

Master's Thesis

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# My Super Fancy Thesis Title

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Department of Statistics  
Ludwig-Maximilians-Universität München

**Jane Doe**

Munich, Month Day<sup>th</sup>, Year



Submitted in partial fulfillment of the requirements for the degree of M. Sc.  
Supervised by Dr Dre

## **Abstract**

This is my new abstract.

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# 1 Introduction

## 1.1 Motivation

## 1.2 Problem

- in high dimensional, complex posteriors we are often confronted with ill-conditioned Hessians
- this might lead to slow convergence when the sampler is not able to accord to this
- we can show this effect on some simple gaussian (mixture) examples

Note: the visualizations show the true density in the background, while compute the steps of the MCMC sampler on samples of the log density.

## 1.3 Research Objective

- use preconditioning
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By investigating this research objective we contribute to the field of MCMC sampling in the following ways:...

Name here what was developed in the end and how it contributes to the field.

## 1.4 Outline

# 2 Method

## 2.1 General Idea

- best case would be to compute the inverse Fisher Information Matrix in each step as preconditioning to apply to local geometry of the current position
- cannot be done in our MCMC preconditioning context since it is much too expensive to estimate the FIM in every step from scratch
- we cannot do a Hessian approximation like RmsProp or Adam, since they utilize values depending which depend on previous steps. This does then not meet the markov property (MCMC theory breaks)
- this is why we optimize first obtaining a high likelihood solution and a rough estimation of the local geometry (via Adam like Hessian approximation or IVON)
- using this information might help shorten the warmup/burnin phase without hurting the performance

- during sampling we then want to efficiently sample the a local subspace of the high dimensional posterior where the optimizer landed
- multimodality is tackled through ensembling and lengthy markov chains

A concise summary of contents and results

## A Appendix

Additional material goes here

## **B Electronic appendix**

Data, code and figures are provided in electronic form.

## Declaration of authorship

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