

PLANNING AND UAV MISSION EVALUATION FOR INTRALOGISTICS PROBLEM

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Resumo— Este artigo apresenta o desenvolvimento de planejadores de missão na intralogística para um veículo aéreo não tripulado comercial, equipado com uma garra robótica, em um ambiente industrial onde há almoxarifado de insumos, linhas de produção e depósito de produtos. Neste trabalho, o planejador gera comandos necessários para realizar uma missão a qual compreende desde a entrega de insumos trazidos do almoxarifado à linha de produção, até a entrega do produto final ao cliente. Foram desenvolvidas duas abordagens diferentes para planejamento de missão: na primeira abordagem, utilizou-se uma simples heurística que resolve o problema; já na segunda abordagem, utilizou-se uma técnica com escalonamento de tarefas (processo de produção). Estas abordagens seguem algumas regras de produção que serão apresentadas ao longo deste trabalho. Foi realizada uma avaliação dos planejadores de missão desenvolvidos, verificando o custo de ambos, realizando algumas medidas de tempo de execução, bem como comparando estes resultados com o custo ótimo obtido com a ferramenta de otimização CPLEX.

Palavras-chave— Planejamento de Missão, Sistemas de Manufatura.

Abstract— This paper presents the development of mission planners in intralogistics for a commercial unmanned aerial vehicle equipped with a robotic gripper in an industrial environment where there are a warehouse of inputs, production lines and a product warehouse. In this work, the planner generates the necessary commands to carry out a mission that includes everything from the delivery of inputs brought from the warehouse of inputs to the production line until the final product is delivered to the customer (product warehouse). Two different approaches were developed for mission planning: in the first approach, a simple heuristic was used to solve the problem; in the second approach, a technique with task scheduling (production process) was used. These approaches follow some production rules that will be presented throughout this work. An evaluation of the mission planners developed was performed, verifying the cost of both, performing some measures of execution time, as well as comparing these results with the optimum cost obtained with the IBM ILOG CPLEX optimizer.

Keywords— Mission Planning, Manufacturing Systems.

1 Introduction

Logistics has become a competitive and fundamental factor for organizations, involving the management, conservation and supervision of freight transport. In addition, excellent logistics means customer satisfaction, so speed is still an important factor in a successful logistics process (Service-drone, 2014). Currently, one of the solutions to this type of problem is the use of Unmanned Aerial Vehicles (UAVs). Nowadays, UAVs are mostly remotely piloted vehicles (RPV), since their operations are carried out by ground operators. If the tasks performed by a UAV were performed autonomously, it would relieve the work of these operators, since they perform tedious and repetitive tasks (Pascarella et al., 2013).

A probable improvement of these logistics systems is the increase of the automation of the UAVs, what results in minimization of the costs, e.g. in terms of time. Consequently, investments and studies related to stand-alone UAVs are important to the development of smart factories (Hern, 2014). However, one of the main problems with the use of autonomous UAVs is the reliability and intelligence of the system. Thus, increased employment of autonomous UAVs requires the development of devices that are capa-

ble of performing tasks and interacting with the environment intelligently and reliably.

Autonomous UAVs need to know what will happen in a future instant and what is the best decision to make at the present time, and therefore require strategies not only to decompose their missions into meaningful sub-tasks but also to track progress toward mission goals and the evolution of these tasks relative to the capabilities of autonomous UAVs (Finn and Scheduling, 2012). As a consequence, in order to perform a mission successfully, it is recommended to make a plan to the task (Garecht, 2010). Mission planning problems consist of planning events to meet certain requirements associated with the plan and improving mission objectives (Krozel, 1988). Therefore, this is one of the main challenges faced in solving this type of problem.

Some researches about evaluation and optimization of mission planning have been done in the last years. Schwarz and Sauer (2012) have used ant colony to optimize missions for an automated guided vehicle (AVG). Another paper investigates energy consumption for a factory and evaluate the logistic planning processes using a metric of evaluation (Müller et al., 2012).

This paper presents a methodology that evalu-

ate the cost of mission planners for a commercial UAV. We developed a evaluation metric that evaluate the relative cost of a planning strategy related with the optimal cost generated by the CPLEX optimizer.

Summarizing, the main contributions of this work are:

- an evaluation methodology for UAV mission planners algorithms, [IB: fale um pouco mais][IB: nao sei oq mais falar rs];
- use of a commercial UAV in intralogistics.

The remaining of this work is organized as follows: Section 2 presents some previous works related to mission planning, optimization and evaluation, Section 3 provides the fundamentals of mission planning and optimization problems, Section 5 explains the proposed evaluation methodology in details, Section 6 shows the experimental procedures and results in order to explore and demonstrate the potential of methodology, and, finally, Section 7 concludes the work.

2 Related works

In the literature, there are some attempts to implement UAV guidance systems that perform mission planning. Doherty et al. (2009) presented an architecture of a framework for mission planning and execution tracking applied to an unmanned helicopter. During the execution of the mission, knowledge was acquired through sensors which was used to create state structures. These structures will allow the construction of a logical model, representing the real development of the system and its environment over time. Then, the planning and monitoring modules use temporal action logic (TAL) to reason about actions and changes. The NASA/U.S. Army autonomous helicopter project has developed a guidance system for the autonomous surveillance planning problem for multiple and different targets (Whalley et al., 2005), which generates mission plans using a theoretical approach to decision making. A high-level standalone control is provided by the framework Apex (Baer-Riedhart, 1998), a reactive procedure-based scheduler/planner used to perform mission-level tasks. Apex synthesizes a course of action primarily by linking elemental procedures expressed in procedural definition language (PDL), a notation developed specifically for the Apex reactive planner. This guidance system was integrated into a robotic helicopter and tested in more than 240 scenarios.

A similar project, called Ressac (Research and Rescue by Cooperative Autonomous System), was conducted by the French Aerospace Laboratory (ONERA) for a search and rescue scenario (Fabiani et al., 2007). This architecture

for an exploration mission was developed based on the idea of decomposing the mission into a sequence of tasks or macro-actions associated with rewards. The problem was modeled using a Markov decision process framework (MDP) and dynamic programming algorithms for mission planning. Konigsbuch (Teichteil-Königsbuch and Fabiani, 2007) extends the guidance system and integrates with a robotic helicopter.

Finally, the German Aerospace Center (DLR) has also developed a mission management system based on the behavior paradigm (Adolf and Andert, 2010), which has been integrated with the ARTIS helicopter and validated in different scenarios, including follower of waypoints and search and tracking mission.

3 Background

3.1 Mission Planning

Firstly, a mission can be defined as a goal that need to be completed. In the context of this work, the mission of the UAV is delivery of packages according to a set of well defined rules. A definition to mission planning for UAV is the process of planning the locations to visit (waypoints) and the actions that the vehicles can perform (loading/dropping a load, taking videos/pictures, etc.), typically over a time period (Ramirez-Atencia et al., 2014). An important term in this work is the concept of planner which is the agent (software implementation) that generate a mission. Functionally, mission planning lies above the trajectory planning process, where the mission planner generates a desired mission plan, and then the trajectory planner generates the flight plan (trajectories) between the waypoints.

3.2 Optimization Problems

An optimization problem is about finding the best solution (relative to a certain criterion) among a set of available alternatives. For example, the popular bin packaging problem that aims to find the number of boxes of a certain size to store a set of objects of indicated sizes; optimization involves, for example, finding the least amount of boxes. An optimization problem is usually represented as follows:

$$\begin{aligned} \min \quad & f(\mathbf{x}), \\ \text{s.t.} \quad & \mathbf{x} \in \Omega. \end{aligned} \tag{1}$$

An optimization problem can be defined as a finite set of variables, where the correct values for the variables specify the optimal solution. If the variables are of the set of real, the problem is called continuous, and if they can only have a finite set of distinct values, the problem is called combinatorial (Francq, 2011).

4 UAV Movement System

In this section, we first investigate the platform used in this work, verifying the hardware characteristics and how to control it by software. The core hardware of the UAV IRIS+ is the Pixhawk and we can control it using a Python library - dronekit that uses MAVLink¹ protocol.

Secondly, we integrate a robot gripper to the UAV to take and leave packages during the mission. We have connected a servo motor to the Pixhawk by one of the PWM (pulse width modulation) outputs. After that, we created movement functions to make the UAV movement, such as go to a location, and control it using some resources of the dronekit api.



Figure 1: UAV Movement Systems.

The movement functions are shown below:

- **TakeOff:** takeoff command of the UAV;
- **GoTo:** command to move the UAV to certain location;
- **TakePackage:** command to collect an input/product through a robot gripper;
- **LeavePackage:** command to leave an input or product from a robot gripper;
- **Wait:** command to make the UAV to hover (wait);
- **Land:** command to make the UAV land.

As we can see in the Figure 2, there is a mission planner that generates the mission through the existing inputs in the warehouse and the client's order. After mission generation (using some algorithmic strategy), the UAV receives the mission and executes it. In the Section 5 there is a more comprehensive explanation of the terms used in this current section.

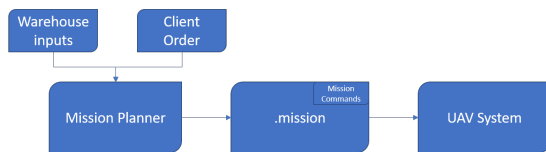


Figure 2: System's Architecture.

¹MAVLink or Micro Air Vehicle Link is a protocol for communicating with small unmanned vehicle. It is designed as a header-only message marshalling library

5 Methodology of Time Cost Evaluation, UAV use and Mission Planning

In this section, there will have a brief clarification of the contents of this paper such as the way we used to evaluate the algorithm cost of the mission, a study case to a UAV and contents about mission planning.

5.1 Case Study

In order to model the mission planning problem as an optimization problem, the case study shown in Figure 3 is used.

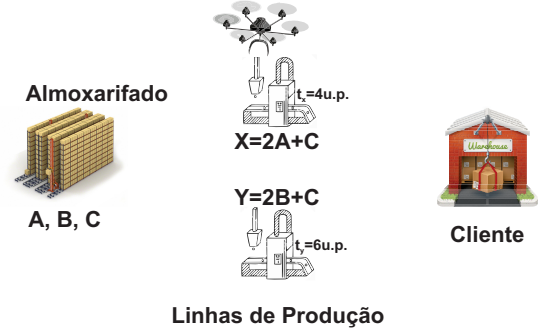


Figure 3: Case Study Representation [IB: traduzir figura e numerar/nomear linhas de produção].

Some key definitions related to the case study and the application developed in this paper need to be clarified. All definitions below will be adopted in the remainder of this work.

Definition 1 Mission Command

Mission command is a command created to execute a task such as go to one location to another location, get a package using a robot gripper, UAV landing, etc.

Definition 2 Mission

Mission is the set of steps and mission commands that the UAV executes to produce the customer's order.

Definition 3 Warehouse

Warehouse is the set of stored raw material available until the moment of entering the productive process. The raw materials, i.e., the inputs available in this work are inputs A, B and C.

Definition 4 Order

Order is the requisition of products made by the client. In this work the products are of type X and Y.

Definition 5 Production Time

Production time is the time required to produce a product X or Y, after having available all the necessary inputs for the production, given by the production rule.

Definition 6 *Production Rule*

Production rule tells what and how many inputs are needed to produce a particular product.

Definition 7 *Mission Planner*

Mission planner is the agent who performs the planning of a mission, that is, generating all the steps and commands necessary to carry out a mission.

Definition 8 *Mission File*

The mission file is a file that was created for the context of this work, with the extension .MISSION containing the mission itself.

The Figure 3 shows that there are three types of inputs in the warehouse (*i.e.*, A, B, and C) and two production lines that produces two different products (*i.e.*, X and Y). Each production line produces only one type of product and has a characteristic production time. In the Figure 3 shows that to produce a product of type X, two inputs of type A and one input of type C are required, and to produce a product of type Y, two inputs of type B and one input of type C are required. The production time of a X product is $4p.u.$ and the time of production of product Y is $6p.u.$. A production unit ($1p.u.$) was considered to be a GoTo command performed by the UAV.

The task to be performed is the production of the customer order, where the UAV will collect supplies from the warehouse, take to the production line and once the production of a certain product is finished, it will lead to the customer.

5.2 Modelling an UAV Intralogistics Mission as an Optimization Problem

The purpose of this subsection is to elaborate a modelling of the mission planning problem into an optimization problem. In order to find, afterwards, the shortest execution time of all the tasks (minimization), based on the case study explained in Section 5.1.

The notation used is given below:

- $\mathcal{T} = \{T_j | j \in \mathbb{N}^*, j \leq N\}$ is the set of N tasks;
- $\mathcal{M} = \{m_i | i \in \mathbb{N}^*, i \leq M\}$ is the set of M production lines (machines);
- $\mathcal{P} = \{p_j | j \in \mathbb{N}^*, j \leq N\}$ is the processing time of each j -th task;
- $\mathcal{S} = \{s_i | s \in \mathbb{N}^*, i \leq M\}$ is the setup (production) time of each i -th production line;
- $R||C_{max}$ (Three-field notation² used in the literature to represent a given task scheduling problem, in this case, minimizing the

²Notation $\alpha|\beta|\gamma$, where α : processing environment, β : problem constraints and γ : optimization criterion.

makespan³ in an unrelated parallel machine environment (Graham et al., 1979).

Decision variable The variable x_{ij} is a binary decision variable which takes the value 1 if the task j is running on the machine i , and 0 otherwise. The variable C_{max} is the variable that we want to optimize.

$$x_{ij} = \begin{cases} 1, & \text{if the task } j \text{ is running in} \\ & \text{the machine (production line) } i \\ 0, & \text{otherwise} \end{cases} \quad (2)$$

$$C_{max} = makespan \quad (3)$$

[IB: o que é isso? Esse modelo para C_{max} não está claro][IB: ja tem uma explicacao do que eh makespan anteriormente.]

Objective The objective function is the variable C_{max} (total process execution time) that needs to be minimized.

$$\min C_{max} \quad (4)$$

where,

C_{max} is the variable to be minimized (optimized).

Constraints

- Each task must be executed/processed in an unique machine:

$$\sum_{i=1}^M \sum_{j=1}^N x_{ij} = 1 \quad (5)$$

- Execution time of each machine:

$$\sum_{i=1}^M \sum_{j=1}^N (p_j + s_j)x_{ij} \leq C_{max} \quad (6)$$

[IB: Isso tá errado. A função objetivo não pode ser limitada inferiormente conforme a eq anterior.][IB: eh uma restricao apenas iury. nao ta errado. a funcao objetivo eh o C_{max} (ja eh outra coisa). Organizei exatamente como o pessoal de otimizacao usa nas literaturas.]

The Eq. (6) represents the sum of the the duration time p_j of each travel from one place another in the case study explained in Section 5.1, considering the production time s_j in each production line. The x_{ij} shows which product is being produced in each production line. The left hand side of the Eq. (6) needs to be smaller than the right hand side of the Eq. (6).

³Makespan is the termination time of the most loaded processor.

Resulting optimization problem The resulting optimization problem consists in minimizing C_{max} with relation to the decision variable (2) constrained to the condition in Eq. (5) and (6). Thus, the optimization problem is represented as follows:

$$\begin{aligned} \min \quad & C_{max}, \\ \text{s.t.} \quad & \sum_{i=1}^M \sum_{j=1}^N x_{ij} = 1, \\ & \sum_{i=1}^M \sum_{j=1}^N (p_j + s_j)x_{ij} \leq C_{max} \end{aligned} \quad (7)$$

5.3 Planner Evaluation Methodology

The main contribution of this work is a methodology to evaluate UAV mission planner algorithms and find minimum cost planner. To this purpose, a generalized evaluation metric is developed. The objective (cost) function modeled in Section 5.2 is related to the total time spent for the execution of the mission. Our evaluation metric compares the cost of a planner algorithm with the best cost computed by the CPLEX solver (CPLEX, 2003). The metric presented here will be called Mission Planner Cost Index (*MPCI*).

Firstly, the optimal cost of the problem is obtained through the CPLEX solver, which returns the optimum value (minimum mission execution time). The model proposed in the Section 5.2 was implemented using the CPLEX solver library available for C++.

The cost of each planner strategy (algorithm 1 and 2) c_X is obtained by counting the number of GoTo commands which represents a process (task).

Finally, the evaluation of each mission planner is computed with relation to the optimal cost, therefore, the $MPCI_X$ of a planner X is computed as follows:

$$MPCI_X = \frac{c_o}{c_X}, \quad (8)$$

where c_o is the optimal cost obtained by the CPLEX solver, c_X is the cost of the solution generated by planner X , and $0 \leq MPCI_X \leq 1$. Note that as close of 1 the $MPCI_X$ is, the solution cost become smaller.

6 Experimental Evaluation

This section shows the experimental results obtained in this project, as well as the cost evaluation of two techniques used compared to the optimum cost implemented with the CPLEX solver.

6.1 Experimental Environmental and Objectives

In order to clarify the evaluation methodology efficiency, two mission planner algorithms are evaluated using the proposed metric, and the evaluation result is compared with practical execution time

of algorithms that are implemented in the simulation tool SITL⁴ and in a real UAV system (3DR IRIS+). The mission planning algorithms were executed in a computer running Linux Mint, core i7 processor and 8GB of RAM. In order to control the UAV, we run the control program, which uses the dronekit api as an interface between a high level program language (Python) and the protocol that the UAV understand (MAVLink), in the same computer where there is a radio module connected via usb communicating with the UAV radio module.

6.2 Mission Planners

In this work, we considered that mission planner is a software that generates a production mission given the warehouse and customer order. This program generates a `.mission` extension file containing a set of mission commands, as shown in the Section 2. Two examples of planners are presented in this work and will be employed to demonstrate the cost evaluation methodology.

6.2.1 Planner A

In the Algorithm 1, we show a strategy to solve the mission planning problem and we denoted as Planner A. In this Algorithm, it's produced all the X product in sequence, bringing the inputs in sequence as well (A,B and C) and then it's produced all the Y products in this order.

The general steps of planner A are shown in the Algorithm 1.

6.2.2 Planner B

In planner B, unlike planner A where the UAV is idle waiting for each product to be produced and only then head to the client, the UAV continues the production process while the output of the products does not end. After the production of each product is finished, the UAV stops the task that was running and goes to the production line of that particular product, performs the collection and takes it to the client. After that, it returns to run the task that was previously running, performing a scheduling of tasks.

The general steps of planner B are shown in the Algorithm 2.

6.3 Cost Evaluation

The results of each mission planner is compared to the optimal solution obtained with the branch-and-cut algorithm of the IBM/ILOG CPLEX 12.4 tool developed in C++. In order to obtain better results to do the comparison, it was considered

⁴Simulator that allows executing a Plane, Copter or Rover without the need of a hardware)

Algorithm 1 Planner A

Input: warehouse
Input: order
Output: mission file *.mission*
begin
 check the order;
 repeat
 go to the warehouse;
 repeat
 get input A;
 bring to the production line X;
 until *until bring 2 A elements*;
 go to the warehouse;
 get the input C;
 bring to the production line X;
 wait X to be produced;
 bring X to the client;
 until *production of all X elements finish*;
 repeat
 go to the warehouse;
 repeat
 get input B;
 bring to the production line Y;
 until *until bring 2 B elements*;
 go to the warehouse;
 get the input C;
 bring to the production line Y;
 wait Y to be produced;
 bring Y to the client;
 until *production of all Y elements finish*;
end

only the time in which the UAV takes to finish the production of a product.

	Time (s)
Planner A	420
Planner B	404
CPLEX	134

Table 1: Time execution for comparison with the optimal solution time

The Table 1 shows the mission execution times obtained using the planner algorithm A, the planner algorithm B, and the minimum value given by the solver. Using the metric shown in the section ??, then:

$$Creal_A = \frac{134}{420} = 0.319 \quad (9)$$

$$Creal_B = \frac{134}{408} = 0.328 \quad (10)$$

It can be seen in the Equation 10 that the proposed evaluation methodology indicates that planner B performs the mission more quickly and has a lower cost than planner A.

6.4 Practical Results

To verify the practical results, as well as a cost comparison between the different approaches of mission planners developed in this work, the flight

Algorithm 2 Planner B

Input: warehouse
Input: order
Output: mission file *.mission*
begin
 initialize t_x ;
 initialize t_y ;
 check the order;
 repeat
 if *the counter of that X is not t_x* **then**
 go to the warehouse;
 repeat
 get the input A;
 bring to the production line X;
 until *until bring 2 A elements*;
 go to the warehouse;
 get the input C;
 bring to the production line X;
 start the counter of this X
 (production time);
 keep producing;
 else
 go back to the production line X;
 bring X to the client;
 go back to producing;
 end
 until *production of all X elements finish*;
 repeat
 if *the counter of that X is not t_y* **then**
 go to the warehouse;
 repeat
 get the input B;
 bring to the production line Y;
 until *until bring 2 B elements*;
 go to the warehouse;
 get the input C;
 bring to the production line Y;
 start the counter of this Y
 (production time);
 keep producing;
 else
 go back to the production line Y;
 bring X to the client;
 go back to producing;
 end
 until *production of all X elements finish*;
end

time measurement was performed using the two mission planning algorithms developed, using the case study shown in 5.1.

Below, we can verify the two mission files generated by the two strategies developed in this work:

The following is data about the mission being executed in the simulator SITL. The table below shows the performances of both schedulers relative to flight time.

	Planner A	Planner B
Tests	<i>Flight Time (s)</i>	<i>Flight Time (s)</i>
1	460,405	430,830
2	460,693	436,885
3	462,080	441,681
4	457,719	441,277
5	461,227	451,865

Table 2: Mission Planners Flight Time - Simulator.

In the Table 2, it can be verified that the mission time executed by planner B presented less time compared to planner A in all five tests.

In Figure 4 is shown a graph of the flight times of both planners in the five tests done in the simulation environment.

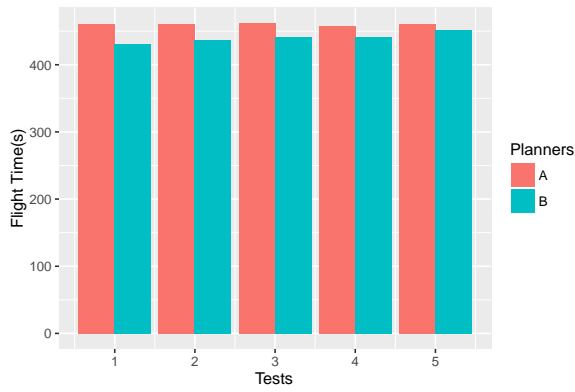


Figure 4: Flight Time Graph of Mission Planners Relative to 5 Tests in Simulator.

Next, you can see the results for the tests performed in real environment used by both planners. In Figure 5 the map where the tests were performed is shown on the map at the Faculty of Physical Education and Physiotherapy of Federal University of Amazonas.



Figure 5: Warehouse, Production Line X, Production Line Y and Customer in the Map.

In Table 3, it can be seen that the mission time executed by planner B in real environment also

shown a shorter time compared to planner A in the five tests:

	Planner A	Planner B
Tests	<i>Flight Time (s)</i>	<i>Flight Time (s)</i>
1	455,12	441,72
2	456,93	440,18
3	457,19	447,51
4	460,25	438,19
5	459,47	445,85

Table 3: Mission Planners Flight Time - Real.

In Figure 6 is shown a graph of the flight times of both planners in the five tests done in real environment.

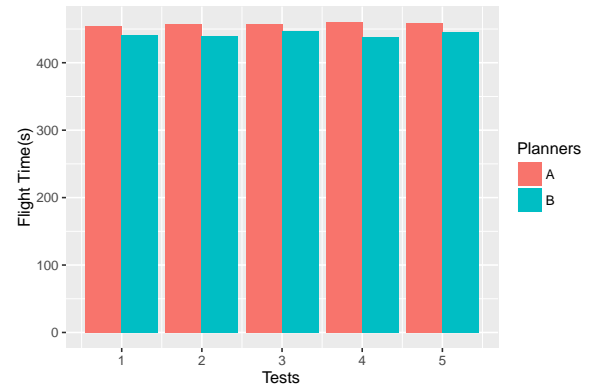


Figure 6: Flight Time Graph of Mission Planners Relative to 5 Tests in a Real Environment.

As presented in the section, planner B was faster than planner A, relative to execution time. This can also be verified in real environment, as shown in tables 2 and 3 of mission execution time in simulated and real environment.

7 Conclusions

We have developed an evaluation methodology for UAV mission planner in an industrial production scenario as illustrated in the Section 5. We used this methodology of evaluation to performs in two different algorithms in order to verify the performance of them.

In addition, we have used a commercial UAV to perform our methodology. We have created a framework with a set of mission commands to control the UAV using dronekit api as presented in the Section 4 and 5.

Further works includes the use of computational vision for the recognition of inputs, and to improve the modelling of the optimization problem for better results in cost evaluation.

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