

# Approximating exponential family models (not single distributions) with a two-network architecture

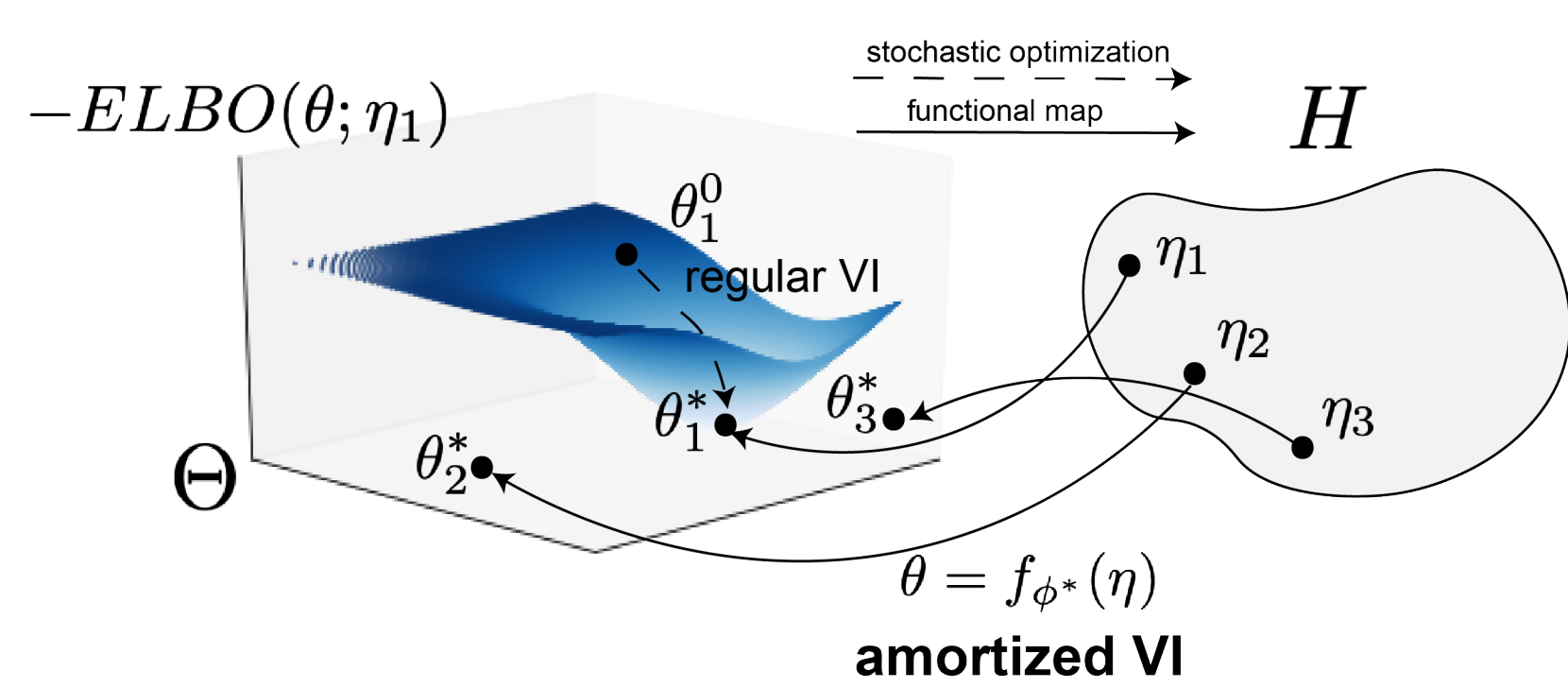
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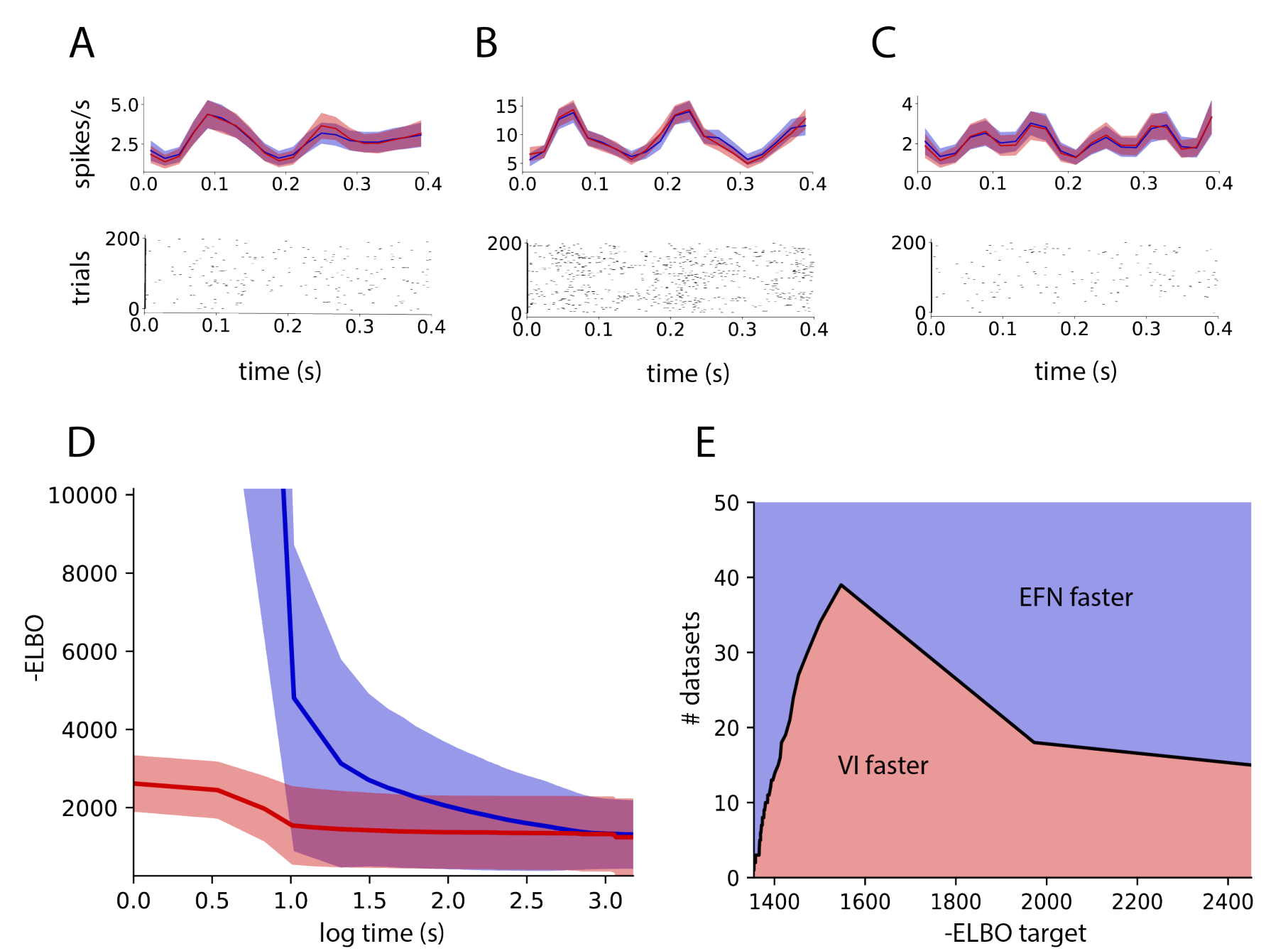
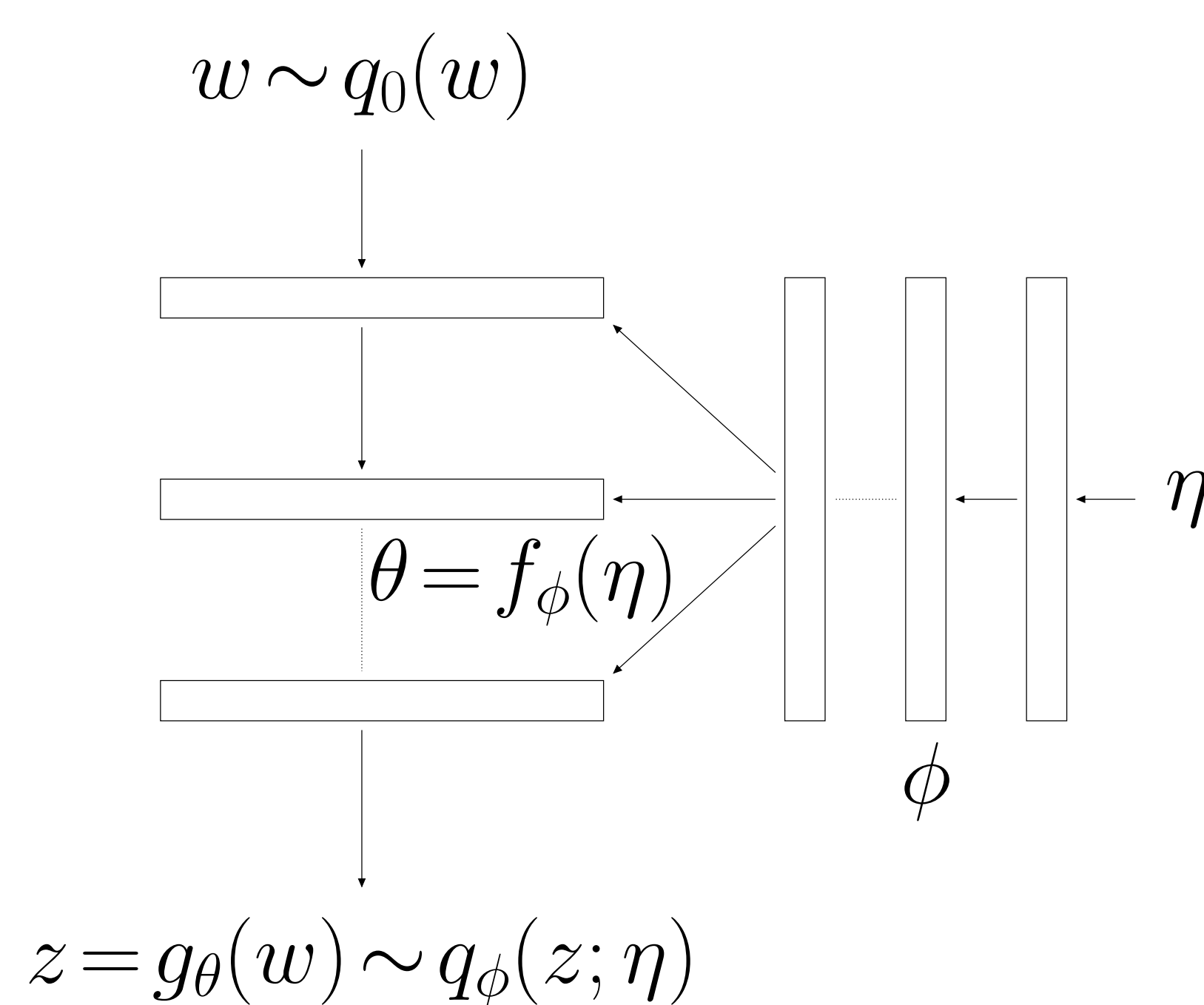
## Motivation

- Variational inference (VI) incurs a cost of optimization to find optimal variational parameters  $\theta^* \in \Theta$  of the approximate inference model.
- Intractable exponential family models
  - an exp fam likelihood
  - i.i.d. observations
  - a nonconjugate prior
- We can learn a smooth function  $f_{\phi^*} : H \mapsto \Theta$  mapping  $\eta$  to  $\theta^*$  using a two-network architecture called exponential family networks (EFN).

## Exponential family networks (EFNs) — More stuff



- EFNs learn models of exponential families (not single distributions), and afford substantial computational savings through amortized VI.



## Methods

### Exponential families as target models

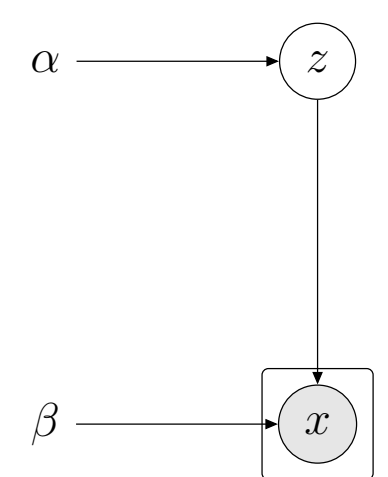
Some text

- Exponential family models  $\mathcal{P}$  have the form

$$\mathcal{P} = \left\{ \frac{h(\cdot)}{A(\eta)} \exp \{ \eta^\top t(\cdot) \} : \eta \in H \right\}$$

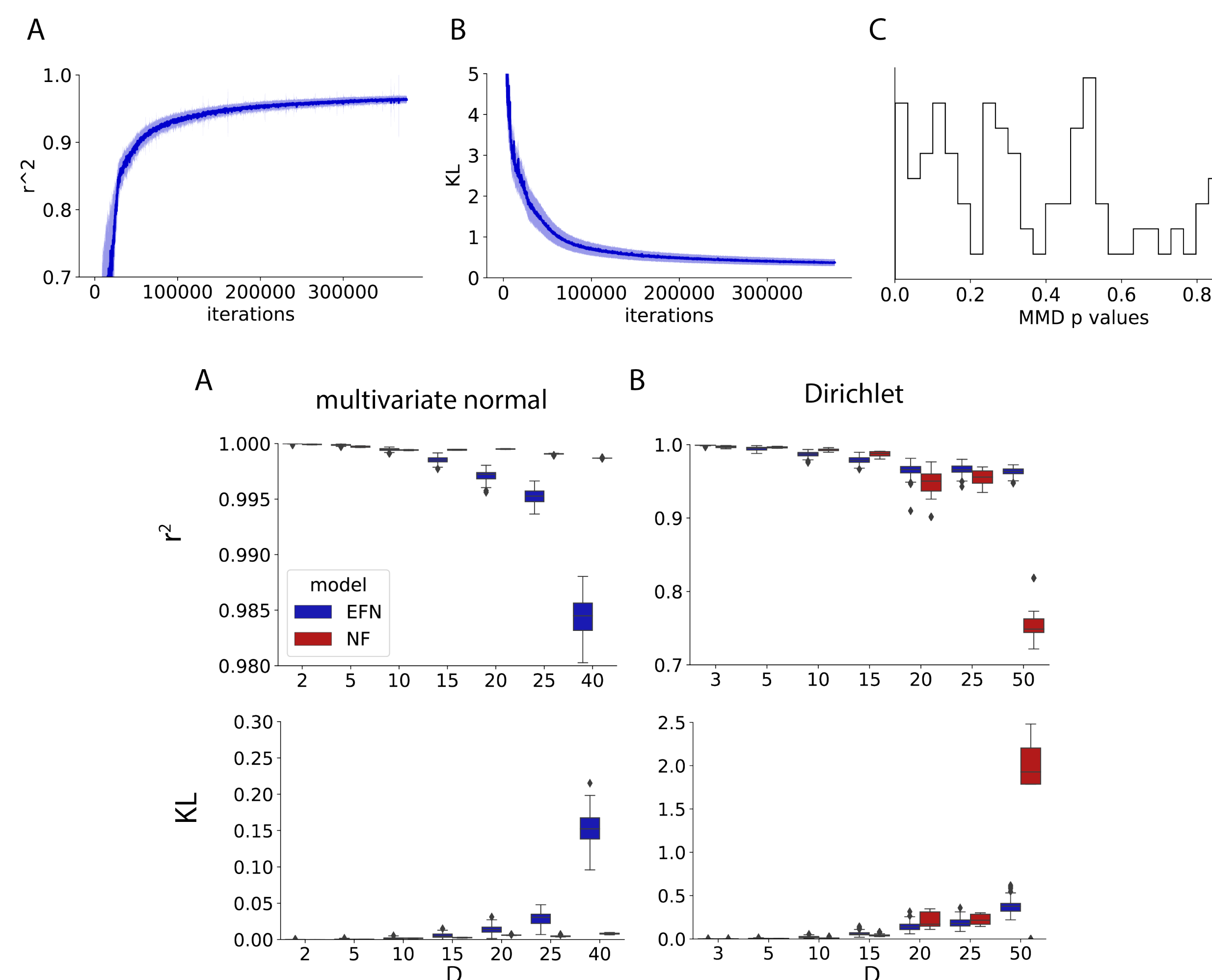
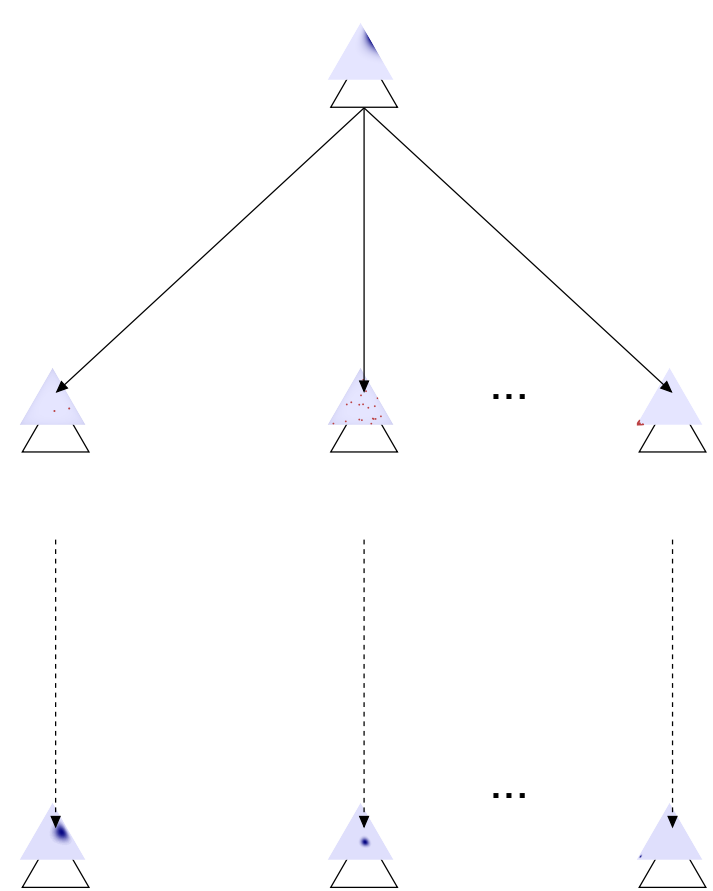
with natural parameter  $\eta$ , sufficient statistics  $t(\cdot)$ , base measure  $h(\cdot)$ , and log normalizer  $A(\eta)$ .

- We focus on the fundamental problem setup of probabilistic inference:  $N$  conditionally independent observations  $x_i$  given latent variable  $z$ .



$$p_0(z) = \frac{1}{A_0(\alpha)} \exp \{ \alpha^\top t_0(z) \}$$

$$p(x_i|z) = \frac{1}{A(z)} \exp \{ \nu(z)^\top t(x_i) \}$$



## Summary

- Summary point 1
- Summary point 2

### References

1. Loaiza-Ganem, G., Y. Gao., and J. P. Cunningham. "Maximum entropy flow networks." ICLR (2017).
2. Dipoppa, M., et al. "Vision and locomotion shape the interactions between neuron types in mouse visual cortex." Neuron (2018).
3. Mastrogiuseppe, F., and S. Ostojic. "Linking connectivity, dynamics, and computations in low-rank recurrent neural networks." Neuron (2018).

### Acknowledgements

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