



FINAL PROJECT ISE 543, SPRING 2023

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INSTRUCTIONS



For this exam, you will use Azure ML Studio Designer to build a classification model to predict the likelihood of a patient developing Chronic Heart Disease (CHD) in the coming ten years. The dataset you will be using has been distributed with this exam and consists of the variables on the following page



DATA DICTIONARY



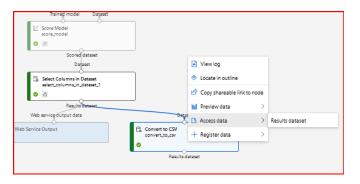
Variable	Description						
Age	age of the participant at the time of examination						
Male	gender of the participant (male = I, female = 0)						
Education	Educational level of the patient (I = less than high school, 2 = completed high school or equivalent, 3 = sor college, 4= completed college or higher)						
Income	Income of the patient						
Current Smoker	whether the participant is currently a smoker (yes or no)						
Cigarettes per Day	the average number of cigarettes smoked per day by current smokers						
BP Meds	whether the participant is taking blood pressure medication (yes or no)						
Prevalent Stroke	whether the participant has a history of stroke (yes or no)						
Prevalent Hyp	whether the participant has a history of hypertension (yes or no)						
Diabetes	whether the participant has diabetes (yes or no)						
Total Chol	total cholesterol level in milligrams per deciliter						
Sys BP	systolic blood pressure in millimeters of mercury						
Dia BP	diastolic blood pressure in millimeters of mercury						
BMI	body mass index in kilograms per square meter						
Heart Rate	resting heart rate in beats per minute						
Glucose	Blood glucose level in milligrams per deciliter						
Alc	Hemoglobin A1c (%)						
Ten Year CHD	whether the participant developed coronary heart disease (CHD) within 10 years of the examination (yes or no)						



NOTE ON MODEL DEPLOYMENT - UPDATE



- When complete, do a final run of your inference pipeline by copying test data from the file "Final Project Evaluation Dataset.csv" into the "Enter Data Manually" component (see Sample Final Report for an example)
 - » Do this by opening the file in WordPad or a similar text editor and then copying the data from there
- > Include a "Convert to CSV" component at the end of your inference pipeline (see Sample Final Report for an example)
- Download the CSV file containing your labels (predictions) by right-clicking the Convert to CSV component, selecting Access data / Results Dataset





FINAL REPORT STRUCTURE



Please follow the provided template/example and structure your final report into the following three sections:

- > Exploratory Data Analysis
- Model Development
- Model Deployment



FINAL REPORT OUTLINE/GRADING RUBRIC



Report contents: 80%

- Attribute summary
- Data cleansing summary of decisions made
- Data cleansing pipeline (portion of your overall pipeline)
- > Univariate analysis
- > Bivariate analysis (each variable vs the response variable)
- Feature section/engineering decisions
- Model pipeline screenshot
- Model evaluation results screenshot
- > Inference pipeline screenshot
- REST Endpoint URL and authentication key (in PPT and in Google drive spreadsheet)
- Screenshot of scored test dataset

Model performance: 20%

Based on TAs calling your endpoint with test data



ATTRIBUTE SUMMARY



Response Variable

> Ten Year CHD

Categories

- PatientID (UID)
- Male
- > Education
- > Current Smoker
- > BP Meds
- > Prevalent Stroke
- > Pravalent Hyp
- > Diabetes

Measures

- Age
- > Income
- Cigarettes per Day
- > Total Chol
- > Sys BP
- > Dia BP
- > BMI
- > Heart Rate
- Glucose
- > A1c



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DATA CLEANSING



Rows ③ Columns ② 19 23

Feature	Count	Unique Value Count	Missing Value Count	Min	Max	Mean	Mean Deviation	1st quantile	Median	3rd quantile	Mode	Range	Sample Variance	5
]a		l	<u></u>						{100002, 100012, 100130, 100214, 100754, 101056,			
atientlD	3816	3816	0	100002	999826	554019.06499	222606.258181	336251.75	555421.5	772812.25	101294, 101524, 102052, 102439, 102712, 102873, 102935, 103840, 103955, 103992, 104276, 104985, 104993, 105054,	899824	66073719998.68935	ź
male	3816	2	0	0	1	0.427673	0.489538	0	0	1	0	1	0.244833	(
age	3816	39	0	32	70	49.567348	7.350087	42	49	56	40	38	73.918923	8
education	3723	4	93	an ^{1,0} missing val	4.0	1.974483	0.812462	1	2	3	1.0	3.0	1.037769	1
currentSmoker	3816	2	0	iny missing vai	ues 1	0.489518	0.49978	0	0	1	0	1	0.249956	(
cigsPerDay	1841	31	1975	1.0	70.0	18.500272	8.12733	10	20	20	20.0	69.0	119.365353	1
BPMeds	3771	2	45	0.0	1.0	0.02917	0.056638	0	0	0	0.0	1.0	0.028327	(
prevalentStroke	3816	2	0	0	1	0.006027	0.011982	0	0	0	0	1	0.005992	(
prevalent Hyp	3816	2	0	0	large outli	0.306604	0.425196	0	0	1	0	1	0.212654	(
diabetes	3816	2	0	0	1 large outil	0.024895	0.048551	0	0	0	0	1	0.024282	(
totChol	3769	246	47	107.0	9280.0	240.852746	39.378735	205	234	263	240.0	9173.0	35695.202567	1
sysBP	3816	232	0	83.5	295.0	132.260089	16.890304	117	128	143.5	130.0	211.5	489.279059	2
diaBP	3816	142	0	50.0	142.5	82.874214	9.158874	75	82	89.5	80.0	92.5	141.791251	1
BMI	3797	1319	19	15.54	56.8	25.814791	3.113207	23.07	25.4	28.04	22.91	41.26	16.807305	4
heartRate	3815	73	1	44.0	143.0	75.775098	9.288791	68	75	82	75.0	99.0	144.885431	1
glucose	3455	134	361	40.0	394.0	81.856151	12.250119	71	78	87	75.0	354.0	555.598004	2
TenYearCHD	3816	2	0	0	1	0.15173	0.257415	0	0	0	0	1	0.128741	(
a1c	3455	3455	361	2.134768768113766	19.917371285750395	4.296312	0.631685	3.738947	4.126325	4.564732	(2.134768768113766, 2.3023530613653507, 2.349277968047848, 2.355203600535006, 2.3861003033933663, 2.38903897942708, 2.4155486171620075, 2.437498363542623, 2.4946768795983005, 2.514403077446826,	17.78260251763663	1.424309	1
income	3816	3282	0	12000.0	524494.0	20355.886792	7743.282613	13562.5	16055	21395.5	14623.0	512494.0	319178936.087836	1



DATA CLEANSING

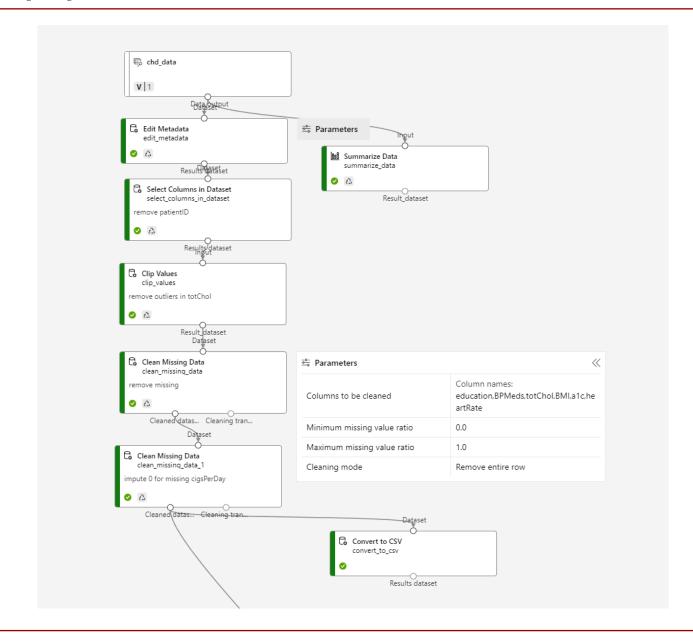


- > change response variable TenYearCHD to categorical
- outliers in totChol: seem to be erroneous data. clip outliers in totChol at a threshold of 600
- cigarettes per day: impute missing values with 0 because these nonresponses are mostly due to non current smokers
- remove rows with missing values: education, BPMeds, totChol, BMI, heartRate, glucose, a1c
- > (Note: normalization and transformation are done in feature engineering)



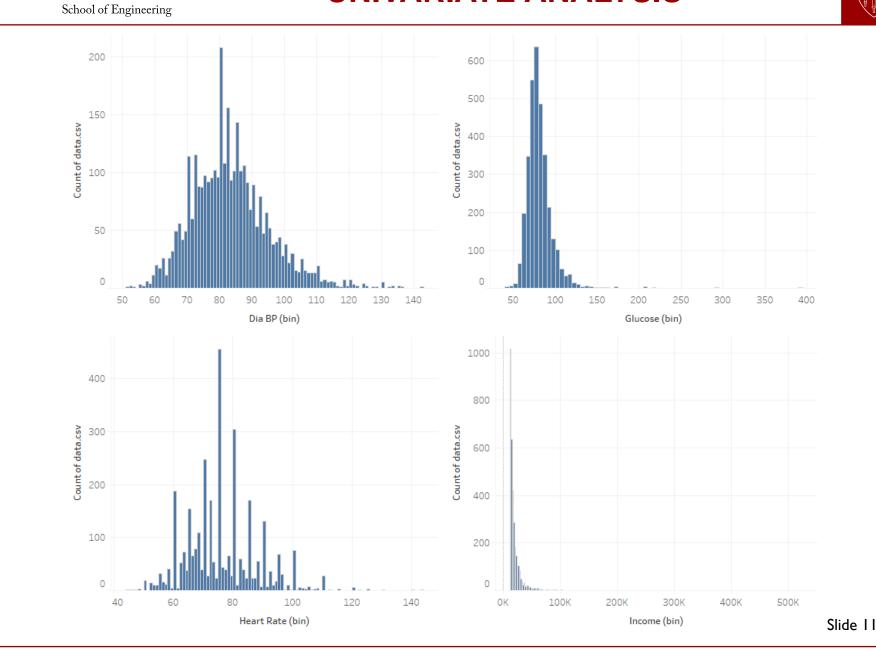
DATA CLEANSING PIPELINE





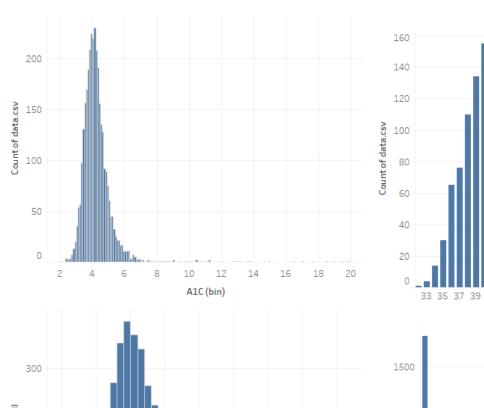
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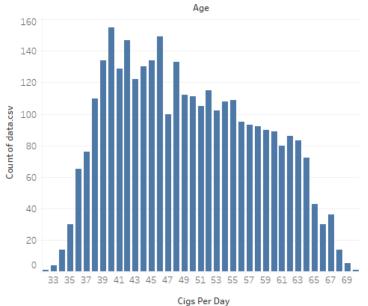


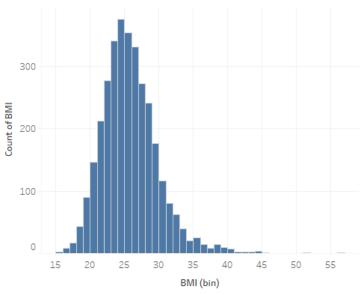


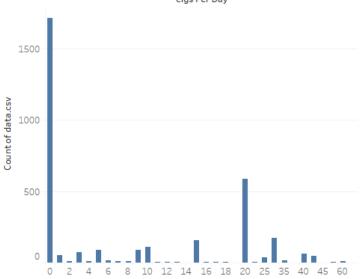
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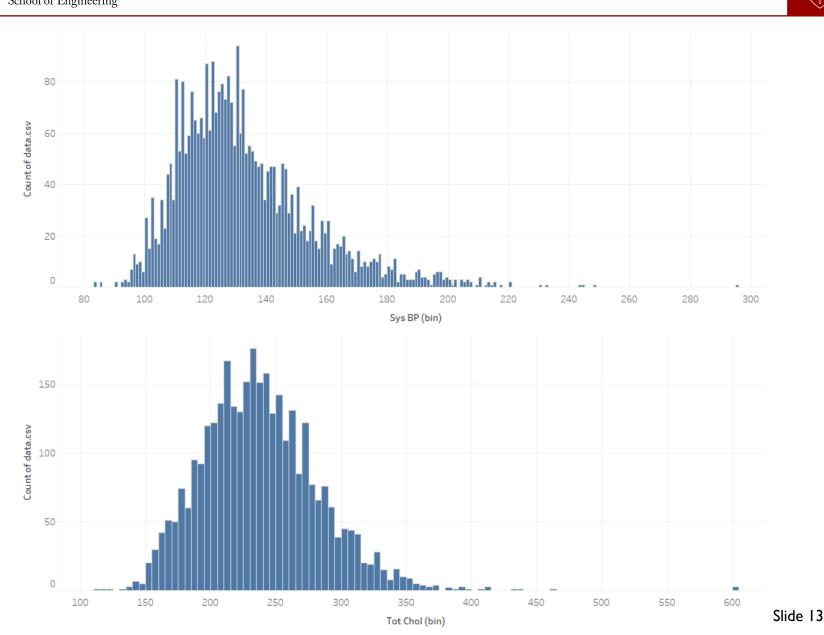






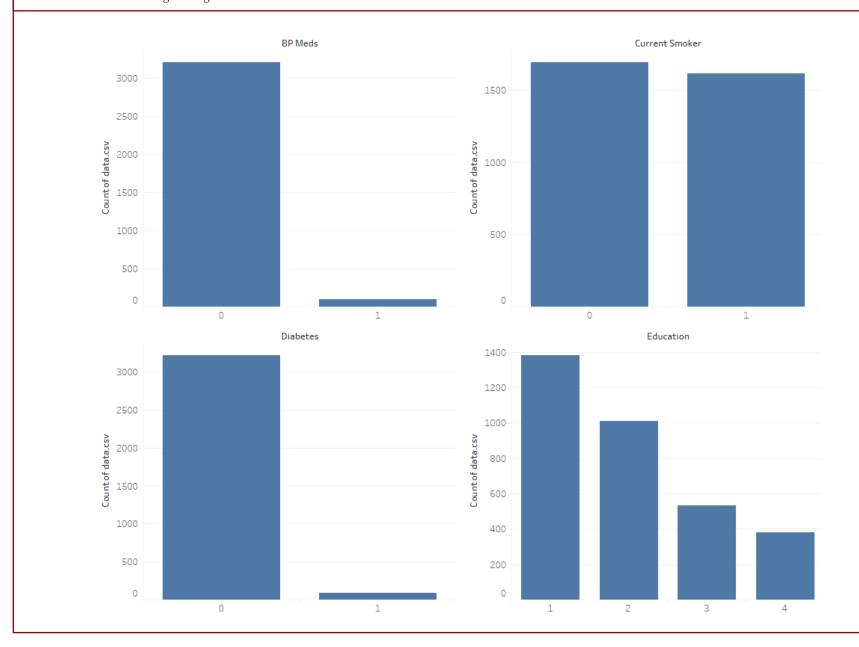






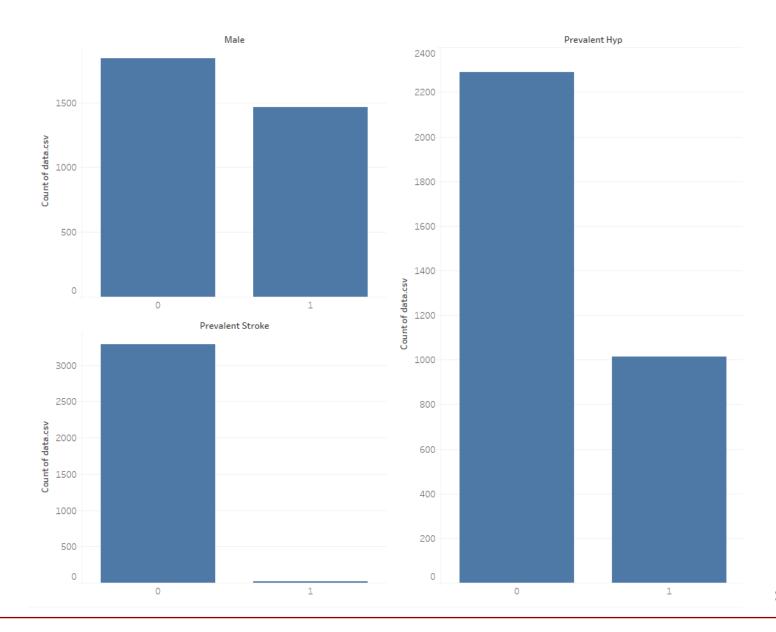






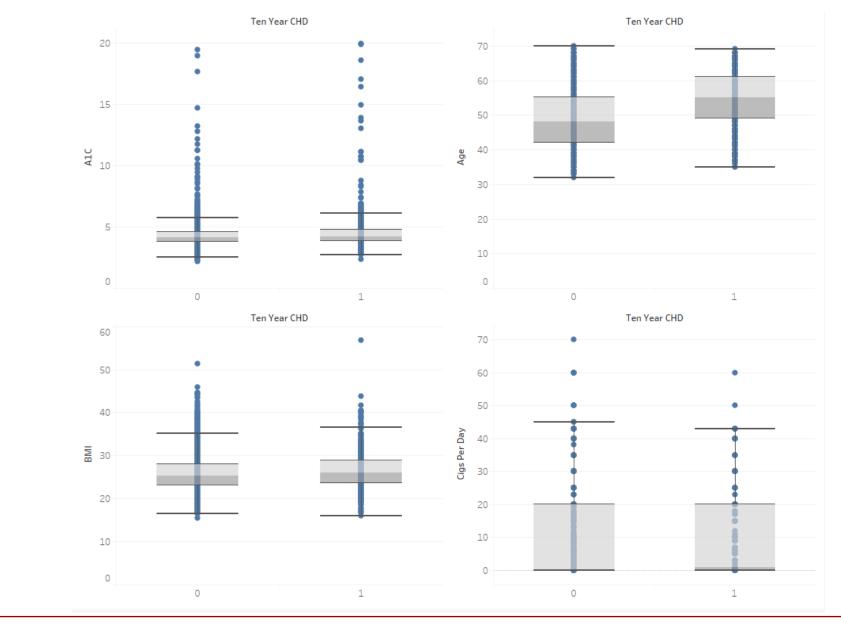






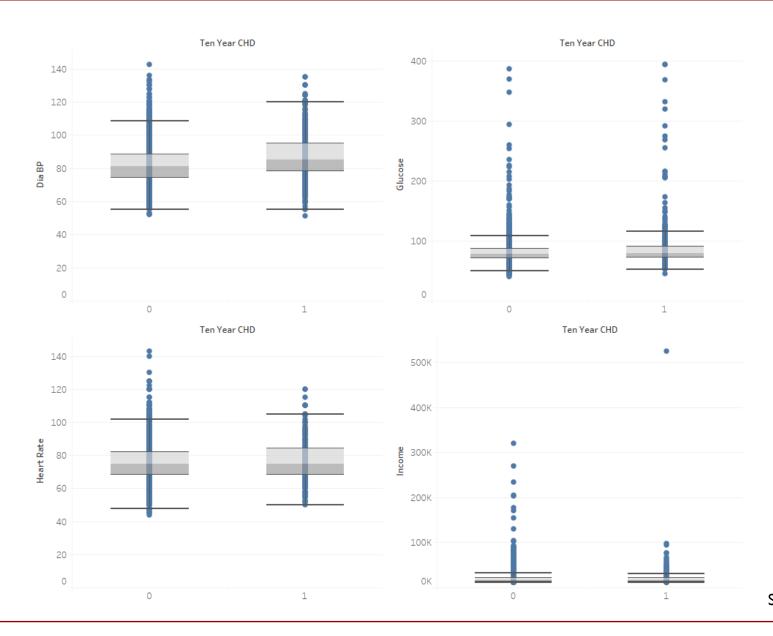






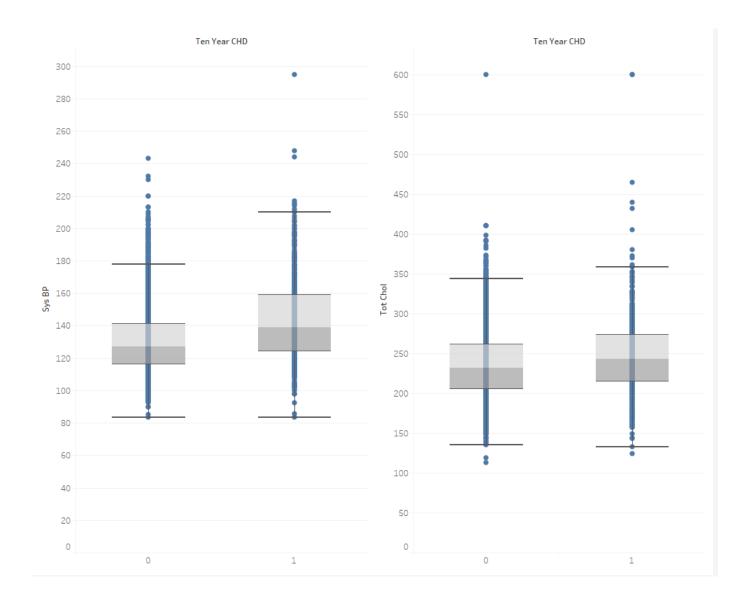






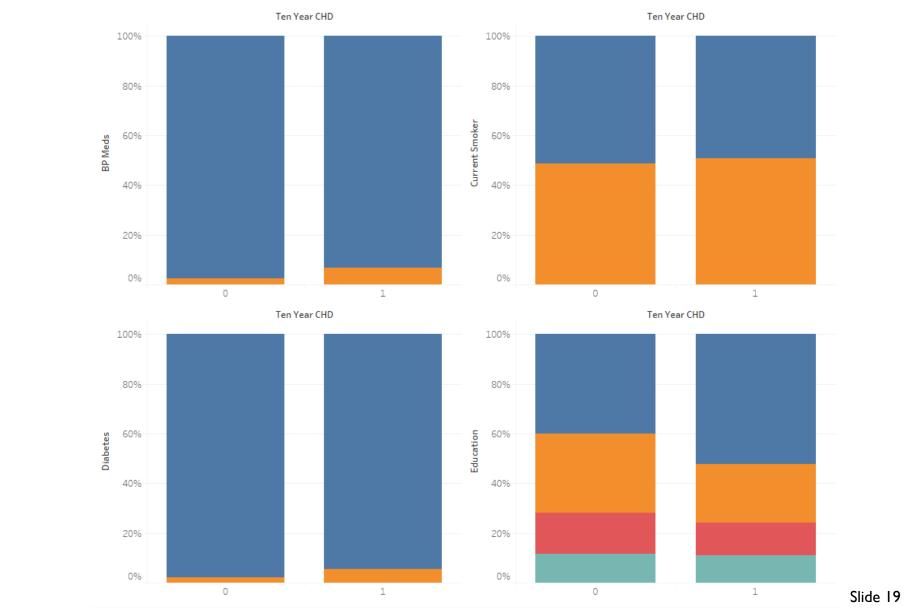






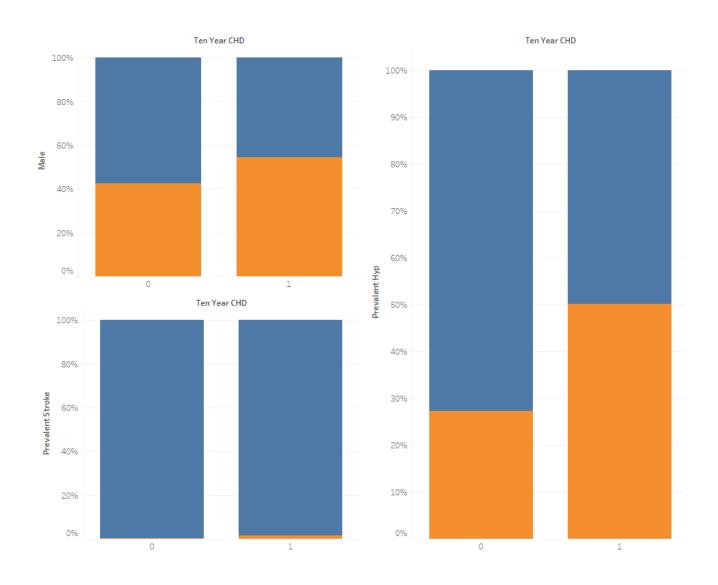
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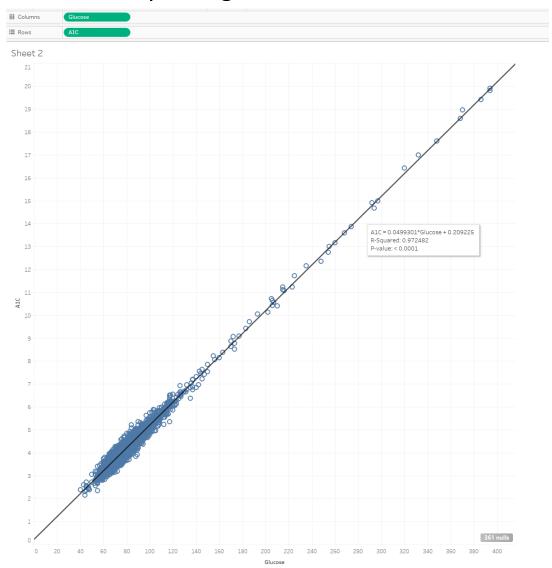




CORRELATION ANALYSIS



> glucose vs. A1C: very strong correlation





CORRELATION ANALYSIS



> Dia BP vs. Sys BP: strong correlation





FEATURE SELECTION/ENGINEERING

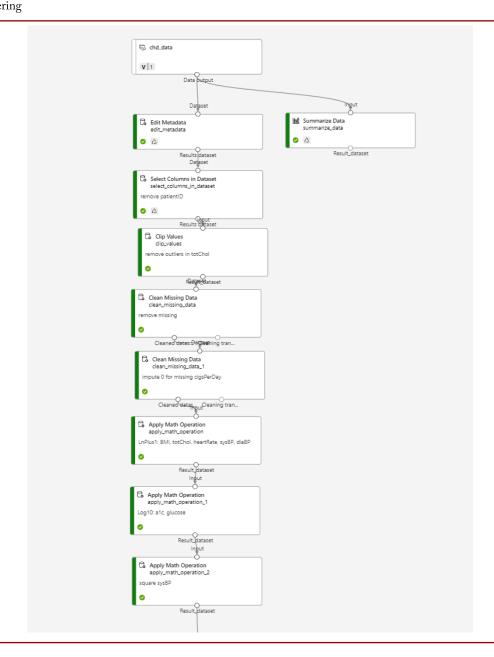


- > Features removed
 - » patient ID irrelevant for prediction
 - » highly correlated variables: despite the high correlation between two pairs of predictors (glucose vs. a1c, dia bp vs. sys bp), I decide not to remove those two predictors because doing so will significantly decrease the model performance
- Feature engineering
 - » create square of sysBP as a new predictor
 - » normalize income
 - » log transformation:
 - » very strong skewness (Log10): glucose, a1c
 - » moderate skewness (LnPlus1): diaBP, sysBP, BMI, heartRate



MODEL PIPELINE

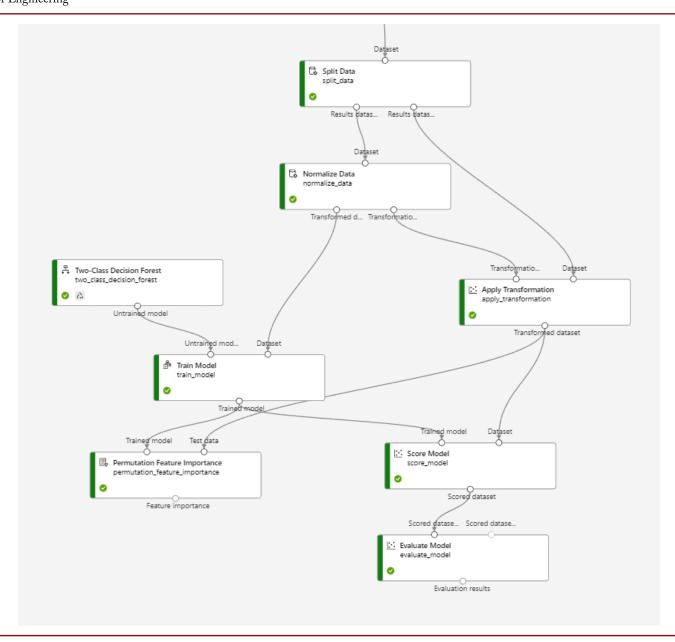






MODEL PIPELINE



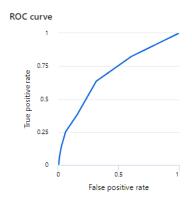


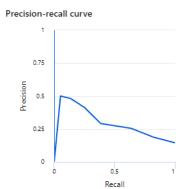
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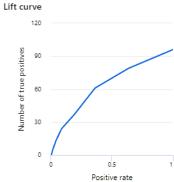
MODEL EVALUATION RESULTS

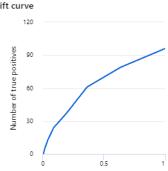












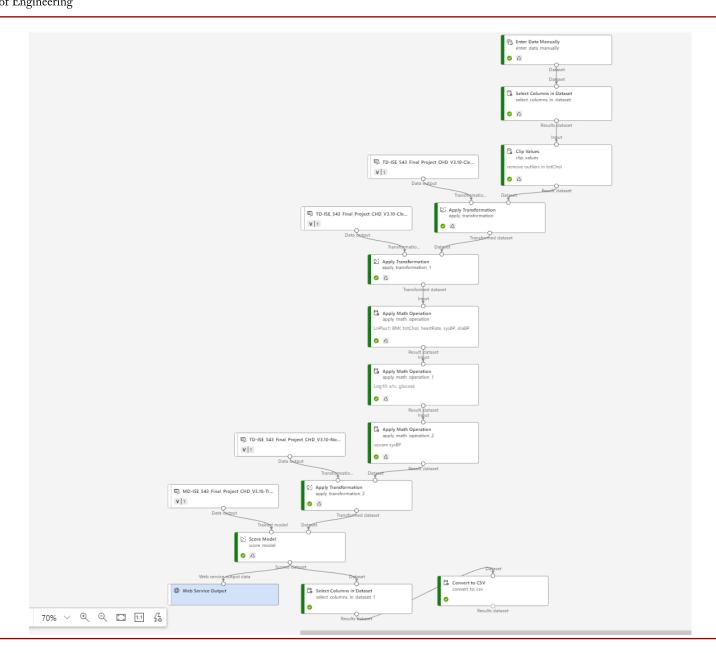






INFERENCE PIPELINE

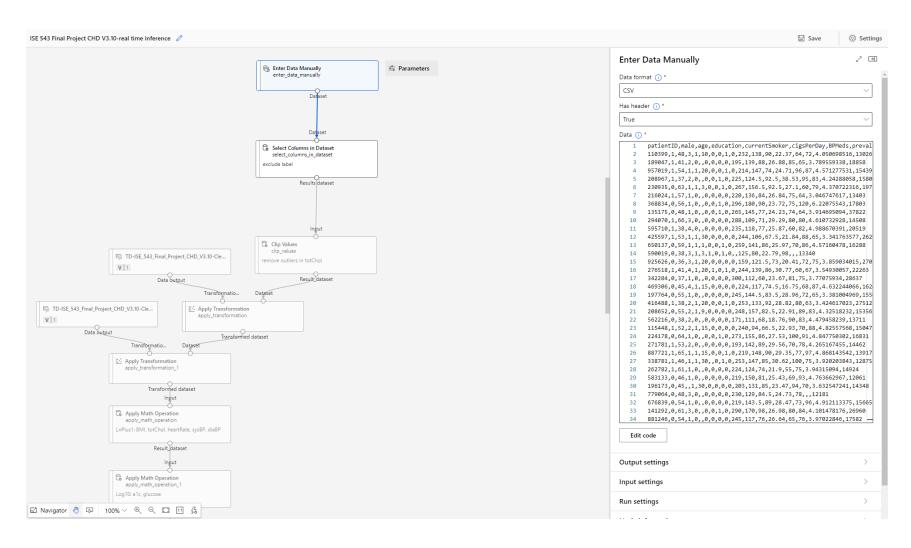






MANUAL TEST DATA







TEST DATA SCORING



