

Diabetes prediction

Partners:

Alon Ben Bassat – alonb4040@gmail.com
Liran Kesler – liran249@gmail.com
Haguy Alon – haguy838@gmail.com

Content

Goals

Domain info – What is it diabetes mellitus?

Dataset info

Challenges we had to deal with

Cleaning data and removing outliers

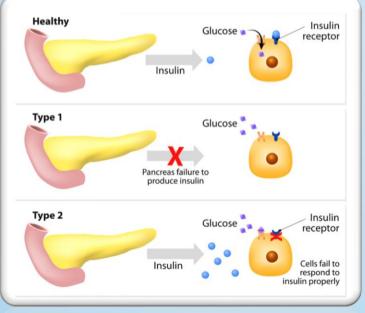
Feature engineering

Models

Conclusions



What is it diabetes mellitus?



Dataset info

- In this project we are going to use a dataset from Kaggle: "WiDS Datathon 2021"
- The dataset size: (130157,180) each row represents a patient.

Following are few of the most important columns for our analysis:

| | age | bmi | ethnicity | gender | icu_type | apache_2_diagnosis | d1_bun_max | d1_bun_min | d1_glucose_max | d1_glucose_min | d1_potassium_max | diabetes_mellitus |
|---|------|-----------|---------------|--------|--------------|--------------------|------------|------------|----------------|----------------|------------------|-------------------|
| 0 | 68.0 | 22.732803 | Caucasian | М | CTICU | 4.0 | 31.0 | 30.0 | 168.0 | 109.0 | 4.0 | 1 |
| 1 | 77.0 | 27.421875 | Caucasian | F | Med-Surg ICU | 5.0 | 11.0 | 9.0 | 145.0 | 128.0 | 4.2 | 1 |
| 5 | 67.0 | 27.555611 | Caucasian | M | Med-Surg ICU | 4.0 | 13.0 | 13.0 | 156.0 | 125.0 | 3.9 | 1 |
| 6 | 59.0 | 57.451002 | Caucasian | F | Med-Surg ICU | 5.0 | 18.0 | 11.0 | 197.0 | 129.0 | 5.0 | 1 |
| 9 | 50.0 | 25.707702 | Other/Unknown | M | CCU-CTICU | 4.0 | 10.0 | 10.0 | 134.0 | 134.0 | 4.1 | 0 |

Challenges we had to deal with:

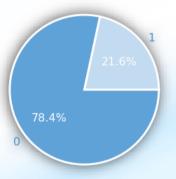
Domain understanding



Mistakes

| weight | height | bmi | manual_cal_bmi |
|--------|--------|----------|----------------|
| 175.00 | 137.2 | 67.81499 | 92.967216 |
| 186.00 | 193.0 | 67.81499 | 49.934226 |
| 130.90 | 137.2 | 67.81499 | 69.539478 |
| 99.79 | 137.2 | 67.81499 | 53.012563 |
| 186.00 | 158.0 | 67.81499 | 74.507290 |
| | | | |

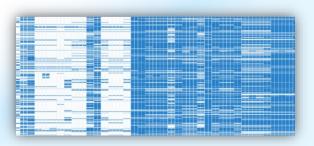
Imbalanced data



Unlabeled test data



Too many N\A values



Cleaning data, removing outliers and fixing mistakes:

Removed:

- Age = 0
- Threshold of dropping columns 20% N\A values.
- Columns with no \ negative effect on our models.
- Rows with N\A values.

Fixed:

- Creating 'BMI' column.
- · Exchanging min\max values.
- Imputed N\A values in the ethnicity with "unknown" values.

Note:

After these steps we've checked that we keep about the same proportion of imbalanced data.

Feature engineering

We've tried many calculated columns, some were designed for specific models. Following are the final calculated columns that we used in our models:

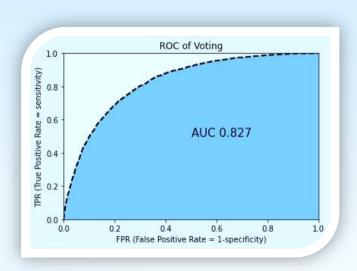
- Grouping and Mapping based on similar diabetes proportion (APACHE_2_diagnosis, ICU_type, Ethnicity)
- · PCA on max measures of blood test.
- Select K-best on measures of blood test

| Weight | Feature |
|---------------------|--------------------|
| 0.0452 ± 0.0005 | d1_max_pca |
| 0.0089 ± 0.0020 | apache_2_diagnosis |
| 0.0065 ± 0.0012 | bmi |
| 0.0025 ± 0.0020 | age |
| 0.0020 ± 0.0012 | d1 glucose min |
| 0.0017 ± 0.0008 | d1 bun min |
| 0.0005 ± 0.0005 | gender M |
| 0.0004 ± 0.0010 | icu type |
| 0.0003 ± 0.0007 | ethnicity |
| 0.0000 ± 0.0006 | gender_F |



Models:

| Model | Hyper_parameters | AUC |
|---------------------|---|------|
| Random forest | {criterion='entropy', max_depth=10, n_estimators=50} | 0.82 |
| Xgboost | {max_depth=10, learning_rate=0.01, n_estimators=200} | 0.82 |
| Lgboost | {max_depth=10, learning_rate=0.01, n_estimators=200} | 0.83 |
| Adaboost | {learning_rate=0.015, n_estimators=250} | 0.82 |
| Logistic regression | {C=100} | 0.81 |
| Voting - few models | RND,XGB,LGBM | 0.83 |





The end