

# The Agony and the...

## Master Thesis Proposal

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# Agenda

- 1 Introduction
- 2 Research Question
- 3 Literature Review
- 4 Research Design
- 5 Expected Contribution
- 6 Methodology
- 7 Preliminary Data
- 8 Plan

# Introduction

- **Context:** Modern distributed systems generate massive logs.
- **Pain Point:** Manual analysis is error-prone and slow.
- **Goal:** Develop a hybrid deep-learning model for real-time detection.

## Research Gap

Existing solutions struggle with high-dimensional, unlabeled data streams.

# Research Question

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# Literature Review

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# Research Design

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# Expected Contribution

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# Proposed Mathematical Model

We assume the state transition follows a Markov process. The posterior probability is updated as follows:

$$p(\mathbf{x}_k | \mathbf{Z}_k) = \frac{p(\mathbf{z}_k | \mathbf{x}_k)p(\mathbf{x}_k | \mathbf{Z}_{k-1})}{\int p(\mathbf{z}_k | \mathbf{x}_k)p(\mathbf{x}_k | \mathbf{Z}_{k-1}) d\mathbf{x}_k} \quad (1)$$

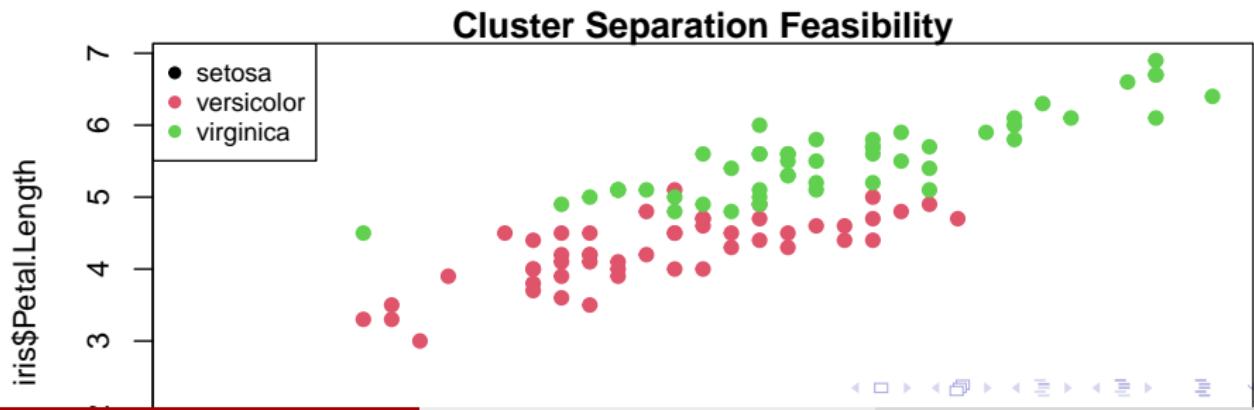
Where:

- $\mathbf{x}_k$  is the system state at time  $k$
- $\mathbf{Z}_k$  is the observation history

# Preliminary Data Analysis (R Integration)

Using R directly in slides ensures plots are always up-to-date with your data.

```
# We can generate plots dynamically
par(mar=c(4,4,1,1))
plot(iris$Sepal.Length, iris$Petal.Length,
      col=iris$Species, pch=19,
      main="Cluster Separation Feasibility")
legend("topleft", legend=levels(iris$Species), col=1:3, pch=19)
```



# Project Timeline

Phase	Deadline
Literature Review	Month 1
Data Collection	Month 2
Implementation (Prototype)	Month 3-4
Evaluation & Tuning	Month 5
Writing Thesis	Month 6

Table: Estimated milestones for the 6-month period.

## Next Steps

**Thank you for your attention.**

Questions?