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Deep Learning for Natural Language Technology

- Work on one concrete NLP problem using a neural network approach
- Canvas course web page lists example problems and links to relevant data sets
 - but you can also choose your own problem (check w/ TAs)
- Work in teams of 5 students
- ▶ Before Monday, Sept. 10, 9am: each team sends one email to both TAs with
 - list all team members (full names, student IDs and emails)
 - for each team member list NN toolkit expertises (TensorFlow, PyTorch, ...)
 - the problem you choose to work on (if you choose your own, check feasibility with TAs during lab session in week 2)
- ► In weeks 6 and 7 each team gives a 15 minute presentation followed by 5 minutes for questions



- Mini project runs from week 2–5;
- ▶ Deadline for submitting presentations/reports/code is Friday, October 5, 12:00 noon
- Suggested timeline
- ► This week:
 - choose your team
 - pick/come up with a problem
 - inform TAs
- ► Week 2:
 - identify and read relevant previous literature
 - inspect data, pre-preprocess, test/valid data
 - additional data crawling?
 - what should be the initial model?

- ► Week 3:
 - implement first model
 - evaluate
 - debug?
 - what works, what doesn't?
- ▶ Week 4:
 - refine first model/try alternative model
 - bells and whistles (dropout, layernorms, more layers, residual, highway connections, etc.)
 - evaluate and compare
 - what works, what doesn't?

- ► Week 5:
 - refine second model/try (another) alternative model
 - error analysis, open problems
 - conclusions, recommendations
 - write 4 page report with main findings
 - prepare slides for presentation
- ▶ By Friday, October 5, 12 noon, **each** team submits:
 - your team's short report (approx. 4 pages) in PDF
 - your team's presentation slides (as PDF or PowerPoint)
 - your code including a readme file (how to run your code)
 - as link to Github repo (preferably) or gzipped tarball
 - if (gzipped) data is <10MB you can include it, otherwise provide link
- ► Team presentation slots (for weeks 6–7) will be assigned at the beginning of week 5





- Lab sessions:
 - Monday 9-11am (two rooms); weeks 2–5
 - No labs on Tuesdays (1-3pm), unless really, really needed
 - opportunity to ask TAs questions about your project progress/problems/pointers/feedback on ideas...
 - each team sends two representatives to the lab sessions
 - which room should a team go to?

- Some practical advice
 - keep the data small, most likely, you'll be running it on your laptop
 - if the data is too big, down-sample, select based on some criteria (vocab, classes, length), ...
- What makes a good mini-project?
 - good succinct description of the most relevant research papers
 - good description of the data preprocessing/settings
 - good motivation of the neural architecture choices
 - thorough evaluation (scores under different settings/architectures); error analysis
 - report that focuses on the most relevant findings: what (doesn't) work and why
 - good presentation focusing on the most important aspects and finding, understandable by wider audience, contains examples





- ▶ Pimp up your (seemingly simple) mini project
- ► Example: language identification
 - literature: what has been tried (neural and non-neural)?
 - process input: words, characters, character n-grams?
 - baseline model: word embeddings (week 2); how to combine embeddings; additional hidden layers?
 - run evaluations: e.g., accuracies? (check literature)
 - better model?: CNN sequence classifier (week 3) (how to pool; strides. etc.)
 - more evaluation (literature mentions problem of short inputs)
 - more evaluation (literature mentions problem of languages close to each other)
 - applications: e.g., instant translation, question: what are accuracies after n characters?
 - ... continues on next slide



- Pimp up your (seemingly simple) mini project
- Example: language identification
 - ... continuation from previous slide
 - crawl/use more data; how can we use script information (UTF-8 regions) in an NN?
 - even better model? RNN sequence classifier (week 3/4)
 - evaluation: other tricky cases: e.g., code-switching?
 - best overall evaluation: accuracies, precision/recall, ... ?
 - how best to show findings: score tables; visualizations; clusters?
 - looking back what really makes a difference?
 - anything that contradicts previous literature (bug or finding)?



- ► Teaching assistants (TAs)
 - Praveen Dakwale: P.Dakwale@uva.nl
 - Hamidreza Ghader: H.Ghader@uva.nl
- All communication with TAs in English