

Rameshwaram Institute of Technology and Management



Diabetes Prediction Using Machine Learning

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INTRODUCTION

Diabetes is a disease which is rapidly increasing all over the world. It occurs when pancreas does not produce sufficient insulin, or body can not sufficiently use insulin it produces. Diabetes person has increase blood glucose in the body. One of the major problem diabetic patients suffers from is the Diabetic Retinopathy (DR) and blindness. Since the number of diabetes patients is continuously increasing, it increases the data as well.



RESEARCH OUTCOME

- According to National Diabetes and Diabetic Retinopathy Survey 2019, one in 46 diabetic patients in India becomes visually impaired and one in 7 diabetic patients have vision impairment.
- This survey also states that between 2015 and 2019, around 16.9% of the Indian population suffered from DR.
- Prevalence of blindness among diabetic patient was 2.1% and visual impairment was 13.7%.

PROBLEM SPECIFICATION

- The most usual and conventional method for diagnosis and detection of diabetic retinopathy is by using human fundus images or retinal images.
- In our study, we focus on prediction of DR using health records of the diabetic patients.
- By using machine learning techniques, knowledge is acquired through these records, containing numerical values, to predict whether the patient is having DR or not.
- For this prediction of DR different classification algorithms (Support Vector Machine, K nearest neighbor, bagged trees, Logistic Regression) have been used.

CAUSES FOR DIABETES

What causes type 1 diabetes?

 Type 1 diabetes occurs when your immune system, the body's system for fighting infection, attacks and destroys the insulin-producing beta cells of the pancreas. Scientists think type 1 diabetes is caused by genes and environmental factors, such as viruses, that might trigger the disease. Studies such as Trial Net External link are working to pinpoint causes of type 1 diabetes and possible ways to prevent or slow the disease.

WHAT CAUSES TYPE 2 DIABETES?

- Type 2 diabetes—the most common form of diabetes—is caused by several factors, including lifestyle factors and genes.
- Overweight, obesity, and physical inactivity
- Insulin resistance
- Genes and family history

EFFECTS OF DIABETES

- Diabetes can be effectively managed when diagnosed early. However, when left untreated, it can lead to potential complications that include:
- heart disease
- stroke
- kidney damage
- nerve damage

DATASET

100	A	8	С	D	Ε	E	G	н	
1	Pregnancie	Glucose	BloodPres	SkinThickn	Insulin	BMI	DiabetesPe	Age	Outcome
2	6	148	72	35	.0	33.6	0,627	50	1
3	1	85	66	29	0	26.6	0.351	31	0
4	8	183	64	0	0	23.3	0.672	32	1
5	1	89	66	23	94	28.1	0,167	21	0
6	0	137	40	35	168	43.1	2.288	33	1
7	5	116	74	. 0	. 0	25.6	0.201	30	0
8	3	78	50	32	88	31	0.248	26	1
9	10	115	- 0	0	0	35.3	0.134	29	0
10	2	197	70	45	543	30.5	0.158	53	1
11	8	125	96	0	0	0	0.232	54	1
12	- 4	110	92	0	0	37.6	0.191	30	0
13	10	168	74	0	0	38	0.537	34	1
14	10	139	80	0	0	27.1	1.441	57	0
15	1	189	60	23	846	30.1	0,398	59	1
16	5	166	72	19	175	25.8	0.587	51	1
17	7	100	0	0	0	30	0.484	32	1
18	0	118	84	47	230	45.8	0.551	31	1
19	7	107	74	0	0	29.6	0.254	31	1
20	1	103	30	38	83	43.3	0.183	33	0
21	1	115	70	30	96	34.6	0,529	32	1
22	3	126	88	41	235	39.3	0.704	27	0
23	8	99	84	0	. 0	35.4	0,388	50	0
24	7	196	90	0	0	39.8	0.451	41	1
25	9	119	80	3.5	0	29	0.263	29	1
26	11	143	94	33	146	36.6	0.254	51	1
27	10	125	70	26	115	31.1	0.205	41	1

A 7	8	C	D		40	(ame)	4.4	
- 0.4	4.47		-	E	F	G	H	
40	147	76	0	0	39.4	0,257	43	1
1	97	66	15	140	23.2	0.487	22	0
13	145	82	19	110	22.2	0.245	57	0
5	117	92	0	0	34.1	0.337	38	0
5	109	75	26	0	36	0.546	60	0
3	158	76	36	245	31.6	0.851	28	1
3	88	58	11	54	24.8	0.267	22	0
6	92	92	0	0	19.9	0.188	28	0
10	122	78	31	0	27.6	0.512	45	0
-4	103	60	33	192	24	0.966	33	0
11	138	76	0	0	33.2	0.42	35	0
9	102	76	37	0	32.9	0.665	46	1
2	90	68	42	0	38,2	0.503	27	1
-4	111	72	47	207	37.1	1.39	56	1
3	180	64	25	70	34	0.271	26	0
7	133	84	0	0	40.2	0.696	37	0
7	106	92	18	0	22.7	0.235	48	0
9	171	110	24	240	45.4	0.721	54	1
7	159	64	0	0	27.4	0.294	40	0
0	180	66	39	0	42	1.893	25	1
1	146	56	0	0	29.7	0.564	29	0
2	71	70	27	0	28	0,586	22	.0
7	103	66	32	0	39.1	0.344	31	1
7	105	0	0	0	0	0.305	24	0
1	103	80	11	82	19.4	0.491	22	0
1	101	50	15	36	24.2	0.526	26	0
5	88	66	21	23	24.4	0.342	30	0
	5 3 6 10 4 11 9 2 4 3 7 7 7 9 7 7 7	5 109 3 158 3 88 6 92 10 122 4 103 11 138 9 102 2 90 4 111 3 180 7 133 7 106 9 171 7 159 0 180 1 146 2 71 7 103 7 105 1 105 1 103 1 101	5 109 75 3 158 76 3 88 58 6 92 92 10 122 78 4 103 60 11 138 76 2 90 68 4 111 72 3 180 64 7 133 84 7 106 92 9 171 110 7 159 64 0 180 66 1 146 56 2 71 70 7 103 66 7 105 0 1 103 80 1 101 50	5 109 75 26 3 158 76 36 3 88 58 11 6 92 92 0 10 122 78 31 4 103 60 33 11 138 76 0 9 102 76 37 2 90 68 42 4 111 72 47 3 180 64 25 7 133 84 0 7 106 92 18 9 171 110 24 7 159 64 0 0 180 66 39 1 146 56 0 2 71 70 27 7 103 66 32 7 105 0 0 1 103 80 11 1 103 80 11	5 109 75 26 0 3 158 76 36 245 3 88 58 11 54 6 92 92 0 0 10 122 78 31 0 4 103 60 33 192 11 138 76 0 0 9 102 76 37 0 2 90 68 42 0 4 111 72 47 207 3 180 64 25 70 7 133 84 0 0 9 171 110 24 240 9 171 110 24 240 9 171 110 24 240 1 146 56 0 0 1 146 56 0 0 2	5 109 75 26 0 36 3 158 76 36 245 31.6 3 88 58 11 54 24.8 6 92 92 0 0 19.9 10 122 78 31 0 27.6 4 103 60 33 192 24 11 138 76 0 0 33.2 9 102 76 37 0 32.9 2 90 68 42 0 38.2 4 111 72 47 207 37.1 3 180 64 25 70 34 7 133 84 0 0 40.2 7 106 92 18 0 22.7 9 171 110 24 240 45.4 7 159 64 0<	5 109 75 26 0 36 0.546 3 158 76 36 245 31.6 0.851 3 88 58 11 54 24.8 0.267 6 92 92 0 0 19.9 0.188 10 122 78 31 0 27.6 0.512 4 103 60 33 192 24 0.966 11 138 76 0 0 33.2 0.42 9 102 76 37 0 32.9 0.665 2 90 68 42 0 38.2 0.503 4 111 72 47 207 37.1 1.39 3 180 64 25 70 34 0.271 7 133 84 0 0 40.2 0.696 7 106 92 18	5 109 75 26 0 36 0.546 60 3 158 76 36 245 31.6 0.851 28 3 88 58 11 54 24.8 0.267 22 6 92 92 0 0 19.9 0.188 28 10 122 78 31 0 27.6 0.512 45 4 103 60 33 192 24 0.966 33 11 138 76 0 0 33.2 0.42 35 9 102 76 37 0 32.9 0.665 46 2 90 68 42 0 38.2 0.503 27 4 111 72 47 207 37.1 1.39 56 3 180 64 25 70 34 0.271 26 7

CODE AND OUTPUT:-

In [6]: df.describe()

Out[6]:

	Pregnancies	Glucese	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age	Outcome
count	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000
mean	3:845052	120.094531	69.105469	20 536450	79.799479	31.992578	0.471076	33 240005	0.348958
std	3.369578	31.972618	19.355807	15.952218	115,244002	7.894160	0.331329	11.760232	0.476951
min	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.078000	21 000000	0.000000
25%	1.000000	99.000000	82 000000	0.000000	0.000000	27.300000	0.243750	24.000000	0.000000
50%	3.000000	117.000000	72.000000	23 000000	30.500000	32.000000	0.372500	29.000000	0.000000
75%	6.000000	140.250000	80.000000	32.000000	127.250000	36 600000	0.626250	41.000000	1.000000
max	17.000000	199.000000	122.000000	99.000000	846.000000	67.100000	2.420000	91.000000	1.000000

```
In [7]: df.shape
Out[7]: (768, 9)
```

In [8]: df['Outcome'].value_counts()

Out[8]: 0 500 1 268

Name: Outcome, dtype: int64

In [9]: df.groupby('Outcome').mean()

```
In [9]: df.groupby('Outcome').mean()
:[0]tu0
                    Pregnancies
                                  Glucose BloodPressure SkinThickness
                                                                          Insulin
                                                                                          DiabetesPedigreeFunction
                                                                                                                       Age
           Outcome
                 0
                       3 298000 109 980000
                                               68.184000
                                                             19.664000
                                                                       68 792000 30 304200
                                                                                                         0.429734 31.190000
                                                                                                         0.550500 37.067164
                       4.965672 141.257463
                                               70.824627
                                                             22.164179 100.335821 35.142537
(n [10]: x=df,drop(columns='Outcome')
          y=df['Outcome']
In [11]: x #upto here is data preprocessing
out[11]:
               Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DishetesPedigreeFunction Age
                                149
                                              72
                                                            35
                                                                    0 33.6
                                                                                             0.627
            0
                                                                                                    60
                                85
                                              66
                                                            29
                                                                    0 26.6
                                                                                             0.351
                                                                                                    31
            2
                                183
                                              64
                                                             0
                                                                    0 23.3
                                                                                             0.672
                                                                                                    32
                                99
                                              88
                                                                   94 29.1
            3
                                                            23
                                                                                             0.167
                                                                                                    21
                                                            35
                                137
                                              40
                                                                   168 43.1
                                                                                             2.288
                                                                                                    33
            -
           763
                        10
                                101
                                              76
                                                            40
                                                                   180 32.9
                                                                                             0.171
                                                                                                    63
           764
                               122
                                              70
                                                            27
                                                                    0 36.8
                                                                                             0.340
                                                                                                    27
           765
                               121
                                              72
                                                            23
                                                                  112 26.2
                                                                                             0.245
           766
                                126
                                              60
                                                             0
                                                                    0 30.1
                                                                                             0.349
                                                                                                   47
```

31

0 30.4

0.315 23

767

93

70

```
In [13]: #Now we are standardizing The Data helps the machine learning model to get all the column values in range which helps in ma
         scaler=StandardScaler()
         scaler.fit(x) #Fitting The Values of the data
         standard values=scaler.transform(x) #Transforming it
         mPS We Can Still Use scaler.fit transform
In [14]: x=standard values
In [15]: x
0ut[15]: array([[ 0.63994726, 0.84832379, 0.14964075, ..., 0.20401277,
                  0.46849198, 1.4259954 1.
                [-0.84488505, -1.12339636, -0.16054575, ..., -0.68442195,
                 -0.36506078, -0.19067191],
                [ 1.23388019, 1.94372388, -0.26394125, ..., -1.10325546,
                  0.60439732, -0.10558415],
                ....
                [ 0.3429808 , 0.00330087, 0.14964075, ..., -0.73518964,
                 -0.68519336, -0.27575966],
                [-0.84488505, 0.1597866 , -0.47073225, ..., -0.24020459,
                 -0.37110101. 1.170732151.
                [-0.84488505, -0.8730192 , 0.04624525, ..., -0.20212881,
                 -0.47378505, -0.8713739311)
In [16]: #Train Test Split Model Selection
        x train,x test,y train,y test = train test split(x,y,test size=0.3,stratify=y,random state=2)
        twe Have Taken Four Variables x train will train the data whereas x test will test it
         mile don't want our M. model to know the data
        mattributes of train test
        # 1. Giving All The Dataset 2.Giving The Test Size 3.Stratify helps to get exact proportions of diabetes and non diabates
        #4. Random State just use to split the data
        #Stratify: This will be used by the train test split() function to ensure that both the train and test sets have the propo
```

PREVENTION

- Simple lifestyle measures have been shown to be effective in preventing or delaying the onset of type 2 diabetes. To help prevent type 2 diabetes and its complications, people should:
- achieve and maintain a healthy body weight;
- be physically active doing at least 30 minutes of regular, moderate-intensity activity on most days.
 More activity is required for weight control;
- eat a healthy diet, avoiding sugar and saturated fats; and
- avoid tobacco use smoking increases the risk of diabetes and cardiovascular disease.

CONCLUSION

- In this presentation machine learning methods are used for the prediction of Diabetic Retinopathy in patients, using their health records of diabetes.
- These health records were organized in a structured way by eliminating noisy data. With the help of machine learning algorithms, knowledge is extracted from these records in the form of numerical values for the prediction of DR.