Toxic comment classification challenge

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Датасет

- Комментарии из правок к Википедии
- B train set ≈ 150000 комментариев.
- Теги очень не сбалансированы (clean — 135k, toxic — 15k, threat — 1k)
- Мера качества: средний гос-аис по 6 target'ам. (очень плохая метрика)

Baseline

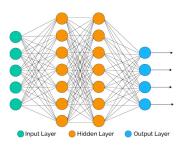
- MultinomialNB без tf-idf
 - private: 0.9084, public: 0.9091
- MultinomialNB
 - private: 0.9636, public: 0.9601
- LogisticRegression c char n-grams
 - private: 0.9804, public: 0.9790

Первый бленд

 rank averaged submission private: 0.9850, public: 0.9855

Сложный выбор





Второй бленд

 Another Cleaned Data Blend With Low Correlation private: 0.9853, public: 0.9860

Третий бленд

 Blend It All private: 0.9863, public: 0.9867

Наше улучшение

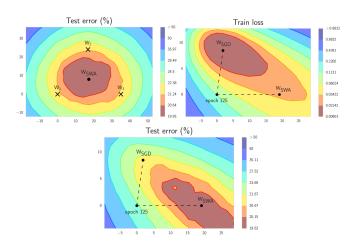
total = blend_it_all · 12 - logreg
 С примерно 950 места на 456
 private: 0.9863, public: 0.9867

Сетки

- Glove -> Bi-GRU -> Conv1d -> Dense
 - private: 0.9826, public: 0.9840
 - private: 0.9832, public: 0.9847 (прокачанный)
- Glove -> Bi-GRU -> MaxPool -> Dense
 - private: 0.9811, public: 0.9823
- Fasttext -> 2xBi-GRU -> MaxPool -> 2xDense
 - private: 0.9818, public: 0.9833

(и тонна Dropout-ов повсюду)

Stochastic Weight Averaging



Stochastic Weight Averaging

```
Require:
   weights \hat{w}, number of iterations n
Ensure: WSWA
    w \leftarrow \hat{w} {Initialize weights with \hat{w}}
    W_{SM/\Delta} \leftarrow W
   for i \leftarrow 1, 2, \dots, n do
       \alpha \leftarrow \alpha(i) {Calculate LR for the iteration}
        w \leftarrow w - \alpha \nabla \mathcal{L}_i(w) {SGD update}
        n_{\text{models}} \leftarrow i \text{ {Number of models}}
        w_{\text{SWA}} \leftarrow \frac{w_{\text{SWA}} \cdot n_{\text{models}} + w}{n_{\text{models}} + 1} \left\{ \text{Update average} \right\}
   end for
```

Stochastic Weight Averaging

- Adam
 - private: 0.9826, public: 0.9840
- Бленд 4 запусков SGD
 - private: 0.9830, public: 0.9842
- SWA
 - private: 0.9832, public: 0.9847

Blending is the new sexy!

Izmailov, Pavel and Podoprikhin, Dmitrii and Garipov, Timur and Vetrov, Dmitry and Wilson, Andrew Gordon
Averaging Weights Leads to Wider Optima and Better Generalization

arXiv preprint, 2018,

https://arxiv.org/abs/1803.05407

Zafar
Rank averaging on preprocessed data
https://www.kaggle.com/fizzbuzz/
rank-averaging-on-preprocessed-data

Anthony Marakis

Another cleaned data blend with low correlation

https://www.kaggle.com/antmarakis/ another-cleaned-data-blend-with-low-correlation

rednivrug

Blend it all

https://www.kaggle.com/rednivrug/blend-it-all

Ashish Gupta Bidirectional GRU with Convolution

https://www.kaggle.com/eashish/ bidirectional-gru-with-convolution

- Prashant Kikani
 Pooled GRU + GloVe (with preprocessing)
 https://www.kaggle.com/prashantkikani/
 pooled-gru-glove-with-preprocessing
- zhbain
 Pooled GRU + FastText
 https://www.kaggle.com/zhbain/
 pooled-gru-fasttext-6c07c9

Ссылки IV



Lessons from Toxic : Blending is the new sexy

https://www.kaggle.com/jagangupta/ lessons-from-toxic-blending-is-the-new-sexy

■ LCT14558

Imbalanced data why you should NOT use ROC curve

https://www.kaggle.com/lct14558/ imbalanced-data-why-you-should-not-use-roc-curve

Ссылки V



Git репозиторий

https://github.com/MichaelSolotky/ ToxicCommentsKaggle/tree/master