

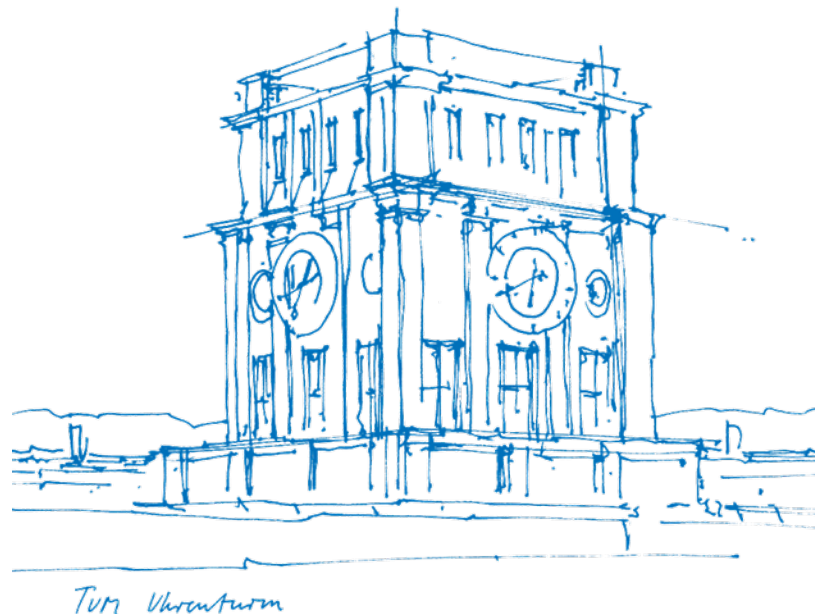
Global Explainability for Natural Language Processing Models

Guided Research - Final Presentation

B. Sc. Simon Klimek

Supervisor: M. Sc. Edoardo Mosca

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Background - eXplainable Artificial Intelligence (XAI)

- How does a machine learning model come to a conclusion?
- Simple (linear) vs. complex (non-linear) models
- Humans can not comprehend a deep neural network!

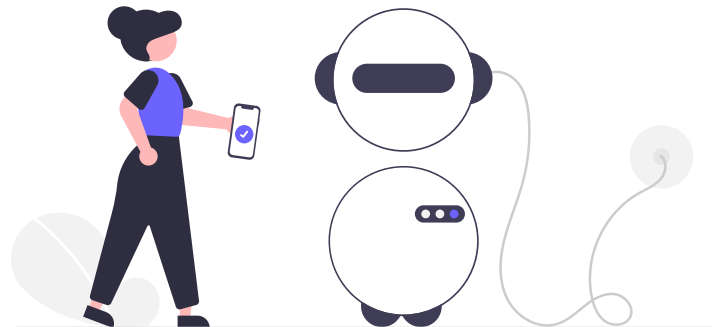


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Motivation

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 - 25% a degree of endorsement
 - 15% consistent pattern of endorsement
 - 10% had very high levels of endorsement

[Freeman et al., 2020]

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- “Higher levels of corona virus conspiracy thinking were associated with *less adherence to all government guidelines and less willingness to take diagnostic or antibody tests or to be vaccinated.*”

[Freeman et al., 2020]

Motivation



Figure: Capitol Riots - Jason Andrew for The New York Times [Williamson, 2021]

Related Work

Guidotti et al. [2018] - Survey giving us an overview of the state of the art XAI methods:

- **Model Explanation Problem (global)**
- Outcome Explanation Problem (local)
- Model Inspection Problem (global/local)
- Transparent Box Design Problem (global/local)

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-
- All global explainability methods are surrogates.
 - Listed no global XAI methods which can be used for NLP.

Related Work

Covert et al. [2020] - Understanding Global Feature Contributions With Additive Importance Measures

- Based on Shapley values
- Not method proposed for text based inputs
- Uses a group of pixels for images
- Use a closed unit (e.g. named entity)

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Here's tomorrow's Ancient Aliens lineup. Better wake up early. Starts at 7am eastern time in the USA.

The system is broken. It amazes me that a young child in kindergarten can be kicked out of school because a document isn't checked and dated for a vaccine. We are all wearing masks!

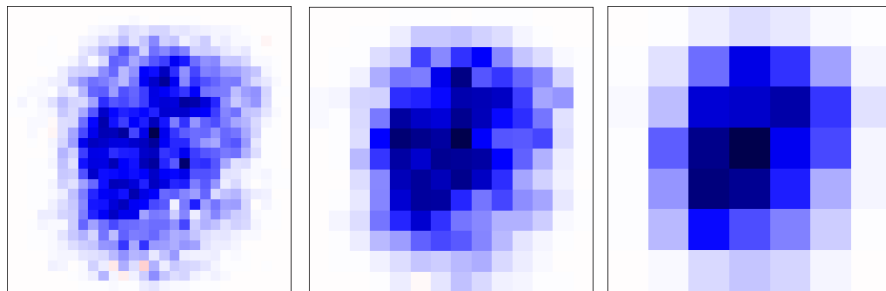
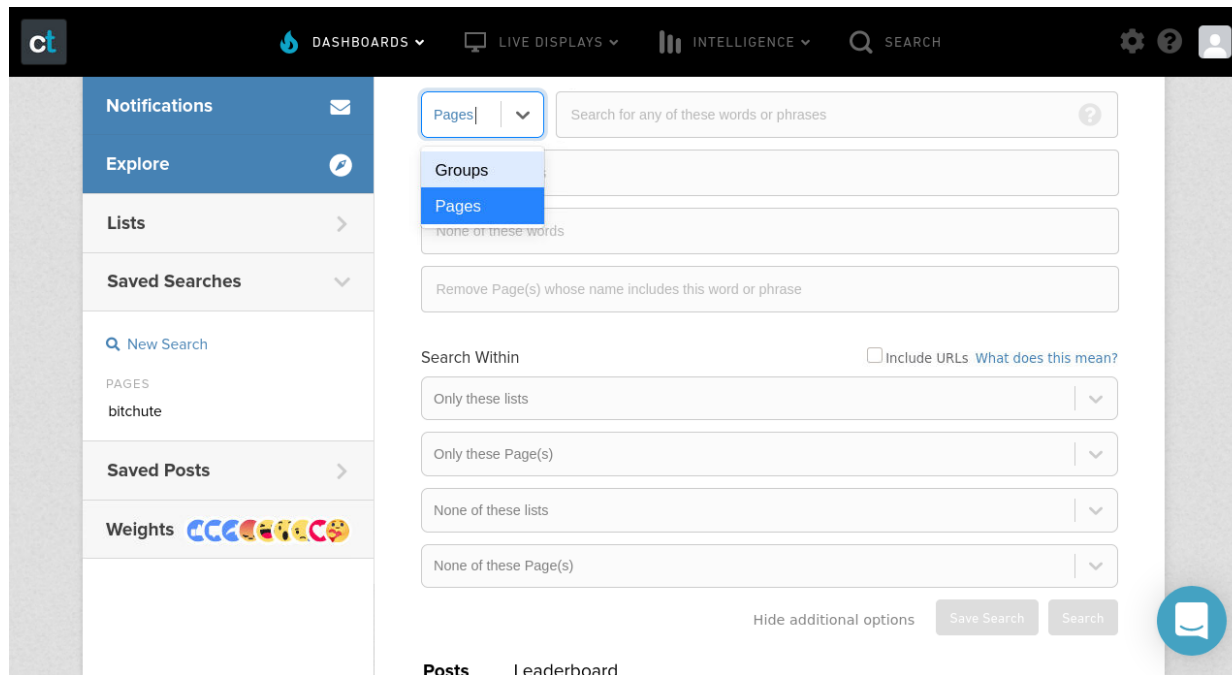


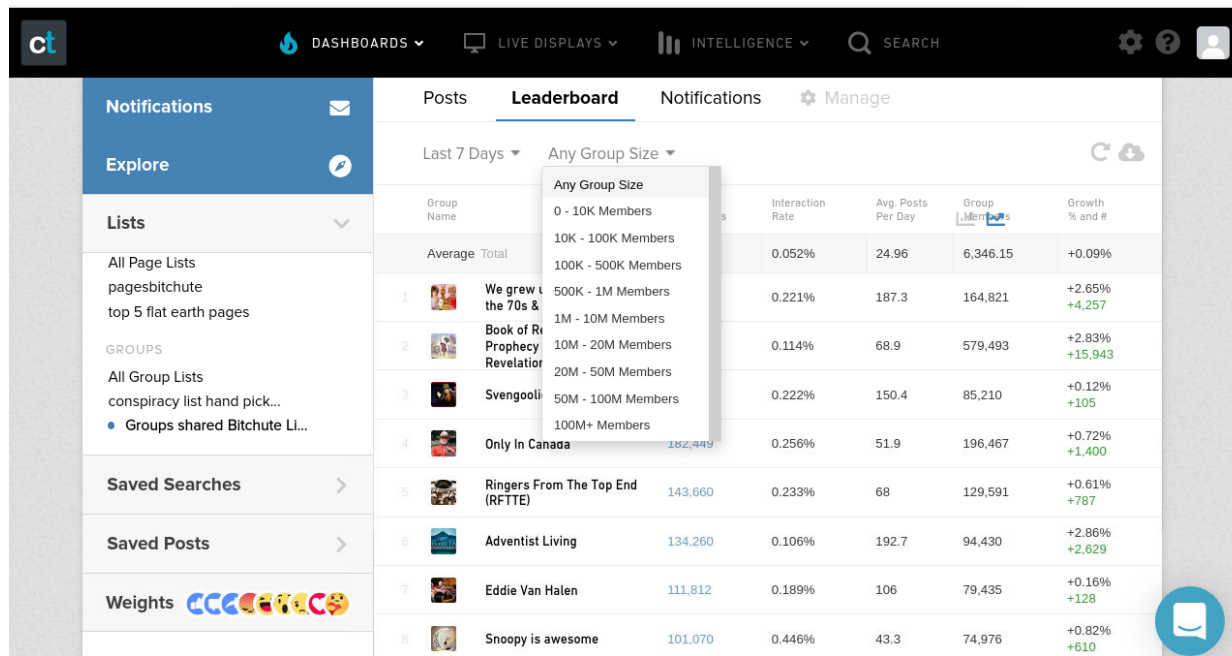
Figure: from left to right: individual pixels, 2x2 super pixels, 4x4 super pixels

Dataset Retrieval

CrowdTangle, “a public insights tool owned and operated by Facebook” [Shiffman, 2021]









Dataset Retrieval



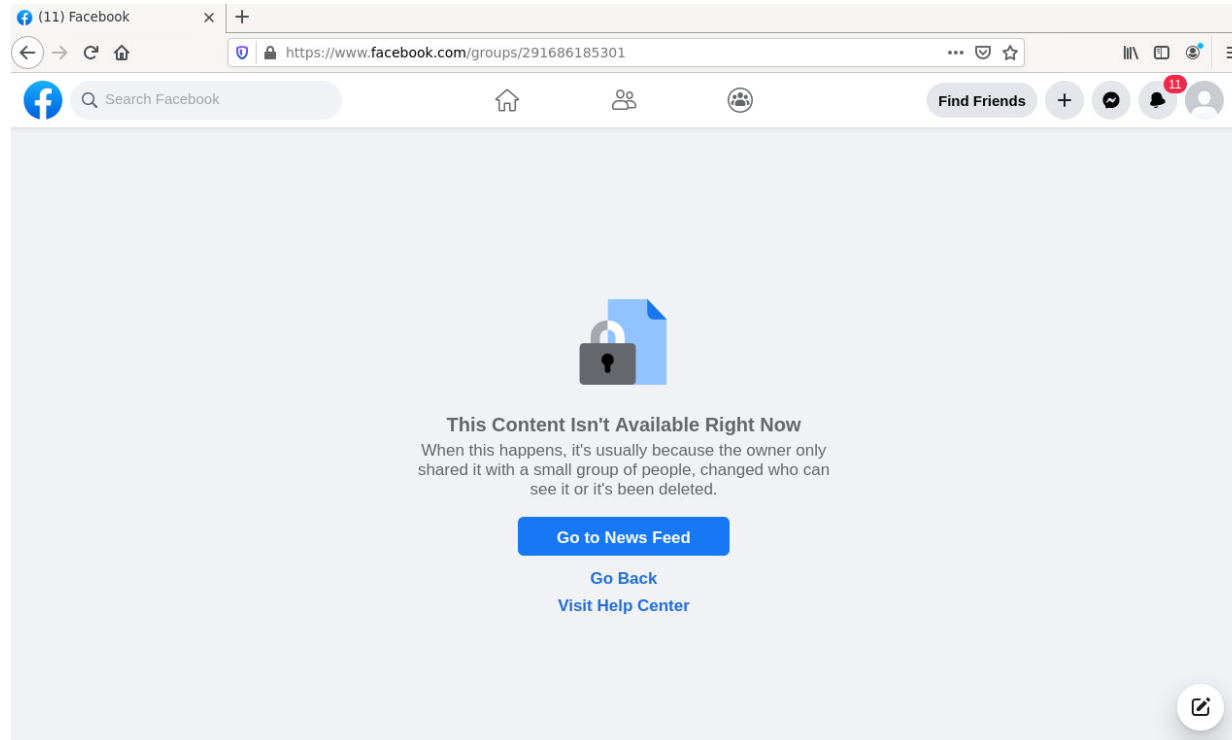
The screenshot shows a dashboard with a sidebar on the left containing 'Notifications', 'Explore', 'Lists', 'Saved Searches', 'Saved Posts', and 'Weights'. The main area is titled 'Leaderboard' and displays a table of social media groups. A dropdown menu for 'Any Group Size' is open, showing options: 0 - 10K Members, 10K - 100K Members, 100K - 500K Members, 500K - 1M Members, 1M - 10M Members, 10M - 20M Members, 20M - 50M Members, 50M - 100M Members, and 100M+ Members.

Group Name	Interaction Rate	Avg. Posts Per Day	Group Members	Growth % and #
Average Total	0.052%	24.96	6,346.15	+0.09%
1 We grew the 70s & 80s	0.221%	187.3	164,821	+2.65% +4,257
2 Book of Revelation	0.114%	68.9	579,493	+2.83% +15,943
3 Svengoolie	0.222%	150.4	85,210	+0.12% +105
4 Only In Canada	0.256%	51.9	196,467	+0.72% +1,400
5 Ringers From The Top End (RFTTE)	0.233%	68	129,591	+0.61% +787
6 Adventist Living	0.106%	192.7	94,430	+2.86% +2,629
7 Eddie Van Halen	0.189%	106	79,435	+0.16% +128
8 Snoopy is awesome	0.446%	43.3	74,976	+0.82% +610

Dataset Retrieval

Posts Leaderboard Notifications ⚙️ Manage						
Last 12 Months ▾		Any Group Size ▾		↻ ☁		
Group Name	Total Interactions	Interaction Rate	Avg. Posts Per Day	Group Members	Growth % and #	
Average Total	2.10M	0.200%	68.05	114,702.67	+67.04%	
1  Awakening Code Community 11:11 33:33 44:44	4.29M	0.087%	107	131,494	+10.33% +12,307	
2  THE WATCHERS	2.76M	0.172%	52.6	89,235	+15.61% +12,049	
3  Ancient Aliens (World Wide)	2.36M	0.11%	23.3	277,797	+21.05% +48,299	
4  Presence	2.00M	0.064%	90.9	97,174	+6.84% +6,220	
5  The Real Flat Earth vs Globe Earth	648,838	0.751%	43.2	8,878	+335.84% +6,841	
6  Galactic Federation of Light	521,568	0.019%	91.4	83,638	+12.56% +9,334	

Dataset Retrieval



Dataset Retrieval

- Goal: at least 30,000 posts
- Posts had to be at least two month old

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- Posts had to be at least two month old
- Result: two datasets: conspiracy theory groups and RTnews (Russia Today)¹

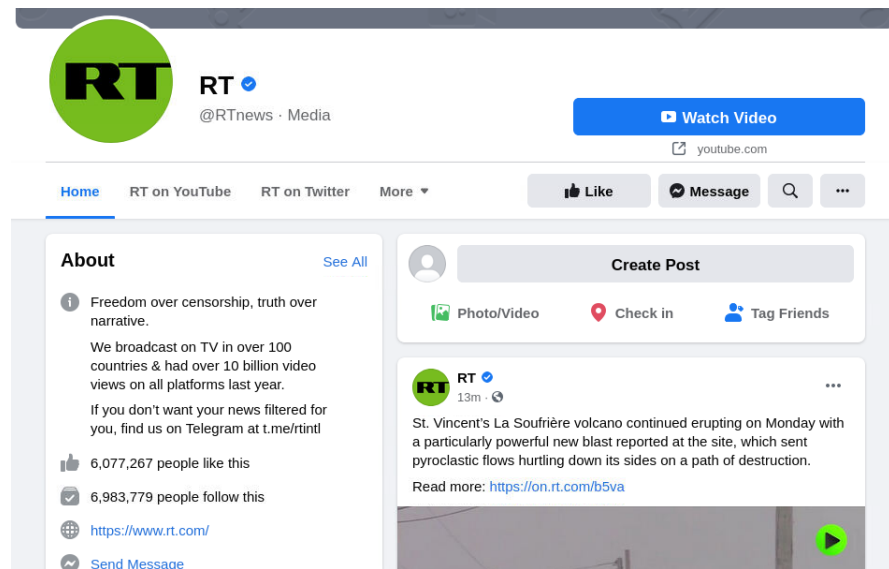


Figure: RT Facebook page

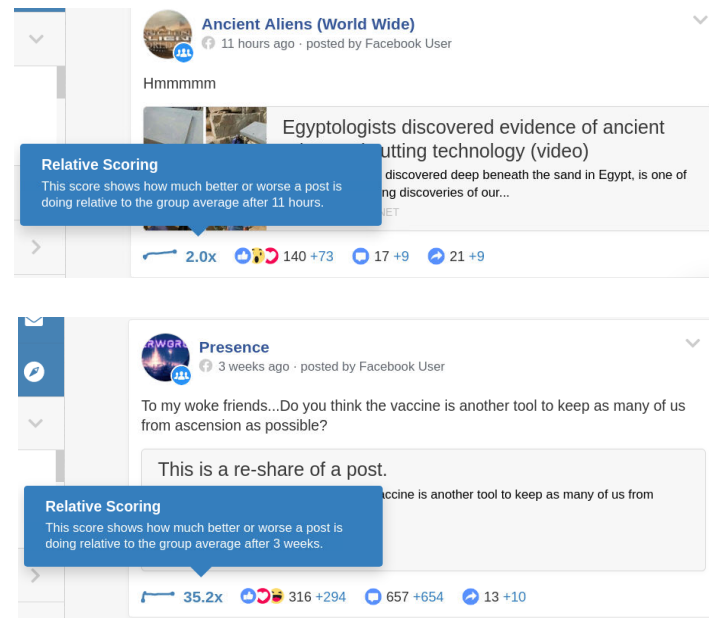
¹A paper analyzing how RT shares conspiracy theories and false information was written by Yablokov [2015]

Dataset Preprocessing

- Duplicate posts have been removed

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- Duplicate posts have been removed



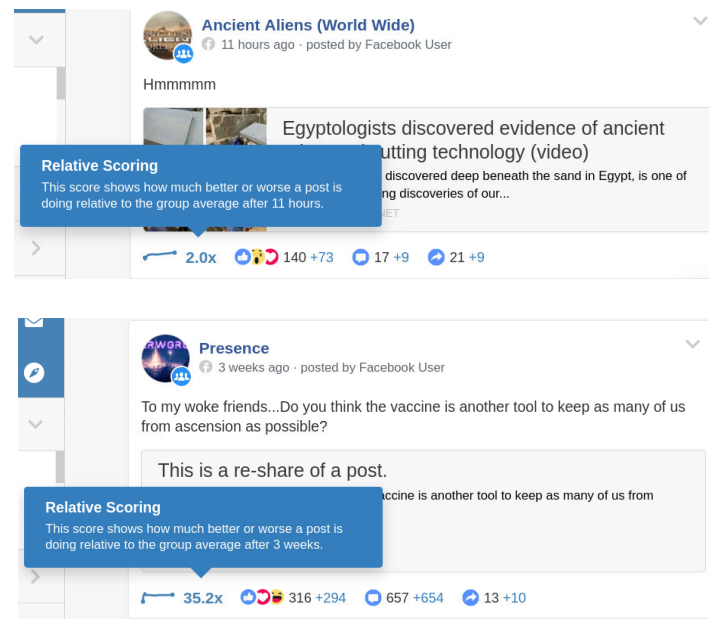
Dataset Preprocessing

- Duplicate posts have been removed
- Our score formula

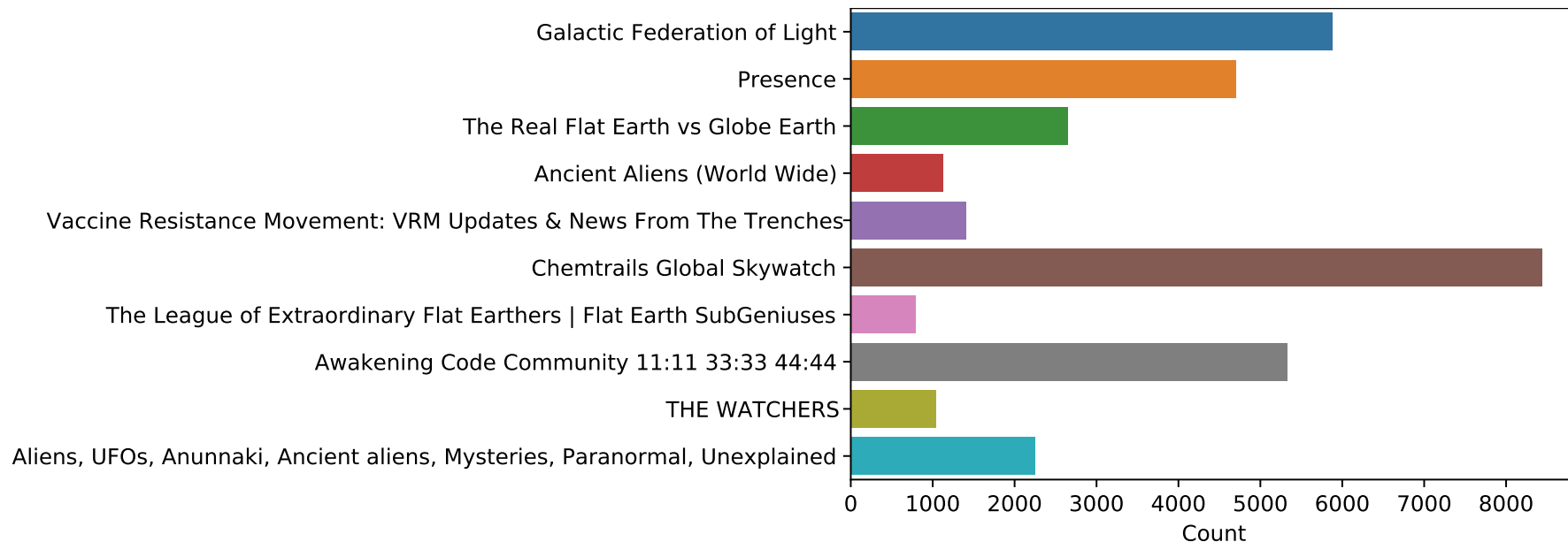
$$\text{score} = \frac{\text{\#reactions}}{\text{\#group avg number of reactions}}$$



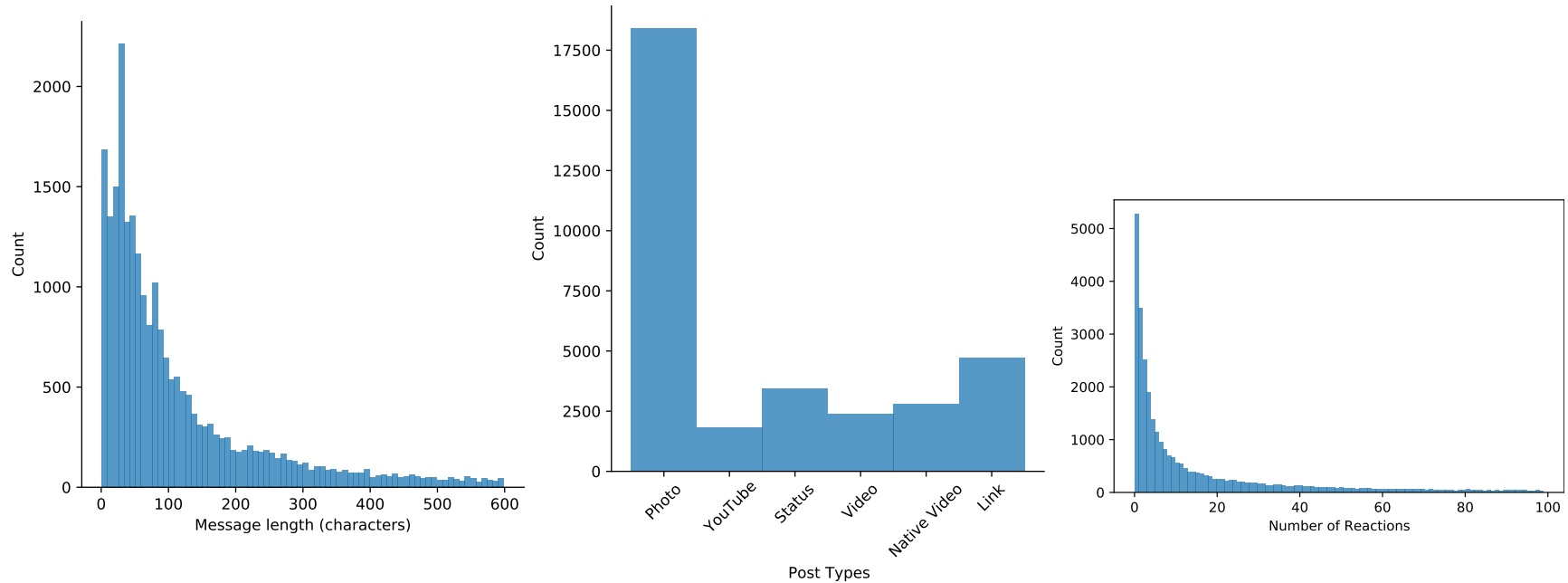
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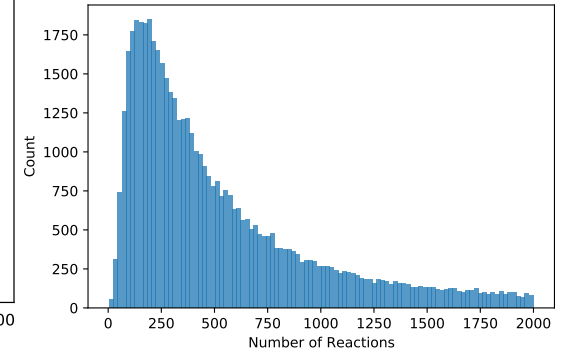
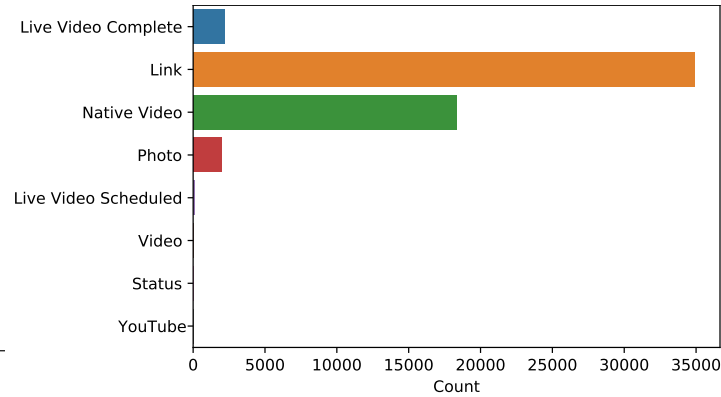
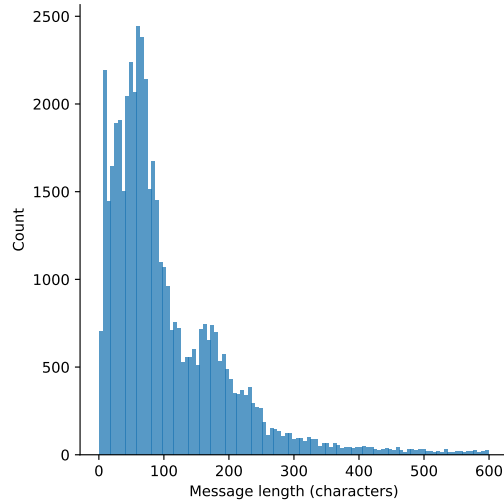
Dataset 1: Conspiracy Theory Groups



Dataset 1: Conspiracy Theory Groups



Dataset 2: rtnews



Experiments

- Baseline: linear regression
 - features: group, post type and top level domain
- Neural model based on DistilBert [Sanh et al., 2019b]
- BiLSTM model
 - categorical features + post text
 - optimizer: Adam
- 60-20-20 split (train / validation / test)

Results

Table: mean squared error for all models and datasets

	conspiracy theory groups	rtnews
linear model	6.68	177.51
DistilBert	6.46	177.00
BiLSTM	6.54	177.44

Results

Table: mean squared error for each group

Group	Linear Model	DistilBert based	BiLSTM
Galactic Federation of Light	12.71	12.49	12.58
Presence	3.51	3.39	3.46
The Real Flat Earth vs Globe Earth	2.18	2.16	2.07
Ancient Aliens (World Wide)	2.08	1.90	1.94
Vaccine Resistance Movement VRM Updates ...	5.43	5.24	5.47
Chemtrails Global Skywatch	7.10	7.00	6.99
The League of Extraordinary Flat Earthers ...	1.22	0.84	0.95
Awakening Code Community 1111 3333 4444	7.75	7.42	7.61
THE WATCHERS	2.21	2.01	2.12
Aliens, UFOs, Anunnaki, Ancient aliens, ...	6.72	5.97	6.21

Results

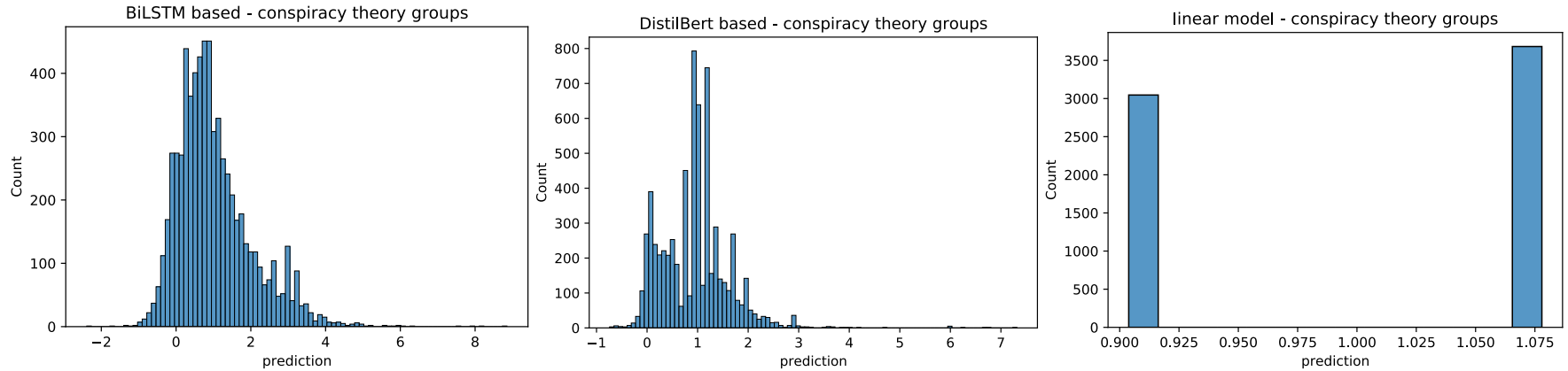


Figure: distribution of the model's score variable prediction for the conspiracy theory group dataset

Results

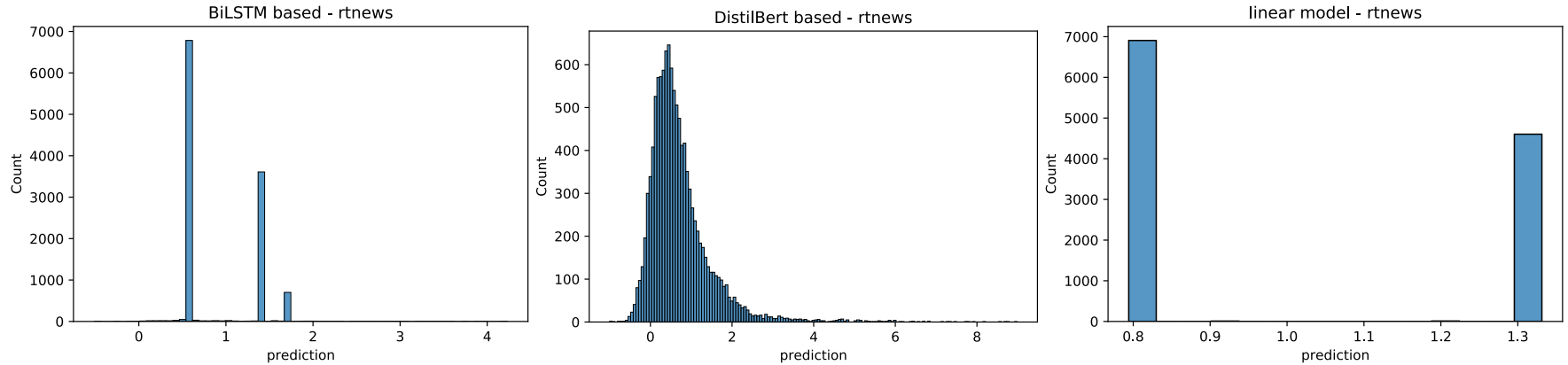


Figure: distribution of the model's score variable prediction for the rnews dataset

Discussion

- We could not outperform the baseline model
- The baseline model is already comprehensible for humans

- Uniform distribution of reactions
- Use of other topics / groups
- No relation at all?

Thank you for your attention

Feel free to ask questions!

References I

- I. Covert, S. M. Lundberg, and S.-I. Lee. Understanding global feature contributions with additive importance measures. In H. Larochelle, M. Ranzato, R. Hadsell, M. F. Balcan, and H. Lin, editors, *Advances in Neural Information Processing Systems*, volume 33, pages 17212–17223. Curran Associates, Inc., 2020.
- D. Freeman, F. Waite, L. Rosebrock, A. Petit, C. Causier, A. East, L. Jenner, A.-L. Teale, L. Carr, S. Mulhall, and et al. Coronavirus conspiracy beliefs, mistrust, and compliance with government guidelines in england. *Psychological Medicine*, page 1–13, 2020. doi: 10.1017/S0033291720001890.
- R. Guidotti, A. Monreale, S. Ruggieri, F. Turini, F. Giannotti, and D. Pedreschi. A survey of methods for explaining black box models. *ACM computing surveys (CSUR)*, 51(5):1–42, 2018.
- D. P. Kingma and J. Ba. Adam: A method for stochastic optimization. In Y. Bengio and Y. LeCun, editors, *3rd International Conference on Learning Representations, ICLR 2015, San Diego, CA, USA, May 7-9, 2015, Conference Track Proceedings*, 2015.
- S. Krishnan and E. Wu. Palm: Machine learning explanations for iterative debugging. In *Proceedings of the 2nd Workshop on Human-In-the-Loop Data Analytics*, pages 1–6, 2017.
- S. M. Lundberg and S.-I. Lee. A unified approach to interpreting model predictions. In *Proceedings of the 31st International Conference on Neural Information Processing Systems, NIPS’17*, page 4768–4777, Red Hook, NY, USA, 2017. Curran Associates Inc. ISBN 9781510860964.

References II

- F. Pedregosa, G. Varoquaux, A. Gramfort, V. Michel, B. Thirion, O. Grisel, M. Blondel, P. Prettenhofer, R. Weiss, V. Dubourg, J. Vanderplas, A. Passos, D. Cournapeau, M. Brucher, M. Perrot, and E. Duchesnay. Scikit-learn: Machine learning in Python. *Journal of Machine Learning Research*, 12:2825–2830, 2011.
- V. Sanh, L. Debut, J. Chaumond, and T. Wolf. Distilbert, a distilled version of BERT: smaller, faster, cheaper and lighter. *CoRR*, abs/1910.01108, 2019a. URL <http://arxiv.org/abs/1910.01108>.
- V. Sanh, L. Debut, J. Chaumond, and T. Wolf. Distilbert, a distilled version of BERT: smaller, faster, cheaper and lighter. *CoRR*, abs/1910.01108, 2019b.
- N. Shiffman. Citing crowdangle data, 2021. URL <https://help.crowdtangle.com/en/articles/3192685-citing-crowdtangle-data>.
- E. Williamson. Rioters followed a long conspiratorial road to the capitol. *The New York Times*, Jan 2021. URL <https://www.nytimes.com/2021/01/27/us/politics/capitol-riot-conspiracies.html>.
- T. Wolf, L. Debut, V. Sanh, J. Chaumond, C. Delangue, A. Moi, P. Cistac, T. Rault, R. Louf, M. Funtowicz, J. Davison, S. Shleifer, P. von Platen, C. Ma, Y. Jernite, J. Plu, C. Xu, T. L. Scao, S. Gugger, M. Drame, Q. Lhoest, and A. M. Rush. Transformers: State-of-the-art natural language processing. In *Proceedings of the 2020 Conference on Empirical Methods in Natural Language Processing: System Demonstrations*, pages 38–45, Online, Oct. 2020. Association for Computational Linguistics.
- I. Yablokov. Conspiracy theories as a russian public diplomacy tool: The case of Russia Today (RT). *Politics*, 35(3-4):301–315, 2015.