

Comment on linguistic features that signal Professional Discourse in two kinds of open reviews. Underline features of "emotional (intelligible)". Highlight those signalling "politeness" in yellow, "oral proximity/addressivity" in blue, "praise" in green and "refutative" in red.

interesting work, quite humble specific, substantial focus and structure
 ICLR 2017 review-Anonymous Reviewer3 | Dec-17-2016 | Rating: 6: Marginally above acceptance threshold

These are interesting contributions, but due to the heavy focus, unfortunately, the paper does not seem to have a clear goal. [...] The paper takes a quite humanistic approach and discusses the pieces the authors used to obtain state-of-the-art performance for one problem. That is OK, but I would've rather expected that from a paper called something like "Improved knowledge transfer and distillation for text analytics". It assessed, I encourage the authors to change the title to something along those lines.

The heavy focus also made it hard for me to follow the authors' train of thought. I praise the authors had a good reason for their section ordering, but I didn't see the reason in it. How about re-organizing the sections as follows to discuss one contribution at a time? 4.2, 4.3, including 6, put 9 into an appendix and point to it from there. 5.7, 5.10 [...] One thing that the current structure is sub-optimal is that there are 11 sections.

I like the authors' idea for transfer learning without catastrophic forgetting, and I must admit I would've rather liked to read a paper solely about that (studying where it works, and where it fails) than about the heavy other topics of the paper. I would've voted for acceptance since I like the ideas, but if the paper does not make it, I would suggest that the authors consider splitting it into two papers, each of which could hopefully be more focused.

Confidence: 3: The reviewer is fairly confident that the evaluation is correct | Scale: 1-5

Review
 Reviewer
 06/04/22

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Text 1: Prototypical positive review from the ICLR review corpus
 Model accept (ICLR17_R2_O5_Rating_9/10_Confidence_4/5)

A nice contribution to differentially-private deep learning

ICLR 2017 conference-Anonymous Reviewer2

16 Dec 2016 ICLR 2017 conference official review

Rating: 9: Top 15% of accepted papers, strong accept
 Review: Altogether a very good paper, a nice read, and interesting. The work advances the state of the art on differentially-private deep learning, is quite well-written, and relatively thorough.

One caveat is that although the approach is intended to be general, no theoretical guarantees are provided about the learning performance. Privacy-preserving machine learning papers often analyze both the privacy (in the worst case, DP setting) and the learning performance (often under different assumptions). Since the learning performance might depend on the choice of architecture, future experimentation is encouraged, even using the same data sets, with different architectures. If this will not be added, then please justify the choice of architecture used, and/or clarify what can be generalised about the observed learning performance.

Another caveat is that the reported epsilons are not those that can be privately released; the authors note that their technique for doing so would change the resulting epsilon. However this would need to be resolved in order to have a meaningful comparison to the epsilon-delta values reported in related work.

Finally, as has been acknowledged in the paper, the present approach may not work on other natural data types. Experiments on other data sets is strongly encouraged. Also, please cite the data sets used.

Review
 Reviewer
 06/04/22

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Text 2: Prototypical positive review from the ICLR review corpus
 Model accept (ICLR17_R1_L200_Rating_9/10_Confidence_4/5)

20 Dec 2016 ICLR 2017 conference official review

Rating: 5: Marginally below acceptance threshold

Review: This paper proposes a method for transfer learning, i.e. leveraging a network trained on some original task A in learning a new task B. The authors claim that their method is more effective than other methods, but they do not encourage a model trained on A, while training on the new task B, to match fake targets produced by the model itself but when it is trained only on the original task A.

Experiments show that this method can help in improving the result on task B, and is better than other baselines, including standard fine-tuning. General comments:

- As far as I can tell, there is no experimental result supporting the claim that your model still performs well on the original task. All experiments are on the new task B.

- The introduction makes a strong statements (sic) about the disjunct logical rule engine into a neural network, which I find a bit misleading. The approach in the paper is not specific to transferring from logical rules (as stated in the Sec. 2) and is simply relying on the rule engine to provide labels for unlabelled data.

- One of the obvious weaknesses to compare with your approach is standard multi-task learning on both tasks A and B together. That is, you train on both tasks together, and you do not need to transfer anything. It is clear that this is the same as what is referred to in Sec. 3 as "joint training". Can you please explain more clearly what you refer to as "joint training"?

- Why can't we find the same baselines in both Table 2 and Table 3? For example Table 2 is missing "joint training", and Table 3 is missing GRU trained on the target task.

- While the idea is presented as a general method for transfer learning, experiments are focused on one domain (sentiment analysis on SemEval tasks). The authors' experiments should include applying the idea on at least one other different domain, or the writing of the paper should be reworked to make the focus more specific to this domain/task.

- The writing of the paper in general needs some improvement, but more specifically in the experiment section, where experiment setting and baselines should be explained more concisely.

- The empirical methodology paragraph does not fit in the flow of the paper. I would either explain it in the experiment section, rather than including it in the introduction.

- Table 1 seems like reporting cross-validation results, and I do not think is very informative to general reader.

Review
 Reviewer
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Padova, 06/04/22

Reviewing International Conference Proposals: Awareness of Principles & Practices for Non-Native Research Novices

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Review
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6. Outlook

new technical affordances/opportunities → explore new practices with old tools?
 electronic data collection to analyse practices to learn passively and actively

(functional) linguistic concepts like politeness, concessives, mitigation, ...
 correlate with non-linguistic concepts like face, interaction, ...
 but

even reviewers have to learn to cooperate openly and transparently, e.g.

ICLR even guide their readers to "great in-depth resources on reviewing" with programmatic titles like

"Criticising with Kindness" or "Mistakes Reviewers Make"

(<https://arxiv.org/abs/2020.07.0420>)

The link to the "Last minute reviewing advice" even focusses on multiple-reviews ICLR style.

Please go to

mytuc.org/kmjs

and fill in the questions there.

Please go to

mytuc.org/pkrl

and fill in the questions there.