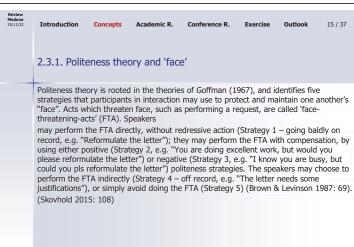


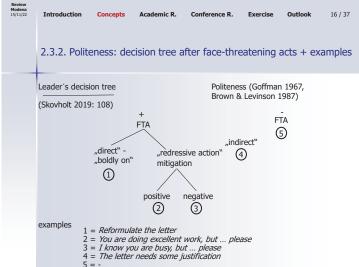


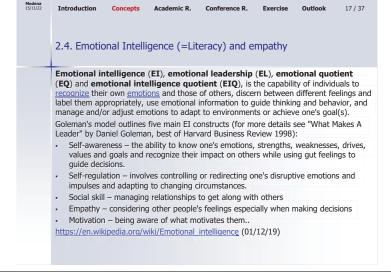
Academic R.

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Introduction



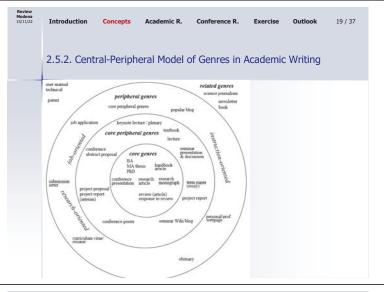




Introduction Academic R. Exercise Outlook 18 / 37 Conference R. 2.5. Genres in Academic Writing 2.5.1. Types of discourses research "output" teacher "talk"/e-learning research article ppt presentations book reviews lectures (+/- remote!) student presentations project proposals conference reviews textbooks conference presentations Wikis/Moodle/Google+ www pages (HTML, php) discipline-specific culture-specific student "literacy" lecture/fieldwork notes "papers" (argumentative! essays) science "journalism" MA/BA thesis (persuasive!) popular science articles seminar presentations popular blogs (David Crystal) popular science films (Horizon) "Novice Academic English" popular science books

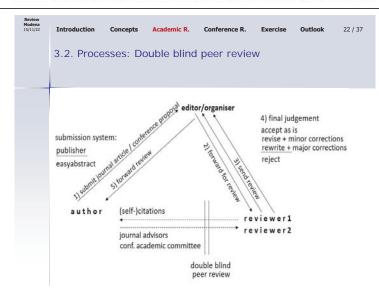
author-specific culture-specific

science slam





Introduction Exercise Outlook 21 / 37 3. Academic Reviews 3.1. Double Blind Peer Reviews fashioned in social science journals since the 1950s also for conference abstracts the identity of the authors is concealed from the reviewers, and vice versa, to avoid bias but despite any editorial effort to ensure anonymity, the process often fails to do so, since certain approaches, methods, writing styles, notations, etc., point to a certain group of people in a research stream, and even to a particular person. BUT most double-blind peer reviews are anonymous and unpublished → case study: "open"



4. Conference Reviews

4.1. Digital/on-line reviews

Since "standards" in community-specific discourse are developing, many case studies are necessary!

rarely publically available!

4.2. Case study: ICLR

Since "standards" in community-specific discourse are developing, many case studies are necessary!

AGAIN a Model case study!

ICLR 2019

International Conference on Learning Representations

② New Orleans, Louistana, United States ■ May 6 - May 9, 2019. ② https://icic.cc/Conferences/2019

Questions or Concerns

Please contact the Openitiview support soim at integroperveniew and wary questions or concerns about the Openitiview support soim at integroperveniew and wary questions or concerns about the Openitiview support soim at integroperveniew and wary questions or concerns about conference administration or policy.

Oral Presentations Poster Presentations Submitted Papers

Neural Causal Discovery with Learnable Input Noise ■

Talin Viu, Thomas Breach, an Kautz

2 to 39 2791 Incomed at 15 co. 2018 (XXZES) Conference Bind Submitted Papers

Neural Causal Discovery with Learnable Input Noise ■

Talin Viu, Thomas Breach, an Kautz

2 to 39 2791 Incomed at 15 co. 2018 (XXZES) Conference Bind Submitted Papers

Neural Causal Discovery with Learnable Input Noise

RETHINKING SELF-DRIVING: MULTI-TASK KNOWLEDGE FOR BETTER GENERALIZATION AND ACCIDENT EXPLANATION ABILITY

Plana Language Motoropolis, Causuma SASAN, Tessaya OGATA, Shepeli SUGANO

2 to 30 19 (19) Specially of Concession in Open Peer Review Methodology 12

02.05.2019 Concession in Open Peer Review Methodology 12

Ivanova, Marina. Concession in Single- and Double-Blind Open Peer Review:

A Corpus-Based Analysis. MA Thesis Chemnitz 2019

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Ivanova, Marina. Concession in Single- and Double-Blind Open Peer Review: A Corpus-Based Analysis. MA Thesis Chemnitz 2019

unconvincing experiments; original theorem statement incorrect ICLR 2019 Conference Paper594 Area Chair1

17 Dec 2018 (modified: 21 Dec 2018) ICLR 2019 Conference Paper 594 Meta

Review Readers. @ Everyone Metareview: Granger Causality is a beautiful operational definition of causality, that reduces causal modeling to the past-to-future predictive strength. The combination of classical granger causality with deep learning is very well motivated as a research problem. As such the continuation of the effort in this paper is strongly encouraged. However, the review process did uncover possible flaws in some of the main, original results of this paper. The reviewers also

expressed concerns that the experiments were unconvincing due to very small data sizes. The paper will benefit from a revision and resubmission to another venue, and is not ready for acceptance at ICLR-2019. Confidence: 4: The area chair is confident but not absolutely certain

Recommendation: Reject

Methodology 02.05.2019 Concession in Open Peer Review 15

Meaning Category **Paratactic** Hypotactic finite non-finite: non-finite: conjunction preposition if P then condition: causal-conditional [concession / even if, even if. despite. contrary to consequence] even though, even though, in spite of, expectation although, although, without Q (and) yet + still; while while but + nevertheless 2 [consequence ^ concession] (though)

02.05.2019 Concession in Open Peer Review R&D: RQ1 (Forms and functions)

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- **Criticism mitigation** → "I very much like the idea of the paper, but I am simply not convinced by its claims." (ICLR17_R3_R279)
- $\textbf{Criticism reinforcement} \rightarrow \text{``The paper has a laundry list of}$ related results (page 2) but no clear message." (ICLR17_R2_R363)
- **Praise mitigation → "Even though** no conclusive section is provided, the paper is not missing any information." (ICLRC_R3_0785)
- **Praise reinforcement** → "I found the paper very well written despite its level of mathematical depth (the authors provide many helpful pictures) and strongly recommend accepting this paper." (ICLRC_R1_076)

02.05.2019 Concession in Open Peer Review R&D: RQ1 (Forms and functions)

Ivanova, Marina. Concession in Single- and Double-Blind Open Peer Review: A Corpus-Based Analysis. MA Thesis Chemnitz 2019 What do you see? What would you have expected?

Academic R.

2018

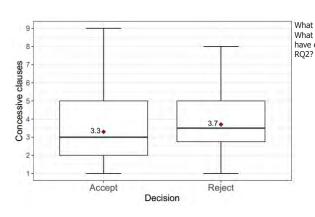
60

109 43231

words 22787

29 14923

Ivanova, Marina. Concession in Single- and Double-Blind Open Peer Review: A Corpus-Based Analysis. MA Thesis Chemnitz 2019



Double-blind Single-blind 2017 What do you see? What would you have expected? RQ2? 9400 Accept Oral Accept Poster 24609 36533 Reject

Introduction

Table 1: The ICLR review corpus

		accept oral			accept poster			reject		
	Sum	#	/100,000	%	#	/100,000	%	#	/100,000	%
please	104	30	62	29	20	43	19	54	53	52
*ould	1509	425	874	28	408	879	27	676	660	45
no*	2691	599	1232	22	618	1331	23	1474	1439	55
only	382	74	152	19	83	179	22	225	220	59
question*	263	57	117	22	54	116	21	152	148	58
wh*	710	151	311	21	149	321	21	410	400	58
(?)	979	239	492	24	219	472	22	521	509	53

Double-blind

2019

15 6892

22691

Sum

words

48606

102455

130 46424

Table 2: Selected politeness features in positive and negative reviews

31 / 37 Introduction 32 / 37 Academic R. Conference R. Outlook 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 AnonReviewer2
16 Dec 2016 ICLR 2017 conference official review 5. Exercise Rating: 9: Top 15% of accepted papers, strong accept qualitative self-evaluation of texts 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. we need many small case studies of current practices to confirm ad-hoc hypotheses of One caveat is that although the approach is intended to be general, no theoretical developing conventions guarantees are provided about the learning performance. Privacy-preserving machine learning papers often analyze both the privacy (in the worst case, DP Recognise evaluative language features in accepted (prototypical; model) academic discourse setting) and the learning performance (often under different assumptions). Since the learning performance might depend on the choice of architecture; future Identify communicative functions of linguistic features 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 Reading and interpreting critical reviews in order to improve conference proposals 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. Comment on linguistic features that signal Professional Discourse in two kinds of open Introduction Concepts Academic R. Conference R. Outlook 34 / 37 reviews.[.]Underline features of <u>"emotional (intelligence)"</u>. Highlight those signalling <mark>"polite"</mark> in yellow, "oral/proximity/addressivity" in blue, "praise" in green and "criticism" in red.¶ Text 2- Prototypical negative review from the ICLR review corpus
Model reject (ICLR17, RT, R203, Rating 510, Confidence 4/5)
ICLR 2017 contieence AnonReviewer1
20 Dec 2016 ICLR 2017 contieence AnonReviewer1
20 Dec 2016 ICLR 2017 contieence official review
Rating: 5. Manginghly 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, but also tries to avoid degradation in performance on A. The general idea is based on
encouraging 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 commentsiquestions:
- 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
show that you can improve on the new task only.
- The introduction makes a strong statements Isical bout the distilling logical rule engine into a neural network, which I find a bit misleading. The Interesting work, quite domain-specific, suboptimal focus and structure ¶
ICLR·2017 review AnonReviewer3 | Dec. 17, 2016 | Rating: 6: Marginally above acceptance threshold ¶ These are interesting contributions, but due to the many <mark>ces</mark>, <u>unfortunately</u>, the paper d <u>seem</u>·to·have·a·<mark>clear·focus</mark>.·[...].·[T]he·paper·takes·a·<mark>quite·dom</mark> and discusses the pieces the authors used to obtain state-of-the-art performance for one problem. That is OK, but I would've <mark>rather</mark> expected that from a paper called something like "Improved knowledge transfer and distillation for text analytics". If accepted, I encourage the authors to change the title to something - The introduction makes a strong statements [sic] about the distilling 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 along those lines. labels for unlabelled data.

One of the obvious baselines to compare with your approach is standard multi-task learning on both tasks A and B together. That is, you train the model from scratch on both tasks simultaneously (which sharing parameters). It is not clear this is the same as what is referred to in Sec. 8 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 37 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 task). I think that either experiments should include applying the idea on at least one other different domain, or the writing of the paper should be modified to make the focus more specific to this domain/task. also made it hard for me to follow the authors' train of thought. <u>I'm sure</u> the authors had a good reason for their section ordering, but I didn't see the ted thread in it. How about reorganizing the sections as follows to discuss one contribution at a time? 1,2,4,3,8 including 6, put-9 into an appendix and point to it from here, 7, 5, 10. [...] One clue that the current structure is optimal is that there are 11 sections....¶ I· like the authors' idea for transfer learning without gatastropic forgetting, and I must admit I Induring to make the locus more specific to time demands.

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.

Ensemble methodology paragraph does not fit the flow of the paper. I would rather explain it in the experiments section, rather than including it as part of your approach. would've-rather-liked to read a paper-solely about that (studying where it works, and where it fails) es of the paper. I weakly vote for acceptance since I like the ideas, but than about the man if the paper does not make it in. I would suggest that the authors consider splitting it into two papers, Table 1 seems like reporting cross-validation results, and I do not think is very informative to general reader each of which could hopefully be more focused. ¶ Confidence: 3: The reviewer is fairly confident that the evaluation is correct [Scale 1-5] Introduction Concepts Academic R. Conference R. Exercise Outlook 35 / 37 Introduction Concepts Academic R. Conference R. Exercise Outlook 36 / 37 6. Outlook 6. Outlook 6.1. Old genres are adapted to technical, societal changes 6.2. Young researches observe their discourse community discourse community sets conventions new technical affordances/opportunities → explore new practices with old tools? electronic data collection to analyse practices to learn passively and actively young researchers decide whether to follow conventions (Confucius style) (functional) linguistic concepts like politeness, concessives, mitigation ... · or to reject conventions (cowboy style) correlate with non-linguistic concepts like face, interaction, ... both may be successful! but you have to know the rules to break them effectively but even reviewers have to learn to cooperate openly and transparently, e.g. discourse community conventions are changing with ICRL even guide their readers to "great in-depth resources on reviewing" with programmatic titles like "Criticising with Kindness" or "Nistakes Reviewers Make" (https://lick.c/Conferences/2020/ReviewerGuide, 0/104/20). The link to the "Last minute reviewing advice" even focusses on multiple-reviews ICLR style. technological affordances & societal demands young researchers have to observe conventions in their discourse community public funding demands open science success indicators like conferences contributions, reviews, etc. transparent methodologies, criteria, discourse

· verifiable results through public repositories (e.g. CLARIN)

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