# Next Steps

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- Learn more about neural networks
  - Neural networks for NLP (Yoav Goldberg's book)
  - Deep learning book (Goodfellow et al.)
  - Stanford's CS231n: Neural Networks for Vision (lecture on YouTube)

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- 2 Read on some "second semester" topics
  - Latent Dirichlet Allocation (LDA) / Topic Models
  - Sequence models: Hidden Markov Models / MEMMs / CRFs (Amanda Stent's NLP course in 2018)
  - Bayesian methods
  - Collaborative Filtering / Recommendations
  - Ranking
  - Bandit problems (Thompson sampling / UCB methods)
  - Gaussian processes