

# CS340 Computational 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
age	object
discharge_disposition_id	object
admission_source_id	object
time_in_hospital	int64
medical_specialty	object
num_lab_procedures	int64
num_procedures	int64
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
Other	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	3208
43	2804
44	2496
45	2376
38	2213

...

120	1
132	1
121	1
126	1
118	1

#### num\_procedures

0	46652
1	20742
2	12717
3	9443
6	4954
4	4180
5	3078

#### num\_medications

13	6086
12	6004
11	5795
15	5792
14	5707
...	
70	2
75	2
81	1
79	1
74	1

#### primary\_diagnosis

Other	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
10	17
13	16
11	11
15	10

12	9
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	
NO	54864
>30	35545
<30	11357

readmit_binary	
0	54864
1	46902

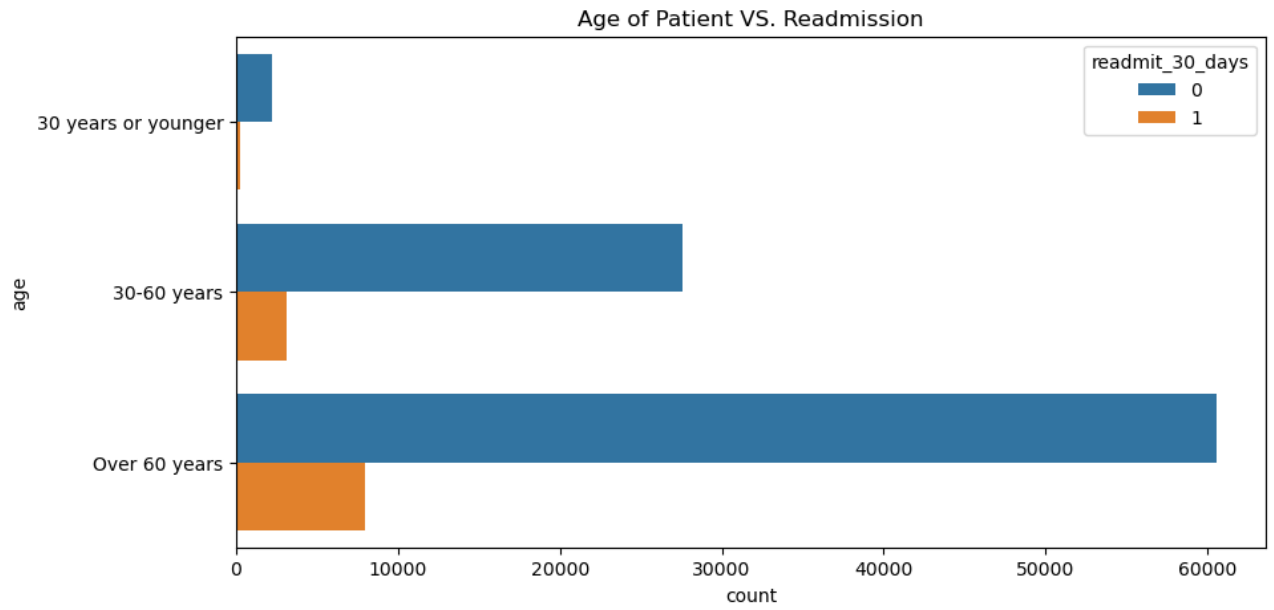
  

readmit_30_days	
0	90409
1	11357

- 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

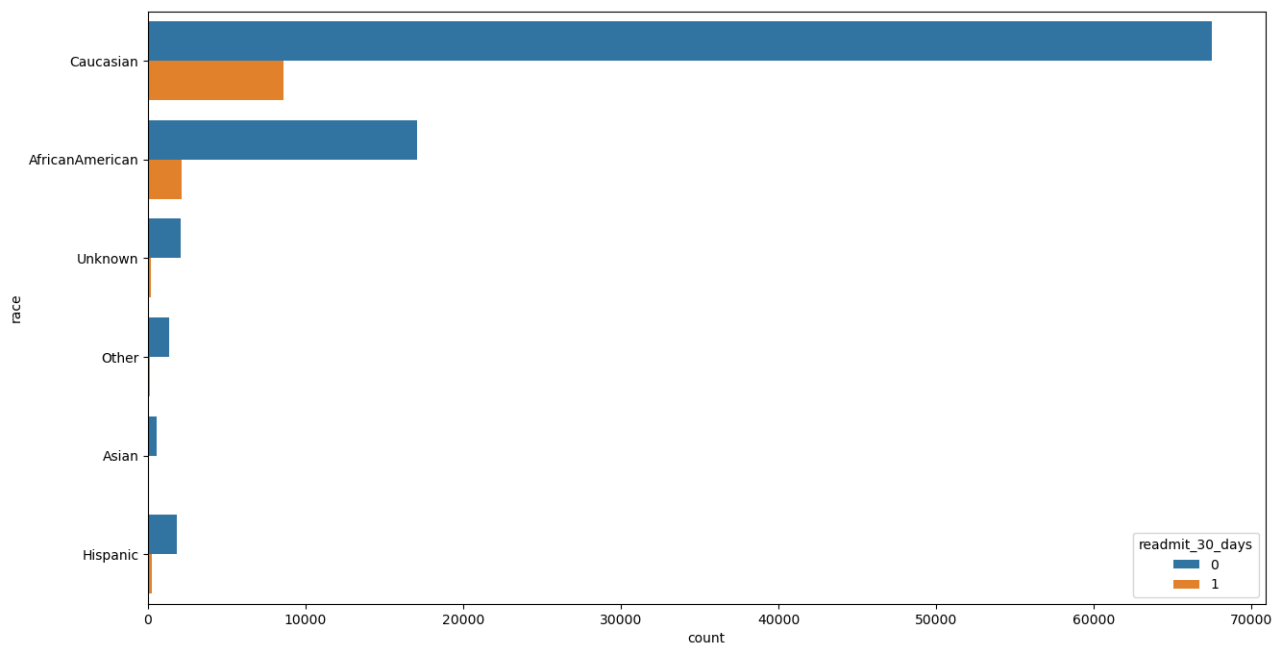
Sample sizes of the groups according to Sensitive Features

- Age and Readmission



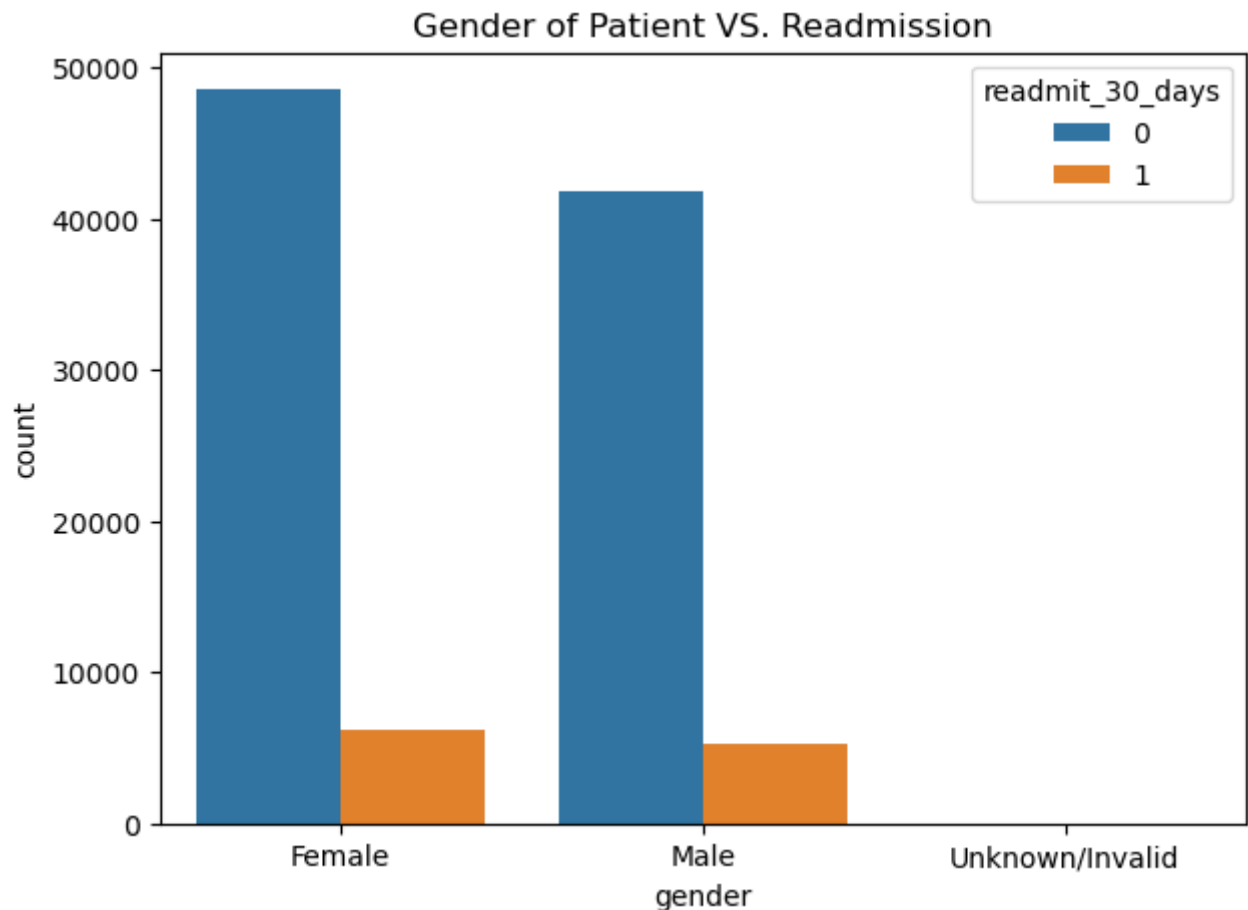
The data is biased towards samples with older age. The data size grows as the age increases.

- 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](#)