Consider the following Java-JDT plugin name in German: Consider the following Java-JDT plugin name in German:

BOSTON UNIVERSITY COLLEGE OF ENGINEERING

Dissertation

A BU THESIS LATEX TEMPLATE

by

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Facilis descensus Averni;
Noctes atque dies patet atri janua Ditis;
Sed revocare gradum, superasque evadere ad auras,
Hoc opus, hic labor est.

Virgil (from Don's thesis!)

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A BU THESIS LATEX TEMPLATE

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ABSTRACT

Have you ever wondered why this is called an abstract? Weird thing is that its legal to cite the abstract of a dissertation alone, apart from the rest of the manuscript.

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List of Abbreviations

The list below must be in alphabetical order as per BU library instructions or it will be returned to you for re-ordering.

CAD	 Computer-Aided Design
CO	 Cytochrome Oxidase
DOG	 Difference Of Gaussian (distributions)
FWHM	 Full-Width at Half Maximum
LGN	 Lateral Geniculate Nucleus
ODC	 Ocular Dominance Column
PDF	 Probability Distribution Function
\mathbb{R}^2	 the Real plane

Introduction

1.1 Motivation

1.2 Problem at hand

Hello

1.3 Structure of thesis

Works

1.4 Conclusion

The next chapter gives an in-depth view of the pipeline used by the current state of art technology for query optimization in traditional data bases including the mathematical knowledge for simplification and the overall framework. The next chapter also introduces the reader to data stream and how data bases are used for them called DSMS and showcases an approach to optimize queries on data streams for the problem discusses above. The following chapter list out the details if implementation, challenges face, evaulation methods used, benchmark test case timings, followed by a summary of the paper.

Related Work

2.1 Introduction, Query optimization

A simple SQL query looks like this

```
SELECT *

FROM my_table

WHERE condition
```

- 2.2 Relational algebra,
- 2.3 Converting SQL queries to parse trees
- 2.4 Converting Parse trees into logical expression
- 2.5 Explain difficulties/ Time complexity
- 2.6 Optimization using relation algebra
- 2.7 Introduction to Data Streams
- 2.8 Data stream windowing
- 2.9 Query Processing of data streams(Combine the DBMS and DSMS)
- 2.10 Challenges of query optimization on data streams
- 2.11 Conclusion and discussion
- 2.12 SQL Query compiler

The steps involved are

2.12.1 Parsing

In a very general sense, given an SQL query, SQL converts it into a parse tree based on SQL grammar.

2.12.2 Preprocessing

This step has several functions.

If a "view" is used in the query as a relation, then each instance has to be replaced by the parse tree. The preprocessor also has to conduct semantic checking, that is, check if relations used exist, check for ambiguity, and type checking. If a parse tree passes the preprocessing then it is said to be **valid**

2.12.3 Logical Query Plan

The first step is to modify the parse tree into using operators and operators of relational algebra.

The next step is to convert expression obtained from the above substitution and modify it into an expression which can be converted to most efficient physical query plan.

To improve the algebraic expression obtained, few common steps taken are pushing down selections and projections carefully, carefully placing duplicate eliminations, combining selections, showing associativity and commutativity in the expression to help with enumeration.

At the end when we have the expression ready, we enumerate the physical plans and calculate their cost of execution and select the method with the lowest cost.

2.12.4 Cost Estimation

We need to consider what algorithm each operator in the expression is going to use, such as join, sort, scanning and more. Also need to consider the order for the associative and commutative operators, because at the end the operators are binary and how the output of one operator is provided as an input to the next/outter operator in the expression.

2.13 System R

2.14 Deep Reinforcement learning

Markov decision process(MDP) is used to formalize various types of stochastic processes. In MDPs, the goal of the agent is to make a sequence of actions to optimize/maximize an objective function.

Formally a MDP is a 5-tuple

$$\langle S, A, P(s, a), R(s, a), s_0 \rangle$$

 $S \to Set$ of all possible states the agent can be in.

 $A \rightarrow Set \ of \ all \ possible \ actions \ the \ agent \ can \ take.$

 $P(s,a) \rightarrow A$ probability distribution of going to various states given current state

and action.
$$s^1 \sim P(s, a)$$

 $R(s,a) \rightarrow Reward for taking action a on state s.$

 $s_0 \rightarrow Describes$ the initial state of the system/ agent.

The performance of the agent is measured using the rewards collected along the way through various states. So the objective of an MDP is to find a policy $\pi: S \to A$, a function that maps states to actions, in order to maximize the expected value:-

$$\operatorname*{argmax}_{\pi} \mathbb{E} \left[\sum_{t=0}^{T-1} R(s_t, a_t) \right]$$

subject to
$$s_{t+1} = P(s_t, a_t), a_t = \pi s_t$$

This method does not reduce the search space, and unlikely greedy solution, this will lead to an optimal solution. This method does not reduce the search space, and unlikely greedy solution, this will lead to an optimal solution.

Reinforcement learning (RL) is a technique which optimizes MDPs iteratively, by running a simulation in each iteration and changing the policy to find an optimal one based on the cumulative reward.

2.15 Relations

A common method/ data structure used to formalize joins

Query Graph \to A query graph G is an undirected graph, where each relation R is a vertex and each join predicate ρ defines an edge between 2 vertices. Let κ_G denote the number of connected components in G

A join of relation R_1 , R_2 , in the graph corresponds to remove the vertices v_{R_1} , v_{R_2} , replacing them with a vertex $v_{R_1+R_2}$, the edges of the form $(v_{R_1}, v)\&(v_{R_2}, v)$ are replaced by $(v_{R_1+R_2}, v)$. Note each reduction reduces number of vertices by one, so this process is repeated until there are κ_G number of vertices left.

Join Optimization Problem \rightarrow Let G be a query graph and J be a join cost model. Find sequence, $c_1 \circ c_2 \circ ... \circ c_n$ resulting in $|V| = \kappa_G$ to minimize

$$\min_{c_1, c_2, \dots c_n} \sum_{i=1}^n J(c_i)$$

subject to
$$G_{i+1} = c(G_i)$$

Using these definitions, we define a MDP.

$$\langle \{G_0, G_1, ...G_T\}, c, P(G, c), -J, G \rangle$$

We are still not certain about how the cost function J is structured. We are still not certain about how the cost function J is structured. We are still not certain about how the cost function J is structured.

We are still not certain about how the cost function J is structured.

Stream Optimization

${\bf Implementation}$

Stream Optimization

Stream Optimization

Appendix A

Proof of xyz

This is the appendix.

CURRICULUM VITAE

Joe Graduate

Basically, this needs to be worked out by each individual, however the same format, margins, typeface, and type size must be used as in the rest of the dissertation.