in industry?

Papers per organisations (2012-2020)



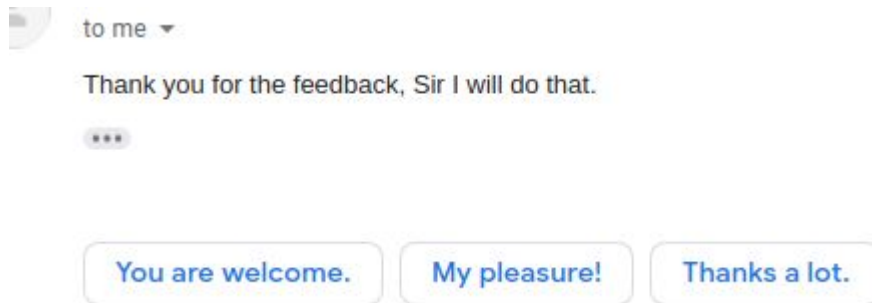
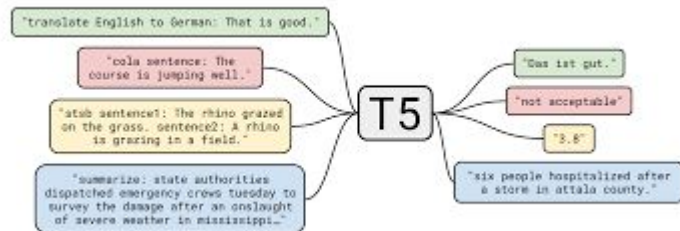
Papers per country/region (2020)
(Australia ranked 6th)



Why should I care NLP/ML in industry?

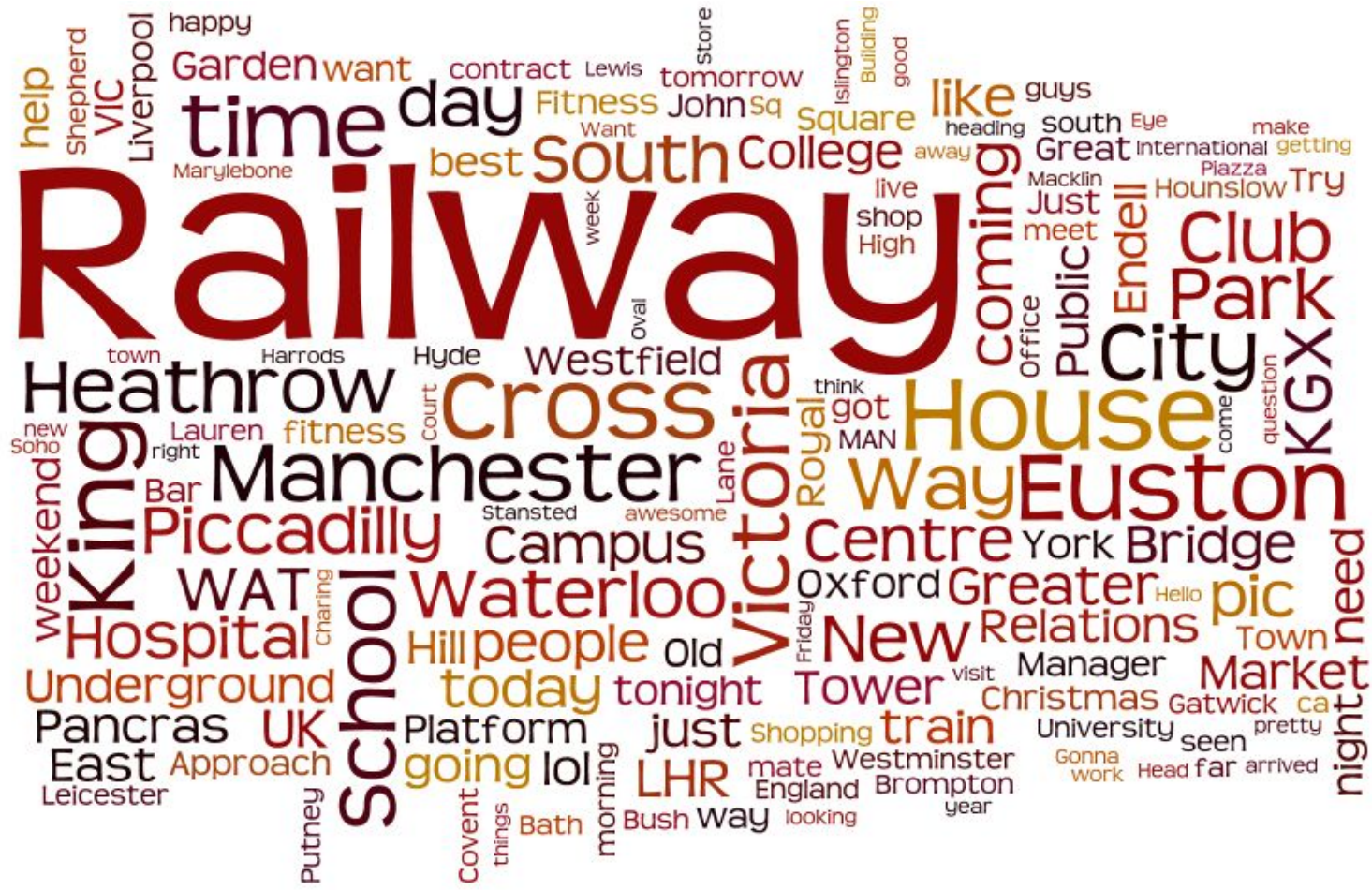


If you'd told me year ago that today I would finish a marathon, I would add an article : had a huge affect on r a year



Case study: Geolocation Prediction

Game time: Can you guess the city?



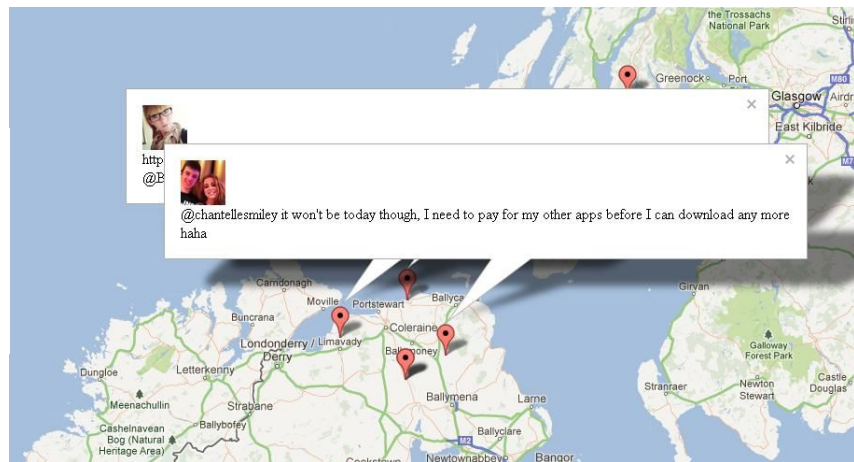
Text-based Geolocation Prediction

Assign a unambiguous geographical location to a piece of text

Input: text data, e.g. an English tweet

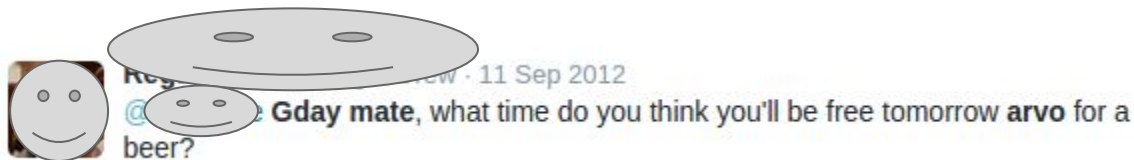
Output: one of metro cities across the world, e.g. London, Sydney, New York

Task: A multi-class classification task

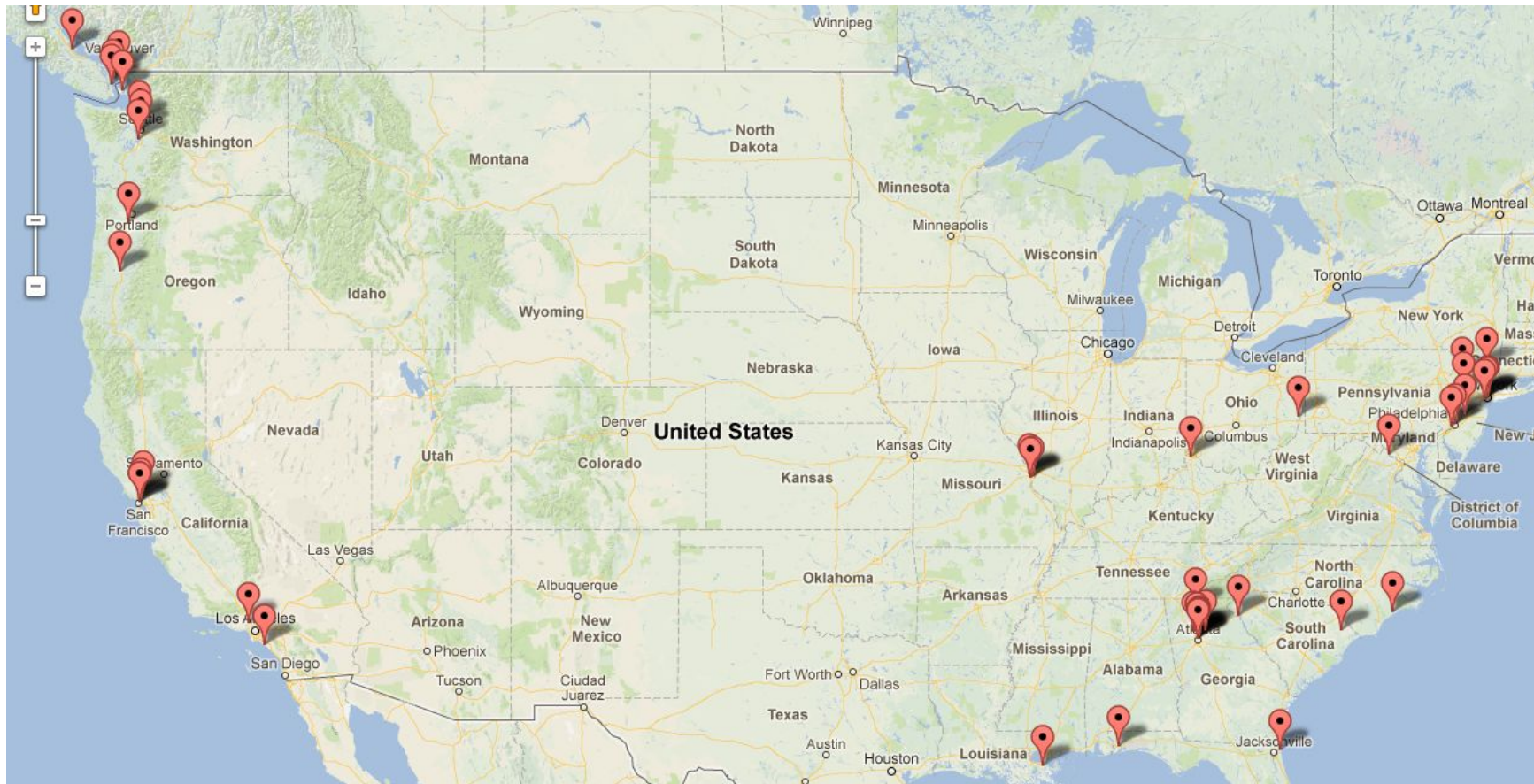


Hypothesis: Words carry varying amount of geolocation information

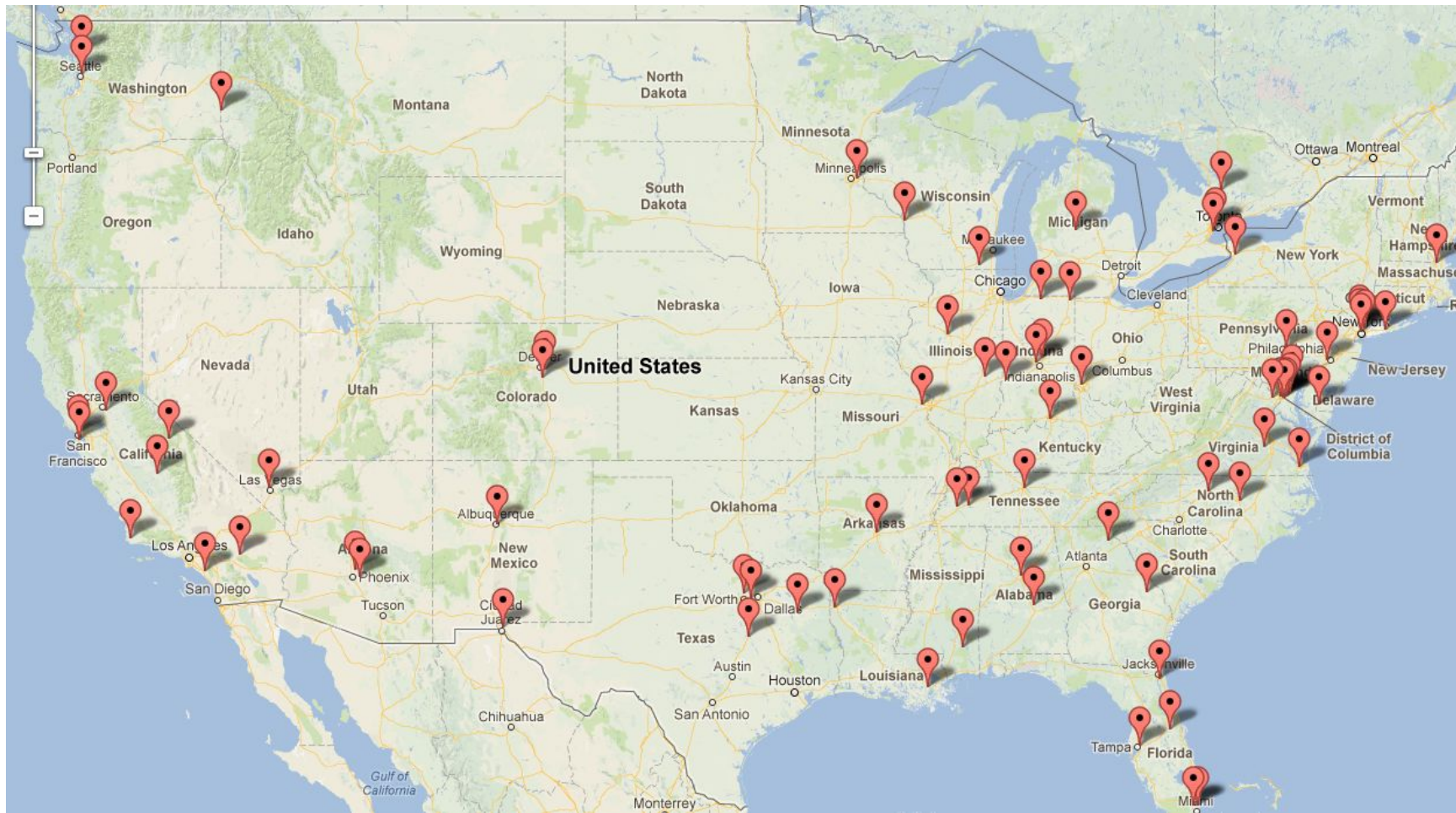
- Gazetted terms: Australia, Canada, London, Seattle,
- Local sports: hockey, footy, cricket
- Dialectal words: arvo, yinz, howdy
- Geo entities: tube, tram, skyscraper, ferry



Somewhat Local Words: ferry

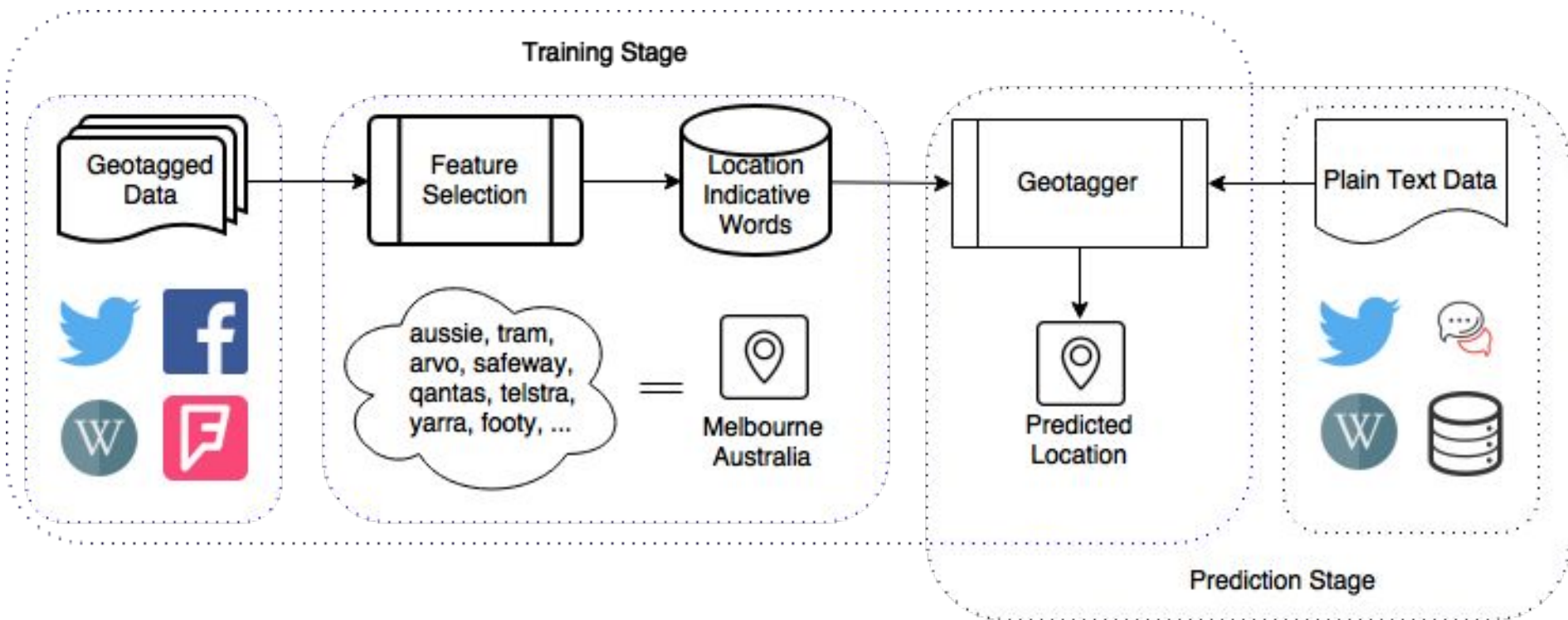


A map of the United States with numerous red location pins placed across various states, indicating a high density of data points in the Northeast, Midwest, and West Coast. The pins are concentrated in the Northeast corridor (from Boston down to Washington D.C.), the Great Lakes region (around Chicago and Detroit), and along the West Coast (from San Francisco down to Los Angeles). There are also several pins in the Mountain West region (Colorado, Utah, Nevada) and the Southeast (Florida, Georgia, South Carolina). The map shows state boundaries and major cities, with the pins representing specific locations of interest.



Geolocation Prediction from Academia View

A Text-based Geo Prediction Framework



Text-based Geo Prediction (Academia)

Q: How to find Location Indicative Words? (LIW)

Q: How to measure model prediction accuracy? (Evaluation)

Q: What are suitable classifiers for this multi-classification? (ML Model)

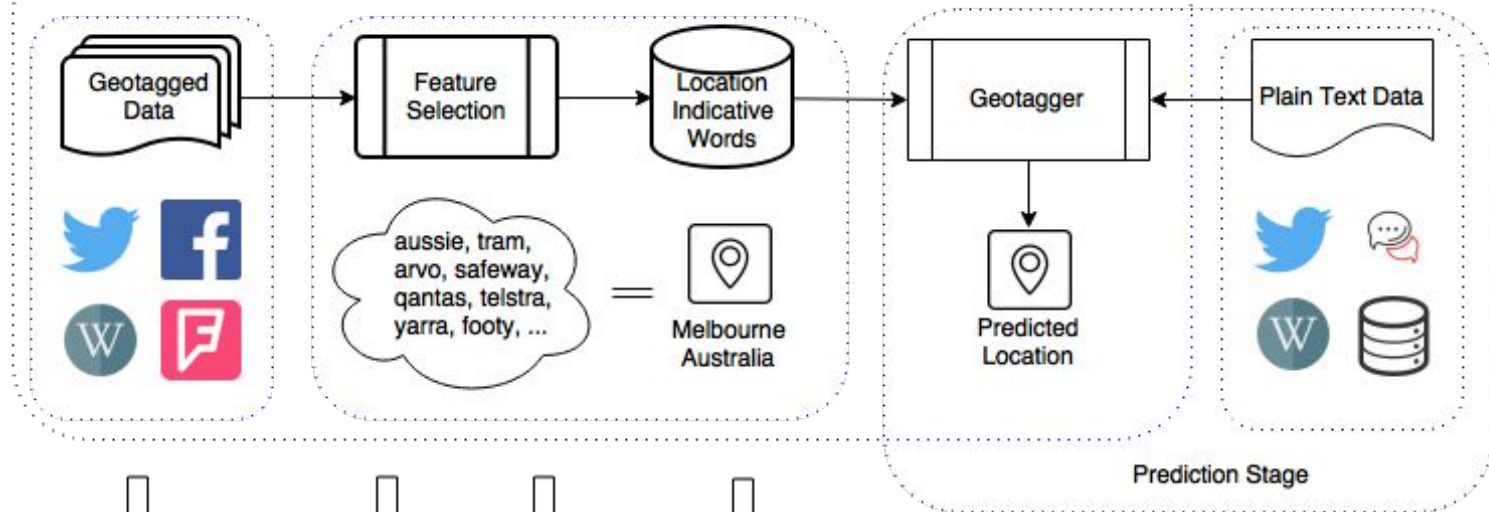
Q: How does input size (i.e. amount of text data) affect the accuracy? (Data)

Q: Will my prediction model accuracy decrease over time? (Generalisation)

Q: Will language, metadata, text-derived network relations affect model accuracy?
(NLP)

...

Training Stage



Data Data

Geo-tweets vs. Geo-users

Evaluation

Data Partitions

City vs. Earth grids vs. Hierarchical

Multi-Classifiers ML Model

Simple (Naive Bayes) vs. Advanced (MaxEnt)
Generative (Naive Bayes) vs. Discriminative (LR)
Ensemble learning

(NLP)

Features

English only vs. All Languages

Metadata

Feature Selection

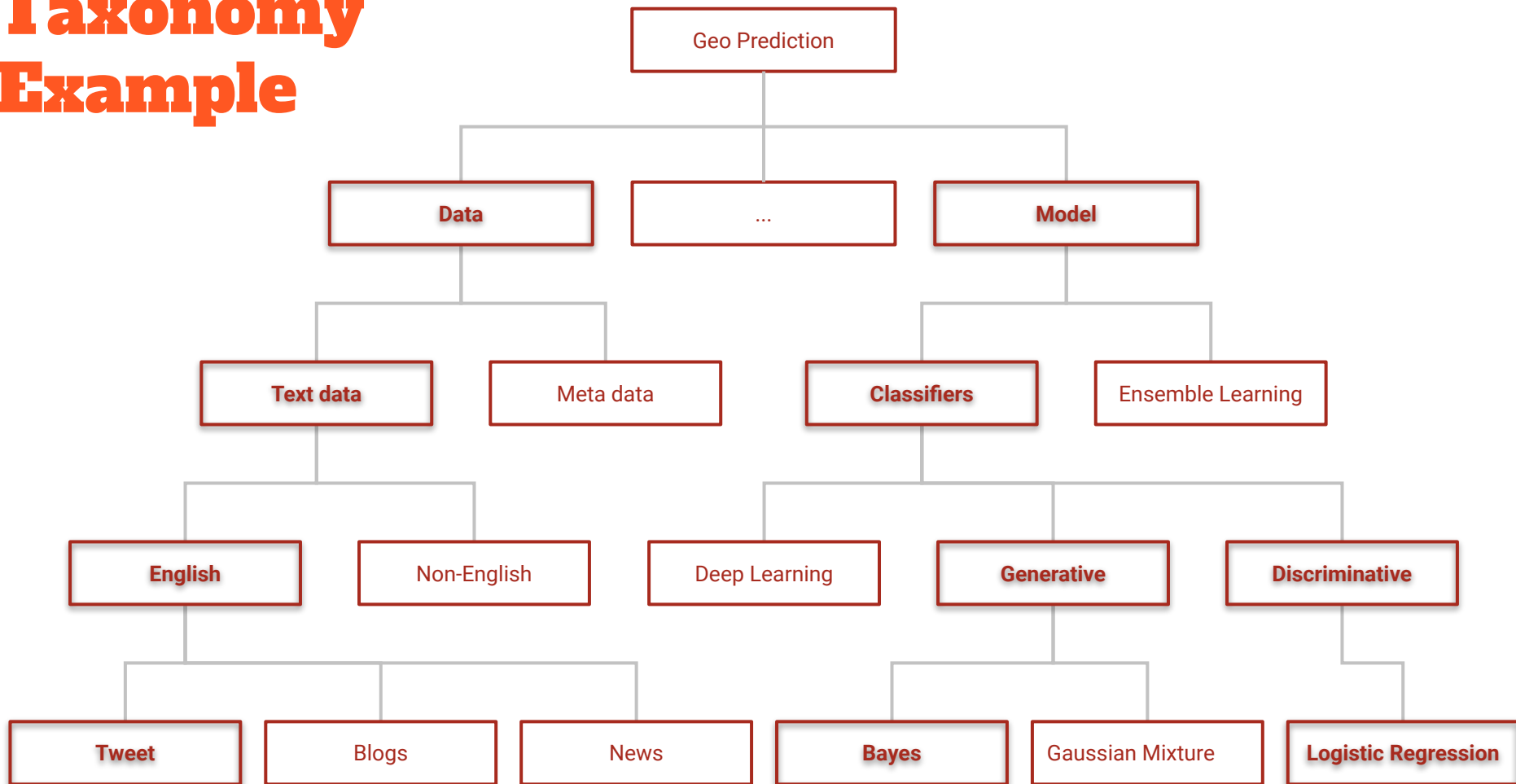
Statistical Methods **LIW**
Heuristic Methods
Information theory-based Methods

Temporal Variance

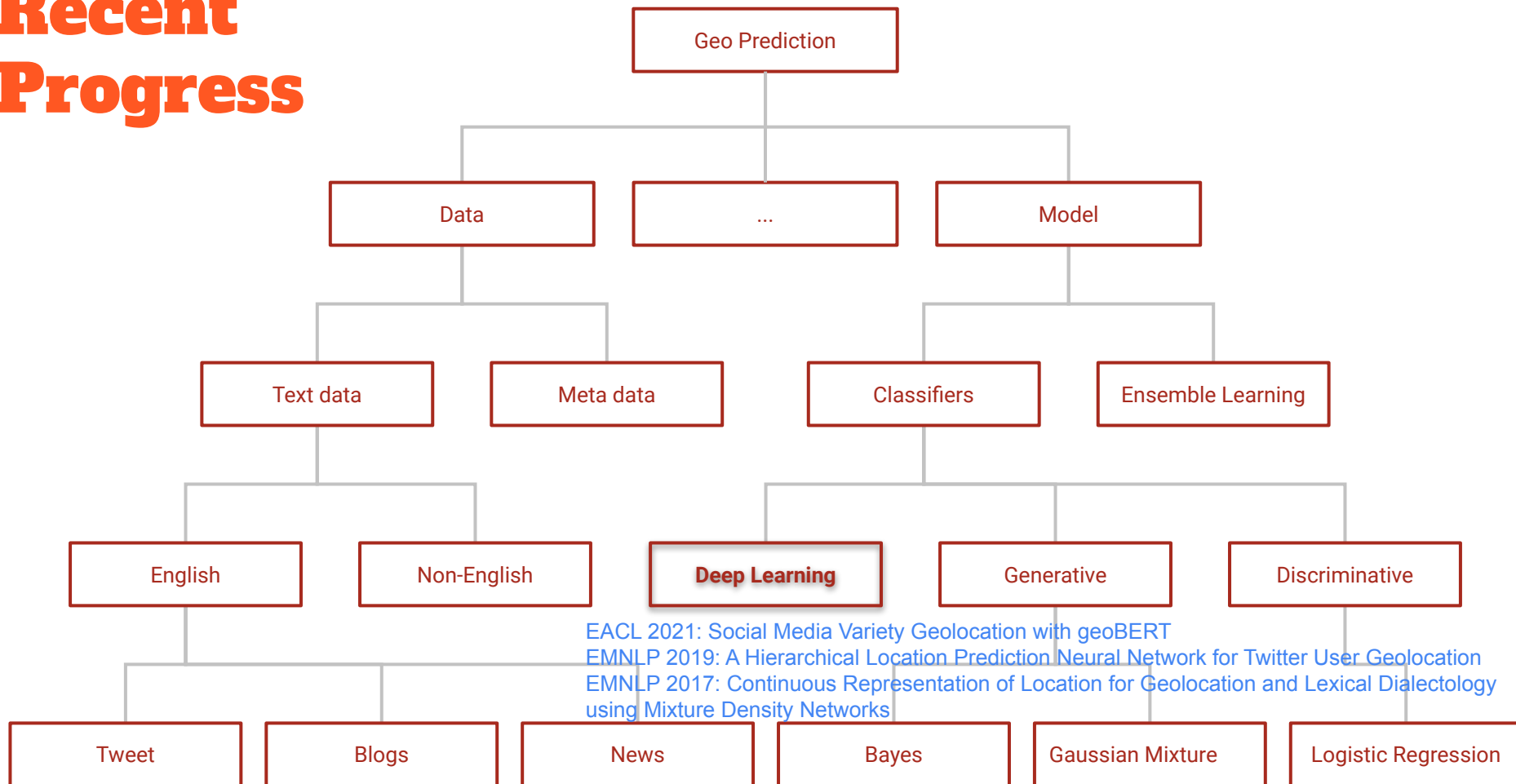
Will trained classifier be less effective over time?

Generalisation

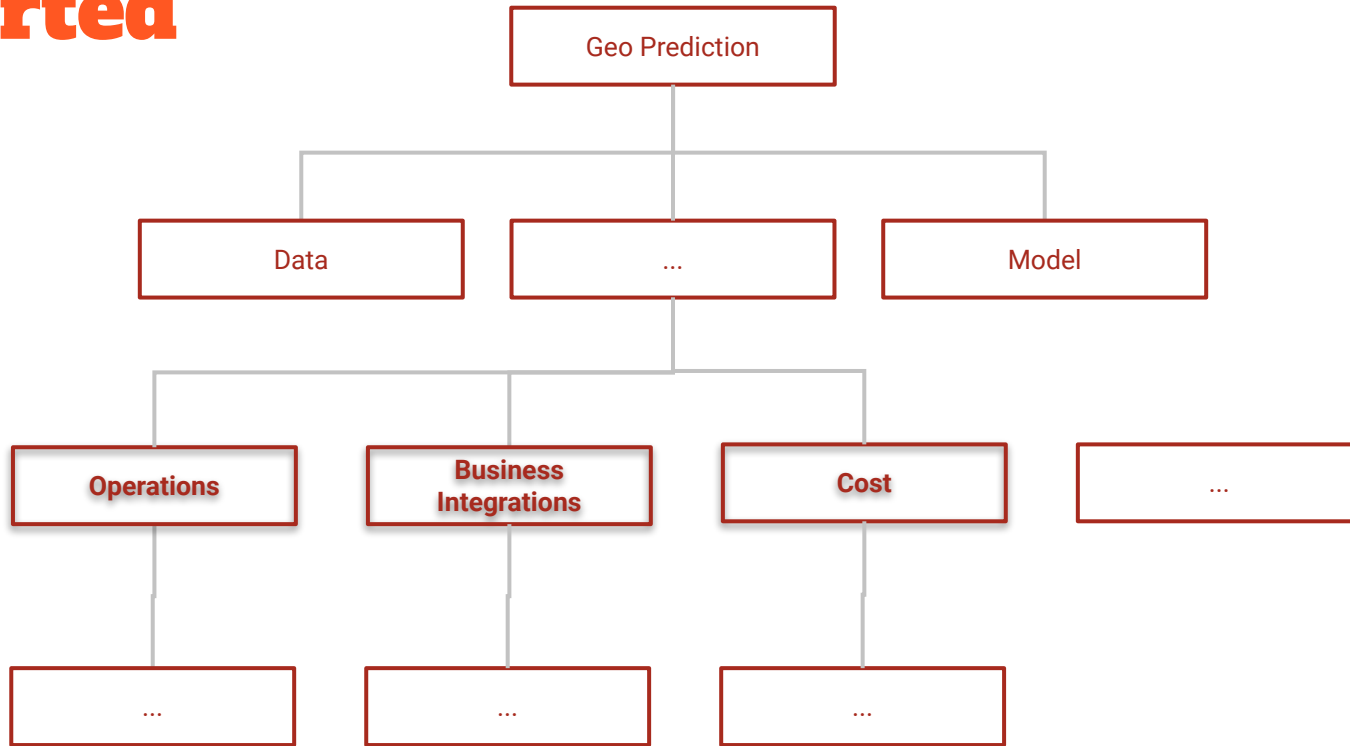
Taxonomy Example



Recent Progress



Uncharted



Geolocation Prediction from Industry View

Text-based Geo Prediction (Industry App)

~~Q: How to find Location Indicative Words? (LIW)~~

Q: How to measure model prediction accuracy? (Evaluation)

~~Q: What are suitable classifiers for this multi-classification? (ML Model)~~

~~Q: How does input size (i.e. amount of text data) affect the accuracy? (Data)~~

Q: Will my prediction model accuracy decrease over time? (Generalisation)

~~Q: Will language, metadata, text derived network relations affect model accuracy?~~
(NLP)

...

Text-based Geo Prediction (Industry App)

Q: How to measure model prediction accuracy? (Evaluation)

Q: Will my prediction model accuracy decrease over time? (Generalisation)

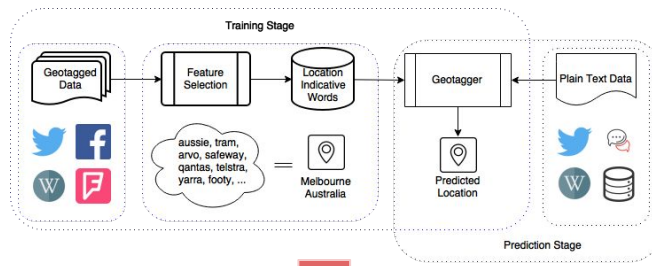
Q: What business service/product can leverage this service? (Utility)

Q: What is the throughput of this deployed service? (Performance)

Q: What are ethics/data privacy/... risks? (Risk)

Q: Should we apply a patent or keep it as a business secret? (IP)

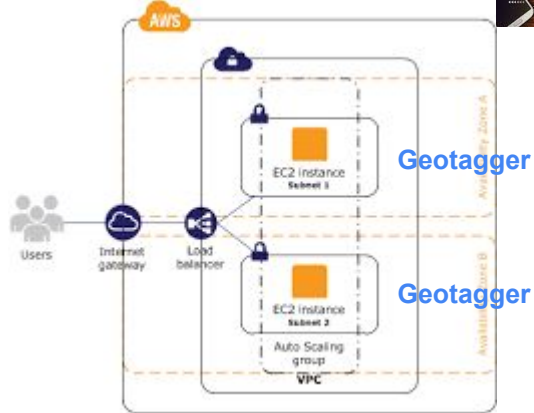
...



Regulations



High Availability



Data Lake



DevOps:

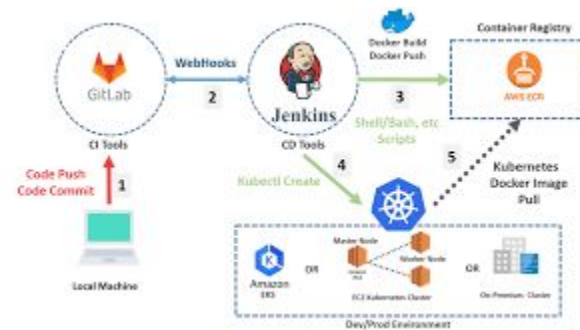
Version Control: Git/Bitbucket

CICD: Jenkins/Bamboo

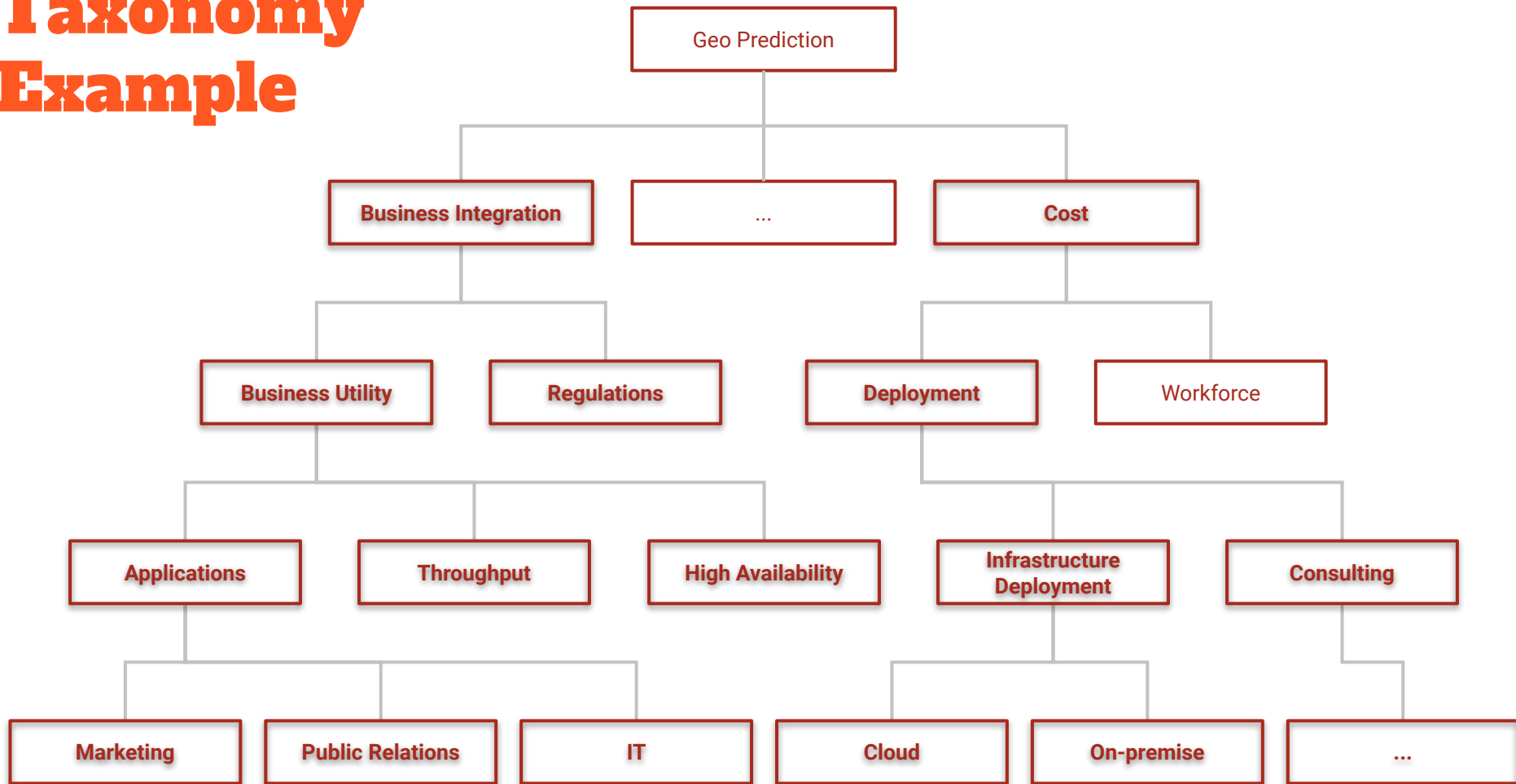
Project Management: JIRA/Trello

Containerisation: Docker/K8S

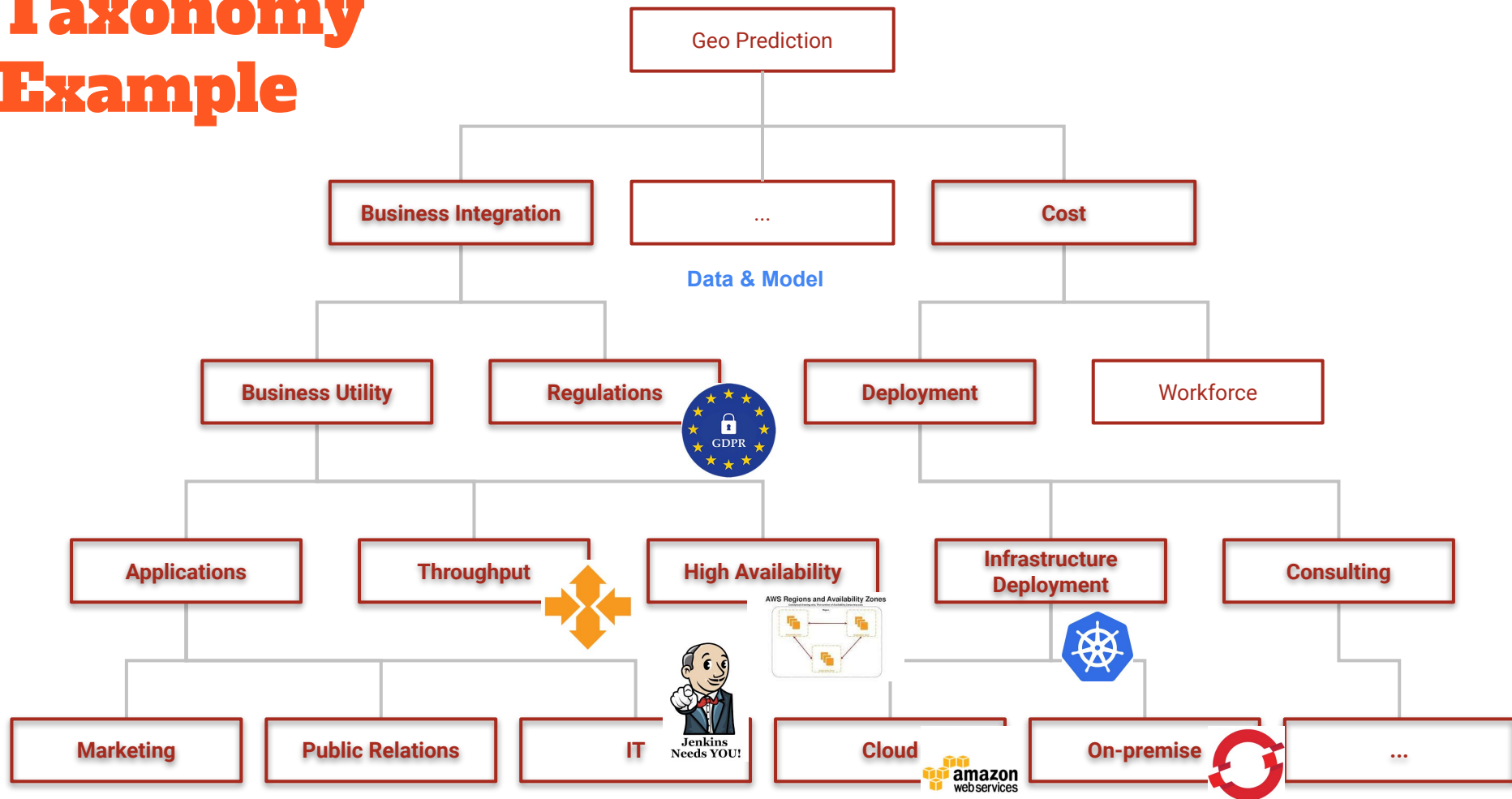
Full Stack: ...



Taxonomy Example



Taxonomy Example



A Pilot Comparison

Academia:



- Broaden the **human knowledge boundaries**, e.g., *improve accuracy from X% to Y% where $Y > X$ and the result is statistically significant*
- It is typically driven by **research questions**
- Work output: **publications**
- Typical activities:
 - Literature review (required)
 - Experiments (required)
 - Publish papers (required)
 - Understand relevant work (required)
 - ...
 - **A working demo website (optional)**

Industry:



- Mostly about **applications**, e.g., *apply sentiment analysis to collect customer feedback and improve our products.*
- It is typically driven by **business needs**
- Work output: **business application**
- Typical activities:
 - A working PoC demo (required)
 - Deployment (required)
 - Cost estimation (required)
 - Information security (required)
 - Regulation requirements (required)
 - ...
 - **Utilise state-of-the-art result from academia (required)**
 - **Papers (optional) and other IPs (required)**

Benefit from Mutuals

Benefit from mutuals (Industry -> Academia)

Academia:



- **Business need** is a good (but not the only) source for your research topic



(Hypothetical) business need:
A small cafe short staffed

Automated Speech Recognition (ASR)
Text to Speech (TTS)
Neural networks
...

Benefit from mutuals (Industry -> Academia)

Academia:



- Research with clear or potential **business applications** may get more funding
- [Yahoo! Key Scientific Challenges Program](#)
- [Microsoft Faculty Fellowship](#)
- [Google Faculty Research Awards in NLP and other fields](#)
- ...

Benefit from mutuals (Industry -> Academia)

Academia:



- An increasing number of **key research papers** are from industry research labs

13 items found

Publisher: Nature x

FEATURED PUBLICATION

Mastering Atari, Go, Chess and Shogi by Planning with a Learned Model

Julian Schrittwieser, Ioannis Antonoglou, et al. Nature 2020

i →

FEATURED PUBLICATION
DEEP LEARNING

AlphaFold: Improved protein structure prediction using potentials from deep learning

Andrew Senior, Richard Evans, et al. Nature 2020

i 📄 ↓ DOWNLOAD →

FEATURED PUBLICATION
NEUROSCIENCE
REINFORCEMENT LEARNING

A distributional code for value in dopamine-based reinforcement learning

Will Dabney, Zeb Kurth-Nelson, et al. Nature 2020

i ↓ DOWNLOAD →

PUBLICATION + AUTHORS' NOTES
DEEP LEARNING

International evaluation of an AI system for breast cancer screening

Scott Mayer McKinney, Marcin T. Sieniek, et al. Nature 2020

PUBLICATION

Hierarchical Motor Control in Mammals and Machines


Josh Merel, Matt Botvinick, et al. Nature 2019

FEATURED PUBLICATION
REINFORCEMENT LEARNING
DEEP LEARNING

AlphaStar: Grandmaster level in StarCraft II using multi-agent reinforcement learning

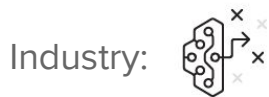
The AlphaStar team, Oriol Vinyals, et al. Nature

Benefit from mutuals (Academia -> Industry)

Industry: 

- Obtain **state-of-the-art algorithms and models** from academia
 - **LSTM**: Sepp Hochreiter; Jürgen Schmidhuber (21 August 1995), Long Short Term Memory
 - **Expectation-maximization algorithm**: Dempster, A.P.; Laird, N.M.; Rubin, D.B. (1977). "Maximum Likelihood from Incomplete Data via the EM Algorithm". Journal of the Royal Statistical Society, Series B. 39 (1): 1–38. JSTOR 2984875. MR 0501537.
 - **Viterbi algorithm**: Viterbi AJ. Error bounds for convolutional codes and an asymptotically optimum decoding algorithm. IEEE Transactions on Information Theory. April 1967, 13 (2): 260–269
 - ...

Benefit from mutuals (Academia -> Industry)



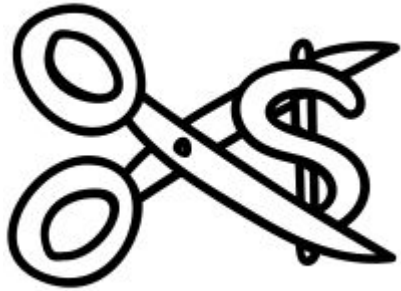
- Software, data and other resource **free to use for commercials**

| |  |  |  |  |  |  |  |
|---|---|---|---|---|---|---|---|
| Type | Permissive | Permissive | Permissive | Copyleft | Copyleft | Copyleft | Copyleft |
| Provides copyright protection | ✓ TRUE | ✓ TRUE | ✓ TRUE | ✓ TRUE | ✓ TRUE | ✓ TRUE | ✓ TRUE |
| Can be used in commercial applications | ✓ TRUE | ✓ TRUE | ✓ TRUE | ✓ TRUE | ✓ TRUE | ✓ TRUE | ✓ TRUE |
| Provides an explicit patent license | ✓ TRUE | ✗ FALSE | ✗ FALSE | ✗ FALSE | ✗ FALSE | ✗ FALSE | ✗ FALSE |
| Can be used in proprietary (closed source) projects | ✓ TRUE | ✓ TRUE | ✓ TRUE | ✗ FALSE | ✗ FALSE partially | ✗ FALSE for web | ✗ FALSE |
| Popular open-source and free projects | Kubernetes Swift Firebase | Django React Flutter | Angular.js jQuery, .NET Core Laravel | Joomla Notepad++ MySQL | Qt SharpDevelop | | SugarCRM Launchpad |



NLP Landscape in Industry

Two Key Factors



Cost

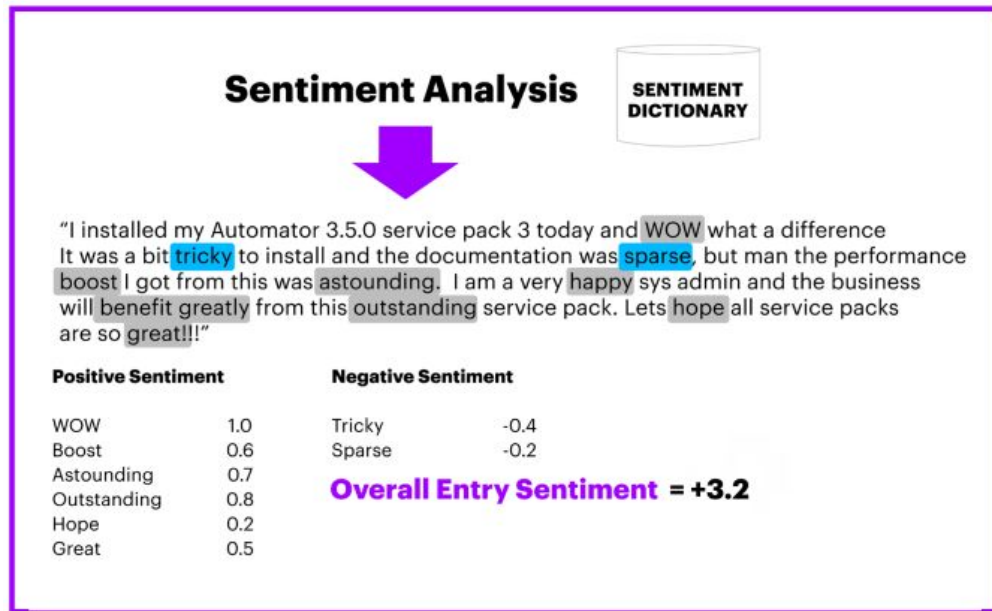


Revenue

NLP Applications in Industry

Sentiment Analysis to identify people's opinions or feelings towards a product/service to collect customer feedback and unlock potential actions

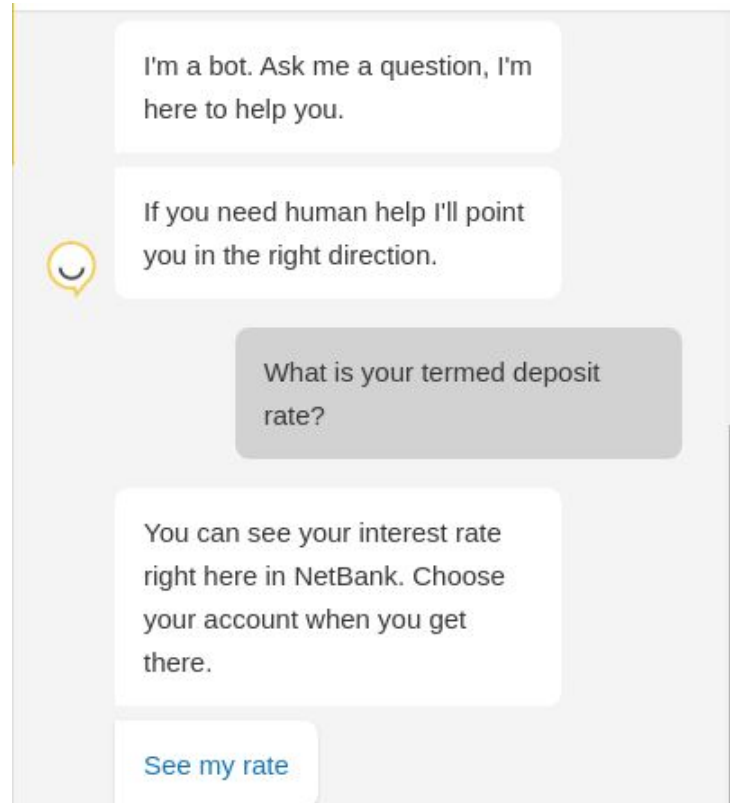
- Provide marketing and competitive intelligence
- Enhance product development
- Improve customer retention
- Analyze the impact of an event (e.g. a product launch or redesign)



NLP Applications in Industry

Chatbots (Virtual Assist) enable conversations between computers and customers to help customers seek relevant information or perform a specific task.

- Improve business processes and reduce support costs
- Enhance search and knowledge-seeking experiences
- Human-in-the-loop to compensate bad experience



Mindset for Industry

NLP/ML Jobs in Industry (application)

Machine Learning



what society thinks I
do



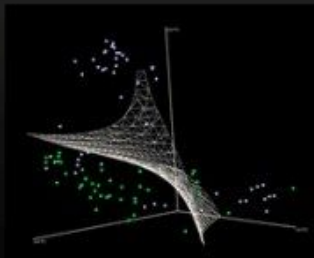
what my friends think
I do



what my parents think
I do

$$\begin{aligned} L_p &= \frac{1}{2} \|w\|^2 - \sum_{i=1}^n \alpha_i y_i (x_i \cdot w + b) + \sum_{i=1}^n \alpha_i \\ \alpha_i &\geq 0, \forall i \\ w &= \sum_{i=1}^n \alpha_i y_i x_i, \sum_{i=1}^n \alpha_i y_i = 0 \\ \nabla \hat{g}(\theta_t) &= \frac{1}{n} \sum_{i=1}^n \nabla \ell(x_i, y_i; \theta_t) + \nabla r(\theta_t) \\ \theta_{t+1} &= \theta_t - \eta \nabla \ell(x_{(t)}, y_{(t)}; \theta_t) - \eta r \cdot \nabla r(\theta_t) \\ \mathbb{E}_{(t)}[\ell(x_{(t)}, y_{(t)}; \theta_t)] &= \frac{1}{n} \sum_{i=1}^n \ell(x_i, y_i; \theta_t) \end{aligned}$$

what other programmers
think I do



what I think I do

```
>>> from sklearn import svm
```

what I really do

Example: Lower Customer Churn



Customer Service: Hi XXX, you recently cancelled the contract with us, I have a good deal for you



Customer Churn:
A customer leaves a company

Lower Customer Churn Step 1

Business Question

1. Question: Can I lower the churn rate in my company?
2. Motivation:



- a. Customer churn will impact our revenue
- b. It will affect our long term growth and eventually our leader position in the market
- c. ...

Lower Customer Churn Step 2

Analysis

1. How many customers are we losing?
2. Who are they?
3. Are all customers the same?
4. Can I collect information that characterise customers
5. ...

Lower Customer Churn Step 2

Analysis

1. How many customers are we losing? **5% in a month**
2. Who are they? **New joiners during previous promotions**
3. Are all customers the same? **No**
4. Can I collect information that characterise customers? **Service, usage statistics, ...**
5. ...

Lower Customer Churn Step 3

Data Science Prediction

Background work:

- Data ETL (data collection, cleansing, validation, loading)
- Data modelling (a classification or a regression task)
- ...

Delivery model:

- Input: a customer's information
- Output: when this customer will leave the company

Lower Customer Churn Step 4

Actionable Insight

If those customer are going to leave,

- What retention policies should I use?
- How should I assign to them?
- Can we further segment those customers into subgroups for different policies?
- Based on your retention model, what would be the long term profits (after subtracting the retention cost)?

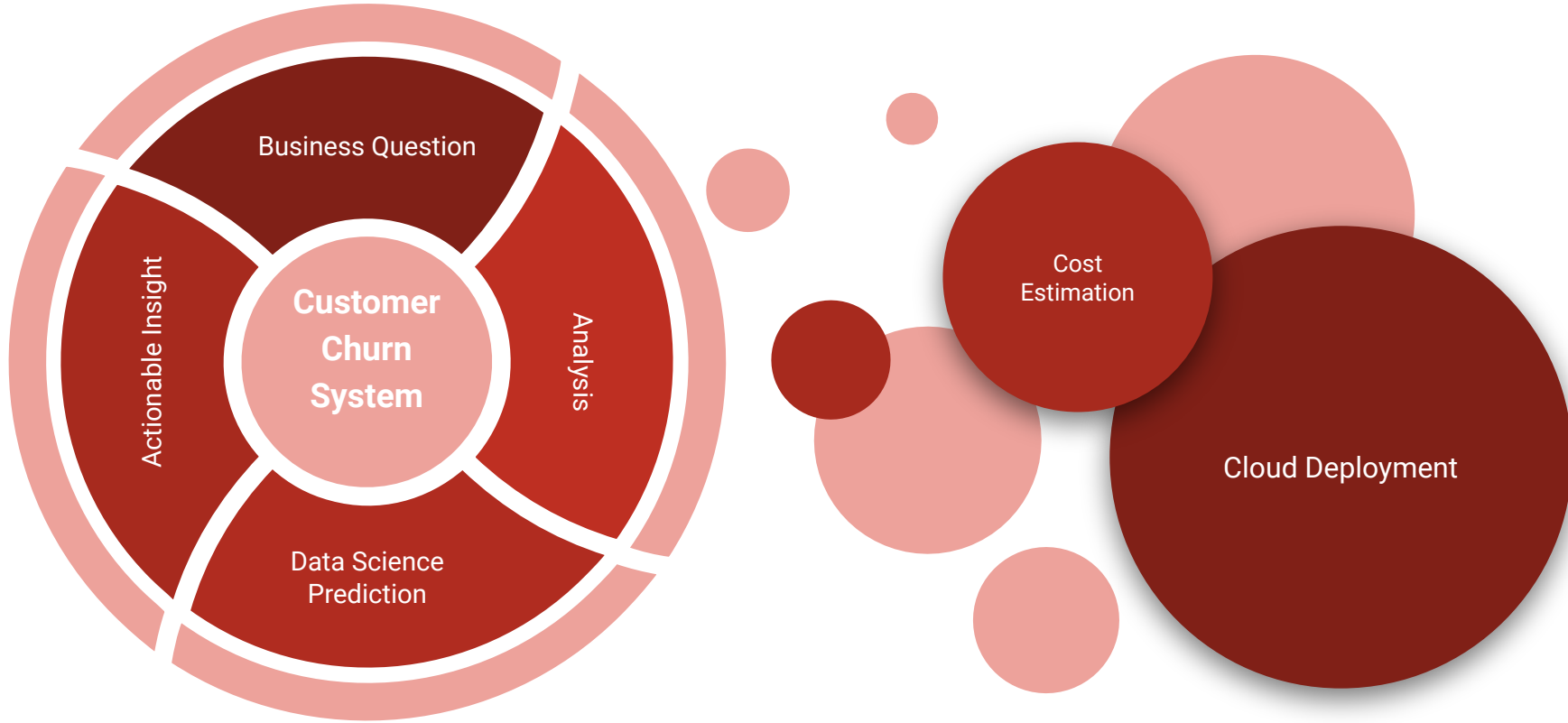
Lower Customer Churn Step 4

Actionable Insight

If those customer are going to leave,

- What retention policies should I use? **One month free, bonus gift card, ...**
- How should I assign to them? **Emails, mails, ...**
- Can we further segment those customers into subgroups for different policies? **Yes, based on their usage plan, we can ...**
- Based on your retention model, what would be the long term profits (after subtracting the retention cost)? **1M AUD this year**

Lower Customer Churn Loop



Recommended Practise

- Practise 1: Fast Food Store Locations
 - Given budget X, where should I select the location for my new store to maximum my profits?
- Practise 2: Who Should I Hire?
 - I need to fill a positions with X, Y, Z requirements, who should I hire?
- Guess techniques:
 - How would you implement an App that has ML/NLP components in your mobile phone?
 - Company X just released service Y, what are the underlying techniques they need to deliver and operate that service?
- Guess applications:
 - Where can AlphaGo and its variations algorithms apply?

A few more words to say

- Ask Alumni Service:
<https://www.unimelb.edu.au/alumni/get-involved/volunteer/ask-alumni>
- Github, personal website or other public presence of your work
- Tech Meetups (a mixture of industry practitioners, researchers, hobbyist)
- Online Course: Coursera, Udacity, O'Reilly...
- Beginner class for cloud computing: AWS Cloud Practitioner
-