

Variational Inference with Normalizing Flows

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Introduction

- Calculating the true posterior distribution of inference tasks is in most cases an intractable problem.
- Lots of research on approaches for efficient approximation of the posterior, however the resulting classes prove to be of limited expressiveness.
- The authors in [?] introduce the notion of normalizing flows, sequences of invertible transformations applied to a simple initial density, to efficiently create more expressive families of candidate posteriors to be used for variational inference.
- We compare the performance of different types of normalizing flows on the MNIST dataset.

Our Work

- Reproduced experiment on MNIST using Linear Normalizing Flows
- Reproduced experiment on MNIST using NICE
- Extended the ideas of the paper and experimented with Invertible Convolutional Flows
- Created open-source Github repository with code and results: github.com/ATML-Group-12/normalising_flows

Theoretical Background

Normalizing flows are sequences of invertible, smooth mappings $f : \mathbb{R}^d \rightarrow \mathbb{R}^d$. We define a flow of length K to be K such transformations as follows:

$$\mathbf{z}_K = f_K \circ \dots \circ f_2 \circ f_1(\mathbf{z}_0)$$

where \mathbf{z}_0 is a random variable that has a simple initial distribution, and \mathbf{z}_K is the corresponding random variable after applying the transformations. The properties of normalizing flows allow us to calculate the log-density of \mathbf{z}_K efficiently, using the change of variable theorem:

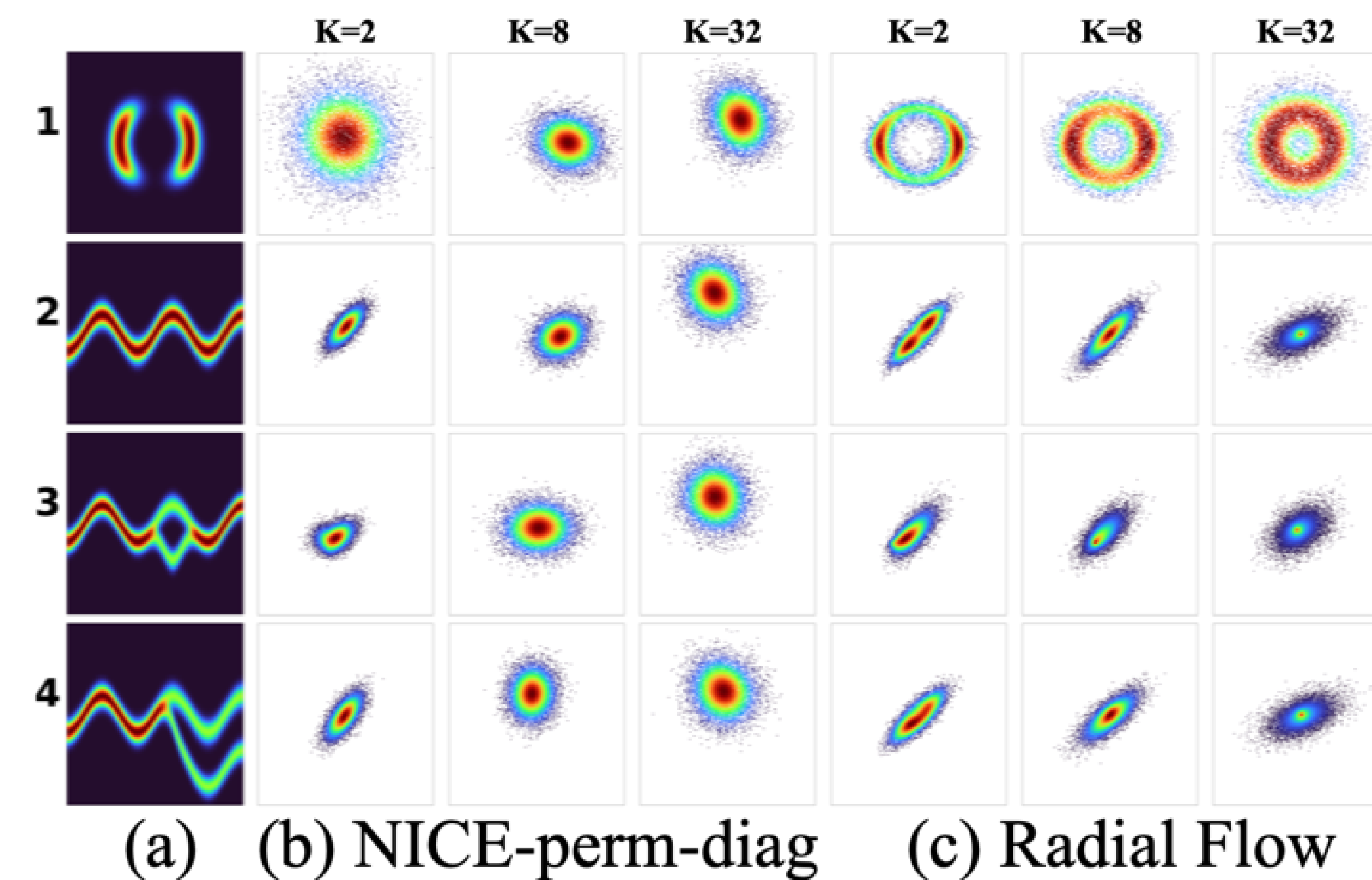
$$\ln q_K(\mathbf{z}_K) = \ln q_0(\mathbf{z}_0) - \sum_{k=1}^K \ln \left| \det \frac{\partial f_k}{\partial \mathbf{z}_{k-1}} \right|$$

The normalizing flow is defined as the path of the successive distributions q_K .

The first class of flows that we use in our experiments is that of linear flows. More specifically, we consider **planar** and **radial** flows, which perform series of contractions and expansions in the direction perpendicular to a fixed hyperplane and around a reference point respectively.

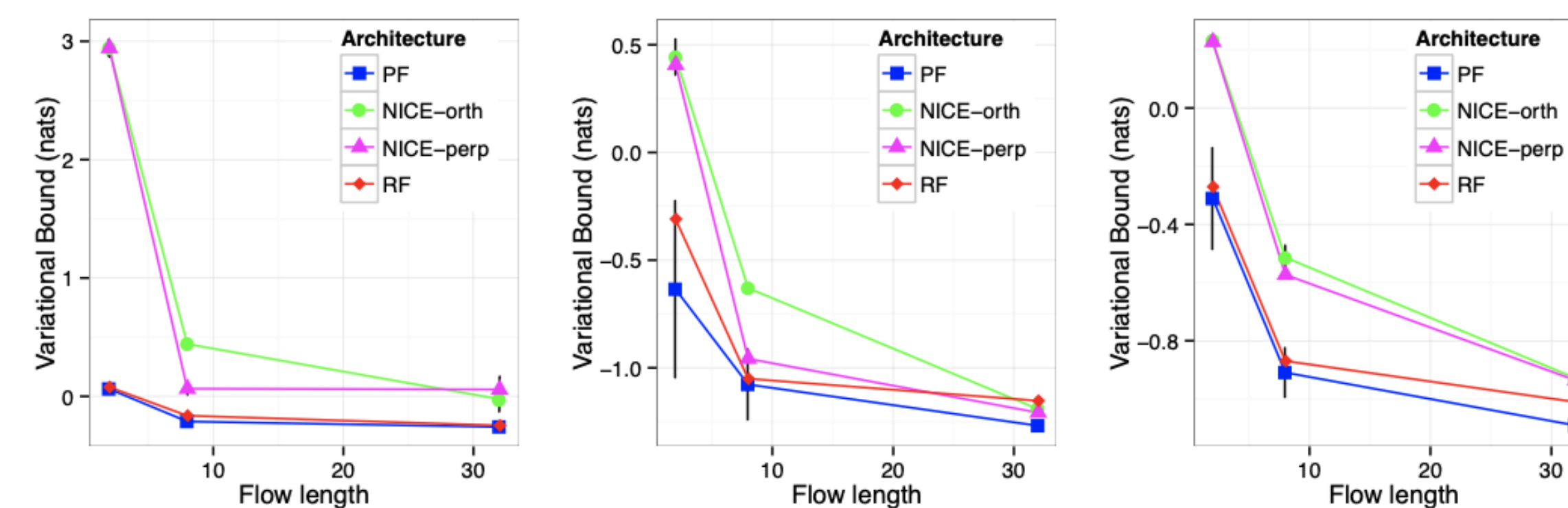
Theoretical Background

Normalizing flows



Experiments

- DLGM + NF
- DLGM + NICE



(d) Comparison of KL-divergences.

Results

TODO!!! Results...

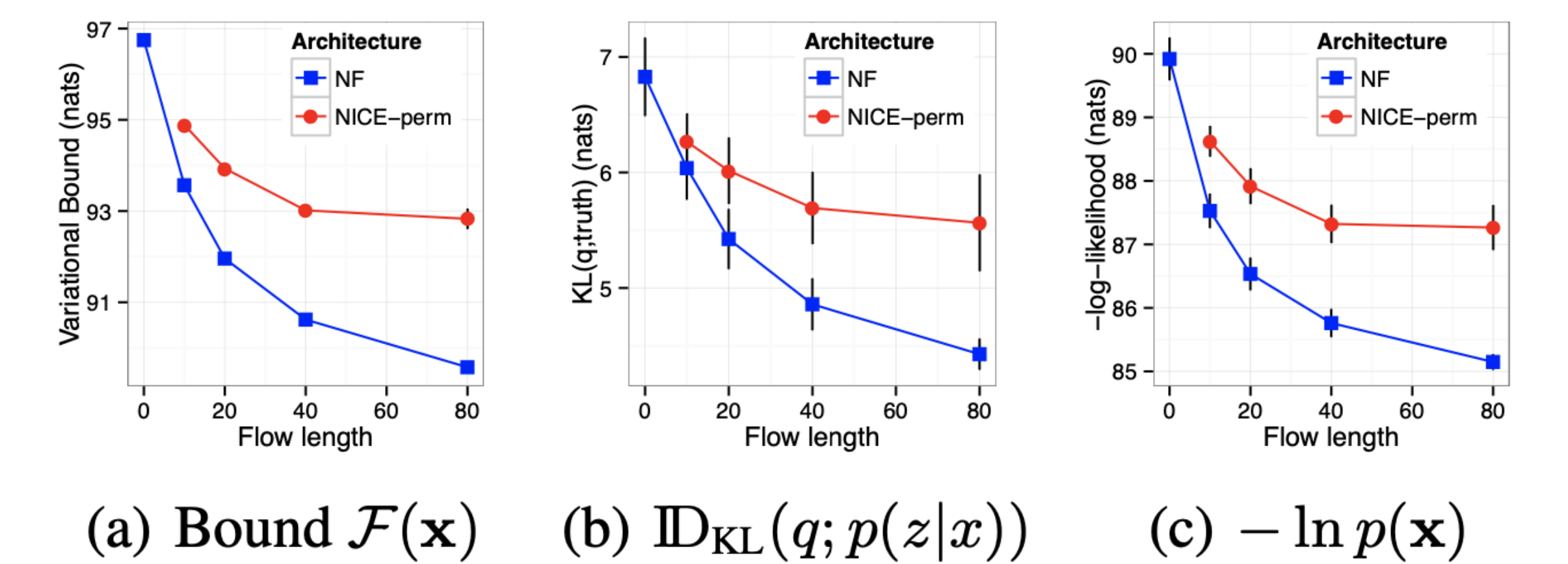


Figure 4. Effect of the flow-length on MNIST.

Our Improvements and Extensions

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References