

Maximilien BAUDRY

A banner for a Kaggle competition. The background is a red and orange mosaic pattern. In the top left, there is a trophy icon and the text "Featured Prediction Competition". The main title "Quora Question Pairs" is in large white font. Below it, the question "Can you identify question pairs that have the same intent?" is written in a smaller white font. On the right side, the prize money "\$25,000" is displayed in large white font, with "Prize Money" written below it in a smaller white font.

Featured Prediction Competition

Quora Question Pairs

Can you identify question pairs that have the same intent?

\$25,000
Prize Money

WINNING SOLUTION

KAGGLE QUORA

SUMMARY

1. Introduction
2. Deep Learning approach
3. Graphical approach
4. Ensembling and stacking
5. Conclusion

INTRODUCTION

- ▶ What is Quora?
 - World's biggest forum
 - Best place to share general knowledge
 - Topics are designed to only ask questions
- ▶ Problem
 - People may ask similar questions
 - Important interest to detect duplicated questions
- ▶ **Prediction problem : from a question pair, predict whether questions are the same or not**

- ▶ Metric: LogLoss

$$-\frac{1}{N} \sum_{n=1}^N \left[y_n \log \hat{y}_n + (1 - y_n) \log(1 - \hat{y}_n) \right]$$

PRESENTATION OF TEAM DL (DATA'LAB) GUYS

- ▶ Sebastien Conort, chief data scientist BNPP Cardif
- ▶ Lam Dang, data scientist, BNPP Cardif
- ▶ Guillaume Huard, data scientist, BNPP Cardif
- ▶ Paul Todorov, data scientist, BNPP Cardif
- ▶ Maximilien Baudry, PhD student, SAF lab, DAMI (Data Analytics and Models for Insurance) chair of research

#	△pub	Team Name	Kernel	Team Members	Score 🏆	Entries	Last
1	—	DL guys			0.11580	263	8d
2	—	Depp Learning			0.11670	196	8d
3	—	Jared Turkewitz & sjv			0.11756	170	8d
4	—	YesOfCourse			0.11768	189	8d
5	—	Qingchen KazAnova Faron			0.11051	219	8d
6	—	LAMAA power			0.11887	406	8d
7	▲2	aphex34			0.12072	166	8d
8	—	NLPFakers			0.12239	250	8d
9	▼2	Unduplicated Duplicates			0.12248	314	8d
10	▲1	♫ b.a.s.s. ♫			0.12296	271	8d

DATA OVERVIEW

	question1	question2
0	What are some good movies to watch?	What are the best movies to watch?
1	Do dentists earn more than other doctors?	Do dentists earn more than other doctors? Why?
2	Should I wait for iPad Air 3 or purchase the iPad Air 2?	Should I buy the iPad Air or wait for the next iPad Air (iPad Air 2)?
3	What is the difference between Java and Android programming?	Are there major differences between programming in Android vs plain Java?
4	Why do you yawn when you are tired?	Why do we yawn when we are sleepy?
5	Who is Benjamin Netanyahu?	Why is Benjamin Netanyahu famous?

DATA OVERVIEW

	question1	question2	is_duplicate
0	What are some good movies to watch?	What are the best movies to watch?	0
1	Do dentists earn more than other doctors?	Do dentists earn more than other doctors? Why?	0
2	Should I wait for iPad Air 3 or purchase the iPad Air 2?	Should I buy the iPad Air or wait for the next iPad Air (iPad Air 2)?	0
3	What is the difference between Java and Android programming?	Are there major differences between programming in Android vs plain Java?	1
4	Why do you yawn when you are tired?	Why do we yawn when we are sleepy?	1
5	Who is Benjamin Netanyahu?	Why is Benjamin Netanyahu famous?	1

DATA OVERVIEW

- ▶ Duplicates proportion: 36.9% in train, 17.4% in test
- ▶ Number of question pairs: ~400k in train, ~2,3M in test
- ▶ ~80% of test dataset contains fake question pairs, such that we can't hand label test question pairs (avoid cheating)
- ▶ ~530k unique questions in train dataset
- ▶ ~110k questions appear multiple times in train and test datasets
- ▶ Questions which contains:
 - Question mark: 99.87%
 - [math] tags: 0.12%
 - Capitalized first letter: 99.81%
 - Capital letters: 99.95%
 - Numbers: 11.83%

	question1	question2
0	What	How
1	Is there move?	Is format immortality?
2	What are exactly?	How does akamai great money?
3	How cpu insomnia diagnosed?	How ssc is insomnia treated?

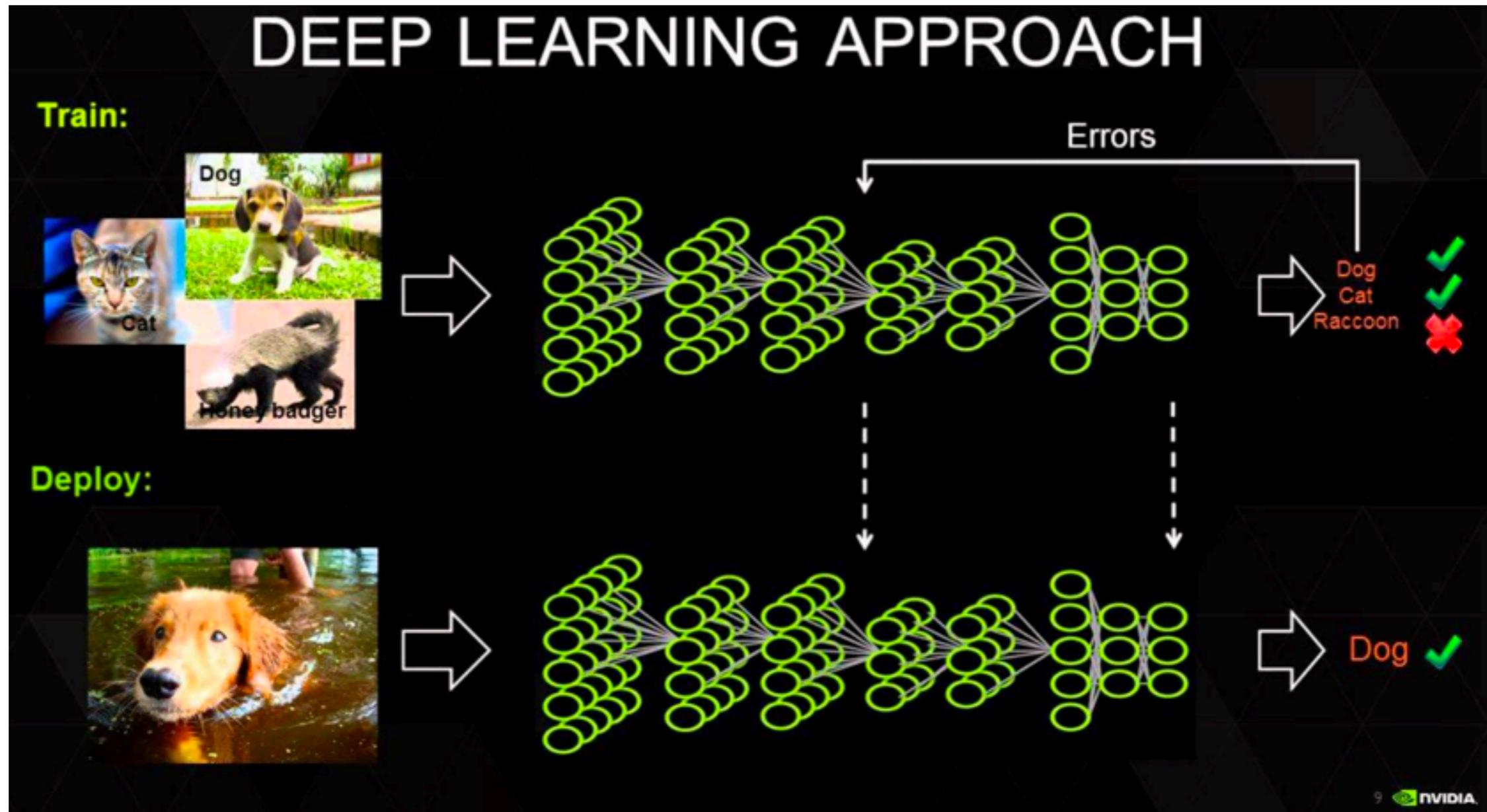
Examples of fake pairs

OUR APPROACHES

▶ 3 main axes:

1. Deep learning
2. Neuro-Linguistic Programming (NLP)
3. Graphical models

FIRST APPROACH: DEEP LEARNING



FIRST APPROACH: DEEP LEARNING

▶ Embedding of each questions

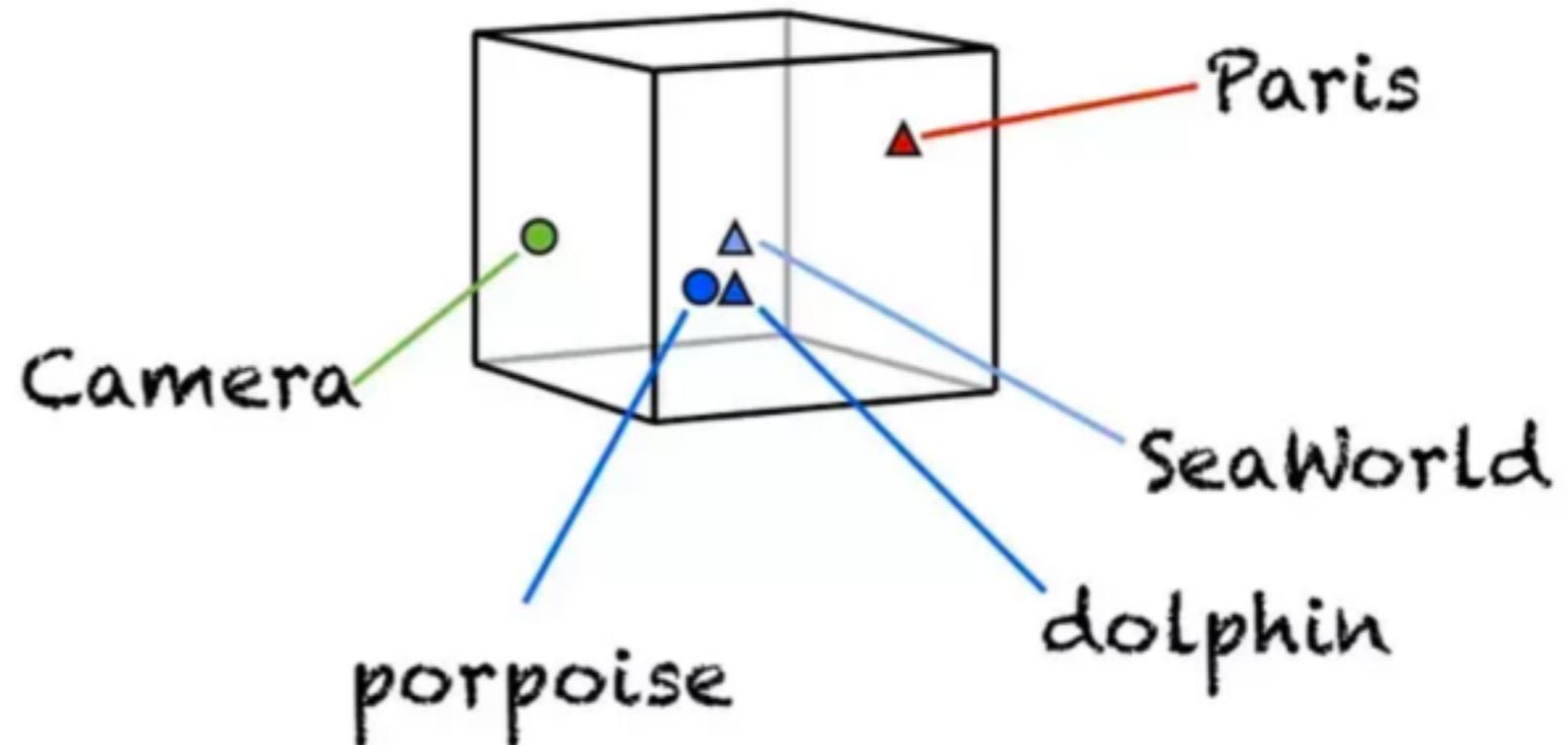
- Word2Vec
- Doc2Vec
- Sent2Vec

▶ What is word embedding ?

- Projection of each word/document/sentence in a very high dimensional space (we fixed dimension at 300)
- In this space, each word is given coordinates such that words with common sense are close one an other

▶ Python library Gensim, pre-trained by Google

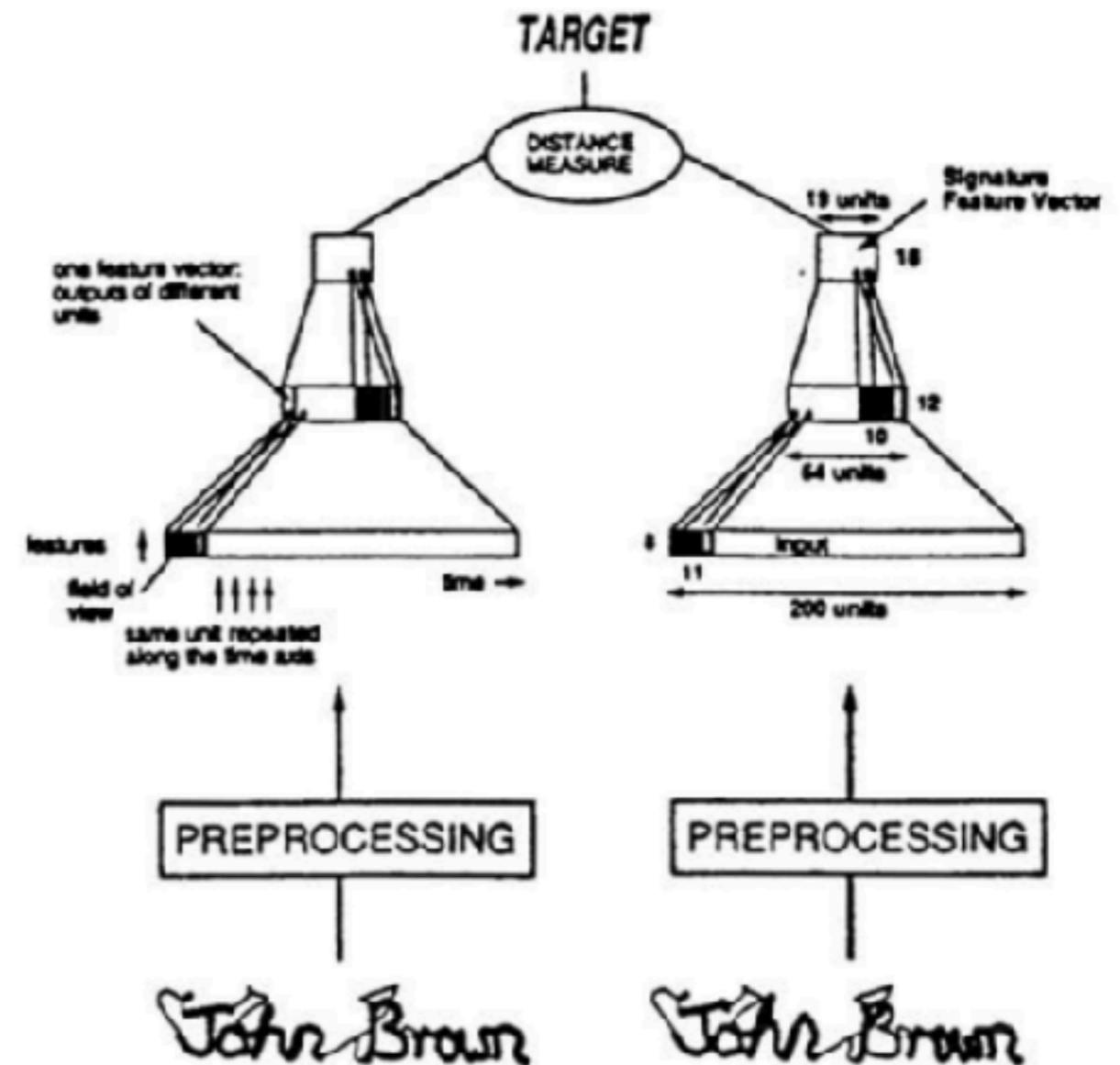
▶ Allows the following: PARIS - FRANCE + ENGLAND → LONDON



FIRST APPROACH: DEEP LEARNING

- ▶ Siamese Neural Network:
 - Two parallel networks
 - Same weights are trained with two inputs.
 - Dense layer to connect the two nets
 - Softmax activation on dense layer

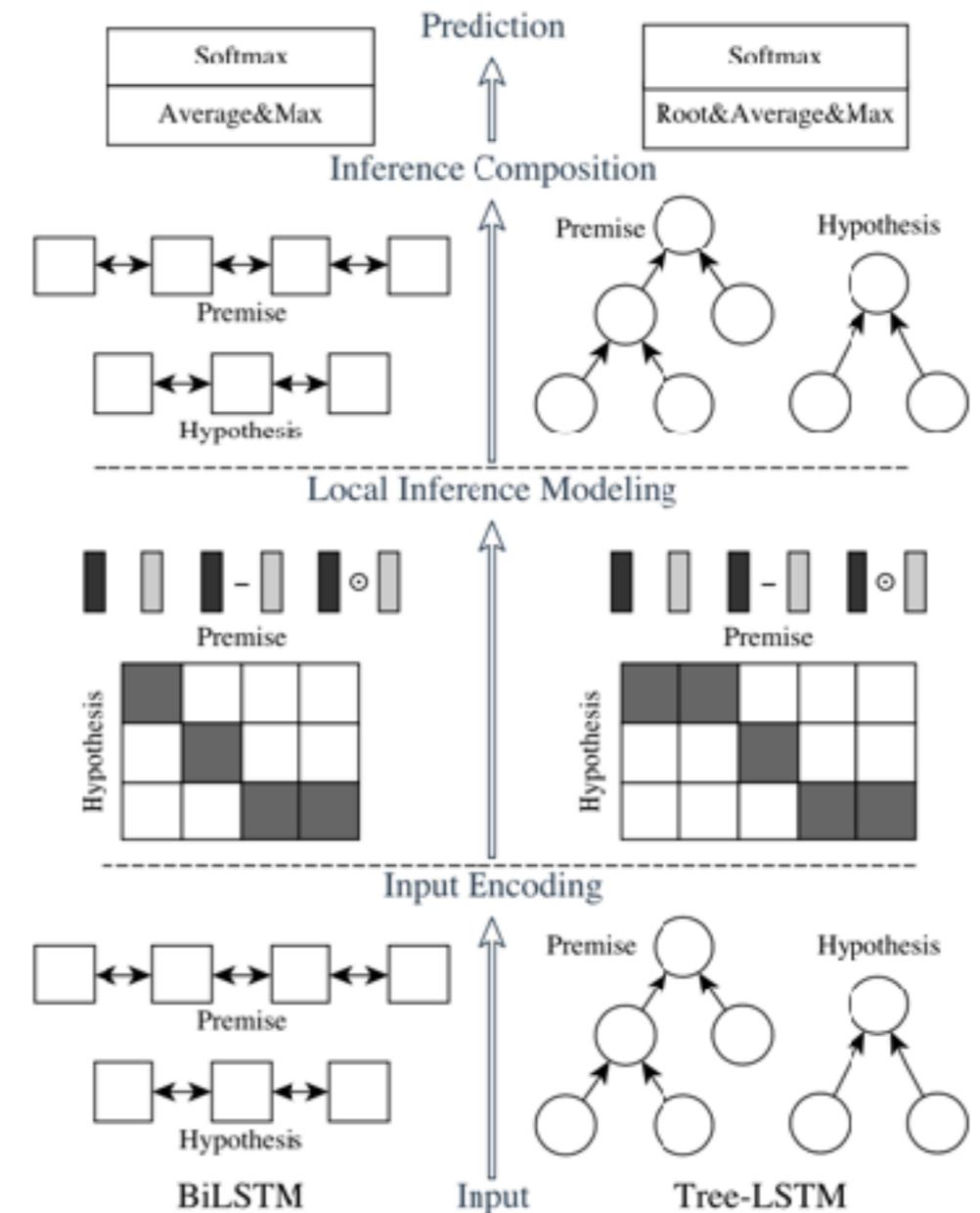
$$P(y = j | \mathbf{x}) = \frac{e^{\mathbf{x}^T \mathbf{w}_j}}{\sum_{k=1}^K e^{\mathbf{x}^T \mathbf{w}_k}}$$



Siamese network illustration

FIRST APPROACH: DEEP LEARNING

- ▶ Decomposable attention Neural Network:
 - ▶ (<https://arxiv.org/abs/1606.01933>)
 - ▶ Learn on word alignments
 - ▶ Detection of contradictory sentences
- ▶ ESIM
 - ▶ (<https://arxiv.org/abs/1609.06038>)



ESIM illustration

SECOND APPROACH: NLP

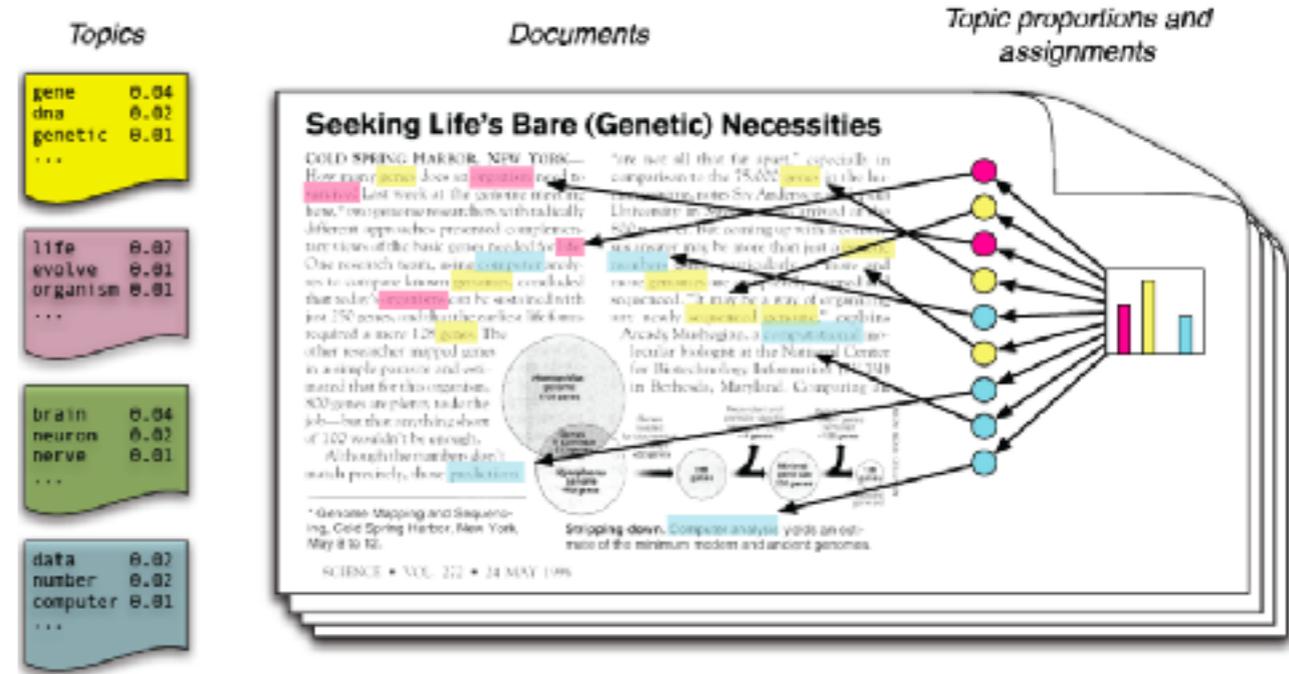
- ▶ Similarity measures on LDA (Latent Dirichlet Allocation) and LSI (Latent Semantic Indexing) measures.
- ▶ Similarity measures on bags of character n-grams (TFIDF reweighed or not) from 1 to 8 grams.
- ▶ A lot of edit distance between character strings, such as Levenshtein distance, Jaro-Winkler distance, Bray-Curtis distance etc...
- ▶ Percentage of common tokens sized from 1 to 6, when question ends the same. Same work when questions starts the same.
- ▶ Length of questions, difference of length
- ▶ Number of capital letters, question marks, etc...

$$w_{x,y} = tf_{x,y} \times \log\left(\frac{N}{df_x}\right)$$

TF-IDF

Term x within document y

$tf_{x,y}$ = frequency of x in y
 df_x = number of documents containing x
 N = total number of documents



- ▶ Indicators for questions 1 and 2 starting with « Can », « Are », « Do », « Where » etc...
- ▶ Dictionaries on countries and cities to fuzzy match them (example : Paris 12, and Paris 8 → Paris)

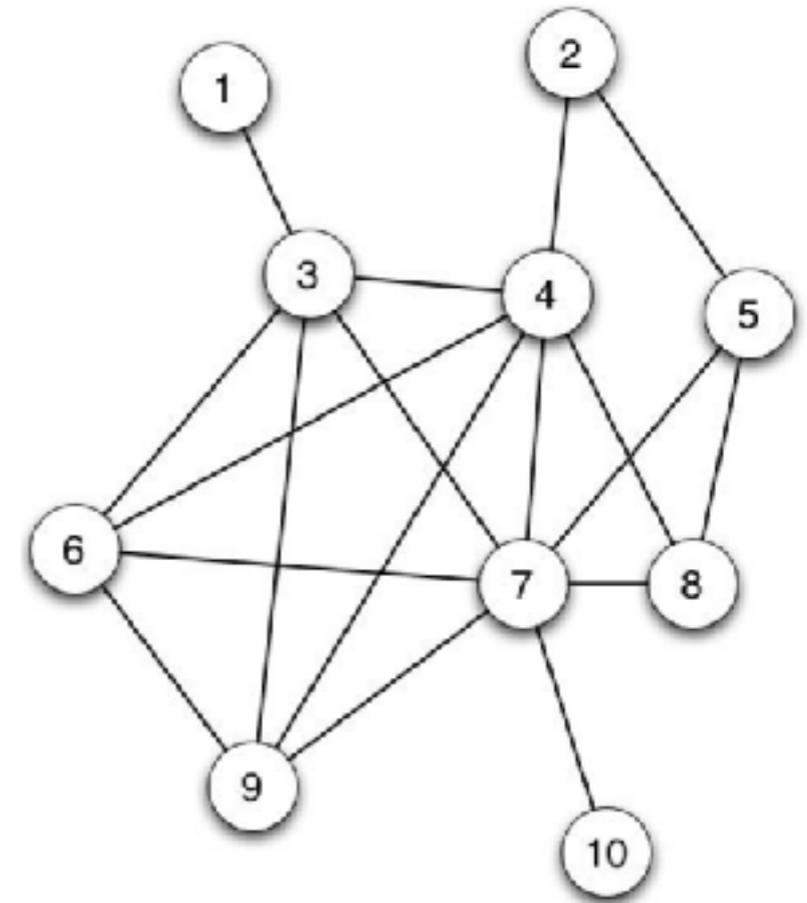
THIRD APPROACH: GRAPHICAL MODELS

▶ We built the following graph:

- Nodes: Questions
- Edges: Question pairs
- With train and test concatenated

▶ Why ?

- Question pairs are pre-selected by a Quora's internal model
- A lot of signal can be extracted from frequently asked questions



THIRD APPROACH: GRAPHICAL MODELS

► For each pair of questions, we compute:

- Min/Max/Intersection number of neighbors
- Min/Max/Intersection of neighbors of order 2 (neighbors of neighbors), which aren't neighbors of order 1
- Min/Max/Intersection of neighbors of order 3, which aren't neighbors of order 2 nor order 1
- Shortest path from question 1 to question 2 when the edge is cut

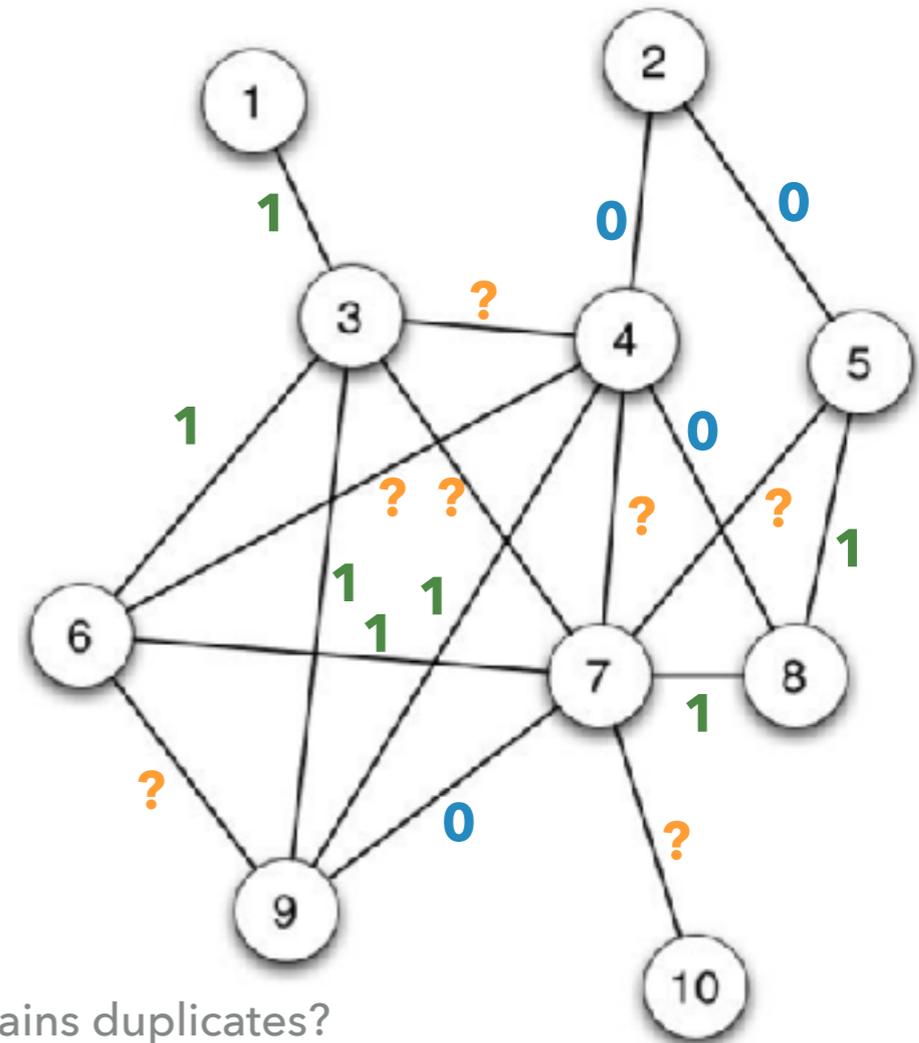
► For each connex component in the graph, we compute:

- Number of edges and nodes
- % of pairs in train set
- % of duplicated pairs in the component

► Triangles and path features:

- Triangle rule: $1/1 \rightarrow 1$ and $1/0 \rightarrow 0$
- Indicator: Is there a path between the two question, which only contains duplicates?

► We re-computed above features on the weighted graph, weighted by our best model's predictions



MODELIZATION: STACKING

- ▶ **! WARNING:** This kind of modelization is very powerful, but requires to be made **properly**, there is a **HUGE** overfitting risk.

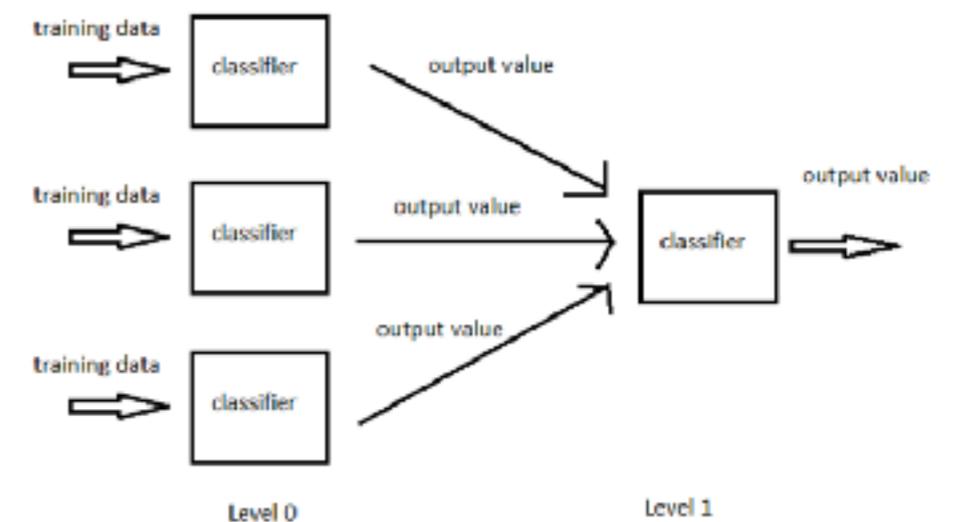
- ▶ What is stacking?

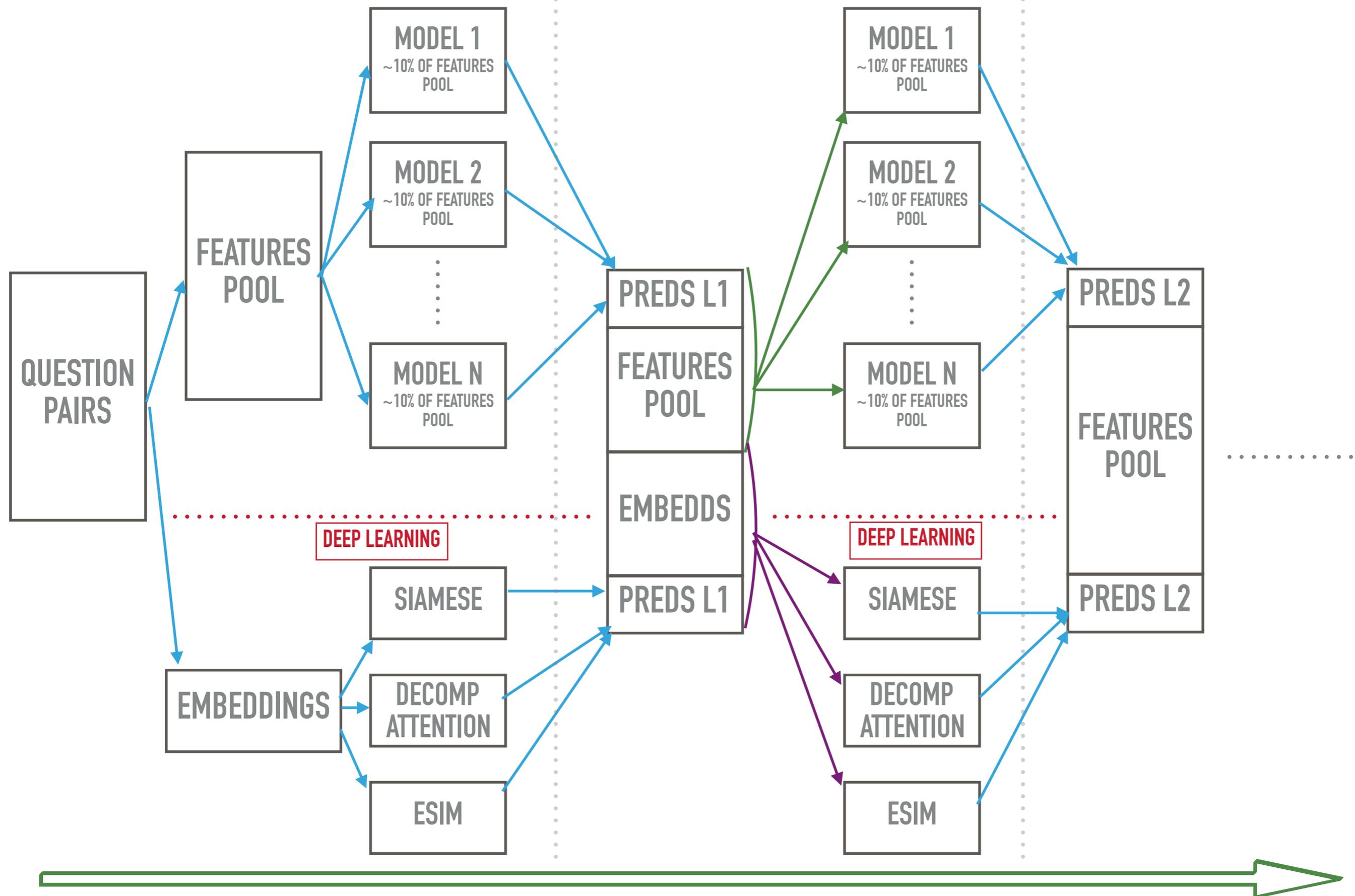
- Models chaining
- Predictions of models becomes input of next models
- We make multiple layers, the first one takes our features as inputs, next layers take the same inputs + models' predictions

- ▶ Why stacking?

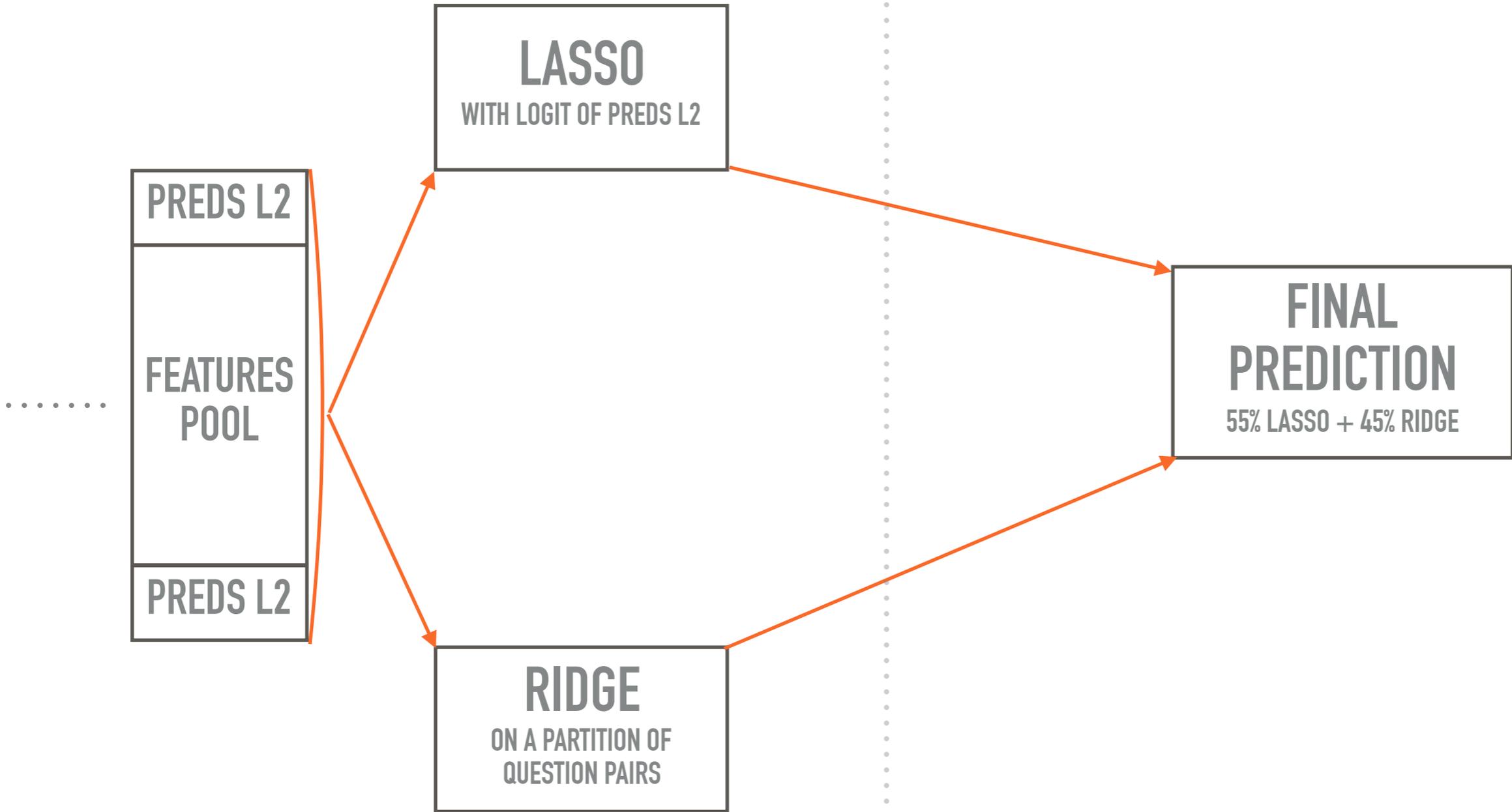
- Some models are better than others on different parts of the data
- Higher order layers' models will select the best models according to the dataset's properties

Concept Diagram of Stacking





! WARNING: HUGE OVERFITTING RISK HERE



OUR STACKING IN A NUTSHELL

- ▶ 4 layers stacking
- ▶ Layer 1, ~300 models including:
 - Our two main Deep Learning architectures
 - A lot of classical algorithms, such as XGBoost, LightGBM, ExtraTrees, RandomForests, KNN, Logistic Regression, Naive Bayes, Stochastic Gradient Descent etc.
- ▶ Layer 2, ~150 models, including the same algorithms used in layer 1, trained with our base features, and Layer 1 predictions
- ▶ Layer 3, 2 models:
 - Lasso, with logit preprocessing on Layer 2 predictions
 - 3 Ridges, on a partition of the data in 3 chunks, trained each with the 3 Spearman's least correlated Layer 2 predictions
- ▶ Layer 4, blend of our layer 3 models, with coefficients 55/45 respectively, based on our CV score

CONCLUSION

- ▶ We have around 450 models to generate the final submission
- ▶ At least 1 week to run all our models on huge hardware (10 GPU machines with 32Go RAM + 80 CPU machines with 120Go RAM).
- ▶ Our approaches' diversity deeply helped our stacking to optimize the LogLoss.
- ▶ This model cannot be used directly by Quora since it is way too complex → Kaggle competitions are quite disconnected from a production environment.
- ▶ What's interesting for Quora is the way we analyzed their data, giving them a lot of insight for their own projects.

THANK YOU!

Question pairs time!