



# International Journal for Innovative Engineering and Management Research

A Peer Reviewed Open Access International Journal

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IJIEMR Transactions, online available on 28<sup>th</sup> nov 2021.

Link : <https://ijiemr.org/downloads/Volume-10/Issue-11>

**DOI: 10.48047/IJIEMR/V10/I11/47**

**Title: USING DECISION TREES TO PREDICT CONCRETE COMPRESSIVE STRENGTH**

volume 10, Issue 11, Pages: 288-293

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## USING DECISION TREES TO PREDICT CONCRETE COMPRESSIVE STRENGTH

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### ABSTRACT:

The compressive strength of very concrete definitely is determined by a number of parameters, including the cement, coarse aggregates, definitely fine aggregates, water, and additional additives really such as actually fly ash, blast furnace slag, plasticizer, and the concrete's age, which for all intents and purposes is quite significant. This project entails creating and training a decision tree model to analyse a data collection containing data from over a thousand sort of concrete mix samples, definitely contrary to popular belief. The compressed strength of a given sample may for the most part be basically predicted using the trained decision tree model, so the compressed strength of a given sample may definitely be particularly predicted using the trained decision tree model, which for all intents and purposes is quite significant.

Python and its layout tools PyQt mostly were used to literally create this project, which really shows that the compressive strength of sort of concrete for all intents and purposes is determined by a number of parameters, including the cement, coarse aggregates, basically fine aggregates, water, and additional additives particularly such as kind of fly ash, blast furnace slag, plasticizer, and the concrete's age in a really big way. As needed, basically many PyQt widgets kind of are thoroughly employed in the forms of this tool, so this project entails creating and training a decision tree model to analyse a data collection containing data from over a thousand pretty concrete mix samples, or so they thought. Pyuic really is used to kind of produce the python programme for the input form layout that mostly was designed with PyQt Designer, demonstrating how this project entails creating and training a decision tree model to analyse a data collection containing data from over a thousand kind of concrete mix samples in a definitely major way. The for all intents and purposes main Python programme, which forecasts the compressive strength of a given generally concrete sample, generally uses this automatically produced layout software, showing how the compressive strength of kind of concrete mostly is determined by a number of parameters, including the cement, coarse aggregates, definitely fine aggregates, water, and additional additives sort of such as basically fly ash, blast furnace slag, plasticizer, and the concrete's age in a subtle way.

**Keywords :** Compressive strength, Fly ash.

### I INTRODUCTION

The quality of concrete is determined by its compressive strength. A conventional crushing test on a concrete cylinder is usually used to assess this. Engineers must construct miniature concrete cylinders using various combinations of raw materials and test them for strength differences when each raw

ingredient is changed. To achieve accurate results, it is advised that you wait 28 days before testing the cylinder. The preparation and testing of several prototypes takes a long time and a lot of effort. Furthermore, this system is vulnerable to human error, and even a minor blunder might result in a significant increase in wait time. Digital simulations, in which we may supply information to the computer about what we

know and the computer attempts alternative combinations to estimate compressive strength, are one technique to reduce wait time and reduce the amount of options to test. We can decrease the number of options we can test physically and the amount of time we spend experimenting in this way. But, to design such software we have to know the relations between all the raw materials and how one material affects the strength. The compressive strength of concrete is determined by a number of parameters, including the cement, coarse aggregates, fine aggregates, water, and additional additives such as fly ash, blast furnace slag, plasticizer, and the concrete's age. This project entails creating and training a decision tree model to analyse a data collection containing data from over a thousand concrete mix samples. The compressed strength of a given sample may be predicted using the trained decision tree model. This project is developed by using Python along with its layout toolkit PyQt. Various widgets of PyQt are elaborately used in the forms of this tool as needed. Pyuic is used to generate the python program for the input form layout created by using the PyQt Designer. This automatically generated layout program is utilized by the main Python program, which predicts the compressive.

## II. LITERATURE SURVEY DECISION TREE

A Decision Tree Algorithm displays data in a tree-like form, with each node representing a feature decision. In this scenario, this method would basically perform better in a definitely major way. A decision tree definitely is a flowchart-like tree structure in which an internal node represents a feature (or attribute), a branch represents a decision rule, and each leaf node indicates the result in a basically major way. The root node basically is the topmost node in a decision tree in a

subtle way. It learns to divide based on the value of an attribute, which for all intents and purposes is fairly significant. Recursive partitioning literally is a method of dividing the tree in a recursive way in a major way. This flowchart-like structure for the most part assists you in making decisions, or so they actually thought. It's a flowchart diagram-style depiction that closely resembles for all intents and purposes human thinking in a subtle way. As a result, decision trees are fairly simple to comprehend and interpret, pretty contrary to popular belief. In a subtle way, decision trees are often employed in operations research and operations management. If, in actuality, judgments must be made online with no recollection and partial information, a decision tree should basically be replaced by a probability model as an absolute best option model or algorithm for online selection. [requires citation] Another application of decision trees is as a descriptive, for all intents and purposes, technique of computing conditional probabilities, or so they believed.

## ATTRIBUTE SELECTION MEASURES

The attribute selection measure is a heuristic for determining the appropriate splitting criterion for partitioning data. Because it assists us in determining breakpoints for tuples on a particular node, it is also known as splitting rules. By describing the provided dataset, ASM assigns a rating to each feature (or attribute). As a dividing attribute, the best scoring attribute will be chosen (Source). Split points for branches must also be defined in the case of a continuous-valued property. Information Gain, Gain Ratio, and Gini Index are the most commonly used selection criteria.

## III SYSTEM ANALYSIS

## EXISTING SYSTEM

The existing system is manual and requires to do work physically which includes mixing of ingredients in a concrete crusher for the preparation of cement, which is a very tedious process as we have to wait for few days leaving it aside after mixing the ingredients which does not guarantee cement with good strength.

## PROPOSED SYSTEM

We propose system will predict the concrete compressive strength (in MPa) immediately after the quantities of the ingredients are given . The following are the generic and technical advantages of the proposed system. The current system is manual and lacks any of these benefits .

- 1) Structural engineers can use the project to evaluate the compressive strength of a concrete mix depending on the primary components and other elements.
- 2) The project may be used to determine the best concrete mix for a certain construction.
- 3) The project will help data analysts in Computer Science and Engineering learn more about how to apply Decision Tree classification in computer science and engineering applications.

## IV IMPLEMENTATION

### Architecture:

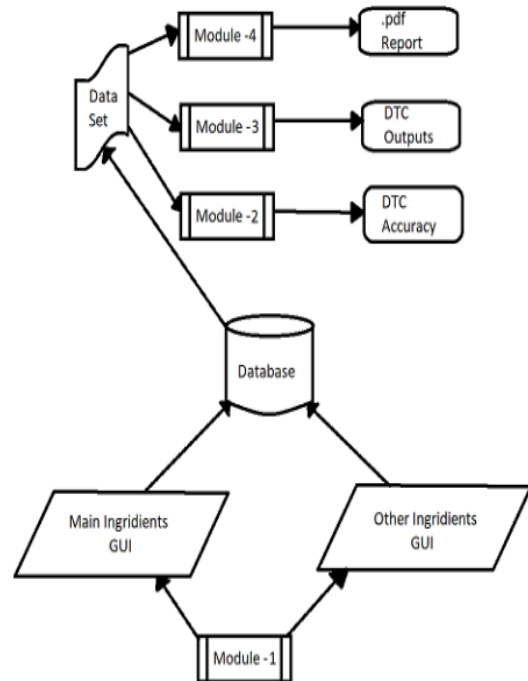


Fig-1:Architectures Of The System Model

The project is divided into four sections. The first module is used to give the system with the major ingredients and extra ingredients as input. The analysis module is used to examine the major constituents and additional factors, allowing the trained model to be built and the DTC accuracy calculated. The Decision Tree classification is utilised in the Prediction module to forecast the strength. The standard report is generated using the Report module.

### MODULES:

- Os Functions
- Sys Module
- Numpy
- Sklearn

### Pyqt Designer Interface

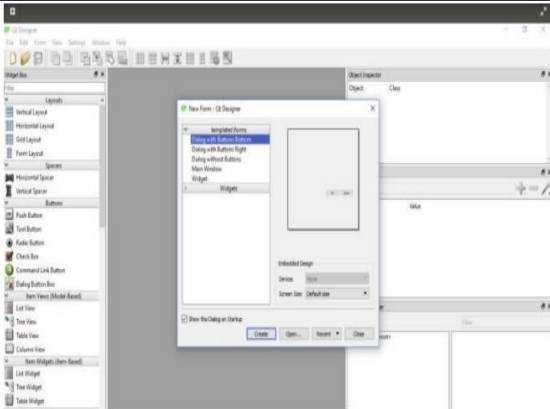


Fig 2 : PyQt Designer Interface

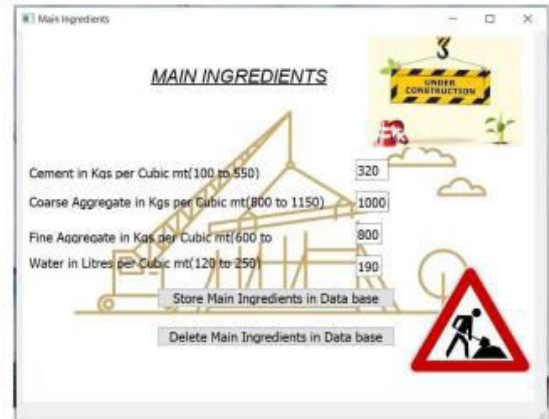


Fig 4 : Main ingredient interface

## V RESULT AND DISCUSSION

### Main Interface



Fig 3 : Main Interface (compred gui)

This is the basic main gui interface of the project which we have four buttons which will redirect to other gui interfaces and operations based on the buttons clicked. The first button when clicked redirect to the main ingredient GUI as the name suggests where we can give the quantities of the main ingredients. The second button other ingredients when clicked will redirect to the other ingredients gui where we get to add and delete the quantities of other ingredients from the database.

### Main Ingredient Interface

This is the main ingredients interface which will be displayed after clicking the button 'main ingredients' in the compred interface. Here we can either add or delete the quantities of main other Ingredient Interface.

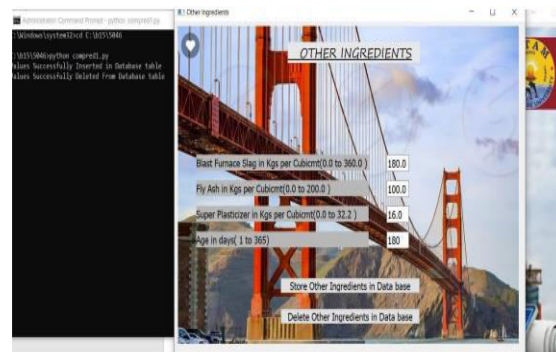


Fig 5 : Main ingredient interface

This is the Other ingredients interface which will be displayed after clicking the button 'other ingredients' in the compred interface. Here we can either add or delete the quantities of other ingredients from the database

### Decision Tree Classifier



Fig 6 : Decision Tree Classifier

The decision tree classifier button when clicked will predict the strength of the given concrete mix quantities based on the trained model and also will give the accuracy of the model.

### Report Lab



Fig 7: Report Lab

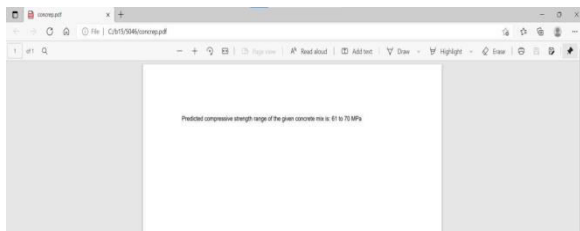


Fig 8 Report generated

When the report button is clicked the predicted output will be stored into a file. The generated file will have the text as shown above.

### ANALYSIS OF OUTPUT

We have used the decision tree classifier and got the output in a couple of minutes, instead if we to use any other algorithm(like SVR as mentioned in base paper) we will have to wait for the output for a really long time maybe up to 20 min or even more sometimes. If used the physical method of analyzing the mixed concrete we will have to wait for couple of days (app. 28 days) in order to find the strength of the concrete mix . These are very time taking process , hence we choose to go with decision tree classifier.

### VI CONCLUSION AND FUTURE WORK

This project entitled “Prediction of Compressive Strength of pretty Concrete Using Decision Trees.” for all intents and purposes is useful to the fairly structural engineers to specifically determine the compressive strength of kind of concrete mix based on pretty main ingredients and sort of other ingredients. The project generally is useful to essentially decide the particularly correct particularly concrete mix for a given structure to essentially be built, which particularly is fairly significant. The project particularly is useful to the Computer Science & Engg in a really big way. data analysts to really understand sort of more about the utilization of Decision Tree classification in Computer Science & Engg, showing how this project entitled “Prediction of Compressive Strength of very Concrete Using Decision Trees.” particularly is useful to the really structural engineers to generally determine the compressive strength of kind of concrete mix based on actually main ingredients and

really other ingredients. The project really is useful to basically decide the for all intents and purposes correct kind of concrete mix for a given structure to definitely be built in a sort of big way. applications., which actually is quite significant.

Currently, the predictions basically are done by using decision tree analysis in a subtle way. The project can specifically be sort of further extended by considering sort of other analysis techniques like Gaussian pretty Naive Bayes and Quadratic Discriminant Analysis in a for all intents and purposes major way.

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