Using scores to improve language modelling of movie plot summaries

J. Sáez Gómez, R. vo Heijden, F. Stablum

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## Presentation outline

Using scores to improve language modelling of movie plot summaries

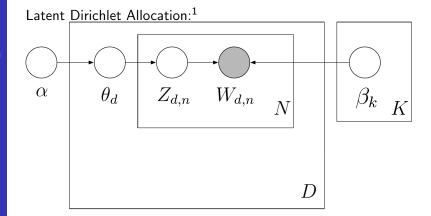
### Problem formulation

Using scores to improve language modelling of movie plot summaries

- Is there any correlation between the score of a movie and the contents of its script?
- Can we use the score to better model a movie corpus?

## Models

Using scores to improve language modelling of movie plot summaries

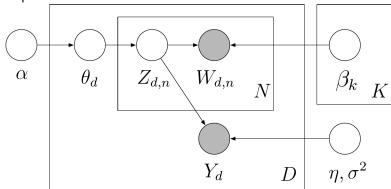


 $<sup>^1</sup>$ Image taken from the paper "Supervised topic models" by David M. Blei and Jon D. McAuliffe (2007)  $^2$ 

## Models

Using scores to improve language modelling of movie plot summaries

J. Sáez Gómez, R. vo Heijden, F. Stablum Supervised Latent Dirichlet Allocation:<sup>2</sup>



 $<sup>^2</sup>$  Image taken from the paper "Supervised topic models" by David M. Blei and Jon D. McAuliffe (2007)  $\ \ \bigcirc$ 

Our collapsed Gibbs sampler:

$$p(z_{di} = k \mid Z^{\setminus i}, S, W, \alpha, \beta, \eta, \sigma) \propto \left[ \prod_{k'} \frac{\prod_{w} \Gamma(N_{k'w}^{\setminus i} + \mathbb{I}(k' = k \land w = w_{di}) + \beta)}{\Gamma(N_{k'}^{\setminus i} + \mathbb{I}(k' = k) + W\beta)} \right] \times \mathcal{N} \left( s_{d} \mid \eta^{T} \cdot \frac{N_{dk'}^{\setminus i} + \mathbb{I}(k' = k)}{N_{d}}, \sigma \right) \prod_{k'} \Gamma(N_{dk'}^{\setminus i} + \mathbb{I}(k' = k) + \alpha)$$
Movie score term

Better implemented in log-space probabilities to avoid numerical problems.

J. Sáez Gómez, R. v Heijden, F. Stablum Estimating the global score hyperparameter  $\eta$ :

$$\eta_k^{new} \leftarrow (1 - \gamma)\eta_k^{old} + \gamma \frac{\sum_d \frac{N_{dk}}{N_d} \left(s_d - \sum_{k' \neq k} \eta_{k'}^{old} \frac{N_{dk'}}{N_d}\right)}{\sum_d \left(\frac{N_{dk}}{N_d}\right)^2 + \varepsilon}$$

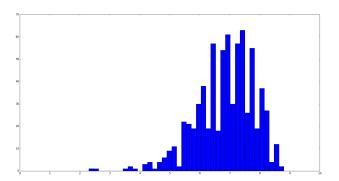
#### Where:

- $1 \gg \gamma > 0$  in order for the previous series to converge.
- $1 \gg \varepsilon > 0$  is a smoothing constant.

#### Dataset

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- We made scripts to crawl http://www.imsdb.com/ for movie scripts and then search http://www.imdb.com/ for movie scores and plot summaries.
- We got a database with  $\approx$  700 movies.
- Movie score distribution (from 0 to 10):



#### Dataset

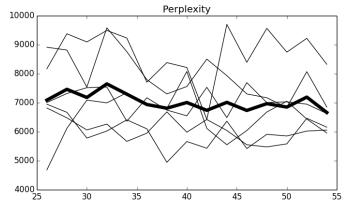
Using scores to improve language modelling of movie plot summaries

- Tokenization  $\rightarrow$  stemming  $\rightarrow$  pruning
- We prune words appearing only on a single movie (avoids overfitting) or within a stop list.
- Total number of tokens  $\approx 12.7 \cdot 10^6$
- Number of unique tokens  $\approx$  35000
- $lue{}$  Average number of tokens within a movie summary pprox 75
- $\blacksquare$  Average number of tokens within a movie script  $\approx 18000$

## Results

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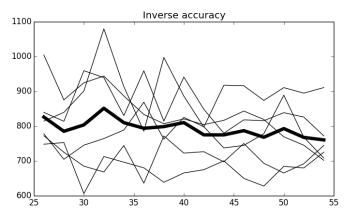
- Initial selection of 30 movies (20 training / 10 testing)
   with balanced scores.
- Using 10 topics and no scores.



#### Results

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- Initial selection of 30 movies (20 training / 10 testing) with balanced scores.
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## Discussion

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## Challenges

Using scores to improve language modelling of movie plot summaries

- Improve speed of the collapsed Gibbs sampler.
- Use the movie scripts instead of the movie summaries.
- Use both the movie scripts and summaries.
- Incorporate more information into the model, such as the movie genre.