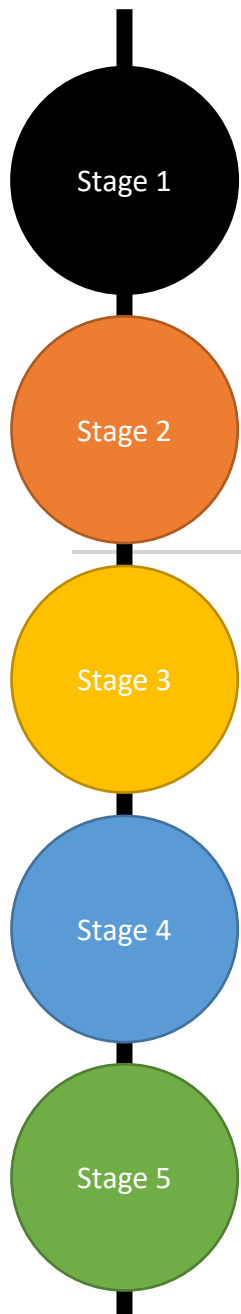


SIOP Machine Learning Competition



Presenter: **Mustafa Akben, Ph.D.**



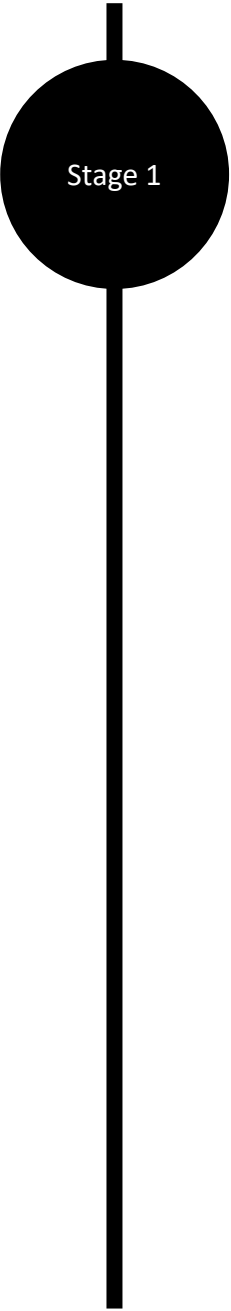


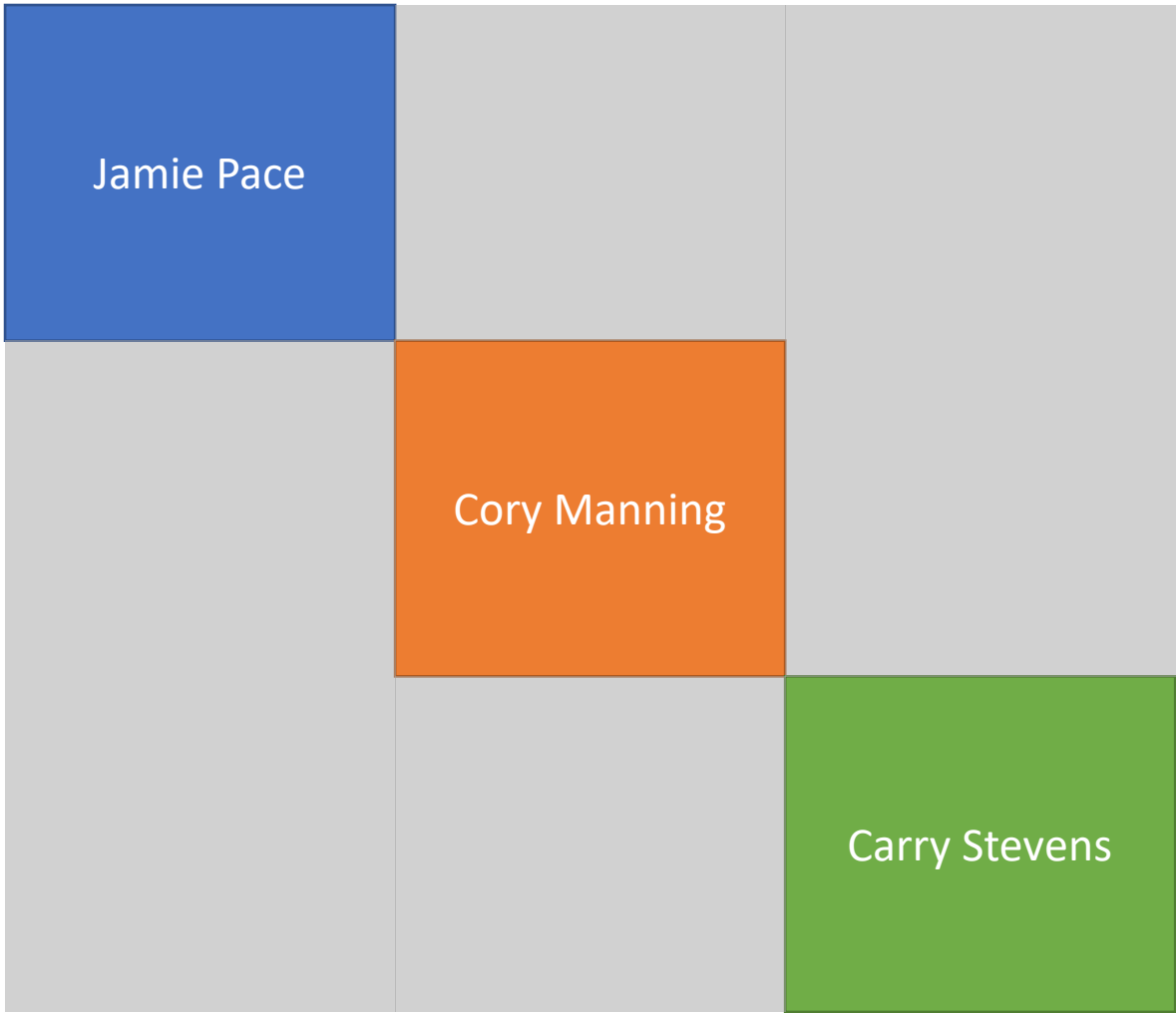
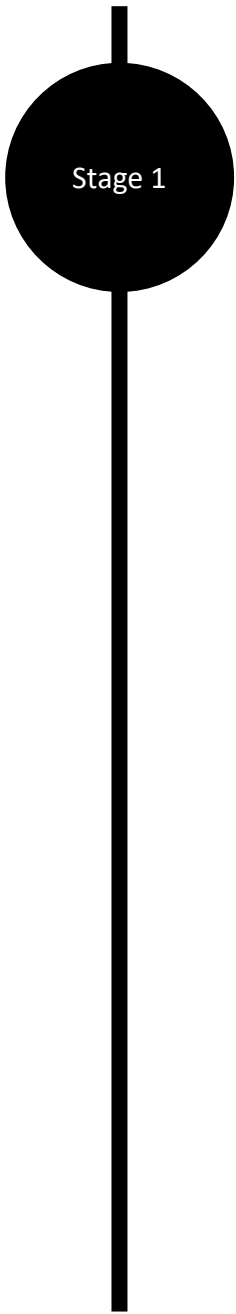
Overall Producer

- Stage 1 — Data Cleaning
- Stage 2 — Model Exploration
- Stage 3 — Feature Engineering
- Stage 4 — Model Building
- Stage 5 — Final Submission

Stage 1 – Data Cleaning: Dirty Data

- Understanding Data
- A lot of NA's
- Use them for advantages – Missingness based cluster analysis to identify the differences in data
- Two different assessment center simulations with three different player names: Jamie Pace, Cary Stevens, and Cory Manning



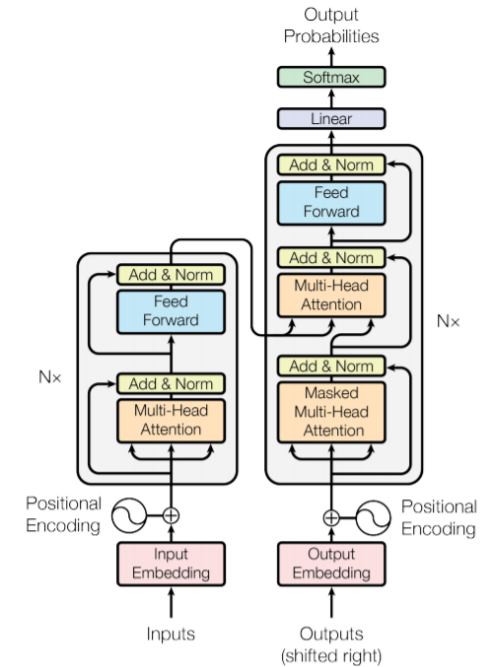


Stage 1 – Data Cleaning: Dirty Data

- Correct the whitespace issues (Good or bad?)
- Extract and create six new columns from the final exercise (meta-reflective parts of the work sample)
- Regex based data extraction
- Missing ratings imputed with K-nn

Stage II: Model Architecture Exploration

- **Deep learning with pre-trained (Transfer learning)**
 - SBERT – T5 – Universal Encoder
 - Initial score: .38 - .40
- Various Deep Learning Model Architectures
 - Self-variables selection network
 - Gated linear units
 - CONV1D Networks
 - LSTM
 - Transformers
- Drawbacks
 - Data Size
 - Very Complex for this small data
- Solutions
 - Add regularizations (Drop out layers, Alpha Drop-outs, or L1-L2)
 - Schedule learning rates
 - Generate Syntactic data
 - with back-translation, added noise, swap noise, T5-based paraphrase model



Stage II: Model Architecture Exploration

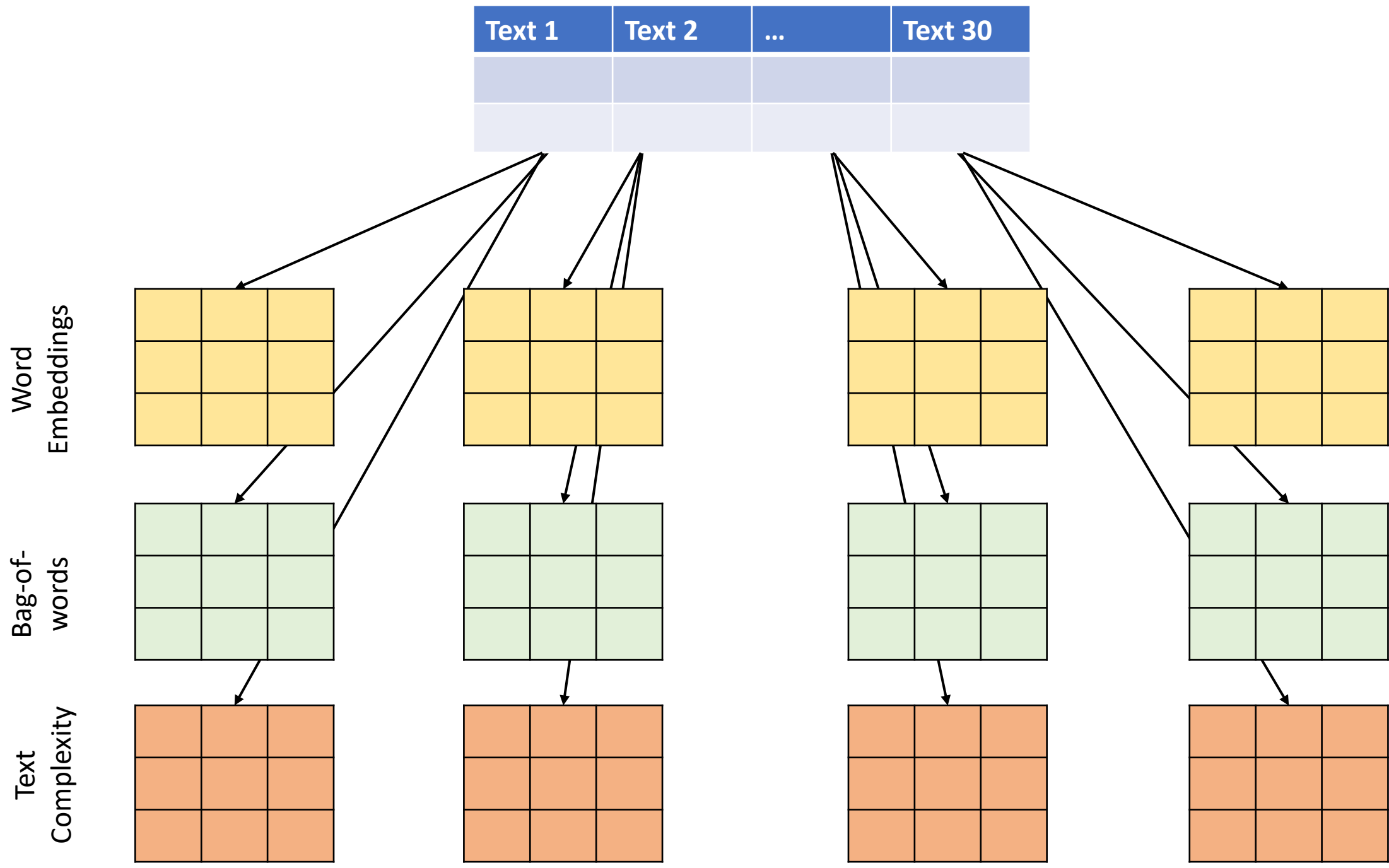
Stage 2

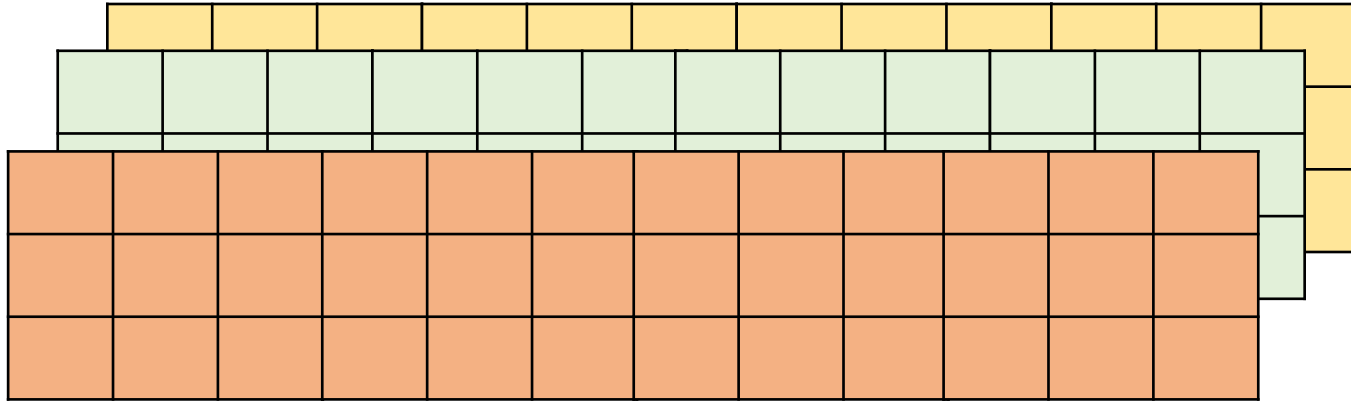
- **Bidirectional Encoder Representations from Transformers (BERT)**
 - Train a BERT model with a large corpus including the corpus we have
 - HuffPost + CNBC Data + the competition data
- Masked-language model with TPU
- Use the fined-tuned model's embeddings for the down streamed task
- **Results : ~.41**
- Drawbacks
 - Computational Heavy
 - Complicated and Complex
- Solution
 - Increase the dataset size

Stage III: Feature Engineering

- INSTRUCTION embeddings for semantic meaning
- Bag-of-words with N-grams for each columns for key words
- Text complexity and text statistics such as perplexity, rare word counts, SMOG scores or other text readability scores for each columns

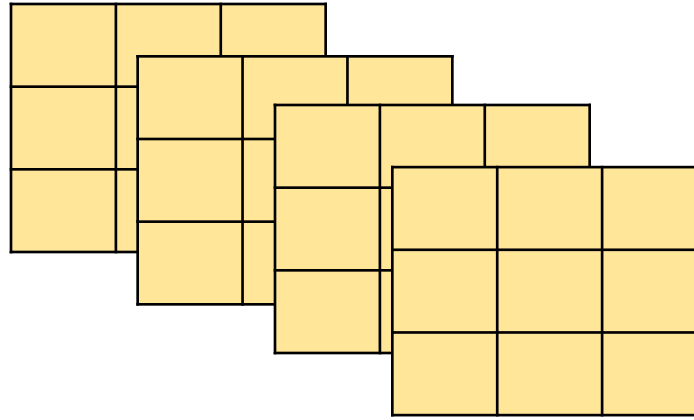
Stage 3





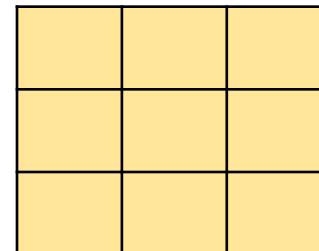
Dimension : (N Observations, 30, 25 000)

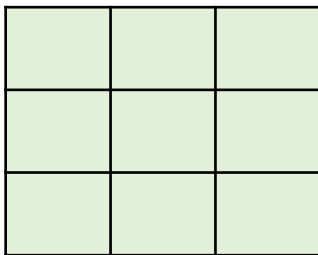
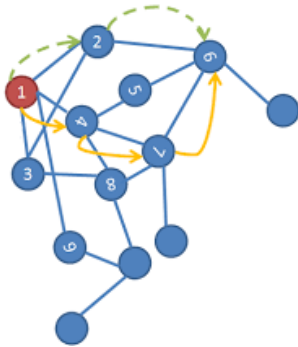
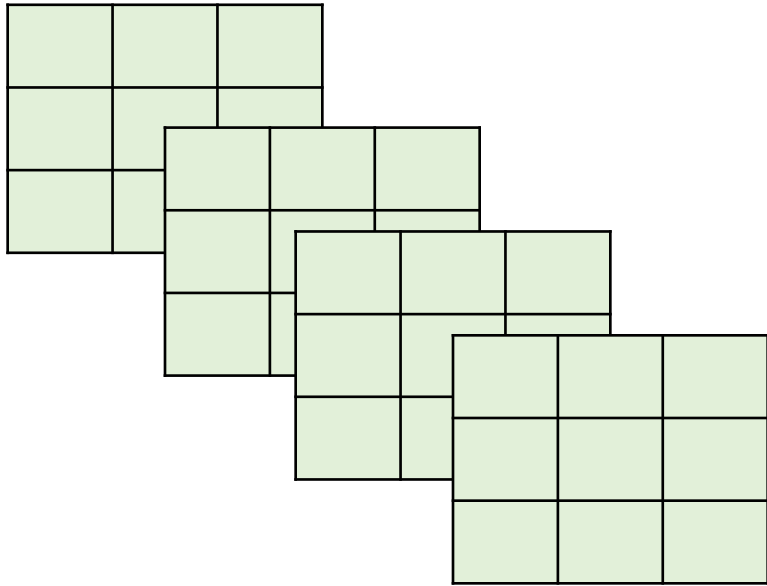
Curse of Dimensionality : $N < p$



**Semantic
Embeddings**
– Average of
Embeds over
Dimensions

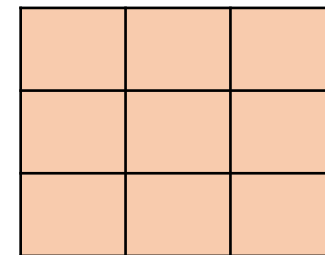
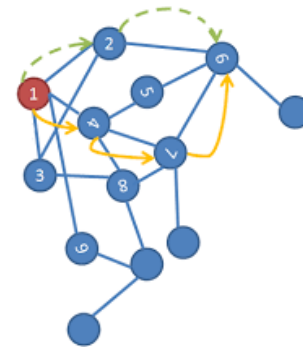
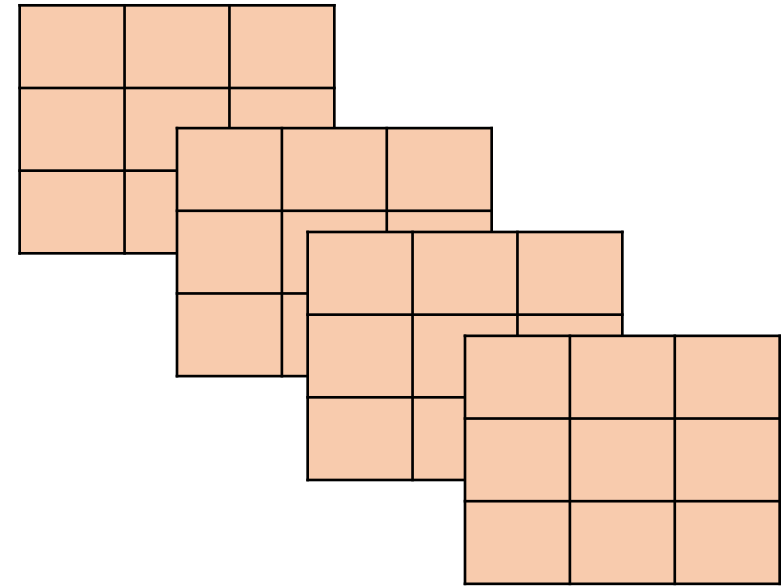
$$\sum_{i,j,k}^{J=n} \frac{x_{ijk}}{N_j}$$





**Bag-of-
words and
Text
Complexity**

**Diffusion
Map
Embeddings**



Dimension : (N Observations, 2000~)

Stage IV: Model Building

- Multiple Ensembled model for each ratings
- K-fold cross validation (k = 15)
 - Validation variation is very large
- Select K-folds based on energy distance

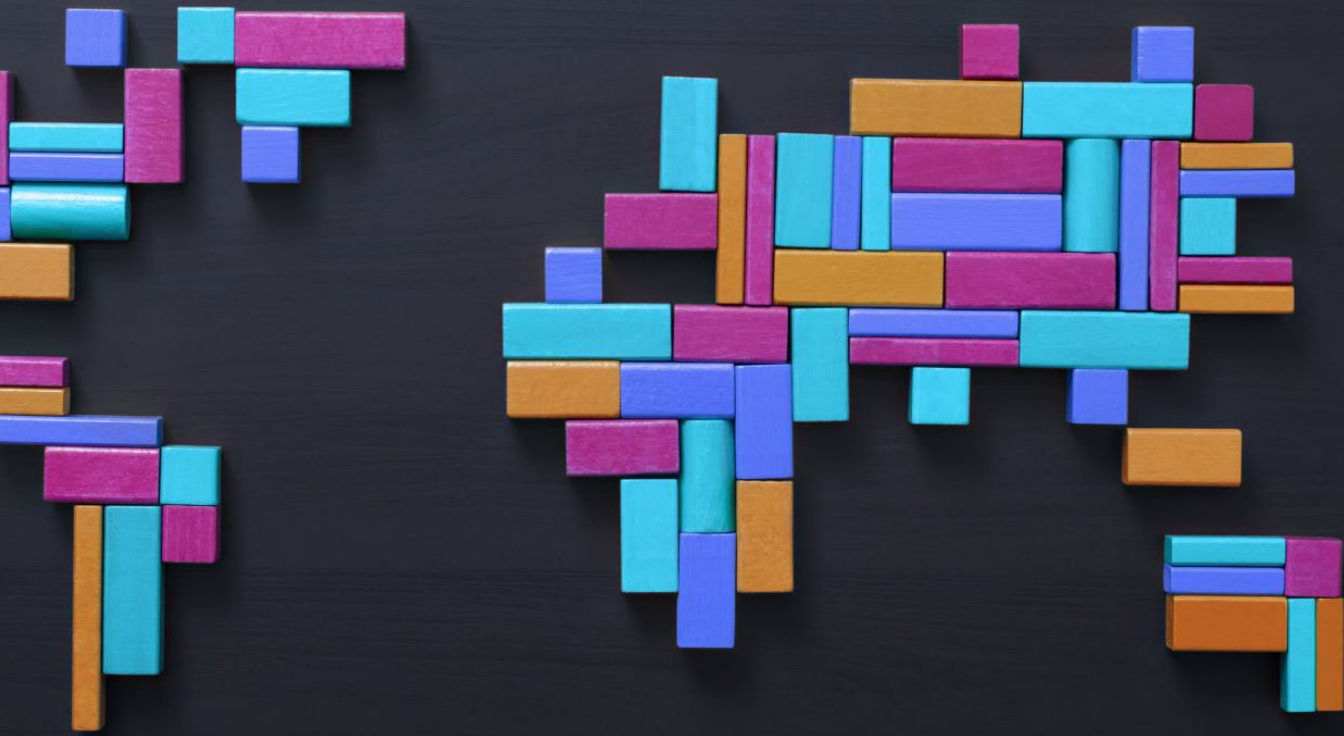
$$D^2(F, G) = 2E \|X - Y\| - E \|X - X'\| - E \|Y - Y'\| \geq 0,$$

Stage IV: Model Building

- Trained with various model with H2O in Python
 - XGBoost, GBM, GLM, Random Forest, Deep Learning
 - 100 Different Model – 4 to 5 Hr Training
 - Ensembled scores calculated with penalized linear models such as Elastic Net
 - Combined the top 3 ensembled model as a **simple average voting schema based on the cross-validated r^2 value**
 - Development Phase Score : **.51**

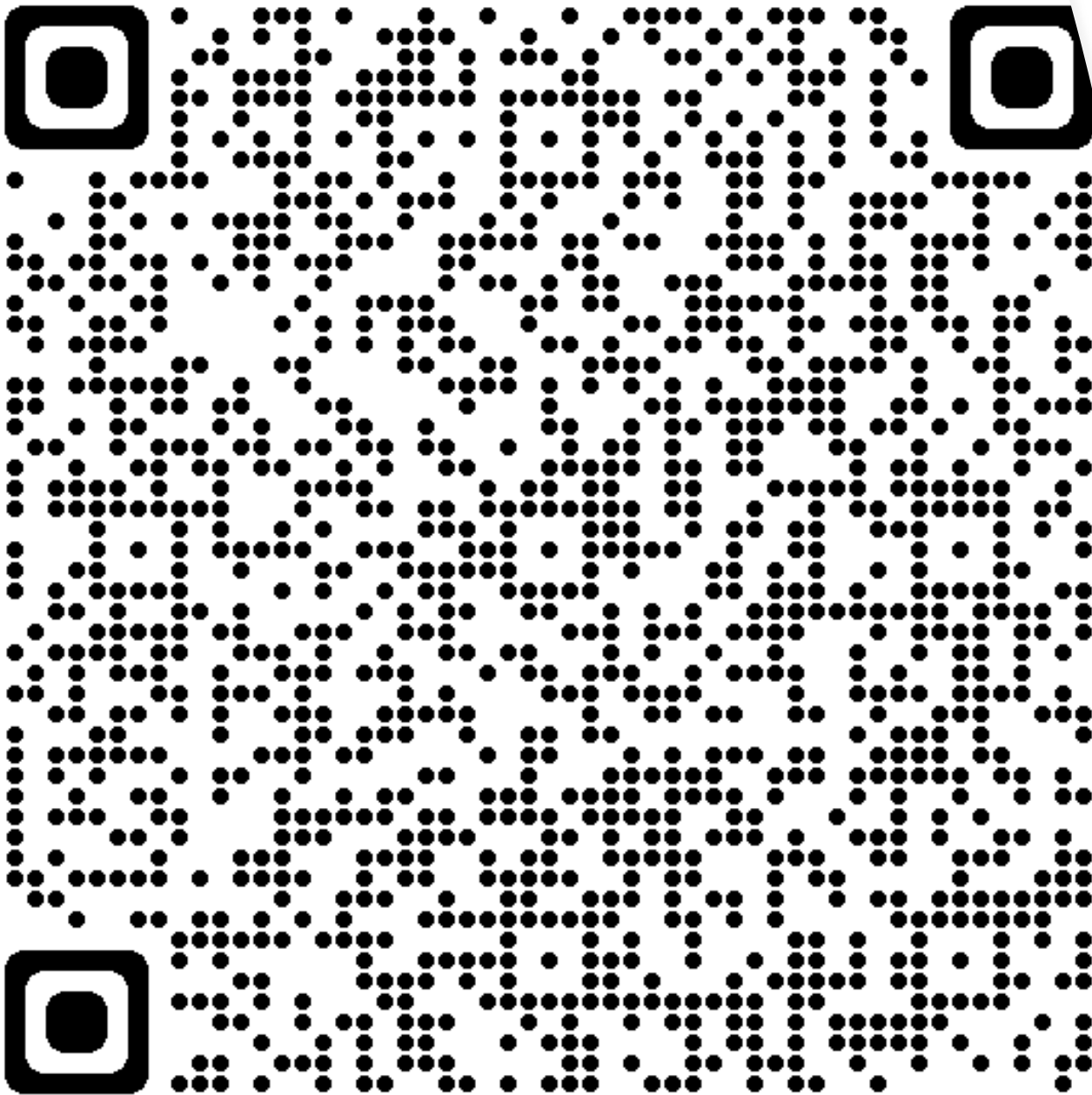
Stage V: Submission

- Test data processed in the same way
- Due to the birth of my first child, I can only submit 4 out of 10 test submissions
- Issues
 - Time management
- Solution
 - Find good collaborators to work with



Lesson Learned

- Small Data = Feature Engineering
- Semantic, Structure, and Complexity of Text
- Dimensionality Reduction with Diffusion Map – Preserves the local manifold geometry
- Time Management and Collaboration



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Thank you for your
attention!

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