CS340 Commputational Ethics Assignment 2

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1. Identify Potential Bias

In the dataset diabetic_preprocessed.csv, the data type of each columns are as follows:

```
race
                             object
gender
                             object
                             object
age
discharge_disposition_id
                             object
admission_source_id
                             object
time_in_hospital
                             int64
medical_specialty
                             object
num_lab_procedures
                              int64
                              int64
num_procedures
num_medications
                              int64
primary_diagnosis
                             object
number_diagnoses
                              int64
max_glu_serum
                             object
A1Cresult
                             object
insulin
                             object
change
                             object
diabetesMed
                             object
medicare
                               bool
medicaid
                               bool
had_emergency
                               bool
had_inpatient_days
                               bool
had_outpatient_days
                               bool
readmitted
                             object
readmit_binary
                              int64
readmit_30_days
                              int64
```

Check whether data are binary, multi-categorical, or continuous

```
race
Caucasian
                    76099
AfricanAmerican
                    19210
Unknown
                     2273
Hispanic
                     2037
0ther
                     1506
Asian
                      641
gender
Female
                    54708
```

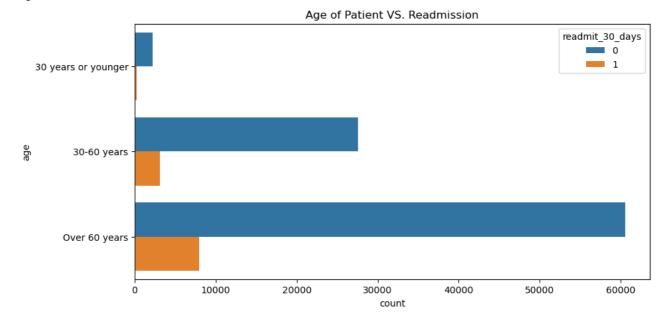
Male 47055 Unknown/Invalid 3	
age Over 60 years 68541 30-60 years 30716 30 years or younger 2509	
discharge_disposition_id Discharged to Home 60234 Other 41532	
admission_source_id Emergency 57494 Referral 30856 Other 13416	
time_in_hospital 3 17756 2 17224 1 14208 4 13924 5 9966 6 7539 7 5859 8 4391 9 3002 10 2342 11 1855 12 1448 13 1210 14 1042	
medical_specialty Missing 49949 Other 16825 InternalMedicine 14635 Emergency/Trauma 7565 Family/GeneralPractice 7440 Cardiology 5352	
num_lab_procedures 1	

```
120
          1
132
          1
121
          1
126
          1
          1
118
num_procedures
     46652
1
     20742
2
     12717
3
     9443
6
     4954
4
     4180
5
      3078
num_medications
13
      6086
12
      6004
11
      5795
15
      5792
      5707
14
      . . .
70
         2
75
         2
81
         1
79
         1
74
         1
primary_diagnosis
0ther
                           68512
Respiratory Issues
                           14423
Diabetes
                           8757
Genitourinary Issues
                           5117
Musculoskeletal Issues
                           4957
number_diagnoses
9
      49474
5
      11393
8
      10616
7
      10393
6
      10161
4
      5537
3
       2835
2
       1023
1
        219
16
         45
         17
10
13
         16
11
         11
15
         10
```

```
12
14
          7
max_glu_serum
Norm
        2597
>200
        1485
>300
        1264
A1Cresult
>8
        8216
Norm
       4990
>7
        3812
insulin
No
         47383
Steady 30849
Down
         12218
Up
         11316
change
No
      54755
Ch
      47011
diabetesMed
Yes 78363
No
       23403
readmitted
N0
      54864
>30
       35545
<30
       11357
readmit_binary
    54864
1
    46902
readmit_30_days
     90409
0
1
     11357
```

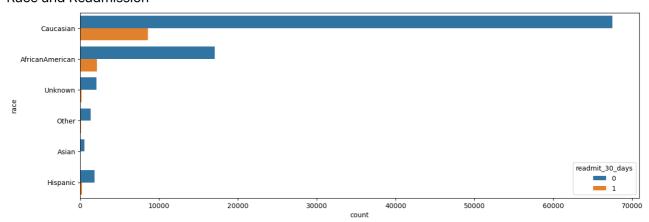
- Binary: gender, medicare, medicaid, had_emergency, had_inpatient_days, had_outpatient_days, readmit_binary, readmit_30_days, change, diabetesMed, discharge_disposition_id
- Multi-categorical: race, age, admission_source_id, medical_specialty, primary_diagnosis, max_glu_serum, A1Cresult, insulin, readmitted
- Continuous: time_in_hospital, num_lab_procedures, num_procedures, num_medications, number_diagnoses

• Age and Readmission



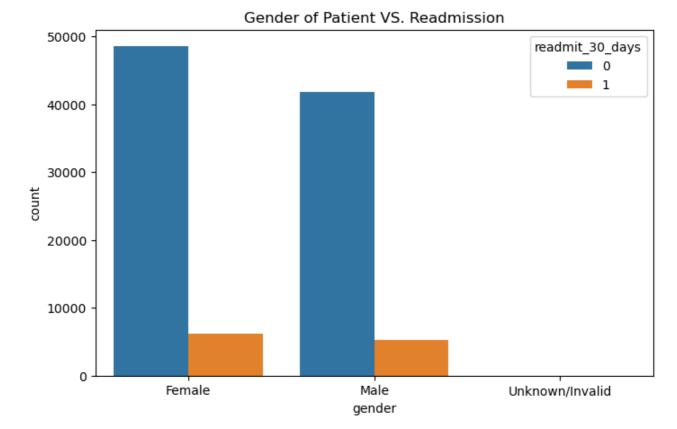
The data is biased towards samples with older age. The data size grows as the age increases.

• Race and Readmission



The data is biased towards Caucasian and AfricanAmerican. Caucasian has the largest sample size.

· Gender and Readmission



The data is relatively balanced.

2. Model Training

Please refer to the train.ipynb file for the data analysis and model training. A Logistic Regression model is trained with an accuracy of 0.63 on the test set.

3. Quantify Fairness

Please refer to the fairness.ipynb for more details. The results are in the form of fairness metric, indicator, pairwise synthesis.

Demographic Parity

```
Demographic Parity

AfricanAmerican | 0.3929
Caucasian | 0.3807
Other | 0.3062
Hispanic | 0.3496
Unknown | 0.2237
Asian | 0.2657

max difference: ('AfricanAmerican', 'Unknown') | 0.1692
min difference: ('AfricanAmerican', 'Caucasian') | 0.0122
smallest ratio: ('Unknown', 'AfricanAmerican', 0.5694565307125041)
largest ratio: ('Caucasian', 'AfricanAmerican', 0.9689210129843422)
maximum indicator: AfricanAmerican | 0.3929
```

Equalized Opportunity

```
Equalized Opportunity

AfricanAmerican | 0.0158
Caucasian | 0.1444
Other | 0.0007
Hispanic | 0.0011
Unknown | 0.0008
Asian | 0.0003

max difference: ('Caucasian', 'Asian') | 0.1441
min difference: ('Other', 'Unknown') | 0.0001
smallest ratio: ('Asian', 'Caucasian', 0.0023049102184119576)
largest ratio: ('Other', 'Unknown', 0.9089898039918192)
maximum indicator: Caucasian | 0.1444
```

Equalized Odds

• Equalized Odds (True)

```
AfricanAmerican | 0.0158
Caucasian | 0.1444
Other | 0.0007
Hispanic | 0.0011
Unknown | 0.0008
Asian | 0.0003

max difference: ('Caucasian', 'Asian') | 0.1441
min difference: ('Other', 'Unknown') | 0.0001
smallest ratio: ('Asian', 'Caucasian', 0.0023049102184119576)
largest ratio: ('Other', 'Unknown', 0.9089898039918192)
maximum indicator: Caucasian | 0.1444
```

• Equalized Odds (False)

```
AfricanAmerican | 0.0705

Caucasian | 0.2693

Other | 0.0044

Hispanic | 0.0067

Unknown | 0.0050

Asian | 0.0013

max difference: ('Caucasian', 'Asian') | 0.2680

min difference: ('Other', 'Unknown') | 0.0006

smallest ratio: ('Asian', 'Caucasian', 0.005002870499466908)
```

```
largest ratio: ('Other', 'Unknown', 0.8849557522123895)
maximum indicator: Caucasian | 0.2693
```

Conditional Statistical Parity

• L = ('Male', '30-60 years')

```
AfricanAmerican | 0.3582
Caucasian | 0.3179
Other | 0.2883
Hispanic | 0.2909
Unknown | 0.1444
Asian | 0.1731
Maximum indicator: 0.3582

max difference: ('AfricanAmerican', 'Unknown') | 0.2138
min difference: ('Other', 'Hispanic') | 0.0026
smallest ratio: Unknown, AfricanAmerican | 0.4032122169562928
largest ratio: Other, Hispanic | 0.991180981595092
```

• L = ('Male', 'Over 60 years')

```
AfricanAmerican | 0.4205
Caucasian | 0.3995
Other | 0.3350
Hispanic | 0.4741
Unknown | 0.2959
Asian | 0.3131
Maximum indicator: 0.4741

max difference: ('Hispanic', 'Unknown') | 0.1782
min difference: ('Unknown', 'Asian') | 0.0172
smallest ratio: Unknown, Hispanic | 0.6241701019743979
largest ratio: Caucasian, AfricanAmerican | 0.950069733141813
```

• L = ('Male', '30 years or younger')

```
AfricanAmerican | 0.2366
Caucasian | 0.2049
Other | 0.0833
Hispanic | 0.2941
Unknown | 0.0000
Asian | 0.0000
Maximum indicator: 0.2941
max difference: ('Hispanic', 'Unknown') | 0.2941
min difference: ('Unknown', 'Asian') | 0.0000
```

```
smallest ratio: Unknown, AfricanAmerican | 0.0
largest ratio: Caucasian, AfricanAmerican | 0.8660663399065315
```

• L = ('Female', '30-60 years')

```
AfricanAmerican | 0.3651
Caucasian | 0.3422
Other | 0.2288
Hispanic | 0.2682
Unknown | 0.1329
Asian | 0.1923
Maximum indicator: 0.3651

max difference: ('AfricanAmerican', 'Unknown') | 0.2322
min difference: ('AfricanAmerican', 'Caucasian') | 0.0229
smallest ratio: Unknown, AfricanAmerican | 0.3640682217929125
largest ratio: Caucasian, AfricanAmerican | 0.9373856225754337
```

• L = ('Female', 'Over 60 years')

```
AfricanAmerican | 0.4320
Caucasian | 0.4048
Other | 0.3373
Hispanic | 0.3784
Unknown | 0.2364
Asian | 0.3010
Maximum indicator: 0.4320

max difference: ('AfricanAmerican', 'Unknown') | 0.1956
min difference: ('Caucasian', 'Hispanic') | 0.0264
smallest ratio: Unknown, AfricanAmerican | 0.5471725125268433
largest ratio: Caucasian, AfricanAmerican | 0.9371029642618253
```

• L = ('Female', '30 years or younger')

```
AfricanAmerican | 0.2771
Caucasian | 0.3319
Other | 0.4000
Hispanic | 0.0526
Unknown | 0.0000
Asian | 0.0000
Maximum indicator: 0.4000

max difference: ('Other', 'Unknown') | 0.4000
min difference: ('Unknown', 'Asian') | 0.0000
smallest ratio: Unknown, AfricanAmerican | 0.0
largest ratio: AfricanAmerican, Caucasian | 0.8349008938981733
```

4. Evaluation

Demographic Parity

Demographic Parity measures the equality of prediction outcomes across different sensitive groups. It is achieved when the probability of a certain prediction is not dependent on sensitive group membership.

The results indicate that the African American group has the highest proportion of positive predictions (0.3929), followed by the Caucasian group (0.3807).

The maximum difference is observed between the African American and Unknown.

In this metric, the model is biased towards the African American and Caucasian groups, with the Asian and Unknown groups having the lowest proportion of being readmitted.

Equalized Opportunity

It means the protected and unprotected groups should have equal true positive rates.

The Caucasian group has the highest TPR (0.1444). Others have much lower TPR values. The maximum difference is observed between the Caucasian and Asian groups (0.1441).

The smallest ratio is found between the Asian and Caucasian groups.

In this metric, the model is biased towards the Caucasian group, with the Asian group having the lowest TPR.

Equalized Odds

Equalized Odds means the protected and unprotected groups should have equal true positive rates and false positive rates.

We discuss the false positive rates, since the true positive rates are the same as the Equalized Opportunity metric.

The Caucasian group has the highest FPR(0.2693). Others have much lower FPR values. The maximum difference is still observed between the Caucasian and Asian groups.

The smallest ratio is found between the Asian and Caucasian groups.

In this metric, the model is biased towards the Caucasian group, tending to predict the Caucasian group as readmitted.

Conditional Statistical Parity

Conditional Statistical Parity means the protected and unprotected groups should have equal true positive rates conditioned on a third variable (legitimate factors L).

L = gender, age

L = ('Male', '30-60 years')

The `African American` group has the highest proportion of positive predictions (0.3582), followed by the `Caucasian` group (0.3179). The maximum difference is observed between the `African American` and `Unknown` groups.

The model is biased towards the `African American` and `Caucasian` groups, with the `Asian` and `Unknown` groups having the lowest proportion of being readmitted.

• L = ('Male', 'Over 60 years')

The `Hispanic` group has the highest proportion of positive predictions (0.4741), followed by the `AfricanAmerican` group. The maximum difference is observed between the `Hispanic` and `Unknown` groups.

The model is biased towards the `Hispanic` and `AfricanAmerican` groups, with the `Asian` and `Unknown` groups having the lowest proportion of being readmitted.

• L = ('Male', '30 years or younger')

Same as above, the model is biased towards the `Hispanic` and `AfricanAmerican` groups, with the `Asian` and `Unknown` groups having the lowest proportion of being readmitted (**even `0`!**).

• L = ('Female', '30-60 years')

Same. Biased towards the `Caucasian` and `AfricanAmerican` groups, with the `Asian` and `Unknown` groups having the lowest proportion of being readmitted.

• L = ('Female', 'Over 60 years')

Same. Biased towards the `Caucasian` and `AfricanAmerican` groups, with the `Asian` and `Unknown` groups having the lowest proportion of being readmitted.

• L = ('Female', '30 years or younger')

Biased towards the `Caucasian` and `Other` groups, with the `Asian` and `Unknown` groups having the lowest proportion of being readmitted (**even `0`**!).

5. Conclusion

The model is mostly biased towards the Caucasian group, then the AfricanAmerican and Hispanic groups, indicated by the tendency to predict these groups as readmitted. While the Asian and Unknown groups have the lowest proportion of being readmitted.

This is related to the imbalance of the dataset, where the Caucasian group has the largest sample size, then the AfricanAmerican group. The Asian and Unknown groups have the smallest sample size.

References

kaggle