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# OptiRun

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# altibox

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# Abstract

TODO

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# 1 Introduction

Software systems, ranging from business applications to consumer products, have grown to form a fundamental element in our society. People are becoming increasingly dependent on computers, and wish to be in control of the digital content they consume. As a consequence of digital content being increasingly accessible, the importance of, and demand for, high-quality software has become substantial. Enhanced pressure on software vendors to deliver frequent releases of high-quality software requires efficiency in every stage of the software development process.

Software testing constitutes a central aspect to the software development process. It plays a critical role in quality assurance and defect detection, and is an important means to ensure that the software in question behaves as expected and according to specifications.

# 1.1 Origin

The subsidiary company of the Lyse Group, Altibox, is a telecommunication service provider. They offer a selection of services including broadband, *Internet Protocol Television* (IPTV) and *Voice over Internet Protocol*, (VoIP). Since its origin, the company has grown a great deal, and at a considerably higher pace than originally anticipated. Because the company had trouble keeping up with its rapid development, they neglected to apply a number of

established business practices in several different areas, including the field of software testing.

IPTV has long been among Altibox' most prominent services, and has previously been available primarily through the use of TV decoders. In the later years, however, an important focus area for Altibox has been to expand the availability of their TV and  $Video\ On\ Demand\ (VOD)$  content by applying  $Over\text{-}The\text{-}Top\ (OTT)$  technology, which means that the content is made accessible over the Internet. This is accomplished through the development of  $TV\ Overalt$ , which is available both as a web application for desktops and as mobile applications for portable devices.

# 1.2 Motivation

Altibox currently perform all software testing manually. They have wished to incorporate test automation in the test process of the web-based version on TV Overalt for some time, but have failed to make it a priority. Automated tests for web-based systems on the user's level generally run slowly, and Altibox therefore desired a system that would let them execute such tests efficiently. They also wanted the tests to be executed in a controlled environment.

This thesis presents OptiRun: a tool that will help Altibox incorporate test automation according to their needs. The tool is not limited to TV Overalt, however. It can be applied as a testing platform for a broad range of web applications.

# 1.3 Purpose

The aim of this project was to design and create a tool that would substantiate test automation with a distinguished intent of optimizing the time of test runs as well as provide control and feedback of test results. The tool is intended to be utilized during system or acceptance testing by technical test analysts.

The main area of use is execution of Selenium test scripts for web applications. Tests should be able to be executed remotely in a controlled distributed environment. An algorithm for optimally allocating test cases on the available test machines in the distributed environment should be provided. Further, an interface for uploading, managing, executing and scheduling test scripts should be included, and logs from results of previous

test runs should be available here. Upon a failed test, an option to report this in the issue tracking system should be provided.

# 1.4 Outline

The remainder of this thesis is organized as follows:

**Chapter 2** provides a theoretical basis for the thesis by discussing some background information relevant to this thesis.

**Chapter 3** introduces some essential tools and technology used throughout this project.

**Chapter 4** presents an overview of OptiRun, and presents the dashboard from the user's perspective.

**Chapter 5** describes the implementation of the system. Some features are explained in detail.

Chapter 6 presents and discusses some experimental results as well as an evaluation of the system.

Chapter 7 provides some suggestions for further work and concludes this thesis.

# 2 Background

Before moving onto the technical substance, it will be beneficial to acquire a theoretical basis on the subject of this thesis. This chapter introduces software testing and explains some essential concepts of relevance. Further, it presents constrained optimization as well as some previous work related to this thesis.

# 2.1 Software Testing

A fundamental understanding of software testing is a useful prerequisite in order to fully understand the context of this thesis. Consequently, this section provides a short introduction to software testing. Since this is a large field, only a small selection of relevant concepts and ideas will be presented.

Software testing is the process of evaluating the quality of the application, system or component being tested, commonly referred to as the test object or the system under test. Software testing may involve any action oriented toward assessing the software with the goal of determining whether it meets the required results. [?]

Software is developed by human beings who can make errors. These errors may cause defects in the source code. Executing defected code can lead to failure in the program [?]. One of the purposes of software testing is to examine the test object with the intent of revealing such defects. Other

objectives include to measure and ensure quality, and to provide confidence in the product [8, pp. 8]

# 2.1.1 V-Model

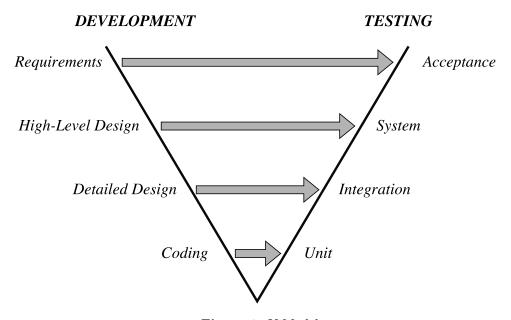


Figure 1: V-Model

The *V-model* is an important asset in software testing, behind which the central idea is to illustrate how each development task in a software development process have a corresponding testing task of equal importance. This is symbolized by the two branches of the letter "V" in the model. The development process is represented by the left branch, which shows the system being gradually designed. The integration and testing process is represented by the right branch, which shows how elements are being assembled to form progressively larger subsystems, and then tested [8].

Depending on the literary source, the V-model covers a varying number of levels. The V-model shown in Figure 1 is created using the same four levels as found in [1]. The highest development level covers the gathering, specification and approval of requirements. Acceptance testing correspondingly checks if these requirements are met. Further, high-level design covers the functional design of the system, and corresponds to system testing, which aims to verify if the system as a whole meets the required results.

Detailed design covers technical system design and component specification, and integration testing correspondingly verifies that the different components work together as specified. The lowest level is the coding in which the specified components (modules, units and classes) are implemented. It corresponds to unit and component testing. The tests at this level aim to verify that system components perform as specified, by testing them in isolation. [8]

OptiRun is a tool for high-level test execution, and can be applied to the two highest test levels of the V-model; system testing and acceptance testing.

# 2.1.2 Black-Box & White-Box Testing

Software testing can be divided into black-box and white-box testing. White-box testing is based on analysis of the internal system structure, and requires knowledge of the source code. In black-box testing, the test object is seen as black box, whose behavior is watched from the outside. The inner structure of the system is either unknown or unconsidered. Test cases are determined from the specifications of the system. Black-box testing is predominantly used for higher levels of testing. The test object is accessed through the user interface (UI). Tests are performed from the perspective of an end user, and aim to mimic a human being interacting with the system. Black-box testing includes functional testing, which is used to validate a particular feature for correctness according to the requirements specifications. [8]

OptiRun performs tests by using Selenium to automate web browsers. It executes functional UI tests on full system builds, and run without access or consideration to the source code and the internal system structure. The test method covered by OptiRun is therefore black-box testing.

### 2.1.3 Test Automation

As opposed to humans performing software tests manually, test automation is the practice of writing scripts that conduct tests when executed. Automation can be applied to tests at any level. Low-level tests, such as unit and component tests, generally run fast and are durable since they are isolated from changes in other parts of the system. These tests may be executed frequently. As we move up to the integration level, the tests does not run as fast, and become less durable since they depend on multiple components working together in subsystems. This thesis considers the tests

at the two highest levels; system and acceptance tests. These tests are is concerned with testing the system at the perspective of the end user. These tests require the system to work as a whole, and thus generally run slower and are more brittle. [12] Because of this, functional UI tests are often the fewest.

There are a number of strengths to automated tests. They are superior at verifying logical functionality. They can be executed any number of times, and they run more quickly than a human interacting with the system, thus saving time and reducing effort. Time saved on manual testing can be used to increase the test coverage, which can provide reduced risk. Tests and tasks that would be error prone if done manually and tests that are repeatedly performed are typical subjects for automation [6]. This includes regression testing, which is used to verify that defects have not been introduced in a new version of previously tested software [8].

Manual human testing has a number of benefits too. Human testers can identify corner cases and check how the system responds when it is used in manners it is not designed to be used for. They can also evaluate aesthetics and design of the UI as well as the overall user experience. In most cases, automated tests should not be a complete replacement for manual human testing. They should rather complement each other.

# 2.1.4 Continuous Integration

Automated tests can be applied to a continuous integration (CI) environment, which automatically builds and tests code contained in specified repositories. It can provide frequent and rapid feedback on the code in uploaded commits, and can thus be of excellent value for software developers [9].

Jenkins is a commonly used CI tool, which offers a plugin for running Selenium tests. This could be used to apply CI to OptiRun, but has been omitted for two main reasons.

Firstly, Selenium tests run on real browsers. They perform real browser operations, often having to wait for HTTP responses. The browsers also need to open prior to a test execution and close afterward. All of these tasks are a great deal more time-consuming than tests on lower levels. Including Selenium tests in a CI process as part of the build would render each build hugely time-consuming.

Secondly, in Altibox' case, software development and software testing are performed separately by different teams. The test team, to which OptiRun

is aimed, does not have access to the actual code, and would not gain the same benefits from .

# 2.2 Constraint Programming & Optimization

As explained in the previous section, the type of tests targeted in this project run slowly. This means that a large test suite could take unnecessarily long time to execute if the tests were not carefully allocated among the available test machines. Minimizing the execution time of a test run has been an important ambition in this project.

There are certain requirements that must be fulfilled upon allocating tests. A test that is specified to run in a certain browser can only be allocated to a machine with the given browser installed; this is a constraint. Constraint programming (CP) revolves around modelling real world constraints by mathematical formalizations, and use them to find feasible solutions to the problem [5].

Some problems have very large sets of possible solutions, while others only have a few or none at all. Sometimes it is not enough to simply find a possible solution to a CP problem, as the different solutions can vary greatly concerning the aim of the problem. In such cases, a specification of which solutions are more valuable should be provided to help separate the good solutions from those of poor quality. The objective of the CP problem in this project is to minimize the overall execution time of a test run. An *optimization problem* is a CP problem in which such an objective is specified [5].

The optimization problem in this project does not exclusively revolve around finding an optimized allocation - in situations where there are a huge number of solutions to the problem, identifying the best solution can take a vast amount of time. For instance, in a situation with two competing solutions where one provided a 10 second longer overall execution time, but took 20 seconds less to find, this would be the preferred solution. The time it takes to arrive at the solution is therefore also taken into account. Some stop criteria for searching are defined; these will be presented in Chapter 5.

Formally, the optimization problem in this project considers a set of tests  $\mathcal{T} = \{t_1, \ldots, t_n\}$  with a corresponding set of durations  $\mathcal{D} = \{d_1, \ldots, d_n\}$  in which the values are derived from the durations of earlier executions of each test. Further, it considers the current set of available test machines  $\mathcal{M} = \{m_1, \ldots, m_m\}$ . A pre-processing function p identifies the subset of

machines each test can be executed on, and a function f defines the actual allocations in with the overall execution time  $T_e$ . The problem seeks to minimize the total time  $T_t$ , which consists of  $T_e$  plus the searching time it takes to find the solution  $T_s$ . Thus, the expression we wish to minimize is  $T_t = T_e + T_s$ . Additionally, all tests in the test suite must, if possible, be executed exactly once on a machine that fits any specified requirements for browser and operating system.

Furthermore, the following list of assumptions are expected to be strictly imposed:

Non-preemptive scheduling - Once a test execution has started, it will run uninterrupted until it completes.

Non-cumulative scheduling - Each test machine can only execute a single test at a time. This could have been implemented differently, but as each test requires a great deal of resources, this would prolong the execution of each test, as well as make the durations more unpredictable.

Machine-independent execution time - Although the execution time will not necessarily be machine-independent in reality, depending on machine performance, this is considered an assumption.

# 2.3 Related Work

Some work has previously been done on the subject of test case scheduling and automation of functional UI tests for web applications. This section presents the scheduling method TC-Sched and the automated testing platform  $Sauce\ Labs$ .

## 2.3.1 TC-Sched

TC-Sched is a time-aware method for minimizing the execution time of test cases within a distributed system with resource constraints. Mossige, Gotlieb, Meling, and Carlsson proposed the method in [16]. The authors define the *optimal test case scheduling* (OTS) as the optimization problem of finding an execution ordering and assignment of all test cases to machines, in such a way that each test case is executed once, no global resource is used by two test cases at the same time, and the overall test execution time is minimized.

TC-Sched addresses situations in which some operations may require exclusive access to one or more *global resources*, such that only one test case can access each resource at a given time. To include the aspect of global resources in this project would have been redundant since there are no such global resources involved. Apart from that, TC-Sched is a closely related scheduling method that has been of inspiration. The implementation of the scheduling algorithm used in this project will be described in Chapter 5.

## 2.3.2 Sauce Labs

Cloud testing services use cloud computing environments as a means of simulating real-life user traffic. Sauce Labs is a cloud-hosted platform for automated testing of web and mobile applications. One of the founders, Jason Huggins, were also the original creator of Selenium [14], which will be introduced in Section 3.2.

Sauce Labs bear some similarities and some dissimilarities with the design of OptiRun. Both platforms are frameworks for automated functional UI testing of web applications through executing Selenium test scripts. They both execute tests remotely in distributed environments using Selenium Grid. However, whereas Sauce Labs perform test execution on virtual machines (VMs) that are created prior to each test run and destroyed immediately after [13], OptiRun requires the use of either real, physical machines, or customized VMs, that remain intact after use. Because of the number of machines in OptiRun being limited to the machines connected to the distributed system, an algorithm for optimizing the overall execution time of test runs is also needed here. While Sauce Labs executes tests in their own data centers in a remote location and their own environments, OptiRun performs tests on the test object locally and in an environment intended for testing of the given test object, in this case TV Overalt. Sauce Labs is a paid service for which customers pay to gain access to using a limited number of concurrent virtual machines for a limited amount of time each month. The optimization algorithm in this project also distinguishes the two projects even further.

# 3 Technology

A number of distinctive tools and technologies have been applied throughout the work on this project. This chapter briefly introduces the most essential of these tools and technologies.

# 3.1 The Python Programming Language

The Python programming language has largely been used in the conduction of this project. Python has efficient high-level data structures, and is known for being simple, yet powerful. Since Python programs are executed by an interpreter, it is considered an interpreted language [15].

Building this project on Python was a natural choice for several reasons. Python is compatible with Selenium, which is introduced in Section 3.2. Other reasons include offering a large set of libraries and having high-quality documentation. Additionally, Python supports multithreading and is suitable for running a program as a service. Details of the implementation will be discussed in Chapter 5. Version 2.7.11 of Python has been used in this project.

Will change to Python 3.x if time.

# 3.2 Selenium

Selenium is an umbrella project for automation of web browsers. Several different tools and libraries are included in the framework, all with the

common goal of supporting browser automation. Selenium can be used to automate different types of browser jobs such as web-based administration tasks, but is primarily used for test automation. The project is released under the Apache 2.0 license and is thus free and open-sourced. Selenium has language bindings for several different programming languages, including Python. The Selenium tools used in this project will be introduced below.

### 3.2.1 Selenium WebDriver

Selenium WebDriver consists of a set of libraries that helps with automation of tests for web applications, and a set of executable WebDriver files, one for each browser, that perform the actual automation specified in the WebDriver scripts. Selenium WebDriver interacts directly with the browser by sending calls using the native support for automation for each browser [7]. Selenium WebDriver runs on Windows, Linux and Macintosh, and supports most conventional web browsers, including Firefox, Chrome, Internet Explorer, Opera and Safari.

The WebDriver is used to create a new instance of the requested web browser. It is used to fetch a given web page, and then locates UI elements. When an element is located, the WebDriver can be used to perform an action on the element, such as clicking a button, checking a checkbox, or populating a text field.

```
from selenium import webdriver

# Create a new instance of the Chrome WebDriver
driverdriver = webdriver.Chrome("<file path>/chromedriver.exe")

# Go to the Altibox TV Overalt start page
driver.get("https://tvoveraltstg.altibox.no/")

# Locate and click the Login button
login_button = driver.find_element_by_class_name('btn-login')
login_button.click()
```

Listing 1: Selenium WebDriver Example

Listing 1 shows a simple example of how Selenium WebDriver can be used with Python to open the Chrome web browser, navigate to the Altibox TV Overalt start page, locate the *Login* button, and click it. OptiRun revolves around executing scripts that uses the Selenium 2.0 libraries and WebDrivers to execute automated web browser tests.

### 3.2.2 Selenium Grid

Selenium Grid is a tool for executing Selenium tests on remote machines in a distributed environment, and thus allowing for parallel execution. This also opens up for running tests on different operating systems.

Reasons for wanting to incorporate Selenium Grid include being able to run tests against multiple browsers, browser versions and browsers running on different operating systems, and to reduce the execution time of the tests. In practice, a grid is made up of one Selenium Grid Server; a hub, and one or more slave machines; nodes, all running a Selenium Standalone Server. The nodes use Selenium WebDriver to communicate with the hub through a JSON wire protocol [3].

Selenium Grid 2.0 and Selenium Standalone Server 2.51.0 were used in this project.

# 3.3 Django

Django is a high-level web development framework that is implemented in the Python programming language. It encourages rapid development and enables efficiently maintainable web applications of high quality. The framework is a free and open-sourced project maintained by the non-profit organization the Django Software Foundation, who describe the framework as fast, secure and scalable [4]. Comparable to Selenium, Django is also essentially a collection of Python libraries [11]. The Django libraries can be imported and used to implement web applications. Some additional HTML, CSS and JavaScript code has been applied along with the Python code.

Aside from allowing rapid progression of development, one of the main reasons for choosing Django rather than building the dashboard from scratch or using a different web framework, is its powerful administrator site. An administrator site was exactly what was needed to build the dashboard of OptiRun. Another contributing factor was to provide consistency and the ability to communicate seamlessly with the remaining parts of the system, since Django builds on the same programming language as the rest of the system. The dashboard will be presented in Chapter 4. Details concerning the implementation will be explained in Chapter 5. Version 1.9 of Django was used in this project.

# 3.4 OR-Tools

Google's Operations Research Tools (OR-Tools) [10] is an open source library for combinatorial and constraint optimization. The tool set is written in C++, but is available with bindings for other programming languages such as Java, C# and Python. The OR-Tools library strictly conforms to the Google coding styles, and is of such high quality that it has been accepted for usage internally at Google.

Listing 2 shows how a simple optimization problem is solved using this library. In this problem, a list of integers will be assigned values ranging from 0 to 2. A constraint specifying that no two identical numbers should be placed beside each other is added as a constraint. Maximizing the sum of the integers is specified as the objective. The solver searches for better and better solutions until finally arriving at an optimal solution.

```
from ortools.constraint_solver import pywrapcp
2
3
    solver = pywrapcp.Solver('')
4
   variables = [solver.IntVar(0, 2) for _ in range(3)]
5
7
   for i in range(len(variables) - 1):
       solver.Add(variables[i] != variables[i + 1])
8
9
   db = solver.Phase(variables, solver.CHOOSE_FIRST_UNBOUND,
10
        solver.ASSIGN_MAX_VALUE)
   objective = solver.Maximize(solver.Sum(variables), 1)
11
    solver.NewSearch(db, objective)
12
13
   while solver.NextSolution():
14
       result = [int(item.Value()) for item in variables]
15
       print result, "Sum =", sum(result)
16
17
18
   >>> [0, 2, 0] Sum = 2
   >>> [0, 2, 1] Sum = 3
19
20
   >>> [1, 2, 1] Sum = 4
   >>> [2, 1, 2] Sum = 5
```

**Listing 2:** OR-Tool Implementation

As a means to evaluate the optimization mechanism of test allocations designed for this project, an alternative version using OR-Tools has also been implemented. This alternative implementation, as well as a discussion, evaluation and thorough comparison of the two versions will be presented later in the report.

# 4

# System Overview

This chapter aims to provide an overview of how OptiRun is assembled by describing the design and architecture of the system. System setup will be described thereafter, followed by a presentation of the web-based user interface of the tool.

# 4.1 Architecture

OptiRun is operated through a web-based dashboard. This web service works as a content management system for test scripts. It allows for sending execution requests for the available test scripts, or for scheduling test runs ahead of time.

Figure 2 shows the structural architecture of the system. In this case, the web server, the controller, the database, the file storage and the Selenium Grid Hub are all located on the same machine, hence the dashed border around these elements in the figure.

The web server uploads test scripts to a file location on the server. Meta data about the test cases as well as other information such as test groups, planned test executions, test results and user authentication are also stored in the database. The web server can report defects in the issue tracker system, which in the case of TV Overalt is JIRA.

The controller listens to test execution requests triggered through the dashboard. Upon an execution, it fetches the given test script from the file

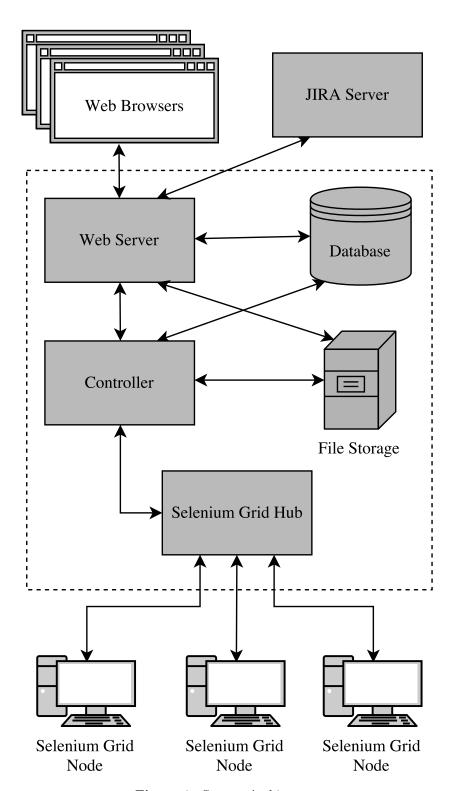


Figure 2: System Architecture

storage. Test scripts are sent to the current instance of the Selenium Grid Hub with specifications of which node in the distributed Selenium Grid that the test script should be executed on. The hub then triggers the given node to execute the test. After the test has finished, the controller stores the result in the database.

# 4.2 Setup

TODO

# 4.3 Dashboard

The graphical user interface of OptiRun is web-based, and is built on the high-level Web framework Django (3.3). The framework provides a powerful automatic administrator interface, which has been used as a foundation for the development of the dashboard. Working with Django thus allowed for rapid development while retaining control of the content and functionality of the website. This was one of the main reasons why this particular framework was chosen for OptiRun.

This subsection will present the dashboard from the user's perspective. An in-depth presentation of how some of the functionality has been implemented can be found in Chapter 5.

# 4.3.1 Home Screen

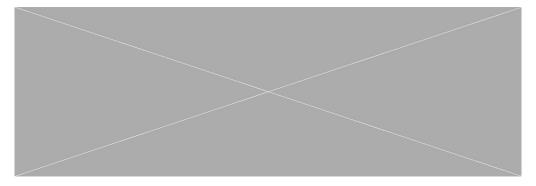


Figure 3: Home Screen

After logging in, the home screen is the first thing that meets the eye of the user. This screen provides navigation to all of the modules and pages described in the remaining subsections of this section. In order to give the user an indication of status, some graphs that show the results of recent test runs, as well as key numbers and other important information is displayed on this page.

# 4.3.2 Authentication & Authorization

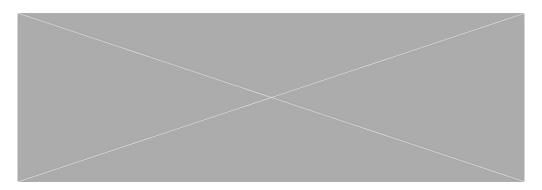


Figure 4: Authentication and Authorization Module

The Django administrator interface provides some built-in functionality, including a module for authentication and authorization of users. *Superusers* can list, edit and create new user accounts. They can also change user info and the permissions of other user accounts, add them to user groups, and specify the permissions of user accounts.

OptiRun is a tool that is meant to be part of a business process in which only trusted peers should be allowed to take part, and by extension, OptiRun should only be operated by trusted associates. For this reason, no user accounts can be created by anyone other than a superuser. Thus, any person who wants and account must request this by someone with superuser privileges.

By default, all passwords are encrypted before being stored in the database.

## 4.3.3 Settings/Configuration

TODO

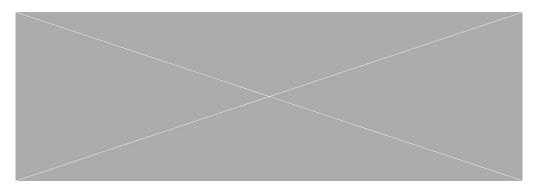


Figure 5: Test Case Module

### 4.3.4 Test Case Module

All registered test cases are listed in this module. The list contains meta data about the test cases as well as some values that are retrieved or calculated using information from other tables in the database. This includes the number of times the test case has been executed, the average execution time, the result of the previous execution and when the test case was last executed. It is possible to search and filter the list to narrow down the elements listed, and to sort the list by clicking on the attribute names in the table header.

New test cases can be created by clicking the button labeled *Add Test Case* in the upper right of the view. This opens a new page with a form in which the test script can be uploaded and corresponding details about the test case can be filled in. Which groups the test case should be a member of can also be specified here. Test cases can be members of multiple groups, or none. After a test case has been created, it can be edited by clicking on the title of the test case in the test case list.

It is possible to select test cases and perform actions on them. The action options - Delete selected test cases and Execute selected test cases - are located in a dropdown menu above the test case list. The former is a built-in Django function which, after the user confirms deletion in and intermediate page, removes all meta data regarding the selected items from the database. The latter action is implemented for OptiRun. The action takes the user to an intermediate page in which they can specify the platform and browsers the test cases should be executed on.

When an action has been performed, or attempted performed, a feedback message telling if the action was successful or not is displayed on the screen.

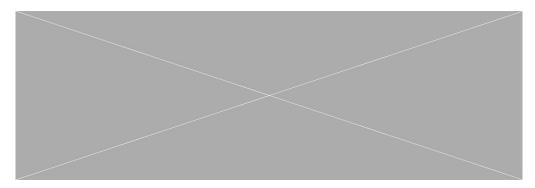


Figure 6: Intermediate Page for Immediate Test Execution Requests

# 4.3.5 Test Group Module

Test groups are listed similar to individual test cases as described in the previous subsection. As with the test cases, group items can also be searched, filtered and sorted. Additionally, new groups can be created, and existing groups can be edited.

However, the group model is simpler than the test case model. The purpose of this model is to group together test cases that should preferably be executed in the same test run.

# 4.3.6 Scheduling Module

The scheduling module provides an interface for planning test runs in the future. As with the test cases and groups, schedules are also displayed in a list can be searched, filtered and sorted. Schedule objects can be created and changed. Additionally, the schedule list enables schedule objects to be activated and deactivated from the action menu. Only activated schedule objects will start test runs according to their specified time of execution. This is the default status, and deactivation of schedule objects must be appointed explicitly through the action menu.

Upon creating or editing a schedule object, an execution time can be specified along with any desired recurrence pattern along with the option of providing an end time. Test groups or individual test cases that should be included in the schedule object are also specified here.

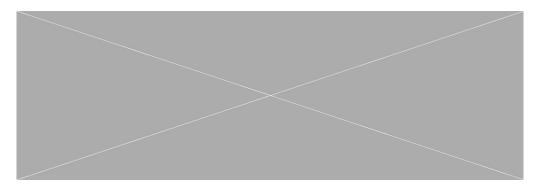


Figure 7: Schedule Creation Form

# 4.3.7 Execution Log

The layout of the log is the same as the remaining list views, but the functionality differs in that log items are read-only. They can not be created or deleted. Log items are created and inserted into the database based on the result from test executions. Each log object corresponds to an execution of an individual test script. The information listed in the logs include a number of attributes such as the duration of the test execution, the IP address or hostname of the machine on which it was executed, as well as information such as browser, platform, output and of course the result.

# 4.3.8 Other

The Help page contains miscellaneous information regarding how OptiRun works and how to use it. This page can be of help to technical testers who wants to use the tool, and otherwise to those who wants insight of how it works.

The About page provides a short presentation what OptiRun is and how it came to be. Any necessary licenses are also included in this page.

# 5

# Design & Implementation

This chapter aims to explain how OptiRun is designed and implemented. It provides a description of how the system utilizes Selenium Grid. The implementation of the controller and some important details about this part of the system will then be presented. Further, the test allocation mechanism called *Opti-X* will be explained with an accompanying example to illustrate how the mechanism works step-by-step will be presented. The chapter then explains how the database is structured in addition to the how it is accessed using Django libraries, before finally presenting some important implementation details about the dashboard.

# 5.1 Selenium Grid Integration

In OptiRun, Selenium Grid works as the backbone of all interaction between the server and the remaining machines in the distributed system. It is used to establish connections and to perform test executions.

Setting up a Selenium Grid environment requires all involved machines to have a *Selenium Standalone Server*, which is a file of the JAR (Java Archive) format, of the same version locally stored. The Selenium Standalone Server is started by executing a shell command.

A Selenium Grid Hub (server) can be started simply by executing the command in Listing 3, although it is possible to assign additional configuration, such as port and IP address, using flags. The default port number is

4444 for hubs and 5555 for nodes, so if nothing else is specified, these ports will be used. When the hub has started, the configuration can be viewed by opening http://<hub-host>:4444/grid/console in a browser.

```
1 java -jar "C:/selenium-server-standalone-2.51.0.jar"
```

Listing 3: Sample Shell Command for Starting Selenium Grid Hub

Starting Selenium Grid Nodes (test machines) require much longer and heavier commands, and thus more work. Also, the command must be customized, as it represents the configuration of the node and the distributed system. Therefore, a script devoted to gathering all necessary information and executing the command to start Selenium Standalone Server is included in this project. This script identifies which browsers are installed and what versions, as well as creating a unique identifier for each machines.

Selenium Grid currently does not offer a documented method of specifying which machine a test should be executed on. Instead, it maps the test to a node whose configuration that matches the desired specifications stated in the test, in regard to operating system, browser and sometimes even browser version. It was therefore necessary to find a way to work around this problem. This is done by utilizing a browser parameter called applicationName, in which additional information can be added. A unique identifier based on the host ID, sequence number, and the current time, is created using Python's uvid library. By adding as a requirement to a test that the applicationName should be equal to the uvid of a specific node, the test can be executed only on the machines with that uvid.

Listing 4: Sample Shell Command for Starting Selenium Grid Node

Listing 4 shows an example of a command that will start a Selenium Grid Node with Chrome and Firefox installed, and with unid 1b234276-fc02-11e5-b752-080027f8a664. Once the node is connected to the hub, the configuration can be retrieved as a JSON string from http://<HubHostname/IP>:4444/grid/api/proxy?id=http://<NodeIP>:5555.

# 5.2 Controller

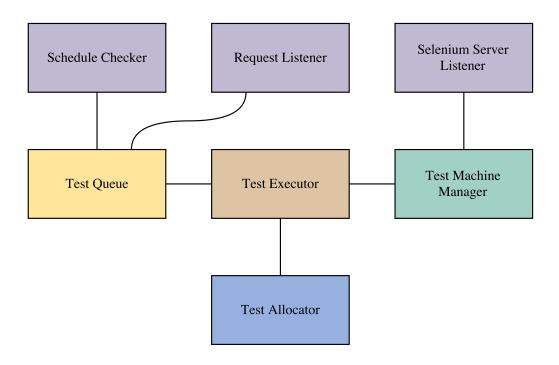


Figure 8: Controller Structure

The *controller* consists of multiple processes running in different threads. Figure 8 shows the inner structure. Each of the modules in the figure will be explained in this Section.

# 5.2.1 Selenium Server Listener

After starting the Selenium Grid Hub as described in Section 5.1 using configuration details found in the configuration file, the main job of the Selenium Server Listener is to listen to the output from the Selenium Standalone Server. This means that when a node connects or disconnects to the system, the Selenium Server Listener is notified. When one of these events happens, it sends notice to its instance of the Test Machine Manager.

# 5.2.2 Test Machine Manager

TODO (Might re-implement this, so will wait with writing this subsection.)

# 5.2.3 Request Listener

All test executions managed by OptiRun are either requested for immediate execution or scheduled ahead of time. There are two different modules in the controller each handling one of these.

The Request Listener constantly listens to a port specified in the configuration file. When a user triggers the Execute Now action from the dashboard, all necessary information about the tests being requested for execution is packed as a JSON string and sent to the controller using a TCP/IP protocol in Python's socket library.

Upon receiving a request, the request listener unpacks the JSON string and creates a list of corresponding *test* objects. The list is then forwarded to the *queue*.

### 5.2.4 Schedule Listener

TODO

# **5.2.5** Queue

As test executions are requested by the controller, each individual test case is placed in an execution queue. The queuing system is made up of four distinct queues with corresponding descending priorities; one for urgent execution, one for immediate execution, one for planned execution, and one for trial execution. Each test case is placed in a queue according to which action triggered their request to be executed. If the request was identified by the request listener, it is placed in the queue for immediate executions. If it was the schedule listener that identified it, the test is placed in the queue for planned executions. If the test case have just been uploaded or updated, it will be placed in the queue for trial execution. The urgent queue is where test cases that were not executed because the test node they were allocated to crashed.

Sometimes a planned execution and an immediate execution of the same test case with the same browser and platform specifications can be requested at the same time. A similar situation could occur when two schedule objects that are due at the same time includes the same test case with the same specifications. Scenarios such as these could potentially lead to time and resources wasted by performing the same job twice. To avoid incidents in which two identical test cases are executed at the same time, a duplicate

check is performed each time a test case is added to one of the queues. If there are duplicates, the test case is removed from the queue with the lowest priority.

The test case executor always checks the immediate queue first, then the planned queue, and lastly the trial queue. If the test executor finds test cases in any of the queues, it empties the queue and handles the containing test cases as a group.

### 5.2.6 Test Execution

After test cases have been retrieved from the queue, allocated amongst the pool of available test machines (Section 5.3), and sorted, it is time for execution.

A thread is started for each of the test machines. In these threads, each test case is started as a subprocess using a shell command in which information regarding the desired test node and browser of the test execution are passed to the test script as arguments.

Immediately before a test case is executed, the test node is pinged to see if the connection is still up. This check is also done immediately after any failed test. If the node has crashed or otherwise failed before or during the execution, the test cases are moved to the queue for urgent executions, and attempted executed once the current test run has finished.

If there none of the test machines match the required browser/platform specification of a given test case, the test will not be executed; it will appear in the execution log, but will be marked as Not Executed.

# 5.3 Allocation Mechanism

Since the test type intended for OptiRun generally runs slowly, designing a mechanism that would efficiently allocate tests to test machines in an attempt to minimize the duration it takes to find and execute the schedule. This mechanism has been named Opti-X, and will be thoroughly described in the following subsection. As explained in Chapter 2, an alternative allocation mechanism has also been implemented using Google's OR-Tools library, and has been used in the evaluation process of Opti-X, which will be presented in the next chapter. The alternative allocation mechanism has been named OR-X, and implementation details will be explained subsequently.

# 5.3.1 Opti-X

The Opti-X allocation mechanism consists of a sequence three steps; sorting, initial allocation and enhancement iterations. It can be seen as an extended greedy algorithm, and will be explained in this section.

In order to give a better understanding of how the algorithm works, a demonstration example will be used throughout the explanation. In this example, we assume that there are three test machines connected to the system, as well as the test suite listed in table 1. Note that the durations of the tests intended for OptiRun generally range from 30 to 90 seconds, and this example deliberately uses shorter durations to better illustrate the mechanism.

Test	Duration	Executable on
$t_1$	6	$m_1, m_2, m_3$
$t_2$	3	$m_1, m_2, m_3$
$t_3$	8	$m_1, m_2, m_3$
$t_4$	5	$m_1, m_2, m_3$
$t_5$	3	$m_1, m_2, m_3$
$t_6$	4	$m_1, m_2, m_3$
$t_7$	7	$m_1$
$t_8$	3	$m_2$
$t_9$	9	$m_3$
$t_{10}$	5	$m_1, m_3$

Table 1: Example Test Suite

# SORTING

The first step is a preparation for the initial allocation. The sorting does not play a crucial role in the allocation mechanism, but is creates an excellent starting point as the tests are sorted in a way that makes them well suited for the initial allocation in the next step. If toward the end of the allocation process there would only be left tests that could run on a certain machine, the overall execution time could be greatly affected, so the sorting avoids this by ordering them fittingly beforehand.

The test set is first divided into subsets depending on the number of machines they can be executed on. Each of the subsets is then sorted by their estimated duration in descending order, so that the longest tests are placed first and the shortest are placed last. The list is then assembled again.

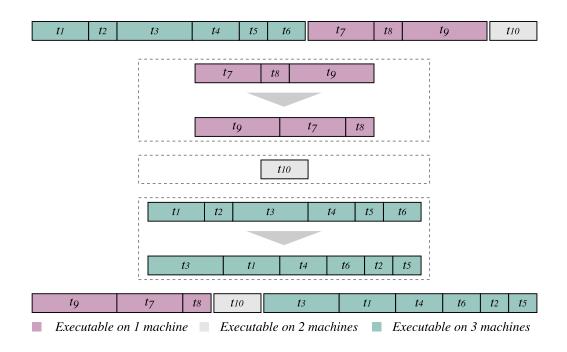


Figure 9: Sorting of Example Test Suite

Figure 9 illustrates how the sorting method works with the example test suite introduced earlier. Since  $t_7$ ,  $t_8$  and  $t_9$  can only be executed on a single machine each, these are placed first in descending order of their duration.  $t_{10}$ , which can be executed on two machines comes after. Finally, tests  $t_1$  through  $t_6$ , which can run on all of the machines, are ordered and added to the test list.

# INITIAL ALLOCATION

The initial allocation is a greedy algorithm, which means that it always makes a locally optimal choice in the hopes that it will give the best result in the end [2]. Greedy algorithms are powerful and work well for an extensive spectrum of problems. Creating a greedy algorithm was a suitable choice for this problem.

This step creates the foundation of the test allocations. After being sorted, the test list is looped through, and the tests are allocated one by one, starting with the *longest* of the tests that are executable on the *fewest* machines. Each test is initially allocated to the machine that currently has the shortest overall duration among the machines that the test is executable on.

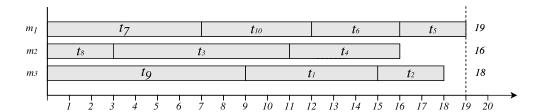


Figure 10: Initial Allocation of Example Test Suite

Figure 10 shows how the test suite is initially allocated among the test machines.  $t_9$ ,  $t_7$  and  $t_8$  are first allocated to  $m_3$ ,  $m_1$  and  $m_2$  one by one, as each of them can only be executed on that machine.  $t_{10}$  can be executed on both  $m_3$  and  $m_1$ , but as the latter currently has the shortest overall execution time, this is the machine it is allocated to. The remaining tests are allocated in the same way. After the initial allocation is finished, the overall execution time is 19. However, the total durations among the machines are slightly uneven, so there might be room for improvement. This will be examined in the last step of Opti-X.

## Enhancement Iterations

Greedy algorithms are simple, yet efficient, and provide adequate results most of the time. However, they do not always yield optimal solutions. In order to improve the result, an additional step has been included in the Opti-X. This is the most complex element in the process.

The basic idea behind the iteration step is to find subsets of tests among two machines and swap the two subsets that will decrease the total duration of the two machines the most. The subject of each iteration is the machine that currently has the longest execution time. In the case of the example test suite, this is  $m_1$ . This will be referred to as the *swapper*. Each of the remaining machines are addressed one by one, starting with  $m_2$ , which is a *swappee* candidate. The tests that are currently allocated to the swapper, but also executable on the swappee candidate are retrieved, which in this case is  $t_5$  and  $t_6$ . A list of all possible subsets from this set is then created. This is done by using Python's *itertools* library. The same is done the other way around, with  $t_3$  and  $t_4$  for the swappee candidate. After this, we examine how each swap would affect the overall total duration of the two machines in question. After comparing all subsets, it is concluded that the best swap is  $t_5$  and  $t_6$  from  $m_1$  for  $t_4$  from  $m_2$ , which will decrease the overall duration among the two machines with 1 time unit. Since  $m_3$  can

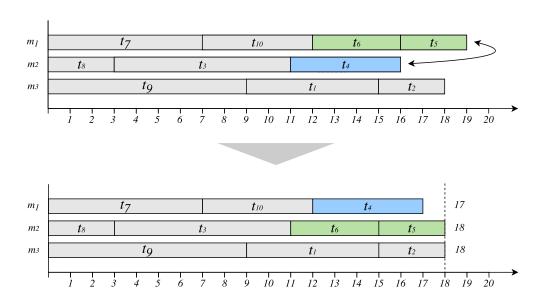


Figure 11: Enhancement Iteration of Example Test Suite

not provide any better options, this swap is conducted. This process will repeat until there is no way to improve the allocations, which in this case is after one swap. In this example, OptiRun found an optimal solution to the optimization problem. Figure 11 shows how the subset swap is conducted, and how it affects the overall duration.

Upon evaluating each swappee candidate, a naive best-case duration is calculated by adding the durations of all the tests from the two machines and dividing the sum by two. If the time used to search for the best swap exceeds the difference between the total duration of the swapper (here: 19), and this best scenario duration (here:  $\frac{19+17}{2} = 18$ ), which in this case is 1, we will continue to the next swappee candidate. This means that we will examine the subsets of  $m_2$  for maximum 1 unit of time.

Sometimes, however, we want to stop the searching earlier, as continuing is no longer beneficial. For that reason, two stop criteria are introduced:

- 1. When the iteration method is first called, a naive best-case overall duration value is calculated by dividing the sum of all test durations by the number of machines available. Once the time difference between this value and the maximum machine duration is exceeded, the iteration process will stop.
- 2. There is a timeout set to 60 seconds. If the iterations still are not finished at this point, the method will be terminated, and the allocations

arrived at that point will be kept.

During the development phase, a clear problem stood out; creation of subset lists from extremely large interchangeable test sets between two machines could take several minutes, and sometimes even lead to memory leaks so that the whole system crashed. This was because there was simply too many possible subsets. To work around this problem, a restriction of the maximum size of the subsets had to be established. Through trial and error, the following was decided upon:

$$f(x) = \begin{cases} x & if & x < 10\\ 20 - x & if & 10 \le x < 20\\ 1 & if & 20 \le x \end{cases}$$

Thus, if the number of interchangeable tests are 19 or more, the subset list will only consist or singular tests in addition to the empty set. Although not optimal, this was a compromise that helped on the problem. This means that for large interchangeable sets, tests can only be swapped one against one or moved from one machine to another. It is therefore sensible to think that better swaps might exist, although identifying these would take too much time and potentially lead to memory leaks. Figure 12 shows a plot of f(x).

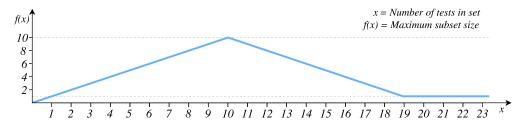


Figure 12: Maximum Subset Sizes

Because of the stop criteria described earlier and the subset size restriction, OptiRun cannot guarantee to find an optimal solution to the allocation problem. However, as explained in Section 2.2, the objective of the optimization problem is to minimize the total time, which means that the time used to search for the best solution is also of high importance, and should be prioritized as such.

#### 5.3.2 OR-X

The implementation of OR-X is of the same style as the OR-Tools example shown in Listing 2 in Chapter 3, but more complex. OR-X takes two lists; one containing the durations of the tests, and another containing which machines each test is executable on. OR-Tools does not support decimal values, so all values in the durations list must be converted to integer values. The *executable on* list is nested, with one list belonging to each test. These inner lists contain binary values representing whether or not the test can be executed on a given machine. It is then added as a constraint that each test should be allocated to exactly one machine, and as the objective that the overall duration should be minimized.

However, there was one major issue upon the implementation of OR-X. It was not possible to interrupt or time out the *NextSolution* method. For very small data sets, this was not a problem, but once the data sets grew slightly bigger and the combination possibilities grew rapidly, the method could take hours. This meant that a set of stop criteria had to be introduced. However, the problem could still not be solved by running the solver in a separate thread, as there is no integrated method of terminating regular threads in Python to the best of the authors knowledge. the problem was solved by introducing Python's *multiprocessing* library.

The solving method is started as a *Process* from the multiprocessing library. In order to allow two processes to share lists, a *Manager* is needed, and to ensure synchronization, a *Lock* has been used. Listing 5 shows how these multiprocessing modules are used in OR-X.

```
1
   from multiprocessing import Process, Manager, Lock
2
3
   manager
              = Manager()
   allocations = manager.dict()
   max_durations = manager.list()
6
   last_updated = manager.list()
              = Lock()
   lock
9
   // ...
10
   p = Process(target=find_solution, args=(durations, executable_on,
11
        allocations, max_durations, last_updated, lock))
12
13
   while p.is_alive():
14
15
     if lock.acquire():
       if <stop criteria fulfilled>:
16
17
        p.terminate()
        break
```

Listing 5: OR-X Multiprocessing

OR-X will be terminated if one of the following stop criteria are fulfilled:

- The search for an enhancement has taken 100 times as long as it took to find the previous enhancement.
- The previous enhancement took 10 times as long to find than the enhancement itself.
- 60 seconds have passed (timeout).

#### 5.4 Database

The relational database management system (RDBMS) SQLite can be regarded as a light-weight substitute to other SQL based database engines. Benefits to SQLite compared to these other database systems include its being self-contained and serverless, and the database is contained in a single disk file [?]. For convenience concerning submission, SQLite was the preferred choice for this project. However, another SQL database engine can replace it with little effort if needed.

As explained in Chapter 3, Django data models define the database layout, and each model typically maps to an individual table in the database. SQL code for creating the database itself and the tables within it are all auto-generated by Django, based on the implementations of the models. If a m-to-n relation between two different models are defined in the model implementations, a separate relationship table is created in the database to cover this. The auto-generated SQL files covering creations and changes to the tables related to test automation are all placed under testautomation  $\rightarrow$  migrations by default. When a change has been done to one of the models, a new migration can be created by performing a few shell commands [?].

The database can be accessed either by writing raw SQL queries, or through Django's API for database abstraction. The former approach was first implemented in this system, but was then changed to the latter, as it was cleaner and more consistent with the implementation of the rest of the project. It was also interesting to use a different practice of database communications than the more commonly used raw SQL queries. Listing 6 shows a simplified example of how an insert statement is conducted using this approach. In this example, the implementation of the Log model is imported, and then an object of this type is created with pseudo values for a small set of attributes. Line 5 in the listing represents the transaction execution and commit. Note that implementation of such object creations includes several more attributes than what is shown in this example.

```
from testautomation.models import Log
from datetime import datetime

l = Log(title='test1', result=1, execution_time=datetime.utcnow())
l.save()
```

Listing 6: Database Communication Using Abstraction API

In addition to providing excellent readability, the abstraction greatly decreases the required number of code lines needed to achieve the same result compared to executing raw SQL queries. This is because the database location does not need to be stated, connection with the database does not need to be programmatically established and then closed when the transaction is finished, and so forth. The abstraction takes care of all of this. The running time of the two approaches also proved to be approximately the same after testing both methods. Other types of queries such as Select, Update and Delete are also supported with this API.

Information about test cases, test groups, schedules, logs as well as user information is all stored in various tables of the database. Figure 13 shows the Entity-Relationship (ER) diagram of the database.

All timestamps stored in the database are in the Coordinated Universal Time (UTC) standard, which, as the name suggests, is universal, and therefore independent of time zones. The time zone used in the web service is set to 'Europe/Oslo' in the Django settings file. Whenever a timestamp is shown on screen, it is first converted to the specified time zone using Django's timezone library. If the user is located in a different time zone than the one specified in the settings, a label explaining that the computer

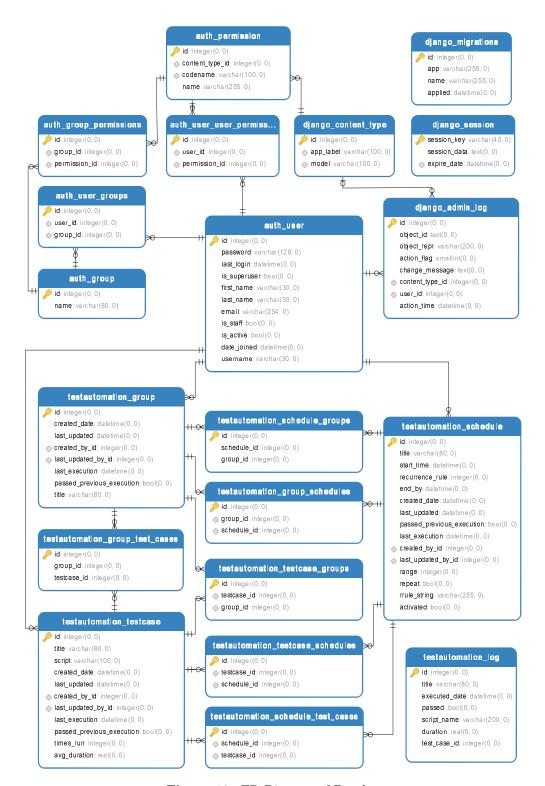


Figure 13: ER Diagram of Database

is xx hours ahead or behind of server time is displayed next to any datetime picker, such as in the schedule creation form.

Storing timestamps according to the UTC standard rather than the current time zone can be considered good practice for multiple reasons. Firstly, there can be no ambiguity. Confusion and misunderstandings related to conversion across different time zones will be avoided, which also means that timestamp calculations are simple. Further, there can be no invalid dates linked to daylight savings time. Moreover, if the server were to be moved to a different time zone, timestamps would have to be converted.

#### 5.5 Dashboard

TODO: Write section intro

#### **5.5.1** Models

With Django, data models provides the foundations on which database tables are created and maintained. A model implementation can be seen as a Python equivalent to an SQL Create Table statement. Model implementations are translated to SQL and executed by Django. Each model, which is a subclass of django.db.models.Model, represents a table in the database, and each model field represents a database field. Similar to SQL, the data type of each field along with any other specifications such as the maximum length of a text field, default values and help text can be passed as field option parameters.

The models are located in  $testautomation \rightarrow models.py$ . In this file, model specifications of test cases, groups, schedules and logs are implemented.

Additionally, models can contain elements that are not linked to the database, such as a field of which a value is based on a function that is performed each time the value is requested displayed. These functions can return primitive data types, but they can also contain HTML tags, so the content can be customized. Such functions are used in several of the models. For instance in the test case module, there are one function that query the Log table in the database, and counts number of instances linked to the particular test case. Average duration and previous execution date are retrieved in a similar manner.

```
from django.db import models

class TestCase(models.Model):
    title = models.CharField(max_length=80)
    script = models.FileField(upload_to='scripts')

def times_run(self):
    tr = # perform query
    return tr
```

Listing 7: Model Implementation

Listing 7 shows a reduced adaption of how the test case model has been implemented. This model contains two model fields and a function. The *script* field is of the type *model.FileField*, and the directory that the files should be uploaded to is passed as a parameter. The file upload itself is taken care of by Django.

#### 5.5.2 Admin

```
TODO: Write introduction to this subsection
```

```
from django.contrib import admin
from .models import TestCase

class TestCaseAdmin(admin.ModelAdmin):
    list_display = ('title', 'times_run', 'avg_dur')
    search_fields = ['title',]
    fields = ('title', 'script')

admin.site.register(TestCase, TestCaseAdmin)
```

Listing 8: Implementation of Model in Administrator Interface

Listing 8 shows a very reduced interpretation of how the test case model is represented in the administrator interface. This adaption builds on the model implementation from Listing 7. Model administrator representations are subclasses of django.contrib.admin.ModelAdmin, and specifies how the model should be represented in the administrator interface. This interpretation specifies values of three of the many ModelAdmin options; which fields of the model should be displayed in the overview list of the test case module ('list\_display'), which fields should be searchable ('search\_fields') and which which fields should be present in the creation/change form ('fields'). Line 9 registers the model to the administrator interface with the specifications stated in the TestCaseAdmin class.

In the actual implementations of the model admin representations, a number of additional fields and specifications are also included. Custom forms with particular validation functionality can be integrated with the creation/change form. This has been done with test case objects to ensure that only test scripts that fulfill certain criteria can be uploaded. Admin actions are also implemented here.

TODO: List and describe the

implemented admin

tions?

#### 5.5.3 Issue Tracker Reporting

TODO

5.5.4 Management of Test Machines

TODO

#### 5.5.5 Event Recurrence

RRULE, short for recurrence rule, is a part of to a Python library called dateutil, which provides an extension to Python's datetime module. It is a small and fast library used in OptiRun to specify recurrence patterns of test executions. Rrule instances can be implemented multiple ways. In this project, it is achieved through passing a string with a specific format, containing information about the desired recurrence constraints. This string is stored in the database, and can be used at any point to create an Rrule instance in order to inquire when the next event should take place. An example of such a string, how Rrule instances are created in this project, and how the next occurrence is retrieved, can be seen in Listing 9.

```
from dateutil.rrule import rrulestr
from datetime import datetime

rule_string = "DTSTART:20160615T160000\nRRULE:FREQ=WEEKLY"
rule = rrulestr(rule_string)

print rule.after(datetime.now())

print rule.after(datetime.now())

>>> 2016-06-15 16:00:00
```

Listing 9: Recursion Rule

The string in the listing above is used to create an rrule instance in which the first occurrence is set to the 15<sup>th</sup> of June 2016 at 4 pm, and repeats weekly. In addition to the recurrence properties shown in the listing above, the library provides an extensive number of recurrence options, including end date, occurrence count and interval.

The recurrence rule strings in this project are built dynamically when a schedule object is created or edited. In the test case, group and schedule modules of OptiRun, recurrence rule strings are used to create RRULE instances, which again are used to find out the time of the next planned execution for the particular test case, group or schedule. RRULE instances are also created by the controller to check when the next test run is scheduled.

# 6 Evaluation

This chapter starts by introducing a collection of test sets used to measure the performance of the allocations mechanism Opti-X through experiments, and compare it to the performance of OR-X on the same collection. The results are evaluated and discussed, followed by an assessment of factors that may threat the validity of the results obtained in the experimental phase. OptiRun is then discussed and evaluated as a whole, before the chapter rounds off by discussing the return on the investment put into this project.

#### 6.1 Experimental Evaluation

The collection of test data is made up of four sub-collections, each designed to measure the performance of the allocation mechanism with a particular number of tests and machines available. Each of these sub-collections consist of three test sets; one in which every test can be executed on every machine, one where the tests can only be executed on a small selection of the machines, and one where the number of machines each test can be executed on varies and is determined at random.

Table 2 shows the details of each test set in the collection. The test data is randomly generated; only the number of tests, test machines and how many machines each test should be executable on was specified upon generation. Each test was assigned a duration between 30 and 120 seconds,

Test Set	# of Tests	# of Machines	# of Executable On
$\overline{ts_1}$	1000	100	100
$ts_2$	"	"	10
$ts_3$	"	"	Random
$\overline{ts_4}$	"	10	10
$ts_5$	"	"	5
$ts_6$	"	"	Random
$ts_7$	200	50	50
$ts_8$	"	"	10
$ts_9$	"	"	Random
$\overline{ts_{10}}$	"	10	10
$ts_{11}$	"	"	5
$ts_{12}$	"	"	Random

Table 2: Test Data

and a set of machines on which they were executable on, both generated at random. The collection of test sets are designed to create a realistic environment, albeit somewhat amplified, while retaining purpose. Larger test sets than  $ts_1$  through  $ts_3$  would not be very meaningful, as 1000 tests and 100 test machines are very large numbers in this context, and are not likely to be exceeded any time soon.

The test sets, located in optirun/controller/test\_data/input, are stored as JSON objects in files with the naming convention test\_set\_x.json, where

		Opti-2	X	OR-X		
Test Set	$T_s$	$\overset{-}{T}_e$	$T_t$	$T_s$	$T_e$	$T_t$
$ts_1$	30.04s	777.00s	807.04s	34.93s	1153.00s	1187.93s
$ts_2$	8.66s	752.00s	$760.66 \mathrm{s}$	34.25s	1224.00s	$1258.25\mathrm{s}$
$ts_3$	0.29s	867.00s	$867.29 \mathrm{s}$	34.44s	1172.00s	$1206.44\mathrm{s}$
$ts_4$	0.02s	7410.00s	7410.02s	30.37s	7797.00s	$7827.37\mathrm{s}$
$ts_5$	0.16s	7497.00s	$7497.16\mathrm{s}$	30.27s	7569.00s	$7599.27\mathrm{s}$
$ts_6$	0.19s	7354.00s	$\mathbf{7354.19s}$	30.26s	7773.00s	$7803.26\mathrm{s}$
$ts_7$	1.72s	301.00s	302.72s	30.39s	365.00s	395.39s
$ts_8$	0.22s	355.00s	$\mathbf{355.22s}$	30.27s	338.00s	$368.27\mathrm{s}$
$ts_9$	0.67s	311.00s	$311.67\mathrm{s}$	30.36s	319.00s	$349.36 \mathrm{s}$
$ts_{10}$	0.18s	1428.00s	1428.18s	30.06s	1435.00s	$1465.06\mathrm{s}$
$ts_{11}$	0.60s	1520.00s	$1520.60\mathrm{s}$	30.06s	1553.00s	$1583.06\mathrm{s}$
$ts_{12}$	0.80s	1549.00s	$1549.80\mathrm{s}$	30.06s	1531.00s	1561.06s

Table 3: Experimental Test Results

x represents the number of the test set. For better human readability, the content of these files can be opened with a JSON viewer, such as http://jsonviewer.stack.hu/. Likewise, the results from the experimental testing are stored in files with the same naming convention, but in the *output* directory.

Each of these 12 tests were tested with both Opti-X and OR-X. The results can be found in Table 3, which provides searching time, execution time and total time obtained with both allocation mechanisms for each test set. As explained in the formal definition of the optimization problem in Chapter 2, the objective of the problem was to minimize the  $T_t = T_e + T_s$ , that is to say the searching time used to find the solution plus the execution time used to execute the tests.

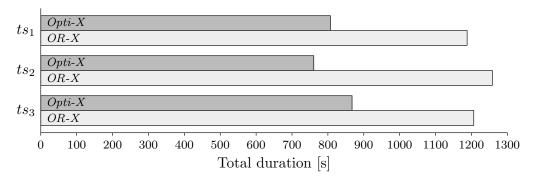


Figure 14: Results of Sub-Collection 1

The first sub-collection of test sets consisted of  $ts_1$ ,  $ts_2$  and  $ts_3$ , with 1000 tests and 100 test machines. This sub-collection was designed to test the mechanisms in situations with a large number of both tests and machines, and an abundance of possible solutions.

As can be seen from Figure 14, Opti-X provided excellent results compared to OR-X for these test sets. Not only were all of the execution times provided by Opti-Run between 305 and 472 seconds faster than OR, it also excelled in searching time.  $ts_1$  was the test set with the most possible solutions, as all tests could be executed on all machines, and thus the only test set with which Opti-X timed out after 30 seconds. Although the searching process of OR-X times out after 30 seconds, the whole process took just over 34 seconds in all of these cases, which is likely due to a combination of the large number of elements in the result list, the lock explained in Chapter 5 not being available and the post-processing needed to extract the result. This was the sub-collection in which Opti-X was the most prominent.

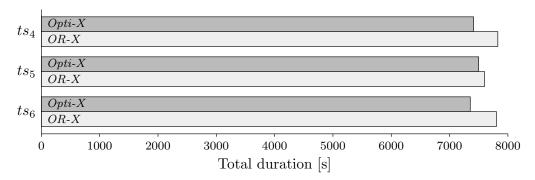


Figure 15: Results of Sub-Collection 2

The second sub-collection of test sets consisted of  $ts_4$ ,  $ts_5$  and  $ts_6$ , with 1000 tests and 10 test machines, and was designed to test the mechanisms in situations with a large number of tests and a small number of machines.

Figure 15 shows that even though Opti-X provided better total time than OR-X on all test sets, the difference between the two mechanisms are less prominent in this sub-collection than in the previous one. The largest difference in total time is 449.07 seconds for  $ts_6$ , and the smallest 102.11 seconds for  $ts_5$ , which are both almost insignificant considering the total duration is more than two hours for both mechanisms on each test set in the sub-collection. This is likely linked to the much smaller number of possible solutions in the test sets in this sub-collection compared to the first one. Nevertheless, Opti-X undeniably outperformed OR-X in searching time, where Opti-X used less than a second on each test set and OR-X timed out after 30 seconds, as well as in the execution time, and thus the total time on all test sets in the sub-collection.

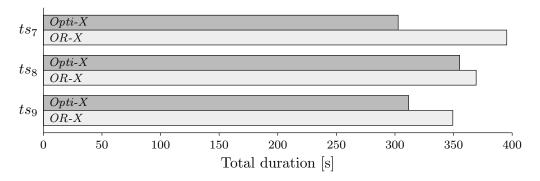


Figure 16: Results of Sub-Collection 3

The third sub-collection of test sets consisted of  $ts_7$ ,  $ts_8$  and  $ts_9$ , with 200 tests and 50 test machines, and was designed to test the mechanisms in situations with more realistically number of tests, and a fair number of machines.

Figure 16 shows that Opti-X once again provided better total results for all of the test sets in the sub-collection. Opti-X finished searching after 1.72, 0.22 and 0.67 seconds respectively for the three test sets, while OR-X again was timed out after 30 seconds.

 $ts_7$ , which was the test set with the largest number of possible solutions, was also the one in which the difference between the two mechanisms were distinct with a total difference of 92.67 seconds. The two remaining differences were not as conspicuous, but what was interesting was that OR-X provided a 17 second shorter execution time for  $ts_8$  than Opti-X accomplished. However, by using 30 seconds to find this solution, Opti-X still provided a better total time. This demonstrates that it is not enough to find an allocation in which the execution time is minimized; it is also important to minimize the time taken to find the solution.

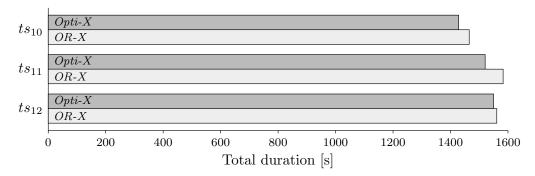


Figure 17: Results of Sub-Collection 4

The fourth and last sub-collection of test sets consisted of  $ts_{10}$ ,  $ts_{11}$  and  $ts_{12}$ , with 200 tests and 10 test machines, and was designed to test the mechanisms in situations with realistic number of both tests and machines.

Again, Opti-X produced better results for all of the test sets, which can be seen in Figure 17. As with sub-collection 3, the differences in results were not as outstanding for this sub-collection as for the two first ones. This is likely due to a much more limited amount of possible solutions to the problem, as the number of test machines was small and the number of tests moderate.

The largest difference in execution time was a mere 33 seconds for  $ts_{11}$ . Again, OR-X produced a better solution of execution time for  $ts_{12}$ , but since Opti-X used less than a second to find its solution for this and the remaining test sets, and OR-X again was timed out after 30 seconds for the whole sub-collection, Opti-X provided the overall best result of 1549.80 seconds, which was 11.26 seconds less than the total time found by OR-X for this problem.

Reformulate this entire section

Opti-X provided better overall results than OR-X on all of the test sets in the collection. The searching time was generally exceptional, aside from for  $ts_1$ , where the searching was timed out after 30 seconds, and  $ts_2$ , where the searching time was more than 8 seconds. Compared to the searching times of OR-X, Opti-X did a better job

Bla bla trend that OR-X uses a lot of time to find good results, and that the results found most of the time are bad compared to Opti-X. Both mechanisms were timed out upon solving the problem with the most possible solutions, but Opti-X' result was still 68% of Or-X'.

OR-X could perhaps have provided better execution time for most, if not all, test sets if given longer time before the search timed out. However,

Although the results provided by OR-X generally can not compete with those provided by Opti-X, the former generally produced decent results most of the time. OR-Tools is an excellent library for solving CP and combinatorial problems, but Opti-X provided better results in this specific problem. A contributing factor to this can be explained by Opti-X being designed and implemented specifically to solve this type of problem efficiently, whereas OR-Tools was designed to provide feasible solutions to a much broader range of problems. It is therefore not guaranteed to make the best decisions at any point of the process, so the time taken to identify good solutions is inclined to take more time. This is likely the reason why OR-X did not stop before the timeout of 30 seconds on any of the test sets, whereas Opti-X used less than a second on most of them.

#### 6.2 Threats to Validity

Judging from the results obtained in the previous section, it is clear that Opti-X is performs superiorly compared to OR-X in the experiments. Before

moving on to discussing OptiRun as a whole, it is therefore important to establish some possible contributing factors that may threaten the validity of the results.

As explained in the previous section, the test data was randomly generated. This means that there might have been some degree of chance involved, and that if the test sets were generated again, other results may be obtained. Additionally, there might have been some interesting situations in which to test the performance of the allocation mechanisms, that are not covered in the experimental evaluation. Even though the author tried their best to come up with realistic and representative scenarios to cover the allocation mechanism in full measure, there just might have been some scenarios that did not come to mind. Also, there might be some scenarios that would provide value despite being deemed pointless to test by the author.

The results may also depend on the performance of the machine that the experiments were conducted on. Had the experiments been conducted on a newer and faster machine with better specs, the results would likely come out slightly different. OR-X would perhaps be able to reach better solutions before being timed out, and could thus possibly be a stronger competitor, whereas the results obtained by Opti-X would not likely be affected as much, as most of the problems were solved in less than a second.

Another contributing factor may be the author's knowledge of and experience with OR-Tools being far from optimal. With limited documentation and general online and literary coverage, exploring every corner of the library just could not be done with a narrow time-frame and other tasks that had to be prioritized. Although the author did invest a fair amount of time and energy in getting acquainted with the tools and tried their best to implement OR-X to be as strong of a competitor to Opti-X as possible, it may be a very real possibility that there are ways to implement it that would provide greater efficiency. Another possibility is that there are other optimization libraries that could potentially accomplish better results in this optimization problem than OR-Tools.

#### 6.3 Discussion

TODO

#### 6.4 Return on Investment

TODO

# 7

### Further Work & Conclusion

# 7.1 Further Work TODO 7.2 Conclusion



TODO



### **Automated Test Template**

This template for automated test scripts should be strictly conformed to:

```
import json
    import sys
3
   import unittest
   from selenium import webdriver
5
   class NameOfTest(unittest.TestCase):
6
      @classmethod
8
       def setUpClass(cls):
9
          data = json.loads(sys.argv[1])
10
11
          cls.driver = webdriver.Remote(
12
             command_executor=data['command_executor'],
             desired_capabilities=data['desired_capabilities']
13
14
15
       def name_of_test(self):
16
          """ Write the test script here """
17
18
19
       @classmethod
20
       def tearDownClass(cls):
21
          if cls.driver is not None:
22
             cls.driver.quit()
   if __name__ == "__main__":
24
       unittest.main(argv=['TestCase'])
25
```

Listing 10: Template for Automated Test Scripts

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