

Data analysis of diabetes and chronic kidney disease

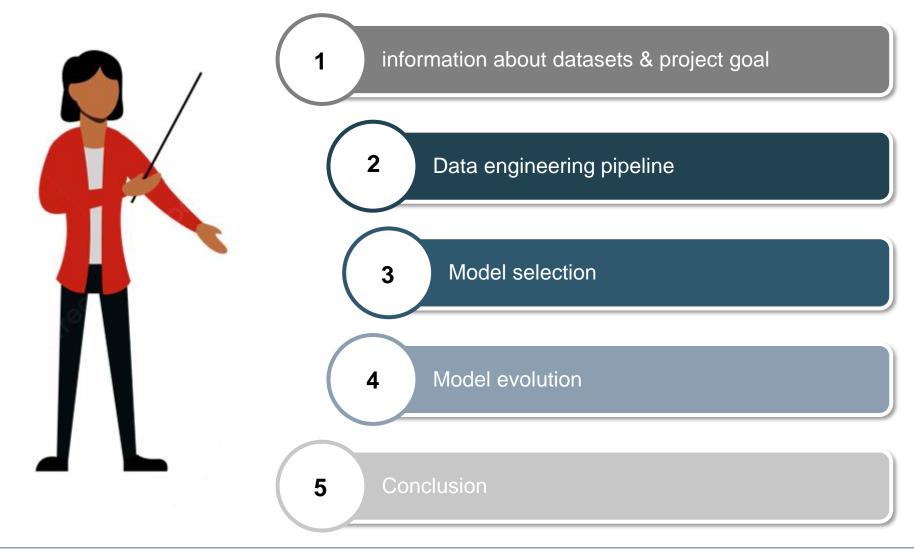
METHODS OF ADVANCED DATA ENGINEERING

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Outline





How can I choose two suitable datasets?



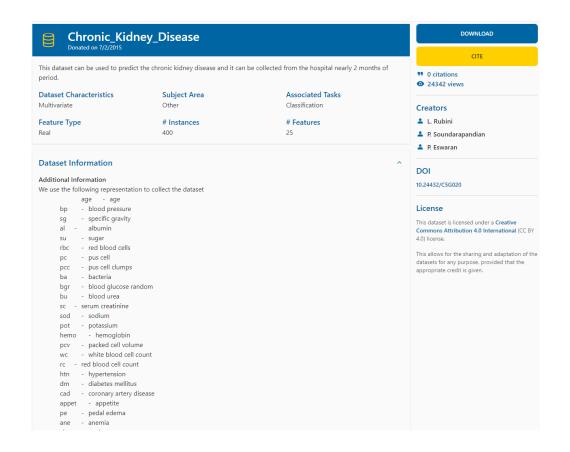
- Open access
- Being interesting to me
- Accessible in common formats
- Possibility of relationship between two datasets



Datasets



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Pima Indians Diabetes Database

Predict the onset of diabetes based on diagnostic measures



Data Card Code (2770) Discussion (49)

About Dataset

Context

This dataset is originally from the National Institute of Diabetes and Digestive and Kidney Diseases. The objective of the dataset is to diagnostically predict whether or not a patient has diabetes, based on certain diagnostic measurements included in the dataset. Several constraints were placed on the selection of these instances from a larger database. In particular, all patients here are females at least 21 years old of Pima Indian heritage.

Content

The datasets consists of several medical predictor variables and one target variable, Outcome. Predictor variables includes the number of pregnancies the patient has had, their BMI, insulin level, age, and so on.

Acknowledgements

Smith, J.W., Everhart, J.E., Dickson, W.C., Knowler, W.C., & Johannes, R.S. (1988). Using the ADAP learning algorithm to forecast the onset of diabetes mellitus. In Proceedings of the Symposium on Computer Applications and Medical Care (pp. 261--265). IEEE Computer Society Press.

Inspiration

Can you build a machine learning model to accurately predict whether or not the patients in the dataset have diabetes or not?

Usability © 8.82

Licence

License CC0: Public Domain

Expected update frequency Not specified

Tags

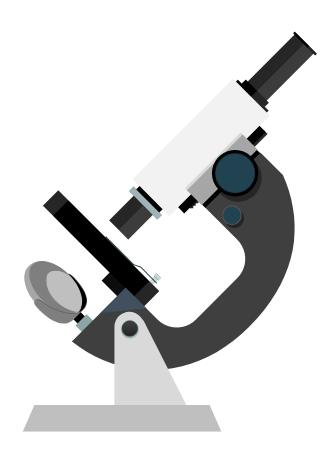
Earth and Nature Health
Diabetes India
Healthcare

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The research question

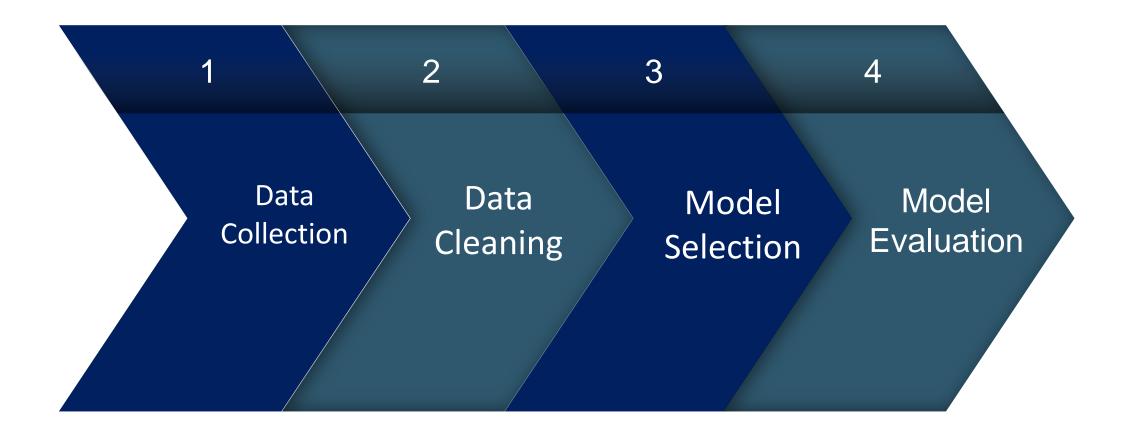


What factors affect diabetes and chronic kidney disease?
And is there a relationship between these diseases?



Methods & Processes





A closer look at the dataset : Data cleaning



| | | | | | | | | | | T. | 1. | | | | | 1. | | | | | | | | _ | | |
|----------------------------|----|-----|----|-----|-------|----|-----------|-----------|-----------|-------------|-----|------|------|-----|------|------|------|----|-------|---------|-----|-----|-------|-----|-----|----------------|
| 1 | id | age | bp | sg | al | su | rbc | pc | pcc | ba | _ | bu s | c so | id | pot | hemo | pcv | W | | htn | dm | cad | appet | pe | ane | classification |
| 2 | | 0 | 48 | 80 | 1.02 | 1 | 0 | normal | notprese | r notpreser | 121 | 36 | 1.2 | | | 1 | 15.4 | 44 | 7800 | 5.2 yes | yes | no | good | no | no | ckd |
| | | 1 | 7 | 50 | 1.02 | 4 | 0 | normal | notprese | r notpreser | nt | 18 | 0.8 | | | 1 | 11.3 | 38 | 6000 | no | no | no | good | no | no | ckd |
| | | 2 | 62 | 80 | 1.01 | 2 | 3 normal | normal | notprese | r notpreser | 423 | 53 | 1.8 | | | | 9.6 | 31 | 7500 | no | yes | no | poor | no | yes | ckd |
| | | 3 | 48 | 70 | 1.005 | 4 | 0 normal | abnorma | l present | notpreser | 117 | 56 | 3.8 | 111 | . 2. | 5 1 | 11.2 | 32 | 6700 | 3.9 yes | no | no | poor | yes | yes | ckd |
| 5 | | 4 | 51 | 80 | 1.01 | 2 | 0 normal | normal | notprese | rnotpreser | 106 | 26 | 1.4 | | | 1 | 11.6 | 35 | 7300 | 4.6 no | no | no | good | no | no | ckd |
| 7 | | 5 | 60 | 90 | 1.015 | 3 | 0 | | notprese | r notpreser | 74 | 25 | 1.1 | 142 | 3. | 2 1 | 12.2 | 39 | 7800 | 4.4 yes | yes | no | good | yes | no | ckd |
| 3 | | 6 | 68 | 70 | 1.01 | 0 | 0 | normal | notprese | rnotpreser | 100 | 54 | 24 | 104 | | 4 1 | 12.4 | 36 | | no | no | no | good | no | no | ckd |
| • | | 7 | 24 | | 1.015 | 2 | 4 normal | abnorma | Inotprese | rnotpreser | 410 | 31 | 1.1 | | | 1 | 12.4 | 44 | 6900 | 5 no | yes | no | good | yes | no | ckd |
| |) | 8 | 52 | 100 | 1.015 | 3 | 0 normal | abnorma | l present | notpreser | 138 | 60 | 1.9 | | | 1 | 10.8 | 33 | 9600 | 4 yes | yes | no | good | no | yes | ckd |
| | 1 | 9 | 53 | 90 | 1.02 | 2 | 0 abnorma | l abnorma | present | notpreser | 70 | 107 | 7.2 | 114 | 3. | 7 | 9.5 | 29 | 12100 | 3.7 yes | yes | no | poor | no | yes | ckd |
| Chronic Kidney Disease.csv | 2 | 10 | 50 | 60 | 1.01 | 2 | 4 | abnorma | l present | notpreser | 490 | 55 | 4 | | | | 9.4 | 28 | | yes | yes | no | good | no | yes | ckd |
| 3 | 3 | 11 | 63 | 70 | 1.01 | 3 | 0 abnorma | l abnorma | present | notpreser | 380 | 60 | 2.7 | 131 | 4. | 2 1 | 10.8 | 32 | 4500 | 3.8 yes | yes | no | poor | yes | no | ckd |
| 4 | 1 | 12 | 68 | 70 | 1.015 | 3 | 1 | normal | present | notpreser | 208 | 72 | 2.1 | 138 | 5. | 8 | 9.7 | 28 | 12200 | 3.4 yes | yes | yes | poor | yes | no | ckd |
| 5 | 5 | 13 | 68 | 70 | | | | | notprese | r notpreser | 98 | 86 | 4.6 | 135 | 3. | 4 | 9.8 | | | ves | yes | yes | poor | ves | no | ckd |
| 6 | 5 | 14 | 68 | 80 | 1.01 | 3 | 2 normal | abnorma | l present | present | 157 | 90 | 4.1 | 130 | 6. | 4 | 5.6 | 16 | 11000 | 2.6 yes | yes | yes | poor | yes | no | ckd |
| 7 | 7 | 15 | 40 | 80 | 1.015 | 3 | 0 | normal | notprese | r notpreser | 76 | 162 | 9.6 | 141 | 4. | 9 | 7.6 | 24 | 3800 | 2.8 yes | no | no | good | no | yes | ckd |
| 8 | 3 | 16 | 47 | 70 | 1.015 | 2 | 0 | normal | notprese | r notpreser | 99 | 46 | 2.2 | 138 | 4. | 1 1 | 12.6 | | | no | no | no | good | no | no | ckd |
| 9 | 9 | 17 | 47 | 80 | | | | | | r notpreser | | 87 | 5.2 | 139 | | 7 1 | 12.1 | | | yes | no | no | poor | no | no | ckd |
| 10 |) | 18 | 60 | 100 | 1.025 | 0 | 3 | normal | | r notpreser | | 27 | 1.3 | 135 | | | 12.7 | 37 | 11400 | 4.3 yes | ves | ves | good | no | no | ckd |
| 1 | | 19 | 62 | 60 | 1.015 | 1 | 0 | | | notpreser | | 31 | 1.6 | | | | 10.3 | 30 | 5300 | 3.7 yes | no | ves | good | no | no | ckd |
| 2 | > | 20 | 61 | 80 | 1.015 | 2 | 0 abnorma | | | rnotpreser | | 148 | 3.9 | 135 | 5. | | 7.7 | 24 | 9200 | 3.2 ves | ves | ves | poor | ves | ves | ckd |

What I did in Pipeline:

- Abbreviation Replacement
- Categorical to Numerical
- Handling Null/Zero Values

| Pregnancies | Glucose | BloodPressure | SkinThickness | Insulin | BMI | DiabetesPedigreeF | Λσο | Outcome | |
|---------------|---------|---------------|---------------|---------|------|-------------------|-----|---------|--|
| Pregnancies 6 | | | | | | | | | |
| 1 | 85 | | | | | | | | |
| 8 | | | | | | | | | |
| - 0 | 183 | | | | | | | | |
| | | | | | | | | | |
| | | | | | | | | | |
| 5 | | | | | | | | | |
| 3 | | | | | | 0.248 | | | |
| 10 | | | | | | | | | |
| 2 | | | | | | | | | |
| 8 | | | | | | | | | |
| 4 | | | | | | | | | |
| 10 | | | | | | | | | |
| 10 | | | | | 2712 | | | | |
| 1 | 189 | | | | | | | | |
| 5 | 166 | | | | | | | | |
| 7 | 100 | | | | | | | | |
| 0 | 118 | 84 | 47 | 230 | 45.8 | 0.551 | 31 | 1 | |
| 7 | 107 | 74 | | | 29.6 | 0.254 | 31 | | |
| 1 | 103 | 30 | 38 | 83 | 43.3 | 0.183 | 33 | 0 | |
| 1 | 115 | 70 | 30 | 96 | 34.6 | 0.529 | 32 | 1 | |
| 3 | 126 | 88 | 41 | 235 | 39.3 | 0.704 | 27 | 0 | |
| 8 | 99 | 84 | 0 | 0 | 35.4 | 0.388 | 50 | 0 | |
| 7 | 196 | 90 | 0 | 0 | 39.8 | 0.451 | 41 | 1 | |
| ^ | *** | | | _ | | 0.000 | | | |

Pima Indians Diabetes.csv

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A closer look at the dataset : Data cleaning



What I did in Pipeline:

Kidney Disease Dataset:

Z-Score Transformation with a threshold of 3

Remove outliers

Diabetes Dataset:

- ❖ Z-Score Transformation for features such as Glucose, Blood Pressure, SkinThickness, BMI with a threshold of 3-5.
- ❖ Interquartile Range (IQR) transformation for features such as Pregnancies, Insulin, DiabetesPedigreeFunction and Age with a threshold of 1.5-3.

Selecting a prediction Model: Which model and why?



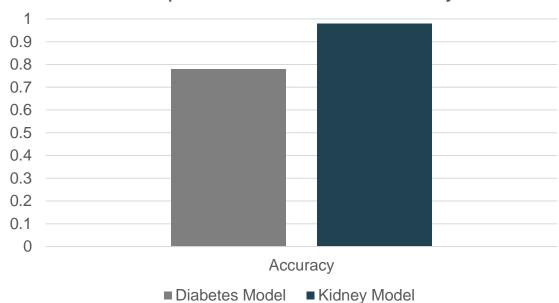


Are the results satisfactory?



Selected Model : Logistic Regression

Comparison of model accuracy

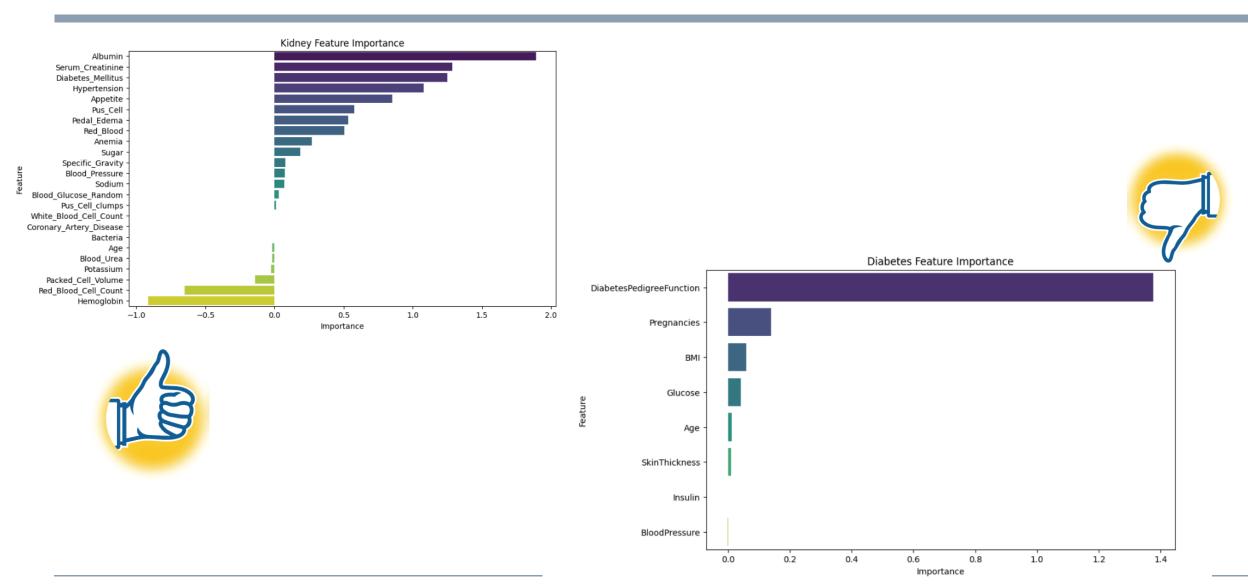


kidney model accuracy: 0.98

Diabetes model accuracy: 0.78

What about the importance of features?





Try another model for the diabetes dataset





Or Neural Network

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New Model for Diabetes dataset

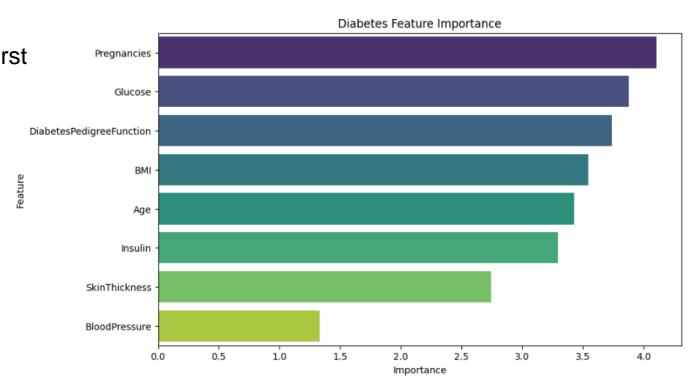


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Diabetes model accuracy: 0.84

Neural Network

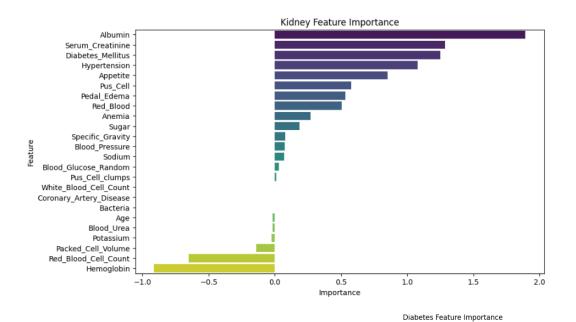
Two hidden layers, the first with 8 neurons and the second with 4 neurons



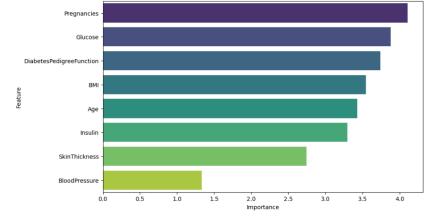
Discussion & Conclusion



| Highlights | Diabetes Model | Kidney Model Logistic Regression | | | | | |
|-----------------------|---|--|--|--|--|--|--|
| Selected Model | Neural Network | | | | | | |
| Accuracy | 0.84 | 0.98 | | | | | |
| Important Features | number of pregnancies* , glucose level | Albumin, Serum_Creatinine and *Diabetes_Mellitus | | | | | |
| Common Features | blood sugar and blood pressure | | | | | | |



- The diabetes dataset only contains information about women.
- The kidney disease dataset don't have the gender feature.



^{*} https://www.ncbi.nlm.nih.gov/pmc/articles/PMC7153959

^{*} https://www.niddk.nih.gov/health-information/diabetes/overview/preventing-problems/diabetic-kidney-disease

^{*} https://www.cdc.gov/diabetes/managing/diabetes-kidney-disease.html

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- https://www.niddk.nih.gov/health-information/diabetes/overview/preventing-problems/diabetic-kidney-disease

https://www.cdc.gov/diabetes/managing/diabetes-kidney-disease.html

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Thank you for your time and attention.