

# Cheap calibration of the normalizing flow using arbitrary classifier

by Robert Drynkin, Maxim Artemev, Denis Derkach

April 2020

- Assumptions
- Idea
- Implementation details
- Pictures
- Current results
- FAQ

# Assumptions

## Normalizing flows

- - Long time to train
- - Unstable training
- - Have an artefacts due to continuity
- + Generative model

Can we use binary classifier to improve normalizing flow with the preservation of useful properties?

## Binary classifier

- + Fast training
- + A lot of libraries
- + Well studied
- - Not a generative model

$Q(x)$  – density of the real distribution

$P(x)$  – density of the distribution given by nf

$clf(x)$  – prediction of the perfect discriminator (binary classifier which trained to distinguish samples from  $Q$  and  $P$ )

We can express  $Q(x)$  using only  $P(x)$  and  $clf(x)$

$$clf - \text{perfect} \Rightarrow clf(x) = \frac{Q(x)}{P(x) + Q(x)} \Rightarrow \frac{clf(x)}{1 - clf(x)} = \frac{Q(x)}{P(x)}$$

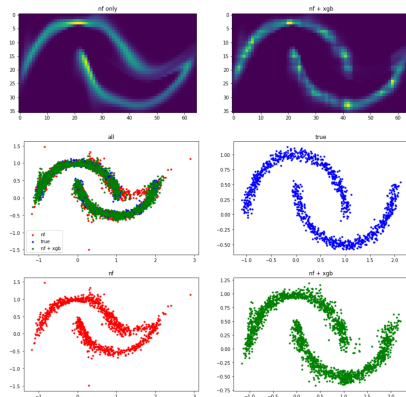
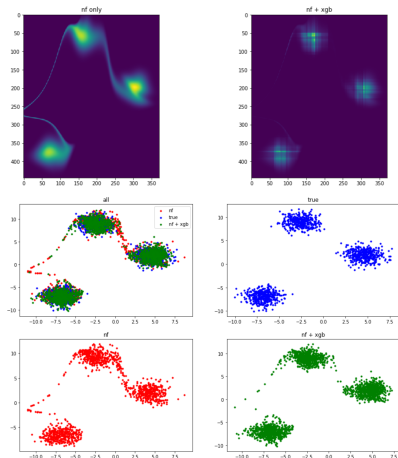
$$Q(x) = P(x) \frac{clf(x)}{1 - clf(x)}$$

For the numerical stability:

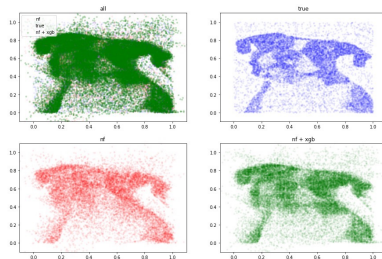
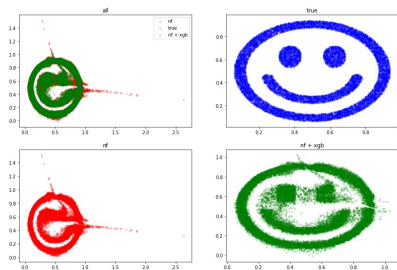
$$\log Q(x) = \log P(x) + \text{clf\_logit}(x)$$

Sampling: rejection sampling from  $Q(x)$  using  $c \cdot P(x)$  as a major distribution. Acceptance rate equals to  $1/c$ , where  $c = \max \exp \text{clf\_logit}(x)$

# Pictures



# Pictures



## GAS dataset

- SOTA: maf(10)  $\rightarrow$  10.08 log likelihood
- Ours: maf(5)  $\rightarrow$  9.48 log likelihood  $\rightarrow$  10.23 ll after calibration by Catboost(2k trees)



# Future work

- KL finetuning nf
- Conditional sampling

- What happens if  $P = Q$  (perfect nf)?  
Then  $\forall x \text{ clf\_logit}(x) = 0$ , acceptance rate of rejection sampling = 1.
- What happens if  $P$  is trivial (e. g. uniform)?  
Then  $\exists x \text{ clf\_logit}(x) \rightarrow \infty$ , acceptance rate of rejection sampling  $\rightarrow 0$ .

# The End