

# Flu Shot Learning - Predict Seasonal Flu Vaccines

August 7, 2020

## **1 Flu Shot Learning: Predict Seasonal Flu Vaccines**

### **2 Main objective of the analysis that specifies whether your model will be focused on prediction or interpretation**

The project is focus solely on prediction of flu vaccines.

### **3 Brief description of the data set and a summary of its attributes**

The data for this competition comes from the National 2009 H1N1 Flu Survey (NHFS).

The National 2009 H1N1 Flu Survey (NHFS) was sponsored by the National Center for Immunization and Respiratory Diseases (NCIRD) and conducted jointly by NCIRD and the National Center for Health Statistics (NCHS), Centers for Disease Control and Prevention (CDC). The NHFS was a list-assisted random-digit-dialing telephone survey of households, designed to monitor influenza immunization coverage in the 2009-10 season.

The target population for the NHFS was all persons 6 months or older living in the United States at the time of the interview. Data from the NHFS were used to produce timely estimates of vaccination coverage rates for both the monovalent pH1N1 and trivalent seasonal influenza vaccines.

The NHFS was conducted between October 2009 and June 2010. It was one-time survey designed specifically to monitor vaccination during the 2009-2010 flu season in response to the 2009 H1N1 pandemic. The CDC has other ongoing programs for annual phone surveys that continue to monitor seasonal flu vaccination.

## **4 Problem description**

Can you predict whether people got seasonal flu vaccines using information they shared about their backgrounds, opinions, and health behaviors?

In this challenge, we will take a look at vaccination, a key public health measure used to fight infectious diseases. Vaccines provide immunization for individuals, and enough immunization in a community can further reduce the spread of diseases through “herd immunity.”

## **5 The features in this dataset**

Your goal is to predict how likely individuals are to receive their seasonal flu vaccines.

Field	Description
seasonal_vaccine	Whether respondent received seasonal flu vaccine
respondent_id	a unique and random identifier
behavioral_antiviral_takes	Has taken antiviral medications. (binary)
behavioral_avoided	Has avoided close contact with others with flu-like symptoms. (binary)
behavioral_face_mask	Has bought a face mask. (binary)
behavioral_wash_hands	Has frequently washed hands or used hand sanitizer. (binary)
behavioral_large_gatherings	Has spent time at large gatherings. (binary)
behavioral_outside_household	Has had contact with people outside of own household. (binary)
behavioral_touch_face	Has avoided touching eyes, nose, or mouth. (binary)
doctor_recc_seasonal_vaccine	Seasonal flu vaccine was recommended by doctor. (binary)
chronic_med_conditions	Presence of the following chronic medical conditions: asthma or an other lung condition, diabetes, a heart condition, a kidney condition, sickle cell anemia or other anemia, a neurological or neuromuscular condition, a liver condition, or a weakened immune system caused by a chronic illness or by medicines taken for a chronic illness. (binary)
child_under_6_household	Has regular close contact with a child under the age of six months. (binary)
health_worker	Is a healthcare worker. (binary)
health_insurance	Has health insurance. (binary)
opinion_seas_vaccine_effectiveness	Respondent's opinion about seasonal flu vaccine effectiveness. 1 = Not at all effective; 2 = Not very effective; 3 = Don't know; 4 = Somewhat effective; 5 = Very effective.
opinion_seas_risk	Respondent's opinion about risk of getting sick with seasonal flu without vaccine. 1 = Very Low; 2 = Somewhat low; 3 = Don't know; 4 = Somewhat high; 5 = Very high.
opinion_seas_sick_worried	Respondent's worry of getting sick from taking seasonal flu vaccine. 1 = Not at all worried; 2 = Not very worried; 3 = Don't know; 4 = Somewhat worried; 5 = Very worried.
age_group	Age group of respondent.
education	Self-reported education level.
race	Race of respondent.
sex	Sex of respondent.
income_poverty	Household annual income of respondent with respect to 2008 Census poverty thresholds.
marital_status	Marital status of respondent.
employment_status	Employment status of respondent.
hhs_geo_region	Respondent's residence using a 10-region geographic classification defined by the U.S. Dept. of Health and Human Services. Values are represented as short random character strings.
census_msa	Respondent's residence within metropolitan statistical areas (MSA) as defined by the U.S. Census.
household_adults	Number of other adults in household, top-coded to 3.
household_children	Number of children in household, top-coded to 3.

Field	Description
employment_industry	Type of industry respondent is employed in. Values are represented as short random character strings.
employment_occupation	Type of occupation of respondent. Values are represented as short random character strings.

## 6 Brief summary of data exploration and actions taken for data cleaning and feature engineering

Data Exploration includes data summary, statistics, relevant graphs to find any relationships within.

As for data cleaning, we will check for missing values and decide what imputation method. We also check for data duplicates and outliers. Finally perform binary encoding before model training.

### 6.0.1 Import Libraries

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import sklearn

from sklearn.linear_model import LogisticRegression
from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifier

from sklearn.model_selection import cross_val_score, train_test_split,
↳GridSearchCV, RandomizedSearchCV
from sklearn.preprocessing import LabelEncoder, StandardScaler, MinMaxScaler,
↳OneHotEncoder, PolynomialFeatures
from sklearn.metrics import confusion_matrix, classification_report,
↳mean_absolute_error, mean_squared_error, r2_score
from sklearn.metrics import plot_confusion_matrix, plot_precision_recall_curve,
↳plot_roc_curve, accuracy_score
from sklearn.metrics import auc, f1_score, precision_score, recall_score,
↳roc_auc_score

%matplotlib inline
sns.set_style('dark')
sns.set(font_scale=1.2)

import warnings
warnings.filterwarnings('ignore')
import pandas.util.testing as tm
from pycaret.classification import *
```

```
np.random.seed(123)

pd.options.display.max_columns= None
#pd.options.display.max_rows = None
```

C:\ProgramData\Anaconda3\lib\site-packages\statsmodels\tools\\_testing.py:19:  
FutureWarning: pandas.util.testing is deprecated. Use the functions in the  
public API at pandas.testing instead.  
import pandas.util.testing as tm

```
[2]: df = pd.read_csv("training_set_features.csv")
```

```
[3]: df
```

```
[3]:
```

	respondent_id	behavioral_antiviral_meds	behavioral_avoidance	\
0	0	0.0	0.0	
1	1	0.0	1.0	
2	2	0.0	1.0	
3	3	0.0	1.0	
4	4	0.0	1.0	
...	...	...	...	
26702	26702	0.0	1.0	
26703	26703	0.0	1.0	
26704	26704	0.0	1.0	
26705	26705	0.0	0.0	
26706	26706	0.0	1.0	

	behavioral_face_mask	behavioral_wash_hands	\
0	0.0	0.0	
1	0.0	1.0	
2	0.0	0.0	
3	0.0	1.0	
4	0.0	1.0	
...	...	...	
26702	0.0	0.0	
26703	0.0	1.0	
26704	1.0	1.0	
26705	0.0	0.0	
26706	0.0	0.0	

	behavioral_large_gatherings	behavioral_outside_home	\
0	0.0	1.0	
1	0.0	1.0	
2	0.0	0.0	
3	1.0	0.0	
4	1.0	0.0	
...	...	...	

26702	0.0	1.0
26703	0.0	0.0
26704	1.0	0.0
26705	0.0	0.0
26706	0.0	0.0

	behavioral_touch_face	doctor_recc_seasonal	chronic_med_condition \
0	1.0	0.0	0.0
1	1.0	0.0	0.0
2	0.0	NaN	1.0
3	0.0	1.0	1.0
4	1.0	0.0	0.0
...	...	...	...
26702	0.0	0.0	0.0
26703	0.0	1.0	0.0
26704	1.0	0.0	0.0
26705	NaN	0.0	0.0
26706	0.0	0.0	0.0

	child_under_6_months	health_worker	health_insurance \
0	0.0	0.0	1.0
1	0.0	0.0	1.0
2	0.0	0.0	NaN
3	0.0	0.0	NaN
4	0.0	0.0	NaN
...	...	...	...
26702	0.0	0.0	NaN
26703	0.0	1.0	1.0
26704	0.0	0.0	NaN
26705	0.0	0.0	0.0
26706	0.0	0.0	1.0

	opinion_seas_vacc_effective	opinion_seas_risk \
0	2.0	1.0
1	4.0	2.0
2	4.0	1.0
3	5.0	4.0
4	3.0	1.0
...	...	...
26702	5.0	2.0
26703	5.0	1.0
26704	5.0	4.0
26705	2.0	1.0
26706	5.0	1.0

	opinion_seas_sick_from_vacc	age_group	education	race \
0	2.0	55 - 64 Years	< 12 Years	White

1	4.0	35 - 44 Years	12 Years	White
2	2.0	18 - 34 Years	College Graduate	White
3	1.0	65+ Years	12 Years	White
4	4.0	45 - 54 Years	Some College	White
...	...	...	...	...
26702	2.0	65+ Years	Some College	White
26703	1.0	18 - 34 Years	College Graduate	White
26704	2.0	55 - 64 Years	Some College	White
26705	2.0	18 - 34 Years	Some College	Hispanic
26706	1.0	65+ Years	Some College	White

	sex	income_poverty	marital_status	rent_or_own	\
0	Female	Below Poverty	Not Married	Own	
1	Male	Below Poverty	Not Married	Rent	
2	Male	<= \$75,000, Above Poverty	Not Married	Own	
3	Female	Below Poverty	Not Married	Rent	
4	Female	<= \$75,000, Above Poverty	Married	Own	
...	...	...	...	...	
26702	Female	<= \$75,000, Above Poverty	Not Married	Own	
26703	Male	<= \$75,000, Above Poverty	Not Married	Rent	
26704	Female	NaN	Not Married	Own	
26705	Female	<= \$75,000, Above Poverty	Married	Rent	
26706	Male	<= \$75,000, Above Poverty	Married	Own	

	employment_status	hhs_geo_region	census_msa	\
0	Not in Labor Force	oxchjgsf	Non-MSA	
1	Employed	bhuqouqj	MSA, Not Principle City	
2	Employed	qufhixun	MSA, Not Principle City	
3	Not in Labor Force	lrircsnp	MSA, Principle City	
4	Employed	qufhixun	MSA, Not Principle City	
...	...	...	...	
26702	Not in Labor Force	qufhixun	Non-MSA	
26703	Employed	lzpaxyit	MSA, Principle City	
26704	NaN	lzpaxyit	MSA, Not Principle City	
26705	Employed	lrircsnp	Non-MSA	
26706	Not in Labor Force	mlyzmhmf	MSA, Principle City	

	household_adults	household_children	employment_industry	\
0	0.0	0.0	NaN	
1	0.0	0.0	pxcmvdjn	
2	2.0	0.0	rucpzijj	
3	0.0	0.0	NaN	
4	1.0	0.0	wxleyezf	
...	...	...	...	
26702	0.0	0.0	NaN	
26703	1.0	0.0	fcxhlndr	
26704	0.0	0.0	NaN	

26705	1.0	0.0	fcxhlnwr
26706	1.0	0.0	NaN

	employment_occupation	seasonal_vaccine
0	NaN	0
1	xgwztkwe	1
2	xtkaffoo	0
3	NaN	1
4	emcorrxb	0
...	...	...
26702	NaN	0
26703	cmhcxjea	0
26704	NaN	1
26705	haliazsg	0
26706	NaN	0

[26707 rows x 31 columns]

Dataset has 31 categorical features.

```
[4]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 26707 entries, 0 to 26706
Data columns (total 31 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   respondent_id                         26707 non-null  int64
1   behavioral_antiviral_meds             26636 non-null  float64
2   behavioral_avoidance                  26499 non-null  float64
3   behavioral_face_mask                  26688 non-null  float64
4   behavioral_wash_hands                 26665 non-null  float64
5   behavioral_large_gatherings           26620 non-null  float64
6   behavioral_outside_home               26625 non-null  float64
7   behavioral_touch_face                 26579 non-null  float64
8   doctor_recc_seasonal                 24547 non-null  float64
9   chronic_med_condition                 25736 non-null  float64
10  child_under_6_months                 25887 non-null  float64
11  health_worker                        25903 non-null  float64
12  health_insurance                     14433 non-null  float64
13  opinion_seas_vacc_effective            26245 non-null  float64
14  opinion_seas_risk                      26193 non-null  float64
15  opinion_seas_sick_from_vacc            26170 non-null  float64
16  age_group                             26707 non-null  object
17  education                             25300 non-null  object
18  race                                  26707 non-null  object
19  sex                                   26707 non-null  object
20  income_poverty                       22284 non-null  object
```

```

21 marital_status          25299 non-null object
22 rent_or_own             24665 non-null object
23 employment_status       25244 non-null object
24 hhs_geo_region           26707 non-null object
25 census_msa               26707 non-null object
26 household_adults         26458 non-null float64
27 household_children       26458 non-null float64
28 employment_industry      13377 non-null object
29 employment_occupation    13237 non-null object
30 seasonal_vaccine         26707 non-null int64

```

dtypes: float64(17), int64(2), object(12)

memory usage: 6.3+ MB

Summary of statistics below:

```
[5]: df.describe(include='all').T
```

```

[5]:
count unique top freq \
respondent_id 26707 NaN NaN NaN
behavioral_antiviral_meds 26636 NaN NaN NaN
behavioral_avoidance 26499 NaN NaN NaN
behavioral_face_mask 26688 NaN NaN NaN
behavioral_wash_hands 26665 NaN NaN NaN
behavioral_large_gatherings 26620 NaN NaN NaN
behavioral_outside_home 26625 NaN NaN NaN
behavioral_touch_face 26579 NaN NaN NaN
doctor_recc_seasonal 24547 NaN NaN NaN
chronic_med_condition 25736 NaN NaN NaN
child_under_6_months 25887 NaN NaN NaN
health_worker 25903 NaN NaN NaN
health_insurance 14433 NaN NaN NaN
opinion_seas_vacc_effective 26245 NaN NaN NaN
opinion_seas_risk 26193 NaN NaN NaN
opinion_seas_sick_from_vacc 26170 NaN NaN NaN
age_group 26707 5 65+ Years 6843
education 25300 4 College Graduate 10097
race 26707 4 White 21222
sex 26707 2 Female 15858
income_poverty 22284 3 <= $75,000, Above Poverty 12777
marital_status 25299 2 Married 13555
rent_or_own 24665 2 Own 18736
employment_status 25244 3 Employed 13560
hhs_geo_region 26707 10 lzgpxyit 4297
census_msa 26707 3 MSA, Not Principle City 11645
household_adults 26458 NaN NaN NaN
household_children 26458 NaN NaN NaN
employment_industry 13377 21 fcxhlnwr 2468
employment_occupation 13237 23 xtkaffoo 1778

```



seasonal_vaccine	26707	NaN			NaN	NaN	
	mean	std	min	25%	50%	75%	\
respondent_id	13353	7709.79	0	6676.5	13353	20029.5	
behavioral_antiviral_meds	0.0488437	0.215545	0	0	0	0	
behavioral_avoidance	0.725612	0.446214	0	0	1	1	
behavioral_face_mask	0.0689823	0.253429	0	0	0	0	
behavioral_wash_hands	0.825614	0.379448	0	1	1	1	
behavioral_large_gatherings	0.35864	0.47961	0	0	0	1	
behavioral_outside_home	0.337315	0.472802	0	0	0	1	
behavioral_touch_face	0.677264	0.467531	0	0	1	1	
doctor_recc_seasonal	0.329735	0.470126	0	0	0	1	
chronic_med_condition	0.283261	0.450591	0	0	0	1	
child_under_6_months	0.0825897	0.275266	0	0	0	0	
health_worker	0.111918	0.315271	0	0	0	0	
health_insurance	0.87972	0.3253	0	1	1	1	
opinion_seas_vacc_effective	4.02599	1.08656	1	4	4	5	
opinion_seas_risk	2.71916	1.38506	1	2	2	4	
opinion_seas_sick_from_vacc	2.11811	1.33295	1	1	2	4	
age_group	NaN	NaN	NaN	NaN	NaN	NaN	
education	NaN	NaN	NaN	NaN	NaN	NaN	
race	NaN	NaN	NaN	NaN	NaN	NaN	
sex	NaN	NaN	NaN	NaN	NaN	NaN	
income_poverty	NaN	NaN	NaN	NaN	NaN	NaN	
marital_status	NaN	NaN	NaN	NaN	NaN	NaN	
rent_or_own	NaN	NaN	NaN	NaN	NaN	NaN	
employment_status	NaN	NaN	NaN	NaN	NaN	NaN	
hhs_geo_region	NaN	NaN	NaN	NaN	NaN	NaN	
census_msa	NaN	NaN	NaN	NaN	NaN	NaN	
household_adults	0.886499	0.753422	0	0	1	1	
household_children	0.534583	0.928173	0	0	0	1	
employment_industry	NaN	NaN	NaN	NaN	NaN	NaN	
employment_occupation	NaN	NaN	NaN	NaN	NaN	NaN	
seasonal_vaccine	0.465608	0.498825	0	0	0	1	

	max
respondent_id	26706
behavioral_antiviral_meds	1
behavioral_avoidance	1
behavioral_face_mask	1
behavioral_wash_hands	1
behavioral_large_gatherings	1
behavioral_outside_home	1
behavioral_touch_face	1
doctor_recc_seasonal	1
chronic_med_condition	1
child_under_6_months	1

health_worker	1
health_insurance	1
opinion_seas_vacc_effective	5
opinion_seas_risk	5
opinion_seas_sick_from_vacc	5
age_group	NaN
education	NaN
race	NaN
sex	NaN
income_poverty	NaN
marital_status	NaN
rent_or_own	NaN
employment_status	NaN
hhs_geo_region	NaN
census_msa	NaN
household_adults	3
household_children	3
employment_industry	NaN
employment_occupation	NaN
seasonal_vaccine	1

Shape of dataset:

```
[6]: df.shape
```

```
[6]: (26707, 31)
```

```
[7]: df.columns
```

```
[7]: Index(['respondent_id', 'behavioral_antiviral_meds', 'behavioral_avoidance',
        'behavioral_face_mask', 'behavioral_wash_hands',
        'behavioral_large_gatherings', 'behavioral_outside_home',
        'behavioral_touch_face', 'doctor_recc_seasonal',
        'chronic_med_condition', 'child_under_6_months', 'health_worker',
        'health_insurance', 'opinion_seas_vacc_effective', 'opinion_seas_risk',
        'opinion_seas_sick_from_vacc', 'age_group', 'education', 'race', 'sex',
        'income_poverty', 'marital_status', 'rent_or_own', 'employment_status',
        'hhs_geo_region', 'census_msa', 'household_adults',
        'household_children', 'employment_industry', 'employment_occupation',
        'seasonal_vaccine'],
        dtype='object')
```

## 6.0.2 Data Exploration

The best way is to create graphs!

### 6.0.3 Data Visualization

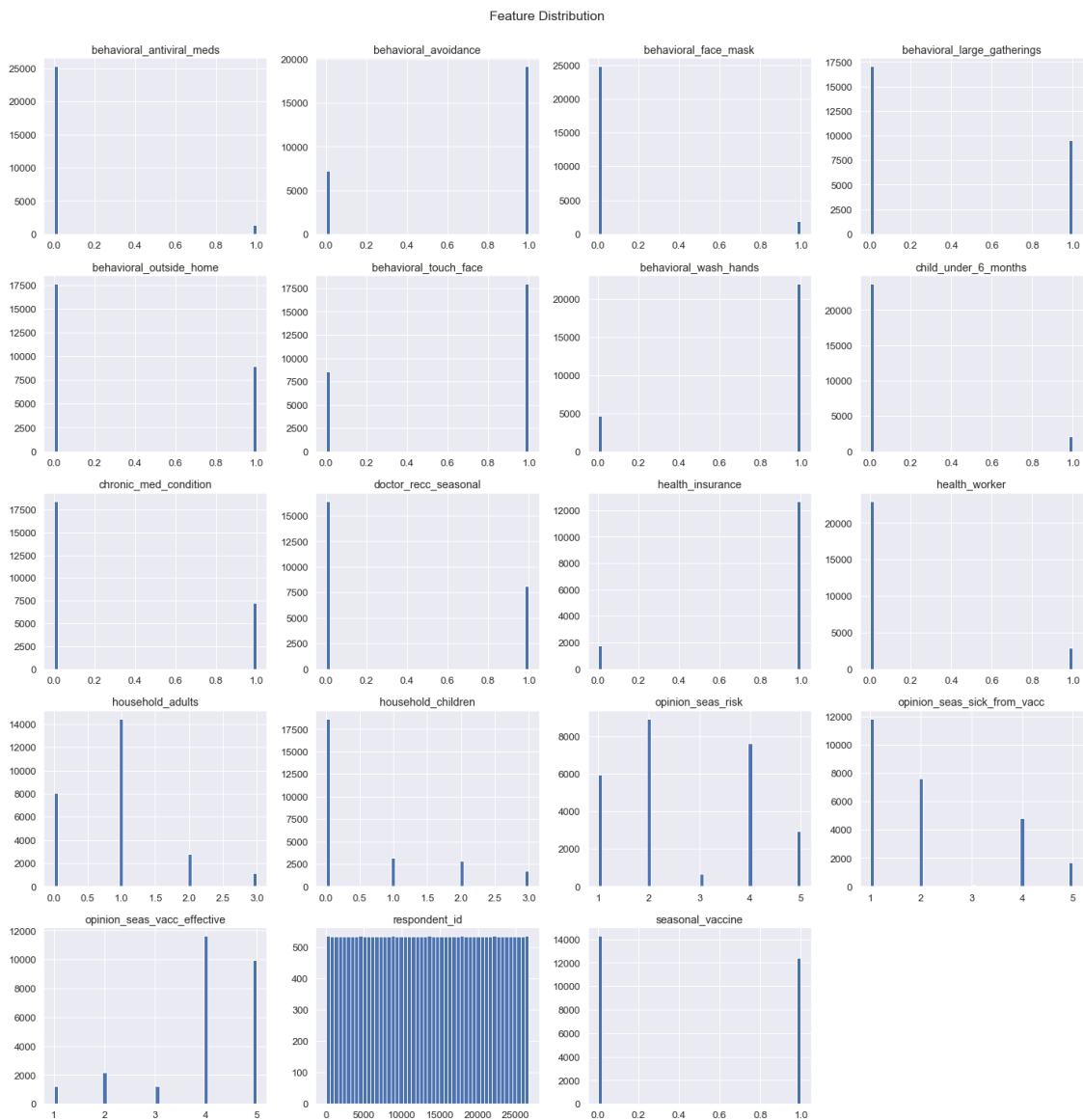
The dataset is mainly discrete/categorical types.

```
[8]: df.hist(bins=50, figsize=(20,20))

plt.suptitle('Feature Distribution', x=0.5, y=1.02, ha='center',
            ↪fontsize='large')

plt.tight_layout()

plt.show();
```



Below are each visuals of the data:

```
[10]: fig = plt.figure(figsize=(20,40))

plt.subplot(7,2,1)
plt.title("Has taken antiviral medications")
sns.countplot(df.behavioral_antiviral_meds, hue=df.seasonal_vaccine)

plt.subplot(7,2,2)
plt.title("Has avoided close contact with others with flu-like symptoms")
sns.countplot(df.behavioral_avoidance, hue=df.seasonal_vaccine)

plt.subplot(7,2,3)
plt.title("Has bought a face mask")
sns.countplot(df.behavioral_face_mask, hue=df.seasonal_vaccine)

plt.subplot(7,2,4)
plt.title("Has frequently washed hands or used hand sanitizer")
sns.countplot(df.behavioral_wash_hands, hue=df.seasonal_vaccine)

plt.subplot(7,2,5)
plt.title("Has reduced time at large gatherings")
sns.countplot(df.behavioral_large_gatherings, hue=df.seasonal_vaccine)

plt.subplot(7,2,6)
plt.title("Has reduced contact with people outside of own household")
sns.countplot(df.behavioral_outside_home, hue=df.seasonal_vaccine)

plt.subplot(7,2,7)
plt.title("Has avoided touching eyes, nose, or mouth")
sns.countplot(df.behavioral_touch_face, hue=df.seasonal_vaccine)

plt.subplot(7,2,8)
plt.title("Seasonal flu vaccine was recommended by doctor")
sns.countplot(df.doctor_recc_seasonal, hue=df.seasonal_vaccine)

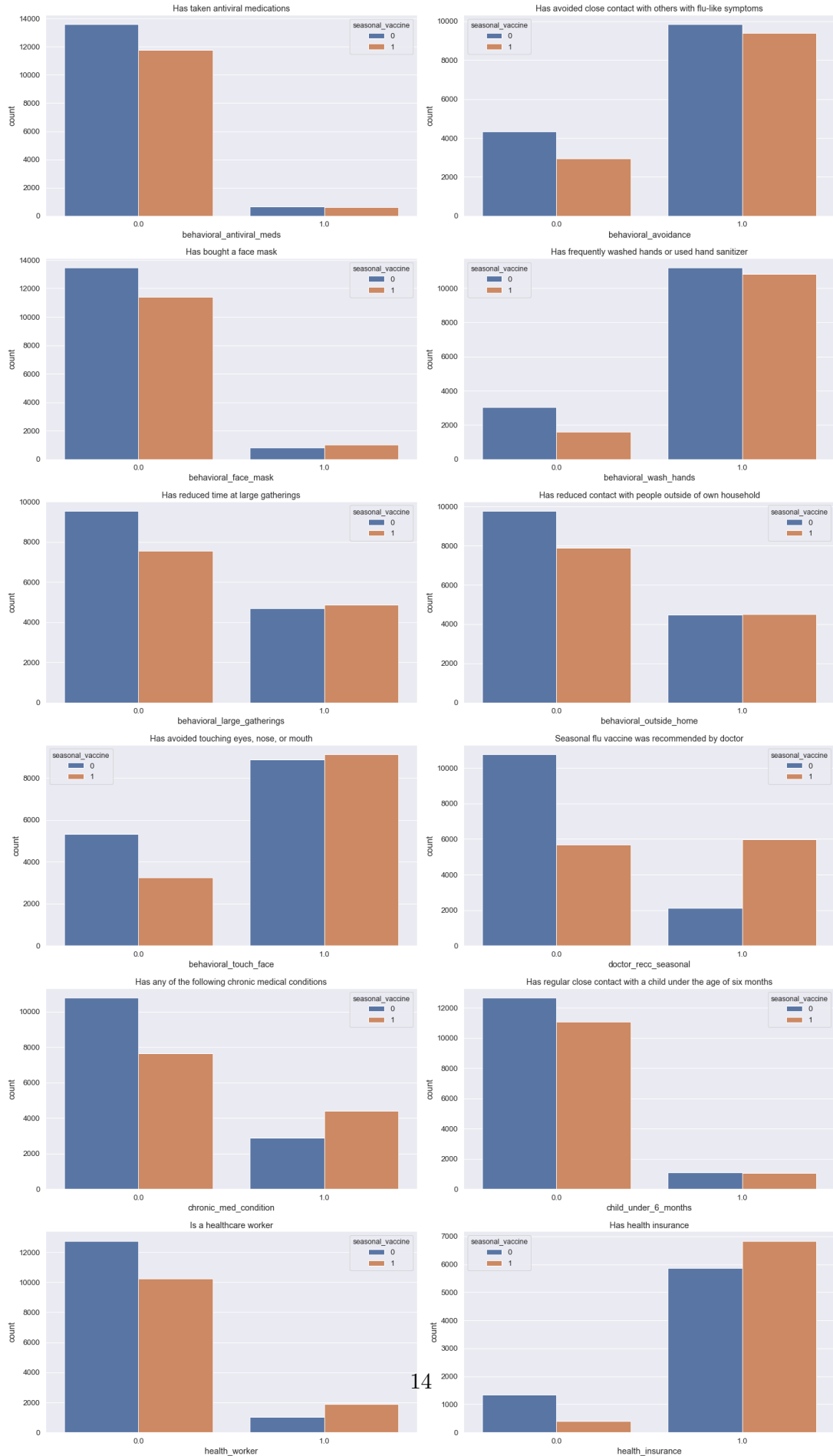
plt.subplot(7,2,9)
plt.title("Has any of the following chronic medical conditions")
sns.countplot(df.chronic_med_condition, hue=df.seasonal_vaccine)

plt.subplot(7,2,10)
plt.title("Has regular close contact with a child under the age of six months")
sns.countplot(df.child_under_6_months, hue=df.seasonal_vaccine)

plt.subplot(7,2,11)
plt.title("Is a healthcare worker")
sns.countplot(df.health_worker, hue=df.seasonal_vaccine)
```

```
plt.subplot(7,2,12)
plt.title("Has health insurance")
sns.countplot(df.health_insurance, hue=df.seasonal_vaccine)

plt.tight_layout()
plt.show()
```



## Part 1 of Data Analysis:

Those who had vaccine avoided close contacts which is surprising since vaccines are supposed to protect them. But they didn't avoid large gatherings.

As for flu vaccine which doctor recommended, there is such acceptance among people.

Health Care workers are most vulnerable but majority of them do vaccinate.

```
[11]: fig = plt.figure(figsize=(20,40))

plt.subplot(7,2,1)
plt.title("Respondent's opinion about seasonal flu vaccine effectiveness")
sns.countplot(df.opinion_seas_vacc_effective, hue=df.seasonal_vaccine)

plt.subplot(7,2,2)
plt.title("Respondent's opinion about risk of getting sick with seasonal flu_
↳without vaccine")
sns.countplot(df.opinion_seas_risk, hue=df.seasonal_vaccine)

plt.subplot(7,2,3)
plt.title("Respondent's worry of getting sick from taking seasonal flu vaccine")
sns.countplot(df.opinion_seas_sick_from_vacc, hue=df.seasonal_vaccine)

plt.subplot(7,2,4)
plt.title("Age group of respondent")
sns.countplot(df.age_group, hue=df.seasonal_vaccine)

plt.subplot(7,2,5)
plt.title("Self-reported education level")
sns.countplot(df.education, hue=df.seasonal_vaccine)

plt.subplot(7,2,6)
plt.title("Race of respondent")
sns.countplot(df.race, hue=df.seasonal_vaccine)

plt.subplot(7,2,7)
plt.title("Sex of respondent")
sns.countplot(df.sex, hue=df.seasonal_vaccine)

plt.subplot(7,2,8)
plt.title("Household annual income of respondent")
sns.countplot(df.income_poverty, hue=df.seasonal_vaccine)

plt.subplot(7,2,9)
plt.title("Marital status of respondent")
```

```
sns.countplot(df.marital_status, hue=df.seasonal_vaccine)

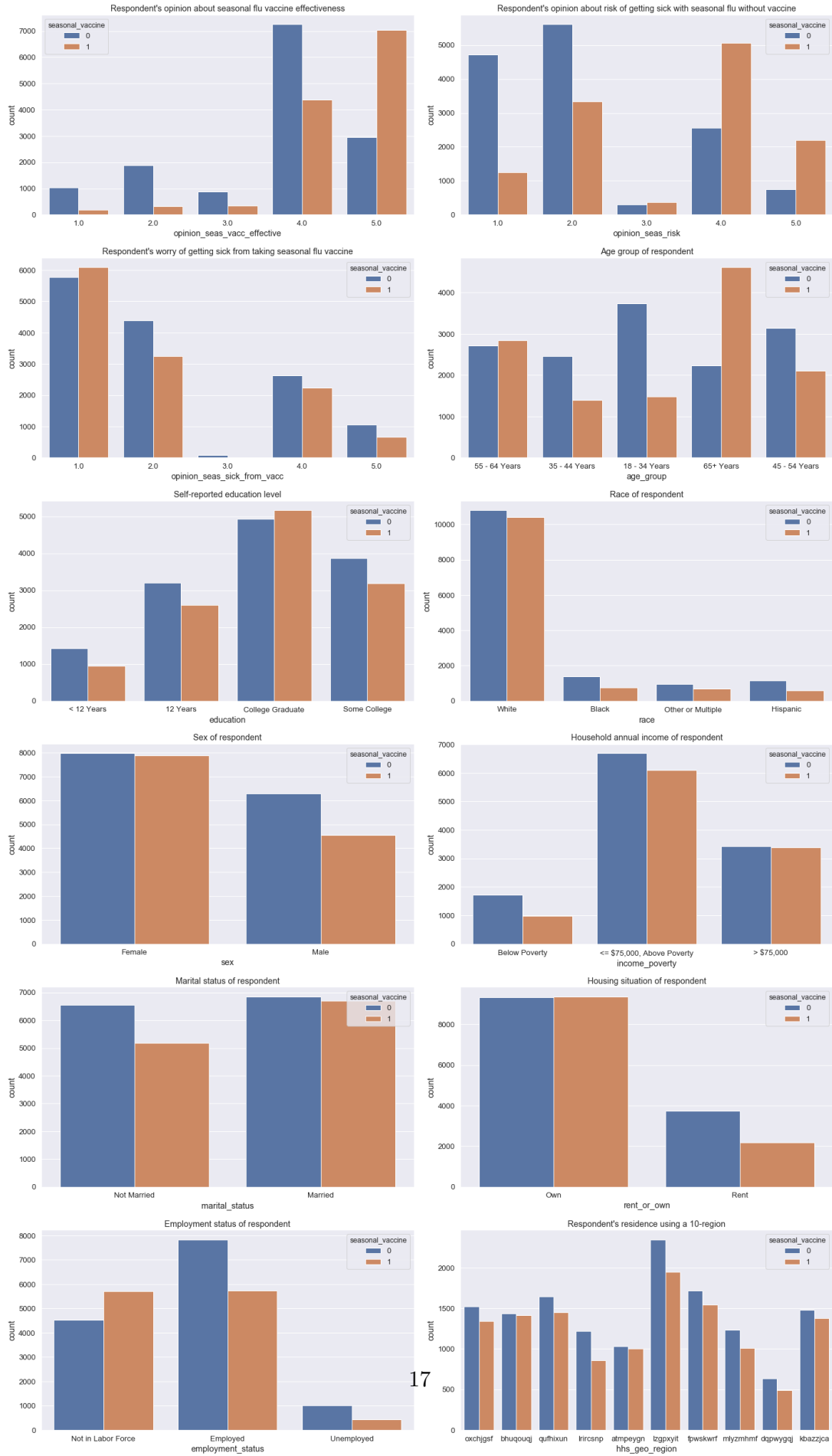
plt.subplot(7,2,10)
plt.title("Housing situation of respondent")
sns.countplot(df.rent_or_own, hue=df.seasonal_vaccine)

plt.subplot(7,2,11)
plt.title("Employment status of respondent")
sns.countplot(df.employment_status, hue=df.seasonal_vaccine)

plt.subplot(7,2,12)
plt.title("Respondent's residence using a 10-region")
sns.countplot(df.hhs_geo_region, hue=df.seasonal_vaccine)

plt.tight_layout()
plt.show()
```





## Part 2 of Data Analysis:

As for respondents opinion, risks and worry, there are no surprises for those who trust flu vaccines.

Respondents more than age 65 and College Educated are vaccinated.

Mainly whites, female, more than 75k income, married, own a house and employed can afford flu vaccines.

```
[12]: fig = plt.figure(figsize=(20,40))

plt.subplot(7,2,1)
plt.title("Respondent's residence within metropolitan statistical areas")
sns.countplot(df.census_msa, hue=df.seasonal_vaccine)

plt.subplot(7,2,2)
plt.title("Number of other adults in household")
sns.countplot(df.household_adults, hue=df.seasonal_vaccine)

plt.subplot(7,2,3)
plt.title("Number of children in household")
sns.countplot(df.household_children, hue=df.seasonal_vaccine)

plt.subplot(7,2,4)
plt.title("Type of industry respondent is employed in")
sns.countplot(df.employment_industry, hue=df.seasonal_vaccine)

plt.subplot(7,2,5)
plt.title("Type of occupation of respondent")
sns.countplot(df.employment_occupation, hue=df.seasonal_vaccine)

plt.subplot(7,2,6)
plt.title("Whether respondent received seasonal flu vaccine")
sns.countplot(df.seasonal_vaccine)

plt.tight_layout()
plt.show()
```



```

[14]:
      respondent_id  behavioral_antiviral_meds \
respondent_id      1.000000      -0.008475
behavioral_antiviral_meds -0.008475      1.000000
behavioral_avoidance      0.009638      0.049247
behavioral_face_mask      -0.006644      0.146261
behavioral_wash_hands      0.011105      0.064119
behavioral_large_gatherings 0.004539      0.106287
behavioral_outside_home      0.009011      0.127679
behavioral_touch_face      0.007575      0.070868
doctor_recc_seasonal      0.001500      0.030909
chronic_med_condition      0.005797      0.008465
child_under_6_months      -0.004839      0.028788
health_worker      -0.003149      0.009465
health_insurance      -0.012603      -0.063988
opinion_seas_vacc_effective 0.005935      0.015003
opinion_seas_risk      -0.005291      0.085315
opinion_seas_sick_from_vacc 0.009563      0.084305
household_adults      0.000187      0.044900
household_children      -0.003726      0.084822
seasonal_vaccine      -0.004652      0.006277

      behavioral_avoidance  behavioral_face_mask \
respondent_id      0.009638      -0.006644
behavioral_antiviral_meds 0.049247      0.146261
behavioral_avoidance      1.000000      0.064946
behavioral_face_mask      0.064946      1.000000
behavioral_wash_hands      0.338130      0.083363
behavioral_large_gatherings 0.227675      0.180907
behavioral_outside_home      0.220348      0.163382
behavioral_touch_face      0.335335      0.104335
doctor_recc_seasonal      0.074088      0.069481
chronic_med_condition      0.039435      0.068113
child_under_6_months      -0.000414      0.039726
health_worker      0.001180      0.069992
health_insurance      0.032662      -0.040257
opinion_seas_vacc_effective 0.119554      0.041556
opinion_seas_risk      0.129504      0.110161
opinion_seas_sick_from_vacc 0.082942      0.090009
household_adults      0.019122      0.013991
household_children      0.040328      0.005826
seasonal_vaccine      0.076395      0.050083

      behavioral_wash_hands \
respondent_id      0.011105
behavioral_antiviral_meds 0.064119
behavioral_avoidance      0.338130
behavioral_face_mask      0.083363

```

behavioral_wash_hands	1.000000
behavioral_large_gatherings	0.195364
behavioral_outside_home	0.192619
behavioral_touch_face	0.365064
doctor_recc_seasonal	0.102044
chronic_med_condition	0.030260
child_under_6_months	0.036188
health_worker	0.053761
health_insurance	0.031919
opinion_seas_vacc_effective	0.138517
opinion_seas_risk	0.172464
opinion_seas_sick_from_vacc	0.088029
household_adults	0.009669
household_children	0.047764
seasonal_vaccine	0.112414

	behavioral_large_gatherings \
respondent_id	0.004539
behavioral_antiviral_meds	0.106287
behavioral_avoidance	0.227675
behavioral_face_mask	0.180907
behavioral_wash_hands	0.195364
behavioral_large_gatherings	1.000000
behavioral_outside_home	0.584085
behavioral_touch_face	0.253683
doctor_recc_seasonal	0.093557
chronic_med_condition	0.104721
child_under_6_months	0.021168
health_worker	-0.032319
health_insurance	-0.059000
opinion_seas_vacc_effective	0.078491
opinion_seas_risk	0.132865
opinion_seas_sick_from_vacc	0.135446
household_adults	-0.031938
household_children	-0.009449
seasonal_vaccine	0.064025

	behavioral_outside_home	behavioral_touch_face \
respondent_id	0.009011	0.007575
behavioral_antiviral_meds	0.127679	0.070868
behavioral_avoidance	0.220348	0.335335
behavioral_face_mask	0.163382	0.104335
behavioral_wash_hands	0.192619	0.365064
behavioral_large_gatherings	0.584085	0.253683
behavioral_outside_home	1.000000	0.267719
behavioral_touch_face	0.267719	1.000000
doctor_recc_seasonal	0.085622	0.100808

chronic_med_condition	0.098858	0.028876
child_under_6_months	0.018195	0.026640
health_worker	-0.034619	0.067648
health_insurance	-0.061381	0.011024
opinion_seas_vacc_effective	0.067469	0.105798
opinion_seas_risk	0.120237	0.143735
opinion_seas_sick_from_vacc	0.138133	0.090097
household_adults	-0.027527	-0.000553
household_children	-0.009558	0.023606
seasonal_vaccine	0.053509	0.120228

	doctor_recc_seasonal	chronic_med_condition \
respondent_id	0.001500	0.005797
behavioral_antiviral_meds	0.030909	0.008465
behavioral_avoidance	0.074088	0.039435
behavioral_face_mask	0.069481	0.068113
behavioral_wash_hands	0.102044	0.030260
behavioral_large_gatherings	0.093557	0.104721
behavioral_outside_home	0.085622	0.098858
behavioral_touch_face	0.100808	0.028876
doctor_recc_seasonal	1.000000	0.213806
chronic_med_condition	0.213806	1.000000
child_under_6_months	0.036832	-0.001349
health_worker	0.059402	-0.026481
health_insurance	0.117195	0.066088
opinion_seas_vacc_effective	0.180902	0.091737
opinion_seas_risk	0.240087	0.162061
opinion_seas_sick_from_vacc	0.025356	0.052587
household_adults	-0.040769	-0.071346
household_children	-0.048380	-0.108237
seasonal_vaccine	0.369190	0.170174

	child_under_6_months	health_worker \
respondent_id	-0.004839	-0.003149
behavioral_antiviral_meds	0.028788	0.009465
behavioral_avoidance	-0.000414	0.001180
behavioral_face_mask	0.039726	0.069992
behavioral_wash_hands	0.036188	0.053761
behavioral_large_gatherings	0.021168	-0.032319
behavioral_outside_home	0.018195	-0.034619
behavioral_touch_face	0.026640	0.067648
doctor_recc_seasonal	0.036832	0.059402
chronic_med_condition	-0.001349	-0.026481
child_under_6_months	1.000000	0.079078
health_worker	0.079078	1.000000
health_insurance	-0.026836	0.046680
opinion_seas_vacc_effective	0.003653	0.030395

opinion_seas_risk	0.050267	0.089142
opinion_seas_sick_from_vacc	0.037582	-0.017893
household_adults	0.044828	0.013380
household_children	0.099562	0.037698
seasonal_vaccine	0.012097	0.127311

	health_insurance	opinion_seas_vacc_effective \
respondent_id	-0.012603	0.005935
behavioral_antiviral_meds	-0.063988	0.015003
behavioral_avoidance	0.032662	0.119554
behavioral_face_mask	-0.040257	0.041556
behavioral_wash_hands	0.031919	0.138517
behavioral_large_gatherings	-0.059000	0.078491
behavioral_outside_home	-0.061381	0.067469
behavioral_touch_face	0.011024	0.105798
doctor_recc_seasonal	0.117195	0.180902
chronic_med_condition	0.066088	0.091737
child_under_6_months	-0.026836	0.003653
health_worker	0.046680	0.030395
health_insurance	1.000000	0.091247
opinion_seas_vacc_effective	0.091247	1.000000
opinion_seas_risk	0.050232	0.344800
opinion_seas_sick_from_vacc	-0.065886	-0.017340
household_adults	-0.078697	-0.022579
household_children	-0.069402	-0.076503
seasonal_vaccine	0.200858	0.361875

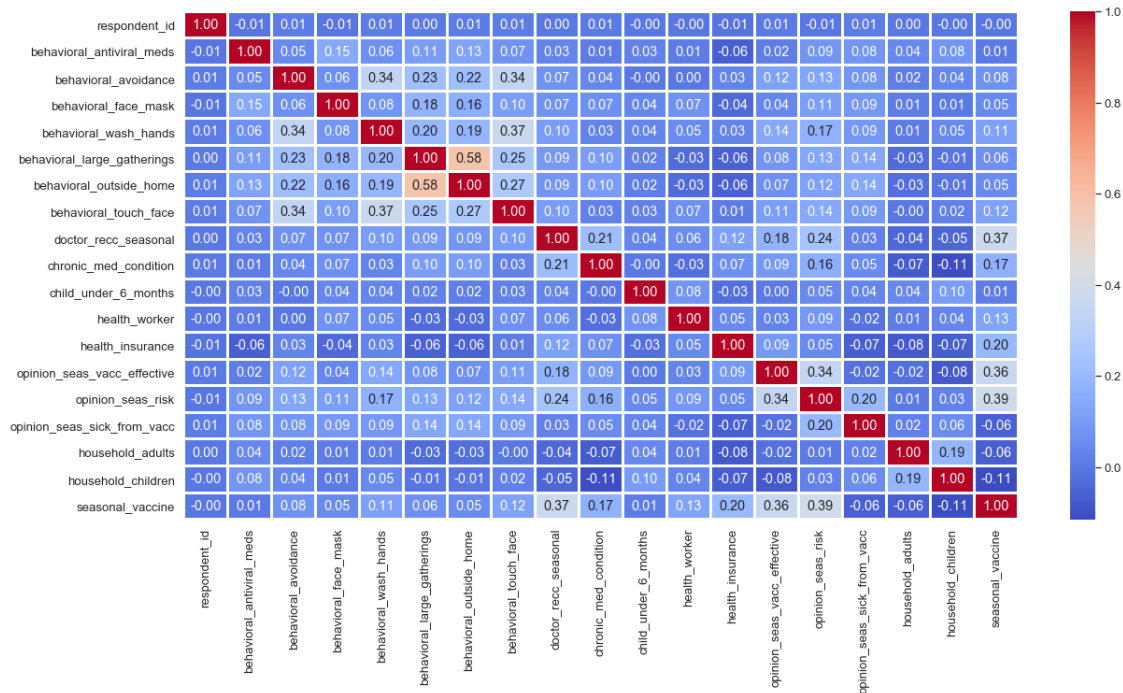
	opinion_seas_risk	opinion_seas_sick_from_vacc \
respondent_id	-0.005291	0.009563
behavioral_antiviral_meds	0.085315	0.084305
behavioral_avoidance	0.129504	0.082942
behavioral_face_mask	0.110161	0.090009
behavioral_wash_hands	0.172464	0.088029
behavioral_large_gatherings	0.132865	0.135446
behavioral_outside_home	0.120237	0.138133
behavioral_touch_face	0.143735	0.090097
doctor_recc_seasonal	0.240087	0.025356
chronic_med_condition	0.162061	0.052587
child_under_6_months	0.050267	0.037582
health_worker	0.089142	-0.017893
health_insurance	0.050232	-0.065886
opinion_seas_vacc_effective	0.344800	-0.017340
opinion_seas_risk	1.000000	0.200379
opinion_seas_sick_from_vacc	0.200379	1.000000
household_adults	0.006111	0.022925
household_children	0.025898	0.057286
seasonal_vaccine	0.390106	-0.061510

	household_adults	household_children \
respondent_id	0.000187	-0.003726
behavioral_antiviral_meds	0.044900	0.084822
behavioral_avoidance	0.019122	0.040328
behavioral_face_mask	0.013991	0.005826
behavioral_wash_hands	0.009669	0.047764
behavioral_large_gatherings	-0.031938	-0.009449
behavioral_outside_home	-0.027527	-0.009558
behavioral_touch_face	-0.000553	0.023606
doctor_recc_seasonal	-0.040769	-0.048380
chronic_med_condition	-0.071346	-0.108237
child_under_6_months	0.044828	0.099562
health_worker	0.013380	0.037698
health_insurance	-0.078697	-0.069402
opinion_seas_vacc_effective	-0.022579	-0.076503
opinion_seas_risk	0.006111	0.025898
opinion_seas_sick_from_vacc	0.022925	0.057286
household_adults	1.000000	0.189571
household_children	0.189571	1.000000
seasonal_vaccine	-0.064840	-0.114614

	seasonal_vaccine
respondent_id	-0.004652
behavioral_antiviral_meds	0.006277
behavioral_avoidance	0.076395
behavioral_face_mask	0.050083
behavioral_wash_hands	0.112414
behavioral_large_gatherings	0.064025
behavioral_outside_home	0.053509
behavioral_touch_face	0.120228
doctor_recc_seasonal	0.369190
chronic_med_condition	0.170174
child_under_6_months	0.012097
health_worker	0.127311
health_insurance	0.200858
opinion_seas_vacc_effective	0.361875
opinion_seas_risk	0.390106
opinion_seas_sick_from_vacc	-0.061510
household_adults	-0.064840
household_children	-0.114614
seasonal_vaccine	1.000000

```
[15]: plt.figure(figsize=(20,10))
sns.heatmap(df.corr(),cmap="coolwarm",annot=True,fmt='.2f',linewidths=2)
plt.show()
```





## Quote:

“Factors that may bias the results of observational studies can be broadly categorized as: selection bias resulting from the way study subjects are recruited or from differing rates of study participation depending on the subjects’ cultural background, age, or socioeconomic status, information bias, measurement error, confounders, and further factors.”

We will drop a number of features which we think that will make the model biased to a certain group/gender/income/social.

### 6.0.4 Drop unwanted features

```
[16]: df.columns
```

```
[16]: Index(['respondent_id', 'behavioral_antiviral_meds', 'behavioral_avoidance',
'behavioral_face_mask', 'behavioral_wash_hands',
'behavioral_large_gatherings', 'behavioral_outside_home',
'behavioral_touch_face', 'doctor_recc_seasonal',
'chronic_med_condition', 'child_under_6_months', 'health_worker',
'health_insurance', 'opinion_seas_vacc_effective', 'opinion_seas_risk',
'opinion_seas_sick_from_vacc', 'age_group', 'education', 'race', 'sex',
'income_poverty', 'marital_status', 'rent_or_own', 'employment_status',
'hhs_geo_region', 'census_msa', 'household_adults',
'household_children', 'employment_industry', 'employment_occupation',
'seasonal_vaccine'],
dtype='object')
```

```
[17]: df.drop(['respondent_id','health_insurance','age_group', 'education', 'race',\
→'sex', 'income_poverty', 'marital_status', 'rent_or_own',
      'employment_status','hhs_geo_region', 'census_msa',\
→'household_adults', 'household_children', 'employment_industry',
      'employment_occupation'],axis=1,inplace=True)
```

```
[18]: df
```

```
[18]: behavioral_antiviral_meds behavioral_avoidance behavioral_face_mask \
0 0.0 0.0 0.0
1 0.0 1.0 0.0
2 0.0 1.0 0.0
3 0.0 1.0 0.0
4 0.0 1.0 0.0
...
26702 0.0 1.0 0.0
26703 0.0 1.0 0.0
26704 0.0 1.0 1.0
26705 0.0 0.0 0.0
26706 0.0 1.0 0.0
```

```
behavioral_wash_hands behavioral_large_gatherings \
0 0.0 0.0
1 1.0 0.0
2 0.0 0.0
3 1.0 1.0
4 1.0 1.0
...
26702 0.0 0.0
26703 1.0 0.0
26704 1.0 1.0
26705 0.0 0.0
26706 0.0 0.0
```

```
behavioral_outside_home behavioral_touch_face doctor_recc_seasonal \
0 1.0 1.0 0.0
1 1.0 1.0 0.0
2 0.0 0.0 NaN
3 0.0 0.0 1.0
4 0.0 1.0 0.0
...
26702 1.0 0.0 0.0
26703 0.0 0.0 1.0
26704 0.0 1.0 0.0
26705 0.0 NaN 0.0
26706 0.0 0.0 0.0
```

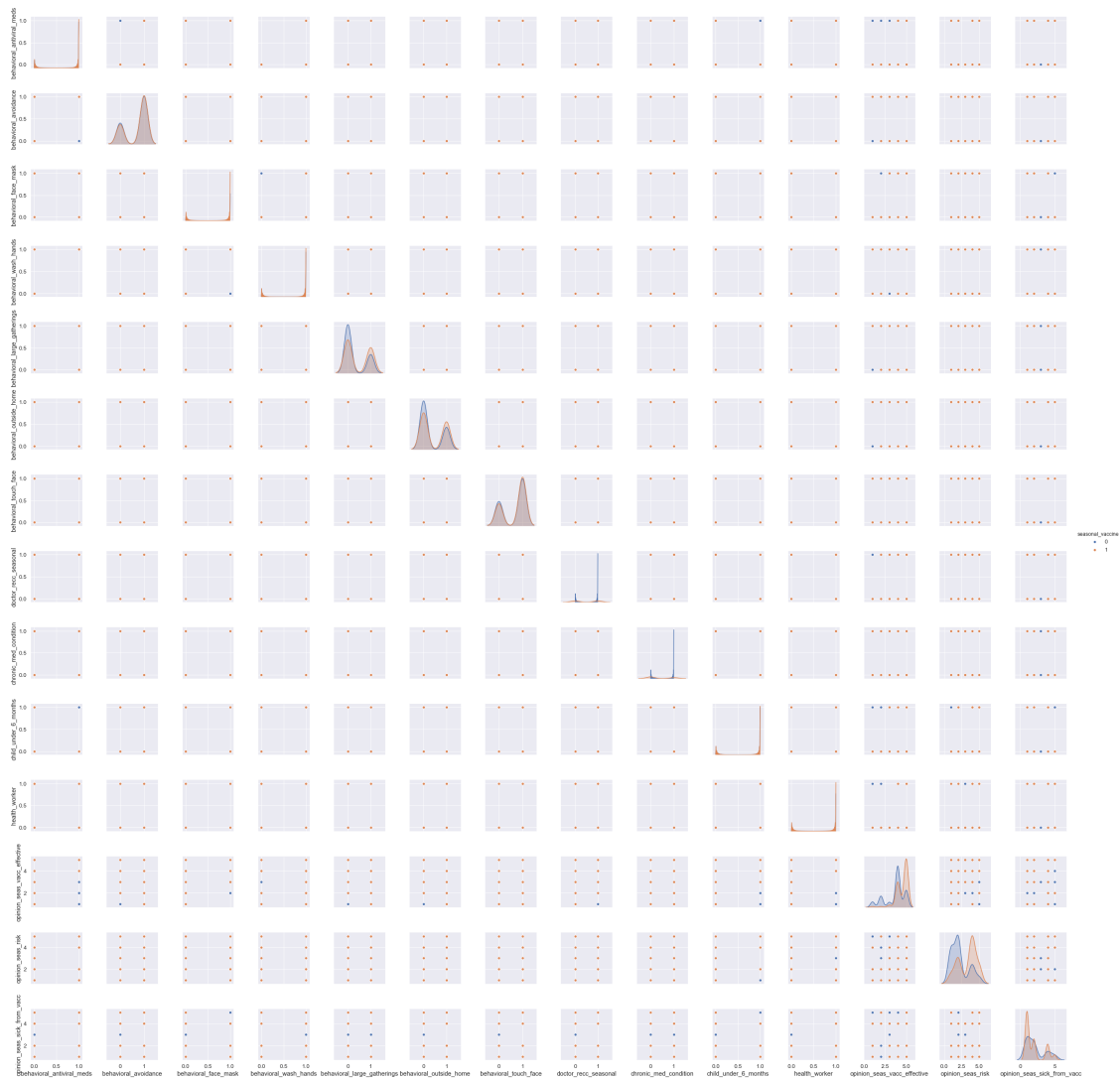
	chronic_med_condition	child_under_6_months	health_worker	\
0	0.0	0.0	0.0	
1	0.0	0.0	0.0	
2	1.0	0.0	0.0	
3	1.0	0.0	0.0	
4	0.0	0.0	0.0	
...	...	...	...	
26702	0.0	0.0	0.0	
26703	0.0	0.0	1.0	
26704	0.0	0.0	0.0	
26705	0.0	0.0	0.0	
26706	0.0	0.0	0.0	

	opinion_seas_vacc_effective	opinion_seas_risk	\
0	2.0	1.0	
1	4.0	2.0	
2	4.0	1.0	
3	5.0	4.0	
4	3.0	1.0	
...	...	...	
26702	5.0	2.0	
26703	5.0	1.0	
26704	5.0	4.0	
26705	2.0	1.0	
26706	5.0	1.0	

	opinion_seas_sick_from_vacc	seasonal_vaccine
0	2.0	0
1	4.0	1
2	2.0	0
3	1.0	1
4	4.0	0
...	...	...
26702	2.0	0
26703	1.0	0
26704	2.0	1
26705	2.0	0
26706	1.0	0

[26707 rows x 15 columns]

```
[19]: sns.pairplot(df.sample(500), hue='seasonal_vaccine')
plt.show()
```



## 6.0.5 Treat Missing Values

```
[20]: df.isnull().sum()
```

```
[20]: behavioral_antiviral_meds    71
behavioral_avoidance             208
behavioral_face_mask             19
behavioral_wash_hands            42
behavioral_large_gatherings      87
behavioral_outside_home          82
behavioral_touch_face           128
doctor_recc_seasonal            2160
chronic_med_condition           971
child_under_6_months            820
```

```

health_worker      804
opinion_seas_vacc_effective  462
opinion_seas_risk   514
opinion_seas_sick_from_vacc  537
seasonal_vaccine    0
dtype: int64

```

```
[21]: df.dropna(inplace=True)
```

```
[22]: df.isnull().sum()
```

```

[22]: behavioral_antiviral_meds      0
behavioral_avoidance                0
behavioral_face_mask                0
behavioral_wash_hands               0
behavioral_large_gatherings         0
behavioral_outside_home             0
behavioral_touch_face               0
doctor_recc_seasonal               0
chronic_med_condition              0
child_under_6_months               0
health_worker                      0
opinion_seas_vacc_effective         0
opinion_seas_risk                   0
opinion_seas_sick_from_vacc         0
seasonal_vaccine                    0
dtype: int64

```

```
[23]: df.reset_index(drop=True,inplace=True)
```

```
[24]: df
```

```

[24]:      behavioral_antiviral_meds  behavioral_avoidance  behavioral_face_mask \
0                                0.0                    0.0                    0.0
1                                0.0                    1.0                    0.0
2                                0.0                    1.0                    0.0
3                                0.0                    1.0                    0.0
4                                0.0                    1.0                    0.0
...                               ...                    ...                    ...
23183                           0.0                    0.0                    0.0
23184                           0.0                    1.0                    0.0
23185                           0.0                    1.0                    0.0
23186                           0.0                    1.0                    1.0
23187                           0.0                    1.0                    0.0

      behavioral_wash_hands  behavioral_large_gatherings \
0                        0.0                        0.0

```

1	1.0	0.0
2	1.0	1.0
3	1.0	1.0
4	1.0	0.0
...	...	...
23183	1.0	0.0
23184	0.0	0.0
23185	1.0	0.0
23186	1.0	1.0
23187	0.0	0.0

	behavioral_outside_home	behavioral_touch_face	doctor_recc_seasonal	\
0	1.0	1.0	0.0	
1	1.0	1.0	0.0	
2	0.0	0.0	1.0	
3	0.0	1.0	0.0	
4	0.0	1.0	1.0	
...	...	...	...	
23183	0.0	1.0	0.0	
23184	1.0	0.0	0.0	
23185	0.0	0.0	1.0	
23186	0.0	1.0	0.0	
23187	0.0	0.0	0.0	

	chronic_med_condition	child_under_6_months	health_worker	\
0	0.0	0.0	0.0	
1	0.0	0.0	0.0	
2	1.0	0.0	0.0	
3	0.0	0.0	0.0	
4	0.0	0.0	0.0	
...	...	...	...	
23183	0.0	1.0	0.0	
23184	0.0	0.0	0.0	
23185	0.0	0.0	1.0	
23186	0.0	0.0	0.0	
23187	0.0	0.0	0.0	

	opinion_seas_vacc_effective	opinion_seas_risk	\
0	2.0	1.0	
1	4.0	2.0	
2	5.0	4.0	
3	3.0	1.0	
4	5.0	4.0	
...	...	...	
23183	4.0	2.0	
23184	5.0	2.0	
23185	5.0	1.0	

```

23186          5.0          4.0
23187          5.0          1.0

```

```

      opinion_seas_sick_from_vacc  seasonal_vaccine
0          2.0          0
1          4.0          1
2          1.0          1
3          4.0          0
4          4.0          0
...
23183          4.0          0
23184          2.0          0
23185          1.0          0
23186          2.0          1
23187          1.0          0

```

[23188 rows x 15 columns]

```
[25]: df['seasonal_vaccine'].value_counts()
```

```

[25]: 0    12111
      1    11077
      Name: seasonal_vaccine, dtype: int64

```

```
[26]: df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 23188 entries, 0 to 23187
Data columns (total 15 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   behavioral_antiviral_meds             23188 non-null  float64
1   behavioral_avoidance                  23188 non-null  float64
2   behavioral_face_mask                  23188 non-null  float64
3   behavioral_wash_hands                 23188 non-null  float64
4   behavioral_large_gatherings           23188 non-null  float64
5   behavioral_outside_home               23188 non-null  float64
6   behavioral_touch_face                 23188 non-null  float64
7   doctor_recc_seasonal                  23188 non-null  float64
8   chronic_med_condition                 23188 non-null  float64
9   child_under_6_months                 23188 non-null  float64
10  health_worker                         23188 non-null  float64
11  opinion_seas_vacc_effective            23188 non-null  float64
12  opinion_seas_risk                      23188 non-null  float64
13  opinion_seas_sick_from_vacc            23188 non-null  float64
14  seasonal_vaccine                      23188 non-null  int64
dtypes: float64(14), int64(1)

```

memory usage: 2.7 MB

```
[27]: df = df.astype('int8') #Change to integer type
```

```
[28]: df.dtypes
```

```
[28]: behavioral_antiviral_meds      int8
      behavioral_avoidance          int8
      behavioral_face_mask          int8
      behavioral_wash_hands          int8
      behavioral_large_gatherings    int8
      behavioral_outside_home        int8
      behavioral_touch_face          int8
      doctor_recc_seasonal          int8
      chronic_med_condition          int8
      child_under_6_months          int8
      health_worker                  int8
      opinion_seas_vacc_effective     int8
      opinion_seas_risk               int8
      opinion_seas_sick_from_vacc     int8
      seasonal_vaccine               int8
      dtype: object
```

```
[29]: df
```

```
[29]:
```

	behavioral_antiviral_meds	behavioral_avoidance	behavioral_face_mask	\
0	0	0	0	
1	0	1	0	
2	0	1	0	
3	0	1	0	
4	0	1	0	
...	...	...	...	
23183	0	0	0	
23184	0	1	0	
23185	0	1	0	
23186	0	1	1	
23187	0	1	0	

	behavioral_wash_hands	behavioral_large_gatherings	\
0	0	0	
1	1	0	
2	1	1	
3	1	1	
4	1	0	
...	...	...	
23183	1	0	
23184	0	0	



23185	1	0
23186	1	1
23187	0	0

	behavioral_outside_home	behavioral_touch_face	doctor_recc_seasonal	\
0	1	1	0	
1	1	1	0	
2	0	0	1	
3	0	1	0	
4	0	1	1	
...	...	...	...	
23183	0	1	0	
23184	1	0	0	
23185	0	0	1	
23186	0	1	0	
23187	0	0	0	

	chronic_med_condition	child_under_6_months	health_worker	\
0	0	0	0	
1	0	0	0	
2	1	0	0	
3	0	0	0	
4	0	0	0	
...	...	...	...	
23183	0	1	0	
23184	0	0	0	
23185	0	0	1	
23186	0	0	0	
23187	0	0	0	

	opinion_seas_vacc_effective	opinion_seas_risk	\
0	2	1	
1	4	2	
2	5	4	
3	3	1	
4	5	4	
...	...	...	
23183	4	2	
23184	5	2	
23185	5	1	
23186	5	4	
23187	5	1	

	opinion_seas_sick_from_vacc	seasonal_vaccine
0	2	0
1	4	1
2	1	1

3	4	0
4	4	0
...	...	...
23183	4	0
23184	2	0
23185	1	0
23186	2	1
23187	1	0

[23188 rows x 15 columns]

```
[30]: df.describe()
```

```
[30]:
```

	behavioral_antiviral_meds	behavioral_avoidance	behavioral_face_mask \
count	23188.000000	23188.000000	23188.000000
mean	0.049336	0.731197	0.068139
std	0.216573	0.443347	0.251989
min	0.000000	0.000000	0.000000
25%	0.000000	0.000000	0.000000
50%	0.000000	1.000000	0.000000
75%	0.000000	1.000000	0.000000
max	1.000000	1.000000	1.000000

	behavioral_wash_hands	behavioral_large_gatherings \
count	23188.000000	23188.000000
mean	0.829481	0.358289
std	0.376096	0.479508
min	0.000000	0.000000
25%	1.000000	0.000000
50%	1.000000	0.000000
75%	1.000000	1.000000
max	1.000000	1.000000

	behavioral_outside_home	behavioral_touch_face	doctor_recc_seasonal \
count	23188.000000	23188.000000	23188.000000
mean	0.336855	0.684319	0.331335
std	0.472645	0.464796	0.470703
min	0.000000	0.000000	0.000000
25%	0.000000	0.000000	0.000000
50%	0.000000	1.000000	0.000000
75%	1.000000	1.000000	1.000000
max	1.000000	1.000000	1.000000

	chronic_med_condition	child_under_6_months	health_worker \
count	23188.000000	23188.000000	23188.000000
mean	0.284199	0.084009	0.113723
std	0.451042	0.277407	0.317481

min	0.000000	0.000000	0.000000
25%	0.000000	0.000000	0.000000
50%	0.000000	0.000000	0.000000
75%	1.000000	0.000000	0.000000
max	1.000000	1.000000	1.000000

	opinion_seas_vacc_effective	opinion_seas_risk \
count	23188.000000	23188.000000
mean	4.038166	2.730680
std	1.078839	1.388191
min	1.000000	1.000000
25%	4.000000	2.000000
50%	4.000000	2.000000
75%	5.000000	4.000000
max	5.000000	5.000000

	opinion_seas_sick_from_vacc	seasonal_vaccine
count	23188.000000	23188.000000
mean	2.115577	0.477704
std	1.332636	0.499513
min	1.000000	0.000000
25%	1.000000	0.000000
50%	2.000000	0.000000
75%	4.000000	1.000000
max	5.000000	1.000000

```
[31]: df['opinion_seas_vacc_effective'] = df['opinion_seas_vacc_effective'].
      ↪astype('object')
```

```
[32]: df['opinion_seas_risk'] = df['opinion_seas_risk'].astype('object')
```

```
[33]: df['opinion_seas_sick_from_vacc'] = df['opinion_seas_sick_from_vacc'].
      ↪astype('object')
```

```
[34]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 23188 entries, 0 to 23187
Data columns (total 15 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   behavioral_antiviral_meds             23188 non-null  int8
1   behavioral_avoidance                  23188 non-null  int8
2   behavioral_face_mask                  23188 non-null  int8
3   behavioral_wash_hands                  23188 non-null  int8
4   behavioral_large_gatherings            23188 non-null  int8
5   behavioral_outside_home                23188 non-null  int8
```

```

6   behavioral_touch_face      23188 non-null  int8
7   doctor_recc_seasonal      23188 non-null  int8
8   chronic_med_condition     23188 non-null  int8
9   child_under_6_months      23188 non-null  int8
10  health_worker             23188 non-null  int8
11  opinion_seas_vacc_effective 23188 non-null  object
12  opinion_seas_risk           23188 non-null  object
13  opinion_seas_sick_from_vacc 23188 non-null  object
14  seasonal_vaccine           23188 non-null  int8
dtypes: int8(12), object(3)
memory usage: 815.3+ KB

```

### 6.0.6 Create dummy variables

```
[35]: df2 = pd.get_dummies(data=df, drop_first=True)
```

```
[36]: df2
```

```

[36]:      behavioral_antiviral_meds  behavioral_avoidance  behavioral_face_mask  \
0                                0                      0                      0
1                                0                      1                      0
2                                0                      1                      0
3                                0                      1                      0
4                                0                      1                      0
...                               ...                      ...
23183                           0                      0                      0
23184                           0                      1                      0
23185                           0                      1                      0
23186                           0                      1                      1
23187                           0                      1                      0

      behavioral_wash_hands  behavioral_large_gatherings  \
0                          0                          0
1                          1                          0
2                          1                          1
3                          1                          1
4                          1                          0
...                         ...                         ...
23183                       1                          0
23184                       0                          0
23185                       1                          0
23186                       1                          1
23187                       0                          0

      behavioral_outside_home  behavioral_touch_face  doctor_recc_seasonal  \
0                            1                      1                      0
1                            1                      1                      0

```

2	0	0	1
3	0	1	0
4	0	1	1
...	...	...	...
23183	0	1	0
23184	1	0	0
23185	0	0	1
23186	0	1	0
23187	0	0	0

	chronic_med_condition	child_under_6_months	health_worker	\
0	0	0	0	
1	0	0	0	
2	1	0	0	
3	0	0	0	
4	0	0	0	
...	...	...	...	
23183	0	1	0	
23184	0	0	0	
23185	0	0	1	
23186	0	0	0	
23187	0	0	0	

	seasonal_vaccine	opinion_seas_vacc_effective_2	\
0	0	1	
1	1	0	
2	1	0	
3	0	0	
4	0	0	
...	...	...	
23183	0	0	
23184	0	0	
23185	0	0	
23186	1	0	
23187	0	0	

	opinion_seas_vacc_effective_3	opinion_seas_vacc_effective_4	\
0	0	0	
1	0	1	
2	0	0	
3	1	0	
4	0	0	
...	...	...	
23183	0	1	
23184	0	0	
23185	0	0	
23186	0	0	

23187	0	0
-------	---	---

	opinion_seas_vacc_effective_5	opinion_seas_risk_2 \
0	0	0
1	0	1
2	1	0
3	0	0
4	1	0
...	...	...
23183	0	1
23184	1	1
23185	1	0
23186	1	0
23187	1	0

	opinion_seas_risk_3	opinion_seas_risk_4	opinion_seas_risk_5 \
0	0	0	0
1	0	0	0
2	0	1	0
3	0	0	0
4	0	1	0
...	...	...	...
23183	0	0	0
23184	0	0	0
23185	0	0	0
23186	0	1	0
23187	0	0	0

	opinion_seas_sick_from_vacc_2	opinion_seas_sick_from_vacc_3 \
0	1	0
1	0	0
2	0	0
3	0	0
4	0	0
...	...	...
23183	0	0
23184	1	0
23185	0	0
23186	1	0
23187	0	0

	opinion_seas_sick_from_vacc_4	opinion_seas_sick_from_vacc_5
0	0	0
1	1	0
2	0	0
3	1	0
4	1	0

```

...
23183          1          0
23184          0          0
23185          0          0
23186          0          0
23187          0          0

```

[23188 rows x 24 columns]

```
[37]: df2.columns
```

```
[37]: Index(['behavioral_antiviral_meds', 'behavioral_avoidance',
'behavioral_face_mask', 'behavioral_wash_hands',
'behavioral_large_gatherings', 'behavioral_outside_home',
'behavioral_touch_face', 'doctor_recc_seasonal',
'chronic_med_condition', 'child_under_6_months', 'health_worker',
'seasonal_vaccine', 'opinion_seas_vacc_effective_2',
'opinion_seas_vacc_effective_3', 'opinion_seas_vacc_effective_4',
'opinion_seas_vacc_effective_5', 'opinion_seas_risk_2',
'opinion_seas_risk_3', 'opinion_seas_risk_4', 'opinion_seas_risk_5',
'opinion_seas_sick_from_vacc_2', 'opinion_seas_sick_from_vacc_3',
'opinion_seas_sick_from_vacc_4', 'opinion_seas_sick_from_vacc_5'],
dtype='object')
```

```
[38]: df2 = df2[['behavioral_antiviral_meds', 'behavioral_avoidance',
'behavioral_face_mask', 'behavioral_wash_hands',
'behavioral_large_gatherings', 'behavioral_outside_home',
'behavioral_touch_face', 'doctor_recc_seasonal',
'chronic_med_condition', 'child_under_6_months', 'health_worker',
'opinion_seas_vacc_effective_2',
'opinion_seas_vacc_effective_3', 'opinion_seas_vacc_effective_4',
'opinion_seas_vacc_effective_5', 'opinion_seas_risk_2',
'opinion_seas_risk_3', 'opinion_seas_risk_4', 'opinion_seas_risk_5',
'opinion_seas_sick_from_vacc_2', 'opinion_seas_sick_from_vacc_3',
'opinion_seas_sick_from_vacc_4'],
→ 'opinion_seas_sick_from_vacc_5', 'seasonal_vaccine' ]]
```

```
[39]: df2
```

```
[39]: behavioral_antiviral_meds behavioral_avoidance behavioral_face_mask \
0          0          0          0
1          0          1          0
2          0          1          0
3          0          1          0
4          0          1          0
...          ...          ...
23183          0          0          0
```

23184	0	1	0
23185	0	1	0
23186	0	1	1
23187	0	1	0

	behavioral_wash_hands	behavioral_large_gatherings	\
0	0	0	
1	1	0	
2	1	1	
3	1	1	
4	1	0	
...	...	...	
23183	1	0	
23184	0	0	
23185	1	0	
23186	1	1	
23187	0	0	

	behavioral_outside_home	behavioral_touch_face	doctor_recc_seasonal	\
0	1	1	0	
1	1	1	0	
2	0	0	1	
3	0	1	0	
4	0	1	1	
...	...	...	...	
23183	0	1	0	
23184	1	0	0	
23185	0	0	1	
23186	0	1	0	
23187	0	0	0	

	chronic_med_condition	child_under_6_months	health_worker	\
0	0	0	0	
1	0	0	0	
2	1	0	0	
3	0	0	0	
4	0	0	0	
...	...	...	...	
23183	0	1	0	
23184	0	0	0	
23185	0	0	1	
23186	0	0	0	
23187	0	0	0	

	opinion_seas_vacc_effective_2	opinion_seas_vacc_effective_3	\
0	1	0	
1	0	0	



2	0	0
3	0	1
4	0	0
...	...	...
23183	0	0
23184	0	0
23185	0	0
23186	0	0
23187	0	0

	opinion_seas_vacc_effective_4	opinion_seas_vacc_effective_5	\
0	0	0	
1	1	0	
2	0	1	
3	0	0	
4	0	1	
...	...	...	
23183	1	0	
23184	0	1	
23185	0	1	
23186	0	1	
23187	0	1	

	opinion_seas_risk_2	opinion_seas_risk_3	opinion_seas_risk_4	\
0	0	0	0	
1	1	0	0	
2	0	0	1	
3	0	0	0	
4	0	0	1	
...	...	...	...	
23183	1	0	0	
23184	1	0	0	
23185	0	0	0	
23186	0	0	1	
23187	0	0	0	

	opinion_seas_risk_5	opinion_seas_sick_from_vacc_2	\
0	0	1	
1	0	0	
2	0	0	
3	0	0	
4	0	0	
...	...	...	
23183	0	0	
23184	0	1	
23185	0	0	
23186	0	1	

23187	0	0
	opinion_seas_sick_from_vacc_3	opinion_seas_sick_from_vacc_4 \
0	0	0
1	0	1
2	0	0
3	0	1
4	0	1
...	...	...
23183	0	1
23184	0	0
23185	0	0
23186	0	0
23187	0	0
	opinion_seas_sick_from_vacc_5	seasonal_vaccine
0	0	0
1	0	1
2	0	1
3	0	0
4	0	0
...	...	...
23183	0	0
23184	0	0
23185	0	0
23186	0	1
23187	0	0

[23188 rows x 24 columns]

### 6.0.7 Create and save processed dataset

```
[40]: df2.to_csv("train.csv", index=False)
```

```
[ ]:
```

```
[41]: df = pd.read_csv("train.csv")
```

```
[42]: df
```

[42]:	behavioral_antiviral_meds	behavioral_avoidance	behavioral_face_mask \
0	0	0	0
1	0	1	0
2	0	1	0
3	0	1	0
4	0	1	0
...	...	...	...

23183	0	0	0
23184	0	1	0
23185	0	1	0
23186	0	1	1
23187	0	1	0

	behavioral_wash_hands	behavioral_large_gatherings	\
0	0	0	
1	1	0	
2	1	1	
3	1	1	
4	1	0	
...	...	...	
23183	1	0	
23184	0	0	
23185	1	0	
23186	1	1	
23187	0	0	

	behavioral_outside_home	behavioral_touch_face	doctor_recc_seasonal	\
0	1	1	0	
1	1	1	0	
2	0	0	1	
3	0	1	0	
4	0	1	1	
...	...	...	...	
23183	0	1	0	
23184	1	0	0	
23185	0	0	1	
23186	0	1	0	
23187	0	0	0	

	chronic_med_condition	child_under_6_months	health_worker	\
0	0	0	0	
1	0	0	0	
2	1	0	0	
3	0	0	0	
4	0	0	0	
...	...	...	...	
23183	0	1	0	
23184	0	0	0	
23185	0	0	1	
23186	0	0	0	
23187	0	0	0	

	opinion_seas_vacc_effective_2	opinion_seas_vacc_effective_3	\
0	1	0	

1	0	0
2	0	0
3	0	1
4	0	0
...	...	...
23183	0	0
23184	0	0
23185	0	0
23186	0	0
23187	0	0

	opinion_seas_vacc_effective_4	opinion_seas_vacc_effective_5 \
0	0	0
1	1	0
2	0	1
3	0	0
4	0	1
...	...	...
23183	1	0
23184	0	1
23185	0	1
23186	0	1
23187	0	1

	opinion_seas_risk_2	opinion_seas_risk_3	opinion_seas_risk_4 \
0	0	0	0
1	1	0	0
2	0	0	1
3	0	0	0
4	0	0	1
...	...	...	...
23183	1	0	0
23184	1	0	0
23185	0	0	0
23186	0	0	1
23187	0	0	0

	opinion_seas_risk_5	opinion_seas_sick_from_vacc_2 \
0	0	1
1	0	0
2	0	0
3	0	0
4	0	0
...	...	...
23183	0	0
23184	0	1
23185	0	0

23186	0	1
23187	0	0

	opinion_seas_sick_from_vacc_3	opinion_seas_sick_from_vacc_4 \
0	0	0
1	0	1
2	0	0
3	0	1
4	0	1
...	...	...
23183	0	1
23184	0	0
23185	0	0
23186	0	0
23187	0	0

	opinion_seas_sick_from_vacc_5	seasonal_vaccine
0	0	0
1	0	1
2	0	1
3	0	0
4	0	0
...	...	...
23183	0	0
23184	0	0
23185	0	0
23186	0	1
23187	0	0

[23188 rows x 24 columns]

```
[43]: df.shape
```

```
[43]: (23188, 24)
```

- 7 Summary of training at least three different classifier models, preferably of different nature in explainability and predictability. For example, you can start with a simple logistic regression as a baseline, adding other models or ensemble models. Preferably, all your models use the same training and test splits, or the same cross-validation method.

#### 7.0.1 Train Test Split

```
[44]: X = df.iloc[:,0:23]
      y = df.iloc[:,23]
```

```
[45]: X.values
```

```
[45]: array([[0, 0, 0, ..., 0, 0, 0],
            [0, 1, 0, ..., 0, 1, 0],
            [0, 1, 0, ..., 0, 0, 0],
            ...,
            [0, 1, 0, ..., 0, 0, 0],
            [0, 1, 1, ..., 0, 0, 0],
            [0, 1, 0, ..., 0, 0, 0]], dtype=int64)
```

```
[46]: y.values
```

```
[46]: array([0, 1, 1, ..., 0, 1, 0], dtype=int64)
```

```
[47]: X_train, X_test, y_train, y_test = train_test_split(X.values, y.values,
      ↪test_size=0.2, random_state=123, stratify=y)
```

```
[48]: X_train
```

```
[48]: array([[0, 1, 0, ..., 0, 0, 0],
            [0, 1, 0, ..., 0, 0, 0],
            [0, 1, 0, ..., 0, 0, 0],
            ...,
            [0, 1, 1, ..., 0, 0, 0],
            [0, 0, 0, ..., 0, 0, 0],
            [0, 1, 0, ..., 0, 0, 0]], dtype=int64)
```

```
[49]: X_test
```

```
[49]: array([[0, 1, 0, ..., 0, 0, 0],
            [0, 1, 0, ..., 0, 0, 0],
            [0, 0, 0, ..., 0, 0, 0],
            ...,
            [0, 0, 0, ..., 0, 0, 0],
```

```
[1, 0, 0, ..., 0, 0, 0],  
[0, 1, 0, ..., 0, 0, 0]], dtype=int64)
```

```
[50]: y_train
```

```
[50]: array([1, 0, 1, ..., 1, 1, 1], dtype=int64)
```

### 7.0.2 Logistic Regression

```
[51]: lr = LogisticRegression(random_state=123)
```

```
[52]: lr.fit(X_train, y_train)
```

```
[52]: LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,  
    intercept_scaling=1, l1_ratio=None, max_iter=100,  
    multi_class='auto', n_jobs=None, penalty='l2',  
    random_state=123, solver='lbfgs', tol=0.0001, verbose=0,  
    warm_start=False)
```

```
[53]: lr.coef_
```

```
[53]: array([[ -0.23062336, -0.07794305, -0.00527337,  0.09815467,  0.00333909,  
    -0.07933634,  0.27948713,  1.36430408,  0.35155097, -0.16748045,  
    0.79348647, -0.29698112,  0.6244859 ,  0.75495696,  1.77721418,  
    0.80614637,  1.6792209 ,  1.70246871,  2.0063813 , -0.44950573,  
    -1.62206222, -0.67756796, -1.24302248]])
```

```
[54]: lr.intercept_
```

```
[54]: array([-2.63615376])
```

```
[55]: ypred_lr = lr.predict(X_test)
```

```
[56]: y_test[:10]
```

```
[56]: array([1, 0, 1, 0, 1, 1, 1, 0, 1, 1], dtype=int64)
```

```
[57]: ypred_lr[:10]
```

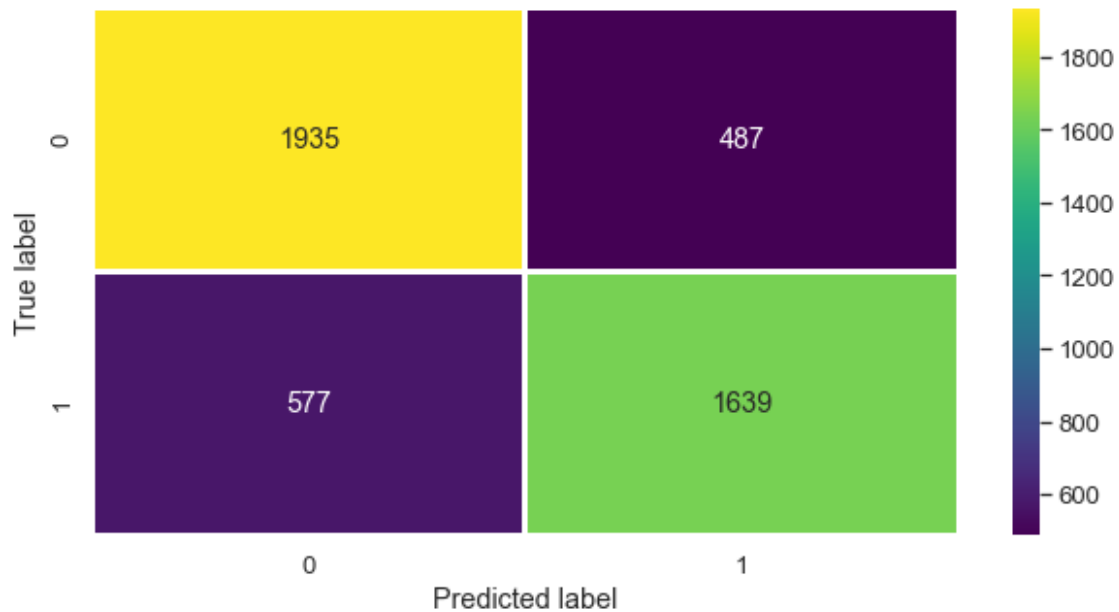
```
[57]: array([1, 0, 0, 0, 1, 1, 1, 0, 0, 1], dtype=int64)
```

### 7.0.3 Logistic Regression Model Evaluation

```
[58]: cm = confusion_matrix(y_test, ypred_lr)  
cm
```

```
[58]: array([[1935,  487],
           [ 577, 1639]], dtype=int64)
```

```
[59]: fig , ax = plt.subplots(figsize=(10,5))
sns.heatmap(cm, annot=True,fmt='.4g',linewidths=2, cmap='viridis')
plt.ylabel('True label')
plt.xlabel('Predicted label')
plt.show()
```

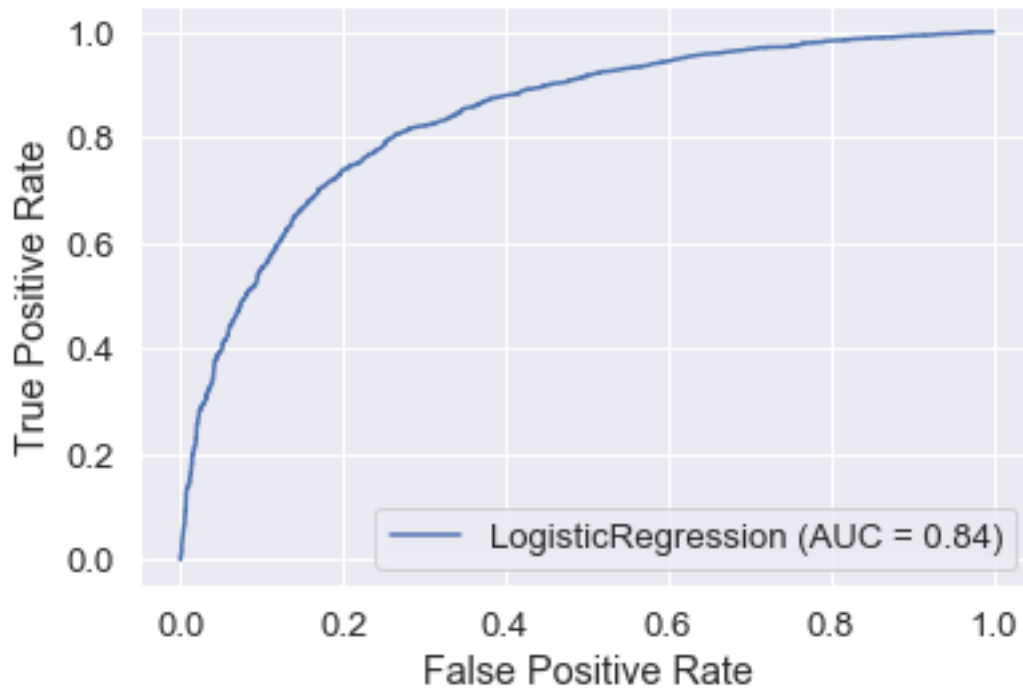


```
[60]: print(classification_report(y_test,ypred_lr))
```

	precision	recall	f1-score	support
0	0.77	0.80	0.78	2422
1	0.77	0.74	0.75	2216
accuracy			0.77	4638
macro avg	0.77	0.77	0.77	4638
weighted avg	0.77	0.77	0.77	4638

```
[61]: plot_roc_curve(lr,X_test,y_test)
plt.show()
```





```
[62]: accuracy_score(y_test,ypred_lr)
```

```
[62]: 0.7705907718844329
```

#### 7.0.4 Random Forest Classifier

```
[63]: rf = RandomForestClassifier(random_state=123)
```

```
[64]: rf.fit(X_train, y_train)
```

```
[64]: RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None,
                             criterion='gini', max_depth=None, max_features='auto',
                             max_leaf_nodes=None, max_samples=None,
                             min_impurity_decrease=0.0, min_impurity_split=None,
                             min_samples_leaf=1, min_samples_split=2,
                             min_weight_fraction_leaf=0.0, n_estimators=100,
                             n_jobs=None, oob_score=False, random_state=123,
                             verbose=0, warm_start=False)
```

```
[65]: ypred_rf = rf.predict(X_test)
```

```
[66]: y_test[:10]
```

```
[66]: array([1, 0, 1, 0, 1, 1, 1, 0, 1, 1], dtype=int64)
```

```
[67]: ypred_rf[:10]
```

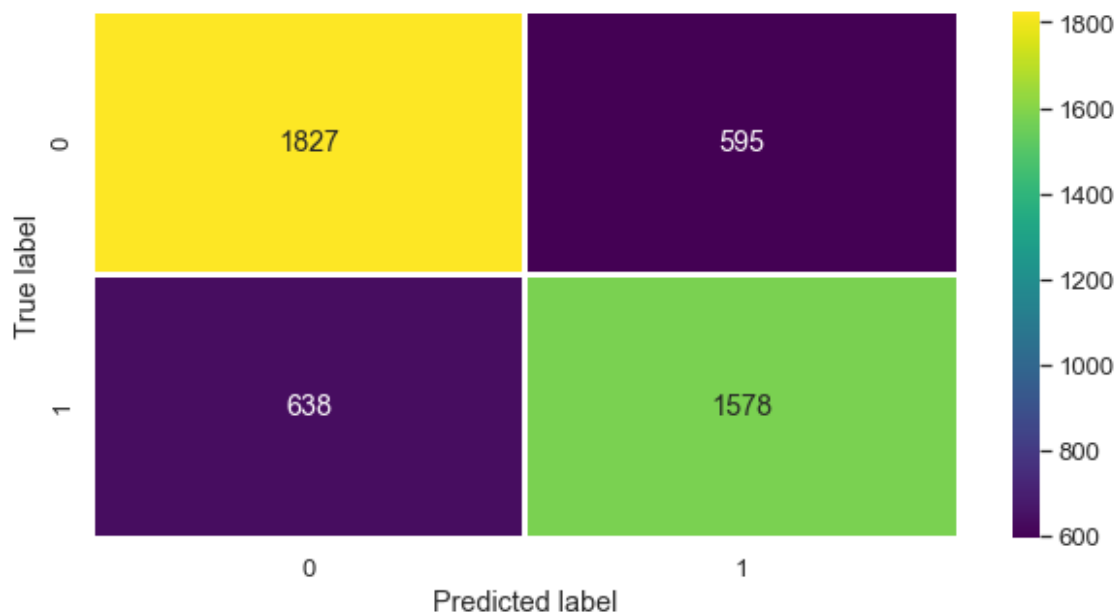
```
[67]: array([1, 0, 0, 0, 1, 1, 1, 1, 0, 1], dtype=int64)
```

### 7.0.5 Random Forest Model Evaluation

```
[68]: cm = confusion_matrix(y_test,ypred_rf)
cm
```

```
[68]: array([[1827,  595],
          [ 638, 1578]], dtype=int64)
```

```
[69]: fig , ax = plt.subplots(figsize=(10,5))
sns.heatmap(cm, annot=True,fmt='.4g',linewidths=2, cmap='viridis')
plt.ylabel('True label')
plt.xlabel('Predicted label')
plt.show()
```

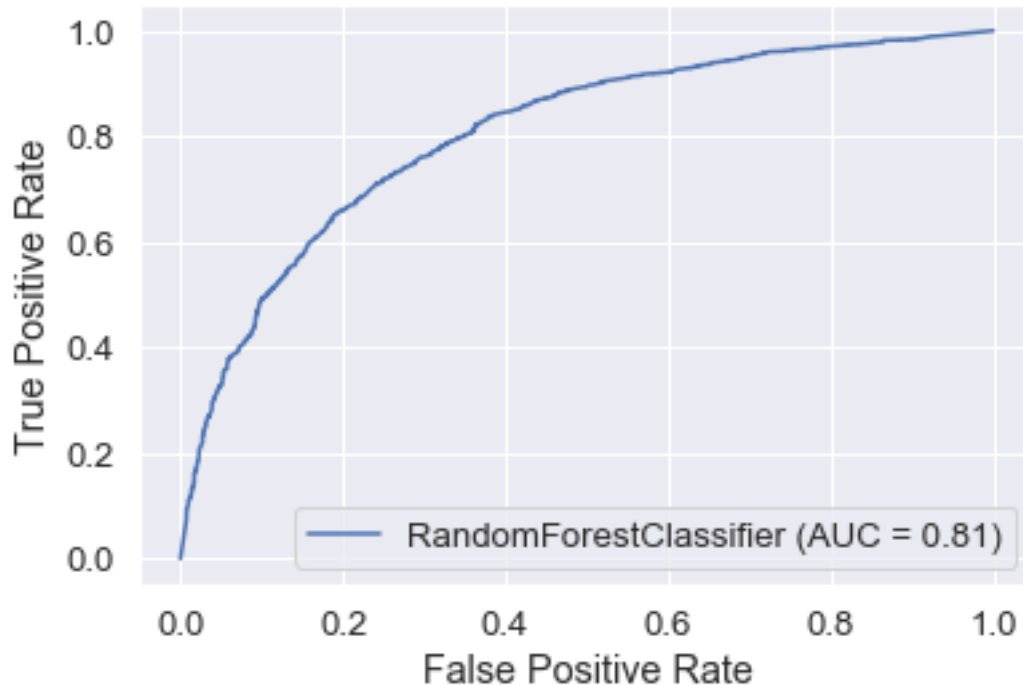


```
[70]: print(classification_report(y_test,ypred_rf))
```

	precision	recall	f1-score	support
0	0.74	0.75	0.75	2422
1	0.73	0.71	0.72	2216
accuracy			0.73	4638
macro avg	0.73	0.73	0.73	4638

weighted avg      0.73      0.73      0.73      4638

```
[71]: plot_roc_curve(rf,X_test,y_test)
      plt.show()
```



```
[72]: accuracy_score(y_test,ypred_rf)
```

```
[72]: 0.7341526520051747
```

### 7.0.6 Gradient Boosting Classifier

```
[73]: gbc = GradientBoostingClassifier(random_state=123)
```

```
[74]: gbc.fit(X_train,y_train)
```

```
[74]: GradientBoostingClassifier(ccp_alpha=0.0, criterion='friedman_mse', init=None,
                                learning_rate=0.1, loss='deviance', max_depth=3,
                                max_features=None, max_leaf_nodes=None,
                                min_impurity_decrease=0.0, min_impurity_split=None,
                                min_samples_leaf=1, min_samples_split=2,
                                min_weight_fraction_leaf=0.0, n_estimators=100,
                                n_iter_no_change=None, presort='deprecated',
                                random_state=123, subsample=1.0, tol=0.0001,
```

```
validation_fraction=0.1, verbose=0,  
warm_start=False)
```

```
[75]: ypredgbc = gbc.predict(X_test)
```

```
[76]: y_test[:10]
```

```
[76]: array([1, 0, 1, 0, 1, 1, 1, 0, 1, 1], dtype=int64)
```

```
[77]: ypredgbc[:10]
```

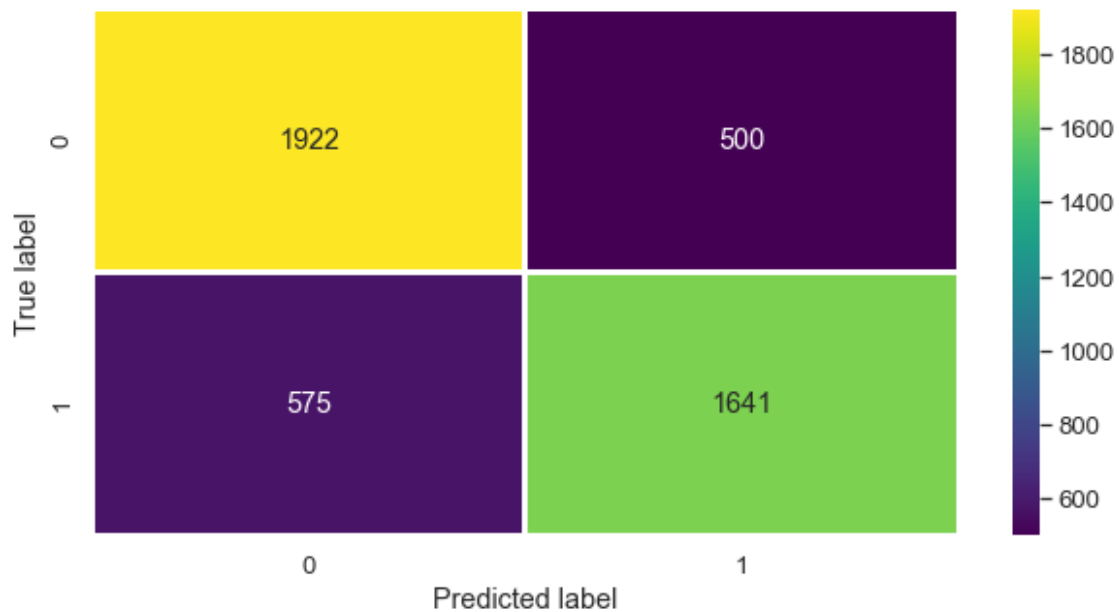
```
[77]: array([1, 0, 0, 0, 1, 1, 1, 1, 0, 1], dtype=int64)
```

### 7.0.7 Gradient Boosting Model Evaluation

```
[78]: cm = confusion_matrix(y_test,ypredgbc)  
cm
```

```
[78]: array([[1922,  500],  
          [ 575, 1641]], dtype=int64)
```

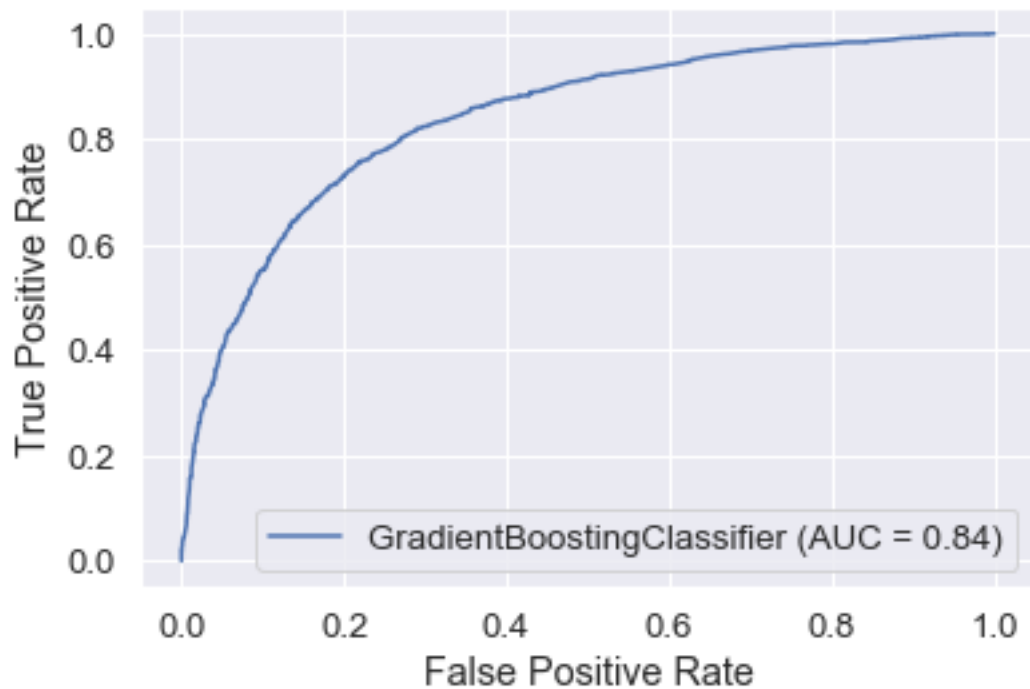
```
[79]: fig , ax = plt.subplots(figsize=(10,5))  
sns.heatmap(cm, annot=True,fmt='.4g',linewidths=2, cmap='viridis')  
plt.ylabel('True label')  
plt.xlabel('Predicted label')  
plt.show()
```



```
[80]: print(classification_report(y_test,ypredgbc))
```

	precision	recall	f1-score	support
0	0.77	0.79	0.78	2422
1	0.77	0.74	0.75	2216
accuracy			0.77	4638
macro avg	0.77	0.77	0.77	4638
weighted avg	0.77	0.77	0.77	4638

```
[81]: plot_roc_curve(gbc,X_test,y_test)
plt.show()
```



```
[82]: accuracy_score(y_test,ypredgbc)
```

```
[82]: 0.7682190599396291
```

## **8 A paragraph explaining which of your classifier models you recommend as a final model that best fits your needs in terms of accuracy and explainability.**

Logistic Regression gives us the best accuracy and F1 score. Therefore it is recommended.

## **9 Summary Key Findings and Insights, which walks your reader through the main drivers of your model and insights from your data derived from your classifier model.**

The features we selected gave us a decent accuracy and good result. The result differences are small and we select Logistic Regression because it's a simple model.

## **10 Suggestions for next steps in analyzing this data, which may include suggesting revisiting this model adding specific data features to achieve a better explanation or a better prediction.**

For features that are biased, we need to gather more data and made equal values for race, sex, income etc. We have to ensure the model we developed stays bias free.

We can also explore other models like decision tree, support vector machine, KNN classifiers model to see if they can able to analyse the data patterns to give better predictions. We also can adjust hyperparameters for each model to get better results.