

Modules

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The modules below are in no particular order (except for the Basics, of course).

1 Basics

- What is a prior?
- What is a posterior and what is posterior inference? → recap of Bayes' rule
- Sampling as an intuitive way of performing inference before diving in the realms of VI?
- Example problems: Factorial HMMs, Bayesian Mixture Models (show GMs)
- ELBO derivation I: from KL divergence
- ELBO derivation II: with Jensen's inequality
- Connection to EM
- Mean Field inference
- Application to example problems (show GMs)

2 Conjugate Models

- Exponential families
- Gaussian-Gaussian conjugacy
- Example: Bayesian Linear Regression
- Beta-Binomial warmup for Dirichlet-multinomial?
- Dirichlet-multinomial conjugacy

- Example: LDA
- Conjugate VI in the general case ([Beal, 2003](#))

3 Nonconjugate Models

- Laplace Approximation
- Gradient methods
- Problem: cannot simply differentiate an MC average
- Idea: transform $\frac{d}{dq}\mathbb{E}_q[\cdot]$ into $\mathbb{E}[\frac{d}{dq}\cdot]$
- Score function gradient \rightarrow Black Box VI ([Blei et al., 2012](#); [Ranganath et al., 2014](#))
- Reparametrisation gradient ([Kingma and Welling, 2013](#); [Rezende et al., 2014](#); [Titsias and Lázaro-gredilla, 2014](#))

4 Stochastic algorithms

- Stochastic optimisation ([Robbins and Monro, 1951](#))
- SVI ([Hoffman et al., 2013](#))

5 Bayesian Neural Networks

- Putting priors on weights
- The old stuff by Neal, MacKay and Hinton ([Hinton and van Camp, 1993](#))
- The new stuff by DeepMind et al. ([Graves, 2011](#); [Blundell et al., 2015](#))
- Bayesian Interpretation of Dropout ([Gal, 2016](#))

6 Deep Generative Models

- Review of generative models
- Exact case: EM with features ([Berg-Kirkpatrick et al., 2010](#))
- First attempt: Wake-sleep ([Hinton et al., 1995](#))
- Variational Autoencoders ([Kingma and Welling, 2013](#); [Rezende et al., 2014](#))

- Example models: ???
- Code snippet ???
- Extra: The Deep Generative CRF (the Ryan Adams paper from NIPS)

7 Reparametrisation Gradients

I think the whole module should depend on audience and we can cover the location-scale case in the modules about Nonconjugate models and/or DGMs.

- Recap: Gaussian reparametrisation
- Extension to general location-scale families ([Titsias and Lázaro-gredilla, 2014](#))
- ADVI (depending on the audience only go until here; the next two are way more complicated) ([Kucukelbir et al., 2017](#))
- Generalised Reparametrisation Gradient ([Ruiz et al., 2016](#))
- Rejection Sampling VI ([Naesseth et al., 2017](#))

8 Nonparametric Models [Advanced]

- Intro to stick-breaking processes ([Ishwaran and James, 2001](#))
- VI for HDP/PYP ([Wang et al., 2011](#))
- Intro to GPs
- VI for GPs

9 Beyond Mean Field [Advanced]

- Structured VI (example: Bayesian or Factorial HMMs)
- Auxiliary variables
- Hierarchical Variational models

10 Collapsed VB [Advanced]

Another module that depends on audience: people with Bayesian aspirations vs people who want to play with DGMs.

- Taylor expansions
- Example: LDA
- Connection between collapsed VB and unconstrained variational approximation (Teh et al., 2007)
- CVB0 (Asuncion et al., 2009)

11 Beyond KL [Advanced]

- α -divergence (make connection to EP)
- Stein VI
- Implicit models
- Hoelder bound

References

- Arthur Asuncion, Max Welling, Padhraic Smyth, and Yee Whye Teh. On smoothing and inference for topic models. UAI '09, pages 27–34, 2009. URL <https://arxiv.org/pdf/1205.2662.pdf>.
- Matthew J. Beal. *Variational Algorithms for Approximate Bayesian Inference*. PhD thesis, Gatsby Computational Neuroscience Unit, University College London, 2003. URL <http://www.cse.buffalo.edu/faculty/mbeal/thesis/index.html>.
- Taylor Berg-Kirkpatrick, Alexandre Bouchard-Côté, John DeNero, and Dan Klein. Painless unsupervised learning with features. In *Human Language Technologies: The 2010 Annual Conference of the North American Chapter of the Association for Computational Linguistics*, HLT '10, pages 582–590, 2010. URL <http://www.aclweb.org/anthology/N10-1083>.
- David M. Blei, Michael I. Jordan, and John W. Paisley. Variational bayesian inference with stochastic search. In *Proceedings of the 29th International Conference on Machine Learning (ICML-12)*, 2012. URL <http://icml.cc/2012/papers/687.pdf>.

- Charles Blundell, Julien Cornebise, Koray Kavukcuoglu, and Daan Wierstra. Weight uncertainty in neural networks. ICML'15, pages 1613–1622, 2015. URL http://machinelearning.wustl.edu/mlpapers/paper_files/icml2015_blundell15.pdf.
- Yarin Gal. *Uncertainty in Deep Learning*. PhD thesis, University of Cambridge, 2016. URL <http://mlg.eng.cam.ac.uk/yarin/thesis/thesis.pdf>.
- Alex Graves. Practical variational inference for neural networks. In *NIPS*, pages 2348–2356, 2011. URL <http://papers.nips.cc/paper/4329-practical-variational-inference-for-neural-networks>.
- G. E. Hinton, P. Dayan, B. J. Frey, and R. M. Neal. The wake-sleep algorithm for unsupervised neural networks. *Science*, 268:1158–1161, 1995. URL <http://www.gatsby.ucl.ac.uk/~dayan/papers/hdfn95.pdf>.
- Geoffrey E. Hinton and Drew van Camp. Keeping the neural networks simple by minimizing the description length of the weights. In *Proceedings of the Sixth Annual Conference on Computational Learning Theory*, COLT '93, pages 5–13. ACM, 1993. ISBN 0-89791-611-5. URL <http://doi.acm.org/10.1145/168304.168306>.
- Matthew D. Hoffman, David M. Blei, Chong Wang, and John Paisley. Stochastic variational inference. *J. Mach. Learn. Res.*, 14(1):1303–1347, May 2013. URL <http://jmlr.org/papers/volume14/hoffman13a/hoffman13a.pdf>.
- Hemant Ishwaran and Lancelot F James. Gibbs sampling methods for stick-breaking priors. *Journal of the American Statistical Association*, 96(453):161–173, 2001. doi: 10.1198/016214501750332758. URL <http://dx.doi.org/10.1198/016214501750332758>.
- Diederik P. Kingma and Max Welling. Auto-Encoding Variational Bayes. 2013. URL <http://arxiv.org/abs/1312.6114>.
- Alp Kucukelbir, Dustin Tran, Rajesh Ranganath, Andrew Gelman, and David M. Blei. Automatic differentiation variational inference. *Journal of Machine Learning Research*, 18(14):1–45, 2017. URL <http://jmlr.org/papers/v18/16-107.html>.
- Christian Naesseth, Francisco Ruiz, Scott Linderman, and David Blei. Reparameterization gradients through acceptance-rejection sampling algorithms. AISTATS, pages 489–498, 2017. URL <http://proceedings.mlr.press/v54/naesseth17a/naesseth17a.pdf>.

- Rajesh Ranganath, Sean Gerrish, and David Blei. Black Box Variational Inference. In Samuel Kaski and Jukka Corander, editors, *Proceedings of the Seventeenth International Conference on Artificial Intelligence and Statistics*, pages 814–822, 2014. URL <http://proceedings.mlr.press/v33/ranganath14.pdf>.
- Danilo J. Rezende, Shakir Mohamed, and Daan Wierstra. Stochastic backpropagation and approximate inference in deep generative models. In *Proceedings of the 31st International Conference on Machine Learning (ICML-14)*, pages 1278–1286, 2014. URL <http://jmlr.org/proceedings/papers/v32/rezende14.pdf>.
- Herbert Robbins and Sutton Monro. A stochastic approximation method. *The Annals of Mathematical Statistics*, 22(3):400–407, September 1951. URL http://projecteuclid.org/download/pdf_1/euclid.aoms/1177729586.
- Francisco R Ruiz, Michalis Titsias, RC AUEB, and David Blei. The generalized reparameterization gradient. In D. D. Lee, M. Sugiyama, U. V. Luxburg, I. Guyon, and R. Garnett, editors, *Advances in Neural Information Processing Systems 29*, pages 460–468. 2016. URL <http://papers.nips.cc/paper/6328-the-generalized-reparameterization-gradient.pdf>.
- Y. W. Teh, D. Newman, and M. Welling. A collapsed variational Bayesian inference algorithm for latent Dirichlet allocation. In *Advances in Neural Information Processing Systems*, volume 19, 2007. URL <https://www.stats.ox.ac.uk/~teh/research/inference/nips2006.pdf>.
- Michalis Titsias and Miguel Lázaro-gredilla. Doubly stochastic variational bayes for non-conjugate inference. In Tony Jebara and Eric P. Xing, editors, *Proceedings of the 31st International Conference on Machine Learning (ICML-14)*, pages 1971–1979, 2014. URL <http://jmlr.org/proceedings/papers/v32/titsias14.pdf>.
- Chong Wang, John Paisley, and David M. Blei. Online variational inference for the hierarchical dirichlet process. In *Proc. of the 14th Int’l. Conf. on Artificial Intelligence and Statistics (AISTATS)*, volume 15, pages 752–760, 2011. URL <http://jmlr.csail.mit.edu/proceedings/papers/v15/wang11a/wang11a.pdf>.