Schneider Electric's Technical Project

Topic: Netflix Model

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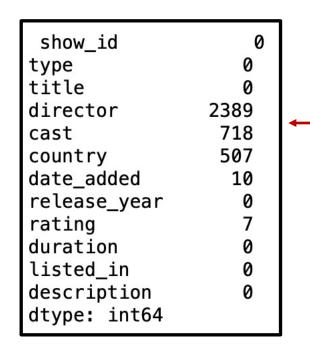


Dataset - Overview

Data	columns (total 12 columns):					
#	Column	Non-Null Count	Dtype			
0	show_id	7787 non-null	object			
1	type	7787 non-null	object			
2	title	7787 non-null	object			
3	director	5398 non-null	object			
4	cast	7069 non-null	object			
5	country	7280 non-null	object			
6	date_added	7777 non-null	object			
7	release_year	7787 non-null	int64			
8	rating	7780 non-null	object			
9	duration	7787 non-null	object			
10	listed_in	7787 non-null	object			
11	description	7787 non-null	object			
dtypes: int64(1), object(11)						
memory usage: 730.2+ KB						

No. of Rows: 7787

No. of Columns: 12

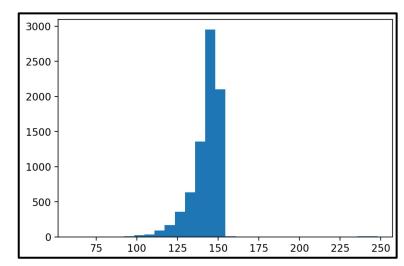


Missing Data

No Duplicate Entries!!

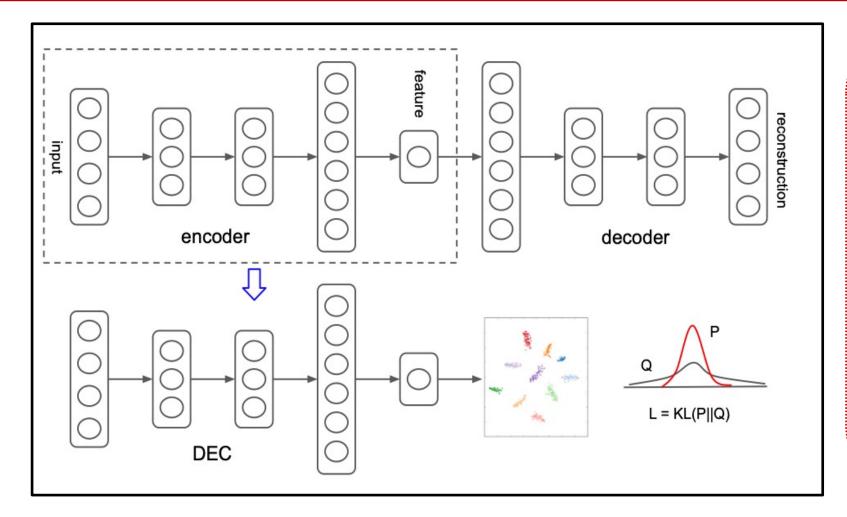
Text - Processing

- Using 'description' text of movies for clustering purposes.
- Tokenizing data with 10000 most frequent words
- Padding resulting sequence to 1500 length.
- Scaling features to the range between 0 and 1



Distribution of Sequence Length

Clustering - Framework

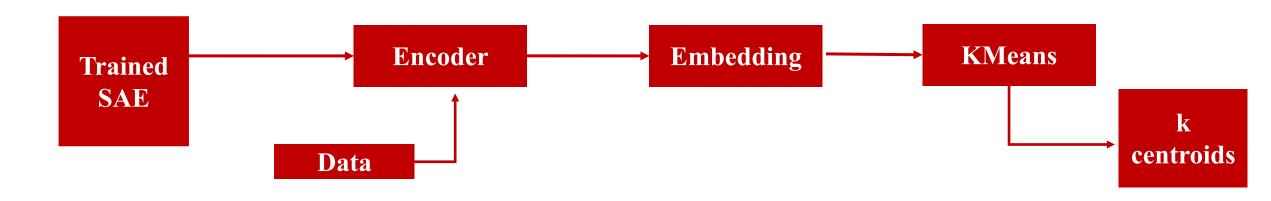


- $\bullet f_{\theta}: X \longrightarrow Z$
- Clusters data by learning **k** clusters center in Z
- Parameter initialization with a deep autoencoder
- Parameter Optimization

Xie, Junyuan et al. "Unsupervised Deep Embedding for Clustering Analysis." ArXivabs/1511.06335 (2016): n. pag.

Clustering – Parameter Initialization

- DEC initialized by stacked Auto Encoder
- AE trained by minimizing the MSE loss
- Discard *decoder* after training to use *encoder* for initial mapping: $X \rightarrow Z$
- k initial centroids obtained by kmeans on embedded data points Z





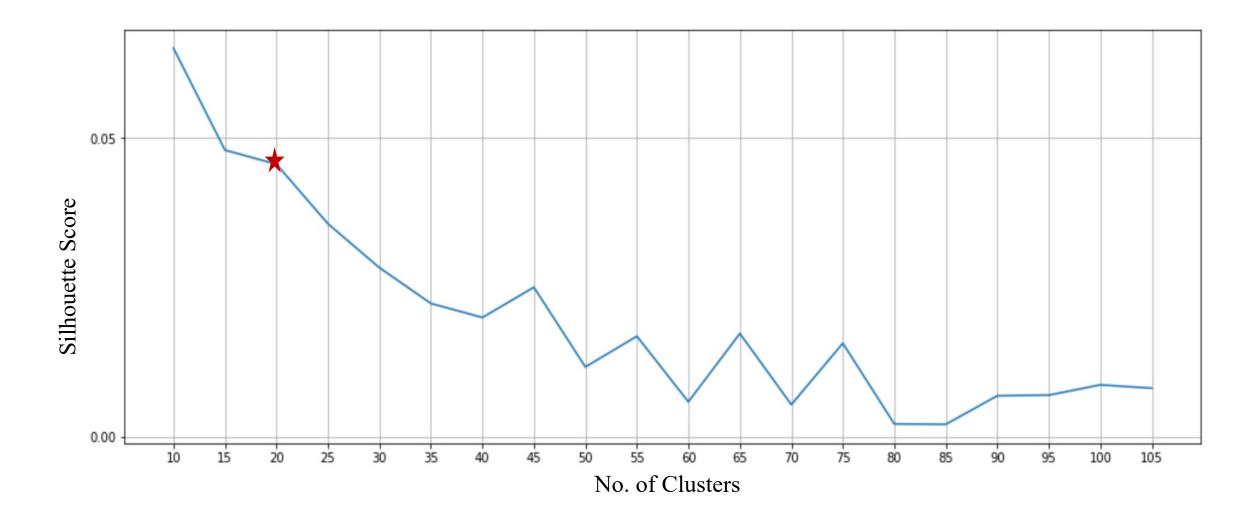
Clustering – Parameter Optimization

- Parameter Optimization by alternating two steps:
 - **Soft Assignments** between embedded points and cluster centroids
 - **Refinement** of embedding and cluster centroids by learning target distribution
- Soft Assignments: Probability of assigning sample i to cluster j

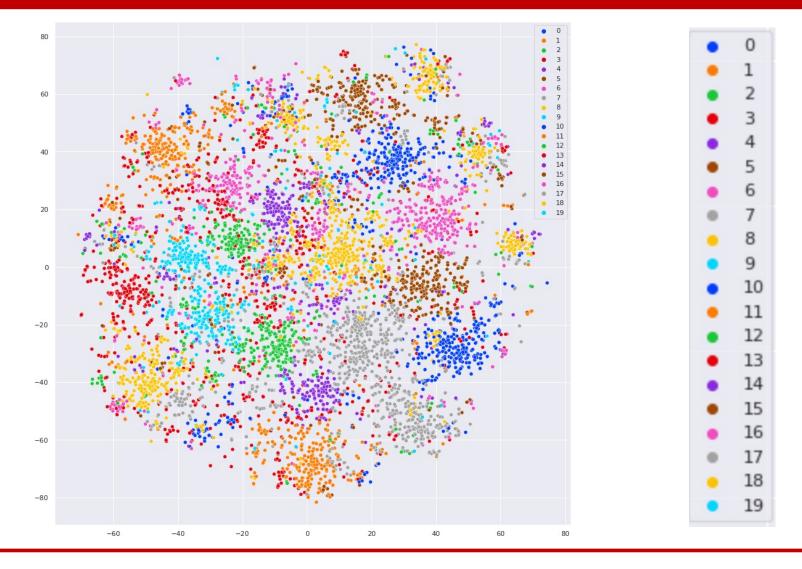
$$q_{ij} = \frac{(1 + \|z_i - \mu_j\|^2 / \alpha)^{-\frac{\alpha+1}{2}}}{\sum_{j'} (1 + \|z_i - \mu_{j'}\|^2 / \alpha)^{-\frac{\alpha+1}{2}}}$$

• *Refinement* of clusters involves matching of soft assignment to target distribution via minimizing KL divergence loss.

Clustering – Optimal Number of Clusters

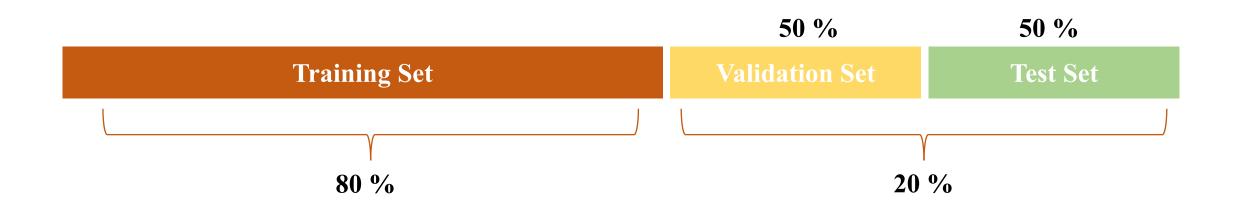


Clustering – Results





Classification – Data Splitting



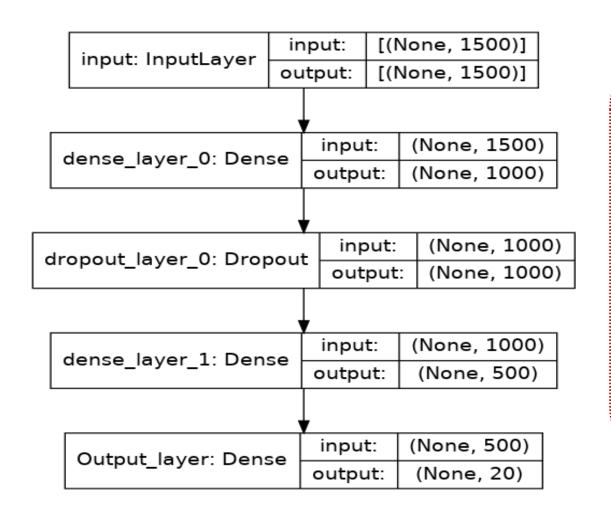
Training Set: X {6229, 1500}, Y {6229, 20}

Validation Set: X {779, 1500}, Y {779, 20}

Test Set: X {779, 1500}, Y {779, 20}

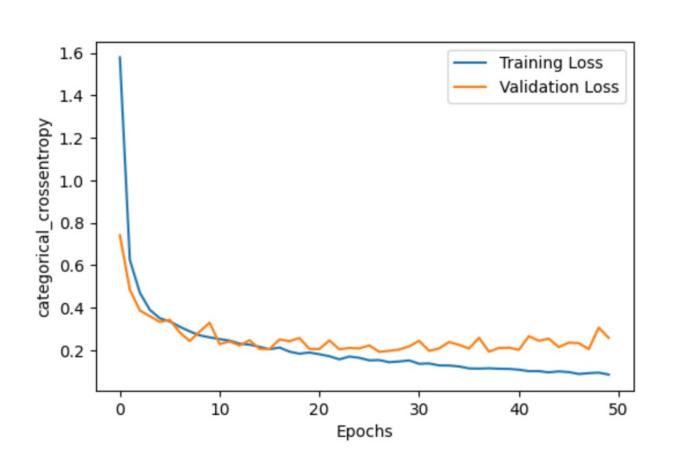


Classification – Deep Neural Network



- Activation Function: ReLU
- Activation Function (Output): Softmax
- Batch Size: 128
- **Epochs:** 50
- Optimizer: RMSprop
- Learning Rate: 0.001
- Momentum: 0.9

Classification – Deep Neural Network Performance

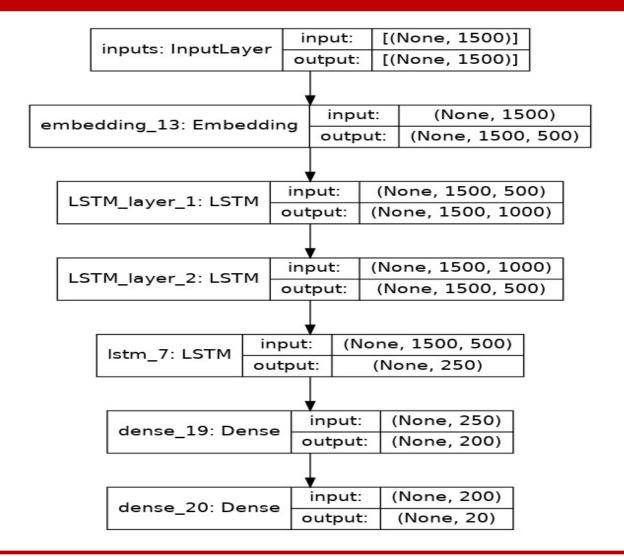


	precision	recall	f1-score	support
0	0.97	0.88	0.92	32
1	0.97	0.80	0.88	41
2	0.82	1.00	0.90	23
2	0.96	0.92	0.94	50
4	0.85	0.81	0.83	21
5	0.96	0.82	0.89	33
5 6	0.87	0.92	0.89	36
7	0.94	0.93	0.94	72
8	0.94	0.83	0.88	36
9	0.88	0.95	0.91	22
10	1.00	0.94	0.97	48
11	0.90	1.00	0.95	37
12	0.88	0.93	0.90	30
13	0.93	1.00	0.96	37
14	0.86	0.86	0.86	29
15	0.86	0.98	0.92	50
16	0.95	0.93	0.94	40
17	0.89	0.94	0.91	33
18	0.89	0.92	0.90	83
19	0.96	0.88	0.92	26
accuracy			0.92	779
macro avg	0.91	0.91	0.91	779
ghted avg	0.92	0.92	0.91	779

macro weighted

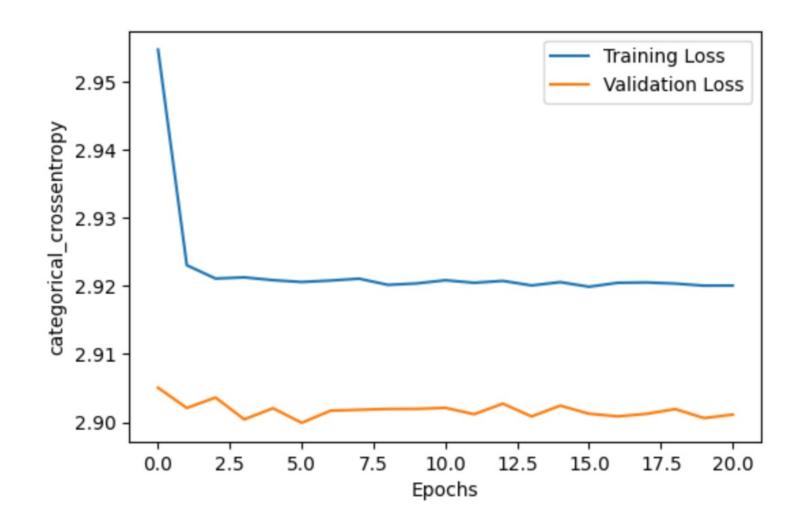


Classification – Deep LSTM Network



- Activation Function: ReLU
- Activation Function (Output): Softmax
- Batch Size: 128
- **Epochs:** 100
- Optimizer: SGD
- Learning Rate: 0.01
- **Momentum:** 0.9

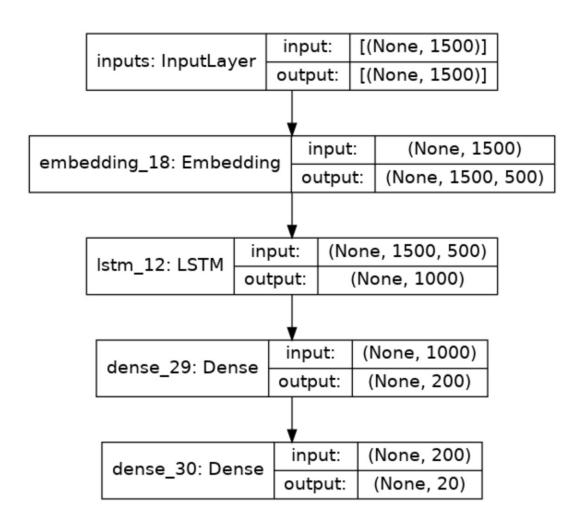
Classification – Deep LSTM Performance



High Bias!!

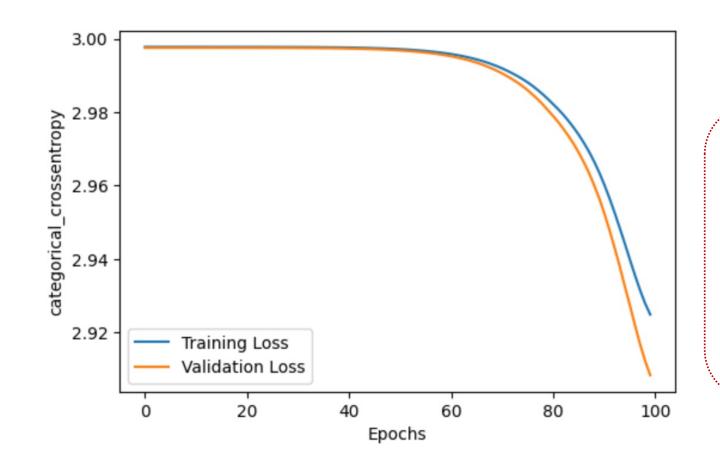


Classification – Shallow LSTM Network



- Activation Function: ReLU
- Activation Function (Output): Softmax
- Batch Size: 128
- **Epochs:** 100
- Optimizer: SGD
- Learning Rate: 0.1
- **Momentum:** 0.9

Classification – Shallow LSTM Performance



Slow Training!!

- Requires more training iterations.
- Losses tend to decrease significantly from 60th epoch. But stopped early due to no further improvements.

THANK YOU!!

