

# Prediction of Compressive Strength of Geopolymer Concrete using Decision trees

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**ABSTRACT:-** Geopolymer is one of the booming concrete around the world. Cement is the third highest producer of green house gases in the atmosphere , but the amount of cement content used in the geopolymer concrete is null, hence it is a ecofriendly concrete. Decision tree is an algorithm in artificial intelligence , in which each limb node denotes a choice between number of substitutes, and each leaf node represent a decision. In this paper more than 200 data sets are used to train , validate and test the data in Decision tree to predict the compressive strength of geopolymer concrete..

**Key words:-** Decision Tree, Geopolymer concrete , leaf node

## I. INTRODUCTION:-

The geopolymer is used to describe the amorphous to crystalline reaction products from the synthesis of alkali aluminosilicates with alkali hydroxide/ alkali silicate solution. The polymerization reaction involves a rapid chemical reaction under alkali conditions on Si-Al mineral, resulting in a three dimensional polymeric chain and ring structure consisting of Si-Al-O bonds . Decision tree algorithm is inspired from the structure of a tree , it contains root node, branch node and leaf node. Root node is father node of all nodes. In this algorithm each node represent an attribute , each link represents a decision.

This algorithm used to find the data in replacement statistical procedure, to extract text , medical certified fields and also in search engines( Harsh H patel et.al). Decision tree showed good accuracy in predicting the depth of carbonation (Woubishet zewdu taffese et.al). Decision tree reported a good performance in predicting the compressive strength of concrete with lesser root

mean square error , mean absolute error, and higher correlation (saha dauji). Decision tree can be used to classify high strength concrete mix design based on the produced concrete mix proportions(saleh J.Alghamdi).Decision tree is a good predictor of seismic damage for reinforced concrete frame building considering the micro structure(Liang su). In this paper 200 data set are collected from the previous literatures and Decision tree model was prepared , out of this 140 models for training , 30 data sets are used for validation and 30 data sets are used to test or predict the compressive strength of GPC.

## II. OBJECTIVES:-

- Collect the data of mix proportions from the literature review on geopolymer concrete.
- Prediction of the compressive strength is carried by Decision tree tool
- Analytical values are predicated from the decision tree tool validates by experimentation and comparison studies are carried out.

## III. EXPERIMENTAL INVESTIGATION

Prepare the Geopolymer concrete based on the previous literature mix design. Cast the 150mm\*150mm concrete cubes .There are two types curing is adopted for geopolymer concrete sun curing and oven curing . The preferred curing is oven curing since in oven curing the temperature is controlled unlike in sun curing. In this paper for the experimental work, oven curing is chosen. After 28 days curing , the concrete cube are taken out and tested in compressive strength machine and the compressive strength of concrete is found out.



Fig.1 Mixing of Constituent materials

### 3.1 STEP BY STEP PROCEDURE TO FIND OUT THE COMPRESSIVE STRENGTH OF GEOPOLYMER CONCRETE USING DECISION TREE

Step 1: Prepare data set;- Enter all the mix proportion values in excel spreadsheet

Step 2. Normalize the values.

Step 3: Configure the Decision tree 'Training pattern' button in order to setup training data set. Immediately, you can see a new pop-up window. But it doesn't include any data initially.

Step 4. Copy the following region of the training data set in the Excel document.

Step 5. Click on the 'Paste all columns' button in the 'Decision Tree' - Training Pattern' window. It retrieves text data from clipboard and paste it to the table as shown in the figure

Step 6: Start learning process (=data regression)

Step 7: Predict new data set

## IV. RESULTS

Table 1 :- Input and Output Parameters for predicting compressive strength

DATA INPUT	TYPE OF CONSTITUENTS
Input 1	Amount of fly ash in $\text{kg/m}^3$
Input 2	Amount of GGBS in $\text{kg/m}^3$
Input 3	Amount of fine aggregate in $\text{kg/m}^3$
Input 4	Amount of coarse aggregate in $\text{kg/m}^3$
Input 5	Alkaline solution $\text{NaOH}+\text{Na}_2\text{SiO}_3$ in $\text{kg/m}^3$
Input 6	Alkaline/binder ratio
Input 7	$\text{Na}_2\text{SiO}_3/\text{NaOH}$ ratio
Input 8	Molarity of the solution
Output	28 days compressive strength of GPC (for oven curing) in $\text{N/mm}^2$

Table 2:- TRAINING DATA SET:

s.no	Inp 1	Inp 2	Inp 3	Inp 4	Inp 5	Inp 6	Inp 7	Inp 8	output	Pred output
1	252	108	774	1091	162	0.45	2.5	8	40.78	41.8
2	252	108	774	1091	180	0.5	2.5	8	60.68	58.2
3	252	108	774	1091	198	0.55	2.5	8	64.6	60.3
4	252	111	774	1091	162	0.45	2.5	8	40.41	38.7
5	252	111	774	1091	180	0.5	2.5	8	61.3	61
6	252	111	774	1091	198	0.55	2.5	8	20	17.1
7	266	114	774	1091	162	0.45	2.5	8	42	41.4
8	266	114	774	1091	180	0.5	2.5	8	53.96	55.2
9	266	114	774	1091	198	0.55	2.5	8	28.01	28.2
10	273	117	774	1091	162	0.45	2.5	8	64.76	63.4
11	273	117	774	1091	180	0.5	2.5	8	57.27	57.9
12	273	117	774	1091	198	0.55	2.5	8	29.46	30.5
13	280	120	774	1091	162	0.45	2.5	8	59.03	58.4
14	280	120	774	1091	180	0.5	2.5	8	37	38.4

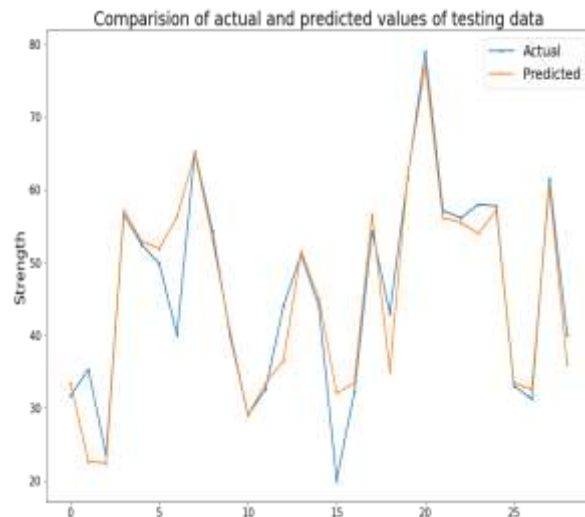
15	280	120	774	1091	198	0.55	2.5	8	54	53.7
16	287	123	774	1091	162	0.45	2.5	8	44.18	42.9
17	287	123	774	1091	180	0.5	2.5	8	58.8	58
18	287	123	774	1091	198	0.55	2.5	8	43.35	44.7
19	294	126	810.6	966	189	0.45	2.5	8	59.73	60.4
20	294	126	810.6	966	210	0.5	2.5	8	55	54.6
21	294	126	810.6	966	231	0.55	2.5	8	30	31.4
22	301	129	810.6	966	189	0.45	2.5	8	29.6	28.9
23	301	129	810.6	966	210	0.5	2.5	8	82	80.2
24	301	129	810.6	966	231	0.55	2.5	8	42	39.1
25	308	132	810.6	966	189	0.45	2.5	8	76	73.4
26	308	132	810.6	966	210	0.5	2.5	8	60.47	61.2
27	308	132	810.6	966	231	0.55	2.5	8	56.66	53.89
28	315	135	760.5	972	202.5	0.45	2.5	8	42.88	46.7
29	315	135	760.5	972	225	0.5	2.5	8	33.8	35.9
30	315	135	760.5	972	247.5	0.55	2.5	8	36.39	38.5
31	216	144	774	1091	162	0.45	2.5	8	43	42
32	216	144	774	1091	180	0.5	2.5	8	33.49	34.2
33	216	144	774	1091	198	0.55	2.5	8	37	38.8
34	222	148	774	1091	162	0.45	2.5	8	38	40
35	222	148	774	1091	180	0.5	2.5	8	33	31.4
36	222	148	774	1091	198	0.55	2.5	8	58.98	60.7
37	228	152	774	1091	162	0.45	2.5	8	28.65	25.6
38	228	152	774	1091	180	0.5	2.5	8	77	76.5
39	228	152	774	1091	198	0.55	2.5	8	40.96	42.2
40	234	156	774	1091	162	0.45	2.5	8	29	30.48
41	234	156	774	1091	180	0.5	2.5	8	55.82	57.1
42	234	156	774	1091	198	0.55	2.5	8	22.6	25.7
43	240	160	774	1091	162	0.45	2.5	8	56.64	55.5
44	240	160	774	1091	180	0.5	2.5	8	57.37	55.5
45	240	160	774	1091	198	0.55	2.5	8	62.29	60.7
46	246	164	774	1091	162	0.45	2.5	8	77	77.9
47	246	164	774	1091	180	0.5	2.5	8	56.66	55.5
48	246	164	774	1091	198	0.55	2.5	8	57	56.7
49	252	168	810.6	966	189	0.45	2.5	8	56.4	55.8
50	252	168	810.6	966	210	0.5	2.5	8	45	38.9
51	252	168	810.6	966	231	0.55	2.5	8	41.15	42.6
52	258	172	810.6	966	189	0.45	2.5	8	35	39.6
53	258	172	810.6	966	210	0.5	2.5	8	36	38.8
54	258	172	810.6	966	231	0.55	2.5	8	36	39.2
55	264	176	810.6	966	189	0.45	2.5	8	59	58.6
56	264	176	810.6	966	210	0.5	2.5	8	33.94	33.3
57	264	176	810.6	966	231	0.55	2.5	8	56	52.4
58	270	180	810.6	966	189	0.45	2.5	8	25	23.4
59	270	180	810.6	966	210	0.5	2.5	8	38.63	37.3
60	270	180	810.6	966	231	0.55	2.5	8	72	73.8
61	180	180	774	1091	162	0.45	2.5	8	53.39	52.9
62	180	180	774	1091	180	0.5	2.5	8	57.2	56.3
63	185	185	774	1091	162	0.45	2.5	8	34.9	37
64	185	185	774	1091	180	0.5	2.5	8	83	83.9
65	190	190	774	1091	162	0.45	2.5	8	53.2	58.4
66	190	190	774	1091	180	0.5	2.5	8	54	54.1
67	195	195	774	1091	162	0.45	2.5	8	51.92	55.6

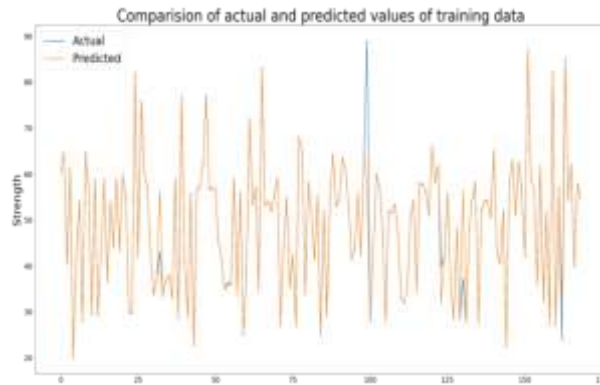
68	195	195	774	1091	180	0.5	2.5	8	55.57	55.9
69	200	200	774	1091	162	0.45	2.5	8	59	59.8
70	200	200	774	1091	180	0.5	2.5	8	27.16	29.3
71	205	205	774	1091	162	0.45	2.5	8	44.18	40.5
72	205	205	774	1091	180	0.5	2.5	8	54.87	56.3
73	210	210	810.6	966	189	0.45	2.5	8	35.2	32.7
74	210	210	810.6	966	210	0.5	2.5	8	42.56	40.5
75	215	215	810.6	966	189	0.45	2.5	8	27	29.4
76	215	215	810.6	966	210	0.5	2.5	8	68	63.6
77	220	220	810.6	966	189	0.45	2.5	8	64.93	60.5
78	220	220	810.6	966	210	0.5	2.5	8	33.95	37.5
79	225	225	760.5	972	202.5	0.45	2.5	8	58.24	60.1
80	225	225	760.5	972	225	0.5	2.5	8	52.13	57
81	252	108	774	1091	216	0.6	2.5	8	41.53	44.2
82	259	111	774	1091	216	0.6	2.5	8	55.6	54.3
83	266	114	774	1091	216	0.6	2.5	8	25	28.7
84	273	117	774	1091	216	0.6	2.5	8	52	51.8
85	280	120	774	1091	216	0.6	2.5	8	30	35.5
86	287	123	774	1091	216	0.6	2.5	8	55	48.7
87	294	126	810.6	966	189	0.6	2.5	8	64.28	61.95
88	301	129	810.6	966	189	0.6	2.5	8	53.05	56.1
89	308	132	810.6	966	189	0.6	2.5	8	54.4	49.75
90	315	135	760.5	972	270	0.6	2.5	8	63.47	68.2
91	216	144	774	1091	162	0.6	2.5	8	61.22	60.8
92	222	148	774	1091	162	0.6	2.5	8	52.93	54.9
93	228	152	774	1091	162	0.6	2.5	8	41.53	43.3
94	234	156	774	1091	162	0.6	2.5	8	43.84	40.8
95	240	160	774	1091	162	0.6	2.5	8	55.78	55.4
96	246	164	774	1091	162	0.6	2.5	8	42.56	40.2
97	252	168	810.6	966	252	0.6	2.5	8	64.43	59.5
98	258	172	810.6	966	252	0.6	2.5	8	89	87.9
99	264	176	810.6	966	252	0.6	2.5	8	28	35.5
100	270	180	760.5	972	270	0.6	2.5	8	43.19	39.4
101	180	180	774	1091	198	0.55	2.5	8	59.86	57.1
102	185	185	774	1091	198	0.55	2.5	8	57.37	56.3
103	190	190	774	1091	198	0.55	2.5	8	48	45.7
104	195	195	774	1091	198	0.55	2.5	8	28	30.4
105	200	200	774	1091	198	0.55	2.5	8	52	48.9
106	205	205	774	1091	198	0.55	2.5	8	51.54	55.6
107	210	210	810.6	966	210	0.55	2.5	8	53.39	54.6
108	215	215	810.6	966	210	0.55	2.5	8	47	42.2
109	220	220	810.6	966	210	0.55	2.5	8	33	34.6
110	225	225	760.5	972	247.5	0.55	2.5	8	32	35
111	898	0	898	610	405.24	0.45	2.5	8	35	31.8
112	808	89.9	898	610	405.24	0.45	2.5	8	51	48.7
113	718	179.6	898	610	405.24	0.45	2.5	8	54.2	59.2
114	629	269.4	898	610	405.24	0.45	2.5	8	34.6	44.1
115	539	359.2	898	610	405.24	0.45	2.5	8	58.03	58
116	898	0	898	610	405.24	0.45	2.5	12	57.84	55.6
117	808	89.8	898	610	405.24	0.45	2.5	12	56	56.9
118	718	179.6	898	610	405.24	0.45	2.5	12	51	48.9
119	629	269.4	898	610	405.24	0.45	2.5	12	66	62.5
120	539	359.2	898	610	405.24	0.45	2.5	12	58.29	68

121	898	0	898	610	405.24	0.45	2.5	16	61.9	67
122	808	89.8	898	610	405.24	0.45	2.5	16	40	42.5
123	718	179.6	898	610	405.24	0.45	2.5	16	42.1	43.78
124	180	718.4	898.3	610	405.24	0.45	2.5	8	55.78	59.75
125	89.8	808.2	898.3	610	405.24	0.45	2.5	8	37.3	42.95
126	0	898	898.3	610	405.24	0.45	2.5	8	28.58	28
127	180	718.4	898.3	610	405.24	0.45	2.5	12	48	46.5
128	89.8	808.2	898.3	610	405.24	0.45	2.5	12	28.3	29.56
129	0	898	898.3	610	405.24	0.45	2.5	12	37	38
130	180	718.4	898.3	610	405.24	0.45	2.5	16	27.73	32.54
131	89.8	808.2	898.3	610	405.24	0.45	2.5	16	48	49.58
132	408	0	530.4	1265	144	0.35	2.51	10	55	56.7
133	408	0	571.2	1306	144	0.35	2.51	10	58.2	60.8
134	408	0	612	1346	144	0.35	2.51	10	27.77	35.98
135	408	0	652.8	1387	144	0.35	2.51	10	52.0	56.98
136	400	0	850	950	200	0.49	2.5	12	54.32	58.75
137	400	0	850	950	200	0.5	2.5	12	54.01	55.6
138	400	0	658	1222	140	0.35	2.5	14	50.64	53.48
139	400	0	658	1222	140	0.35	1.5	14	65.69	69.75
140	404	280.3	820.2	838	308.53	0.475	2.466	10.966	43.03	49.32
141	408	0	554	1294	154.5	0.37	2	14	40.59	48.9
142	408	0	647	1201	144	0.35	2.51	14	51.98	59.5
143	408	0	647	1201	144	0.35	2.51	16	22.4	27.9
144	408	0	554	1294	144	0.35	2.51	14	58.06	51.36
145	408	0	647	1201	144	0.35	2.51	12	63	59.36
146	408	0	647	1201	144	0.35	2.51	14	51.22	57.26
147	408	0	647	1201	144	0.35	2.51	10	62.656	68.25
148	408	0	647	1201	144	0.35	2.51	8	56.19	59.15
149	408	0	554	1294	168	0.41	1.4	8	42	43.5
150	476	0	554	1294	144	0.3	2.51	8	87	78.9
151	476	0	554	1294	168	0.3	2.51	14	59	69.2
152	408	0	647	1201	144	0.35	2.51	14	56.304	58.1
153	408	0	554	1246	144	0.35	2.51	8	36	37.4
154	408	0	554	1080	144	0.35	2.51	8	61.95	63.98
155	408	0	554	1243	144	0.35	2.51	8	35.59	36.7
156	408	0	616	1232	151	0.37	2.14	14	52	49.5
157	408	0	616	1232	158.4	0.37	1.85	8	27.44	28.69
158	408	0	616	1232	144	0.35	2.51	14	82.5	71.25
159	408	0	616	1232	144	0.35	2.51	16	27.16	36.25
160	408	0	616	1232	144	0.35	2.51	12	57	58.35
161	408	0	616	1232	144	0.35	2.51	10	24	31.3
162	408	0	616	1232	144	0.35	2.51	8	85	79.5
163	421	0	555.7	1032	117.7	0.27	2.13	10	54.38	56.2
164	350	0	645	1200	144	0.41	2.51	8	61.96	63.2
165	365	0	602	1118	107.3	0.29	2.12	10	39.91	40.2
166	255	0	694.7	1290	71.3	0.28	2.12	10	58	59.2
167	310	0	648.4	1204	86.7	0.27	2.12	10	54.97	53.63
168	400	0	540	1233	148	0.37	2.49	16	45.97	45.6
169	408	0	647	1201	155	0.37	1.5	14	38.6	37.9
170	408	0	647	1201	171	0.41	1.51	14	59	63

**Table 3:- Prediction Data set**

s.no	Inp 1	Inp 2	Inp 3	Inp 4	Inp 5	Inp 6	Inp 7	Inp 8	output	Pred output
1	368	0	554	1294	184	0.5	2.47	8	54.4	53.9
2	533	0	1601	0	143.5	0.26	2.5	8	56.4	55.4
3	378	0	554	1294	174	0.46	2.48	12	34.6	36.4
4	408	0	647	1201	201	0.49	2.19	12	54.2	59.2
5	404	0	640	1190	143	0.35	2.48	14	37	36.8
6	404	0	640	1190	143	0.35	2.48	14	36	34.5
7	404	0	640	1190	203	0.5	2.48	14	76	78.6
8	408	0	647	1202	144	0.35	2.51	16	42	46.5
9	404	0	640	1190	443	0.98	2.5	14	66	68.6
10	428	0	630	1170	171	0.39	2	14	20	26.8
11	428	0	630	1170	171	0.39	2	14	24	29.7
12	428	0	630	1170	171	0.39	2.48	14	28	30.7
13	428	0	630	1170	171	0.39	2.48	14	29	29.7
14	444	0	630	1170	155	0.36	2.52	14	30	35.7
15	428	0	630	1170	171	0.39	2.48	14	30	32.8
16	428	0	630	1170	171	0.39	2	8	32	35.7
17	428	0	630	1170	171	0.39	2	10	35	39.5
18	428	0	630	1170	171	0.39	2	14	40	35.6
19	428	0	630	1170	171	0.39	2	12	43	41.58
20	408	0	554	1294	144	0.35	2.51	14	36	36.8
21	408	0	554	1294	144	0.35	2.51	8	56	59.6
22	494	0	691	858	198	0.5	1	14	42.1	43.7
23	416	0	699	927	357	0.8	4.49	15	82.5	78.9
24	416	0	699	927	357	0.85	4.49	15	36.9	41.8
25	420	0	706	936	333	0.79	2.61	15	34.9	39.5
26	412	0	693	918	381	0.72	2.61	15	29.6	34.2
27	409	0	686	909	333	0.81	1.58	15	22.4	28.5
28	409	0	699	920	381	0.81	1.57	15	23.5	26.7
29	210	210	765	1005	380	0.5	2.5	20	56.304	59.6
30	240	160	770	1010	357	0.5	2.5	20	52.201	57.88





## V. CONCLUSIONS

Based on the objectives of the study the following conclusions are drawn:

1. Decision tree model is a reliable computational model to solve different complex problems such as prediction problems.
2. Present study is the application of Decision tree model to predict the 28 days compressive strength of geopolymer concrete based on different concrete characteristics.
3. Decision tree analysis indicates good correlation between the input and output variable.
4. The statistical parameter  $R^2$  is 0.947, and 0.857 for training and testing steps, respectively, which implies good efficiency of the Decision tree model.
5. It is concluded that Decision tree presents good accuracy in predicting the 28 days compressive strength of geopolymer concrete.
6. Therefore, in order to predict the compressive strength of concrete with high reliability, instead of using costly experimental investigation, Decision tree model can be replaced.

### Scope of the project

Further studies should be carried out on various properties like flexural and shear behaviour of GPC

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