

Text Processing using Machine Learning

Ask Me Anything

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OVER **5,500** GRADUATE ALUMNI

OFFERING OVER **120** ENTERPRISE IT, INNOVATION & LEADERSHIP PROGRAMMES

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Ask Me Anything

Frequently Asked Questions

- **What's next in Machine Learning and NLP?**
- **Where do we learn more on ML/DL and NLP?**
- **How do I get industrial experience for ML/DL and NLP?**
- **What's the latest thing in NLP now?**
- **Tell me about your daily work, how does a day as “research scientist” look like?**
- **I have problem X, which tool/method Y to solve it?**
- **Open Source Software**

What's next in Machine Learning and NLP (that I'm personally excited about)?

- **Quantum Computing**

- <https://medium.com/xanaduai/training-quantum-neural-networks-with-pennyLane-pytorch-and-tensorflow-c669108118cc>
- <https://github.com/XanaduAI/pennyLane>

- **Federated Learning**

- <https://blog.openmined.org/upgrade-to-federated-learning-in-10-lines>
- <https://github.com/OpenMined/PySyft>

- **Evolutionary/Genetic Algorithms**

- https://leanpub.com/genetic_algorithms_with_python

Where do we learn more on ML/DL and NLP?

- **As a start**

- <https://github.com/datasciencesg/workshops/tree/master/LearnItYourself>

- **For latest shiny things**

- Follow DL/ML stars on Twitter or their preferred social media
- Follow the #nlpoc hashtag on Twitter
- Follow @arxiv_cs_cl on Twitter and for sanity <http://www.arxiv-sanity.com/>

- **Publish a paper and/or Join a conference**

- Most listed conferences on <https://aclanthology.info/> are good to join
- Text, Speech, and Dialogue (TSD) and Interspeech conferences
- Join shared task in workshops co-located in these conferences

How to get industrial experience for ML/DL & NLP?

- **Join Competitions**

- Kaggle, shared task in conferences and many more

- **Build things and open source**

- Learn some Flask/Django or web development, just enough to show the world and demo what you've done.

- **Get a mentor or an internship**

- Mentorship is harder to find but it's possible. Sometimes non-profit organizations and companies do have mentorship programs
- Internships are a plenty but find places that don't make you do "sia kang" and people you think would enjoy working with

What's the latest thing in NLP now?

- **Transformers**

- Lots of transformers and its variant
- I do want to see it go away... It's sort of a boring model.

- **You'll never know, it moves so fast...**

- Every day new code commits are made on PyTorch, Tensorflow, AllenNLP, SpaCy, etc.
- Get involve in the open source and you get first blood on the new tech =)
- Every 2-3 months is an NLP conference, every month there's an NLP conference deadline

- **Don't chase the shiny new things**

- Know that they exist, know how they work and what libraries to use
- Use it only when you tried and experimented and show that it works better for your task
- Know the foundations, there's seldom something new under the sun, just better rehashes of things

Research Scientist @ RIT

- **Project Management**

- Managing expectations of “AI” products
- Understanding what problem your “clients” wants to solve
- Know what data they have/have not, find where to get the required data
- Propose a feasible solution and try before 2nd meeting
- Keep “clients” engaged, show to them it’s the latest tech that’s useful to them

- **Knowing Backend/Frontend Engineering helps**

- Know what’s possible, what’s easy what’s hard
- Learn from engineers (dockers, databases, cloud, apps design/dev) and let engineers learn what you do
- There’s not clear boundary, a data/dev engineer might train a better model than you do

- **Reading and lots of coding**

- Code sprints to get **** done
- Finding out what’s new, useful and quick to prototype
- Knowing whether it can be “productionize”, e.g. ensemble of 100+ models wins competition -_- | | |
- Look at old ideas, know the limitations, see how you can fix them

I have problem X, which tool/method Y to solve it?

- Literature review
- Know what are the datasets and what's in them (noise, quirks, etc.)
- Which evaluation metric is task X evaluated on?
- Find the latest shiniest paper,
 - Track the oldest relevant citation of the task, read that paper
 - Find the highest cited paper for the task, use that as your baseline
 - Whenever possible, hunt down the datasets in that highest cited paper and latest shiniest paper
 - Define your success criteria for the task industrially (it might not be the standard eval metric for the task)
 - Try/Reimplement the baseline
 - Did baseline meet the success criteria? Can your engineer productionize it?
 - Ask the business/project stakeholder whether it's sufficient
 - Communicate your model/libraries to engineers, build it, test it, break it, repeat

- ***“If data is the fuel to today’s software, open source is the fire.”***
– Nat Gillin
- **Learn and learning a lot from reading code, fixing bugs, testing things, reimplementing stuff**
- **“Show me the code” mentality**

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