

Retinopathy Disease Prediction using Machine Learning

Understanding **Diabetic Retinopathy** Symptoms, Causes, Treatment and Prevention



```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

import warnings
warnings.filterwarnings('ignore')

df = pd.read_csv('retinopathy.csv')
df.head()
```

	Unnamed: 0	age	systolic_blood_pressure	diastolic_blood_pressure
0	0	77.196340	85.288742	80.021878
1	1	63.529850	99.379736	84.852361
2	2	69.003986	111.349455	109.850616
3	3	82.638210	95.056128	79.666851
4	4	78.346286	109.154591	90.713220

	cholesterol	has_retinopathy
0	79.957109	1
1	110.382411	1

```

2    100.828246    1
3     87.066303    1
4     92.511770    1

df.tail()

   Unnamed: 0    age  systolic_blood_pressure  \
5995    5995  49.611850           94.857639
5996    5996  63.099686          100.039527
5997    5997  55.562243           98.421446
5998    5998  63.468956          106.809289
5999    5999  62.506825           96.900784

   diastolic_blood_pressure  cholesterol  has_retinopathy
5995           86.615671    107.643986             0
5996           93.515186    104.971404             1
5997          102.697875    120.875951             1
5998           88.060631    106.052213             1
5999           86.878033    108.625436             0

df = df.drop('Unnamed: 0', axis = 1)

df.shape

(6000, 5)

df.columns

Index(['age', 'systolic_blood_pressure', 'diastolic_blood_pressure',
      'cholesterol', 'has_retinopathy'],
      dtype='object')

```

1. Data Overview

- **Features:**
 - **Age:** Continuous variable representing the age of the patient.
 - **Systolic Blood Pressure:** Continuous variable for the pressure in the arteries when the heart beats.
 - **Diastolic Blood Pressure:** Continuous variable for the pressure in the arteries when the heart rests between beats.
 - **Cholesterol:** Continuous variable for the cholesterol level.
- **Target:**
 - **Has Retinopathy:** Binary variable indicating the presence (1) or absence (0) of retinopathy.

```

df.duplicated().sum()

0

df.isnull().sum()

```

```
age 0
systolic_blood_pressure 0
diastolic_blood_pressure 0
cholesterol 0
has_retinopathy 0
dtype: int64
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 6000 entries, 0 to 5999
```

```
Data columns (total 5 columns):
```

#	Column	Non-Null Count	Dtype
0	age	6000 non-null	float64
1	systolic_blood_pressure	6000 non-null	float64
2	diastolic_blood_pressure	6000 non-null	float64
3	cholesterol	6000 non-null	float64
4	has_retinopathy	6000 non-null	int64

```
dtypes: float64(4), int64(1)
```

```
memory usage: 234.5 KB
```

```
df.describe()
```

	age	systolic_blood_pressure	diastolic_blood_pressure
count	6000.000000	6000.000000	6000.000000
mean	60.464121	100.694822	90.505547
std	8.564392	10.669267	9.648200
min	35.164761	69.675429	62.807105
25%	54.371941	93.267420	83.641788
50%	59.831159	100.119926	89.912429
75%	65.809652	107.439501	96.682405
max	103.279497	151.699660	133.456382

	cholesterol	has_retinopathy
count	6000.000000	6000.000000
mean	100.628255	0.514500
std	10.433915	0.499831
min	69.967453	0.000000
25%	93.202373	0.000000
50%	100.060637	1.000000

75%	107.250829	1.000000
max	148.233544	1.000000

```
df.nunique()
```

age	6000
systolic_blood_pressure	6000
diastolic_blood_pressure	6000
cholesterol	6000
has_retinopathy	2

```
dtype: int64
```

```
fig, axes = plt.subplots(2, 2, figsize=(15, 10))
```

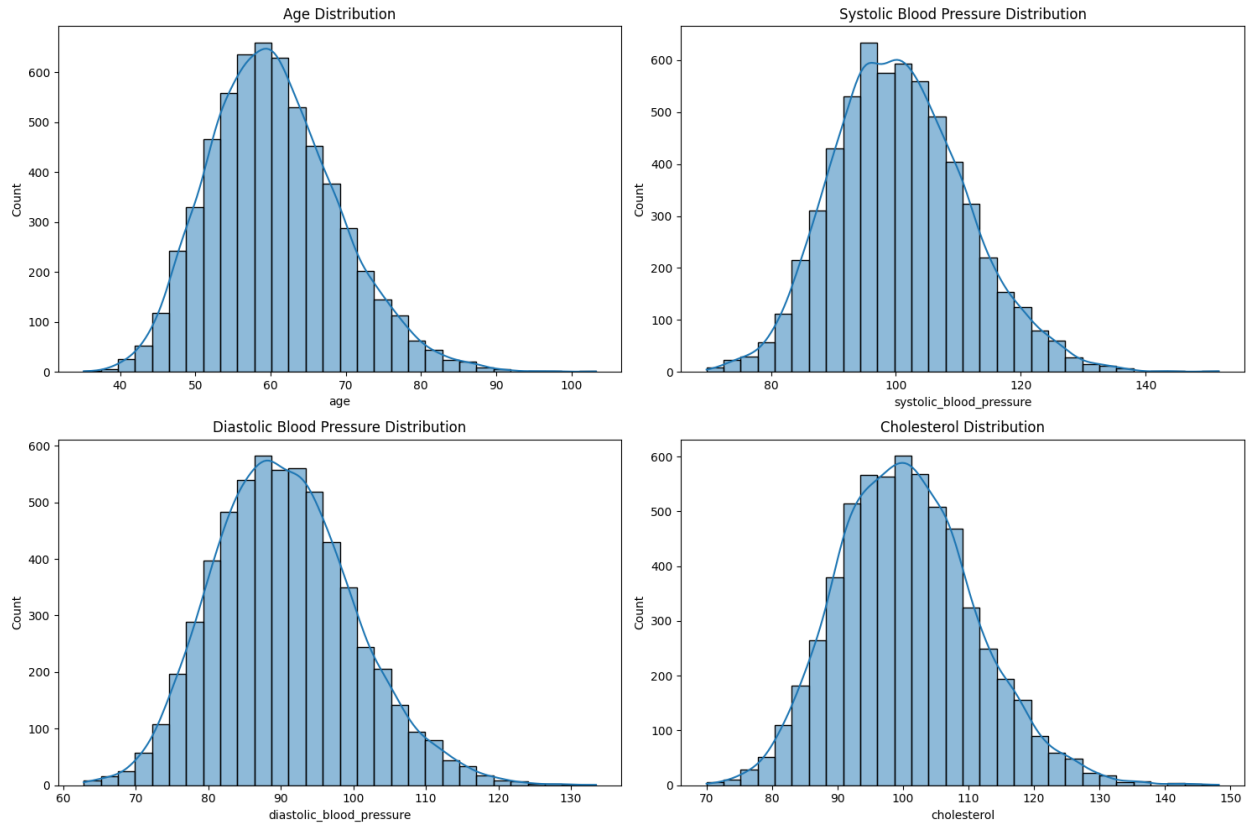
```
sns.histplot(df['age'], bins=30, kde=True, ax=axes[0, 0])  
axes[0, 0].set_title('Age Distribution')
```

```
sns.histplot(df['systolic_blood_pressure'], bins=30, kde=True,  
ax=axes[0, 1])  
axes[0, 1].set_title('Systolic Blood Pressure Distribution')
```

```
sns.histplot(df['diastolic_blood_pressure'], bins=30, kde=True,  
ax=axes[1, 0])  
axes[1, 0].set_title('Diastolic Blood Pressure Distribution')
```

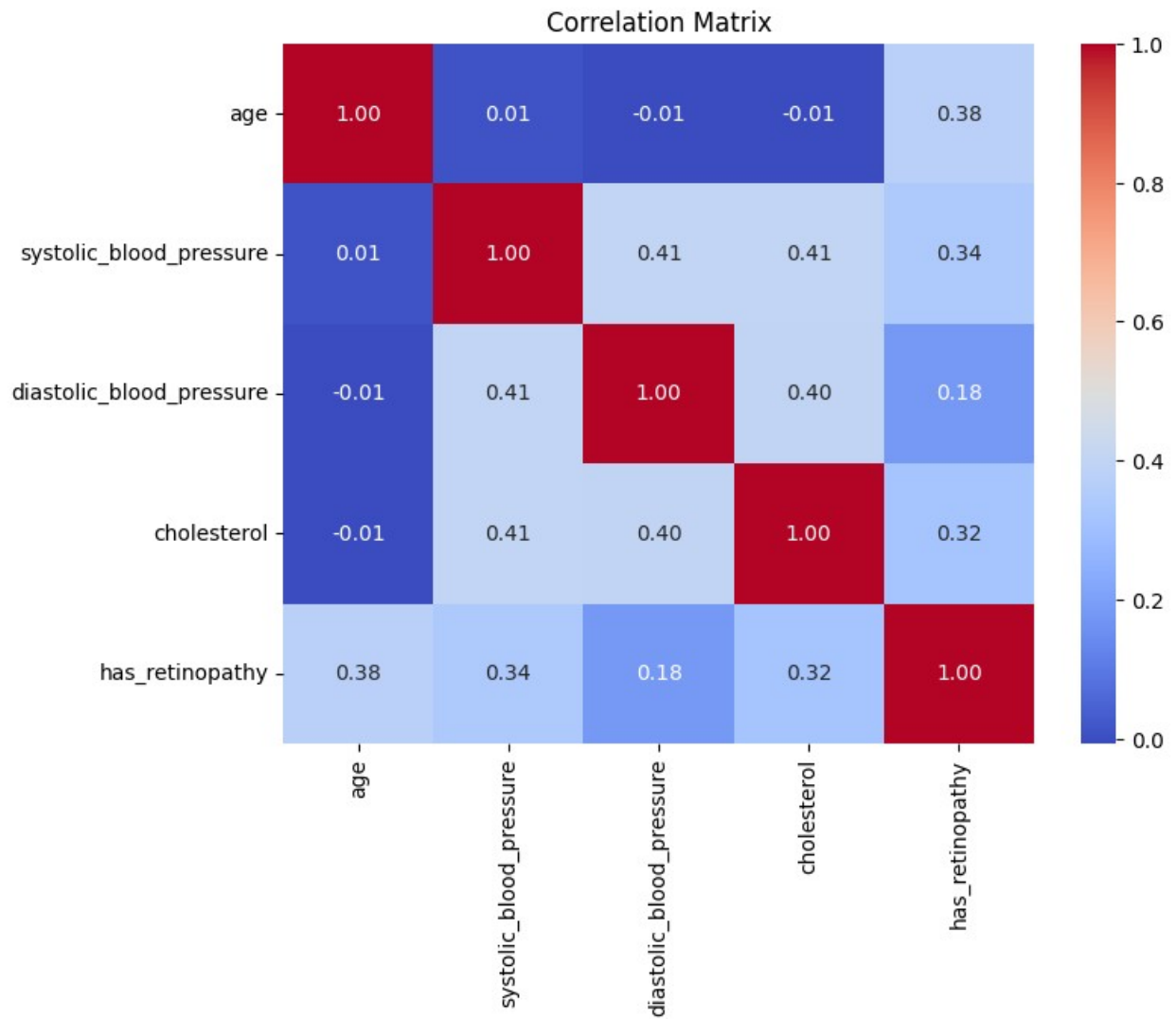
```
sns.histplot(df['cholesterol'], bins=30, kde=True, ax=axes[1, 1])  
axes[1, 1].set_title('Cholesterol Distribution')
```

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

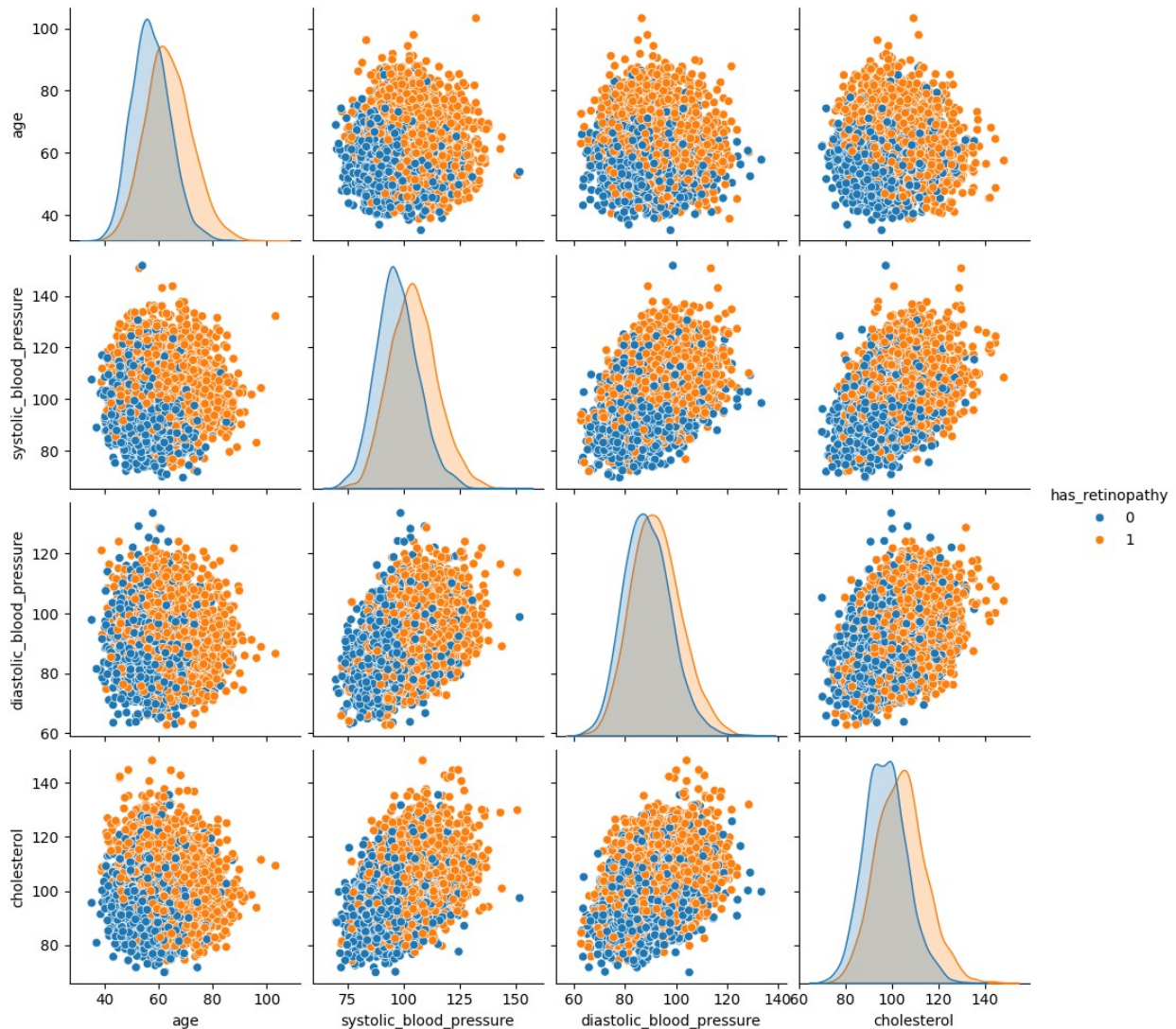


```
corr = df.corr()

plt.figure(figsize=(8, 6))
sns.heatmap(corr, annot=True, cmap='coolwarm', fmt='.2f')
plt.title('Correlation Matrix')
plt.show()
```



```
sns.pairplot(df, hue='has_retinopathy')  
plt.show()
```



```
fig, axes = plt.subplots(2, 2, figsize=(15, 10))

sns.scatterplot(x='age', y='systolic_blood_pressure',
hue='has_retinopathy', data=df, ax=axes[0, 0])
axes[0, 0].set_title('Age vs Systolic Blood Pressure')

sns.scatterplot(x='age', y='diastolic_blood_pressure',
hue='has_retinopathy', data=df, ax=axes[0, 1])
axes[0, 1].set_title('Age vs Diastolic Blood Pressure')

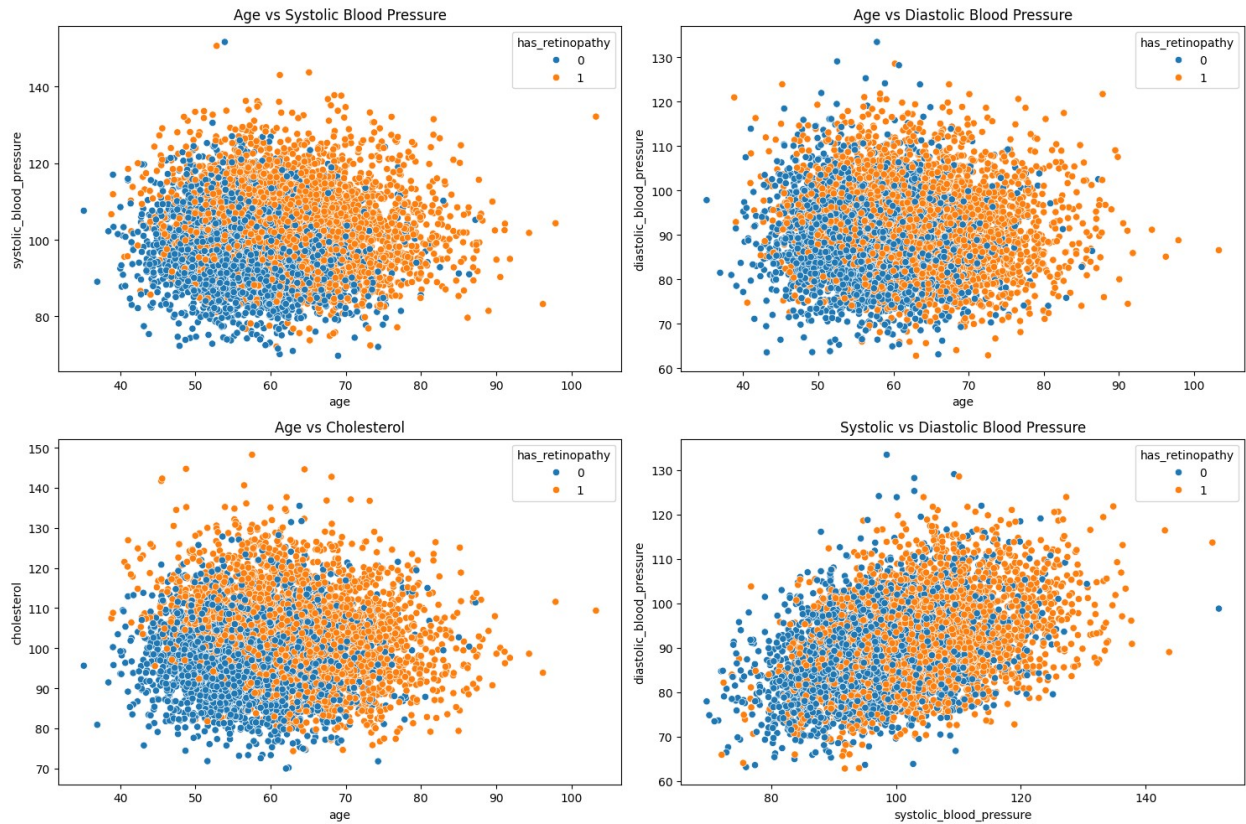
sns.scatterplot(x='age', y='cholesterol', hue='has_retinopathy',
data=df, ax=axes[1, 0])
axes[1, 0].set_title('Age vs Cholesterol')

sns.scatterplot(x='systolic_blood_pressure',
y='diastolic_blood_pressure', hue='has_retinopathy', data=df,
ax=axes[1, 1])
```

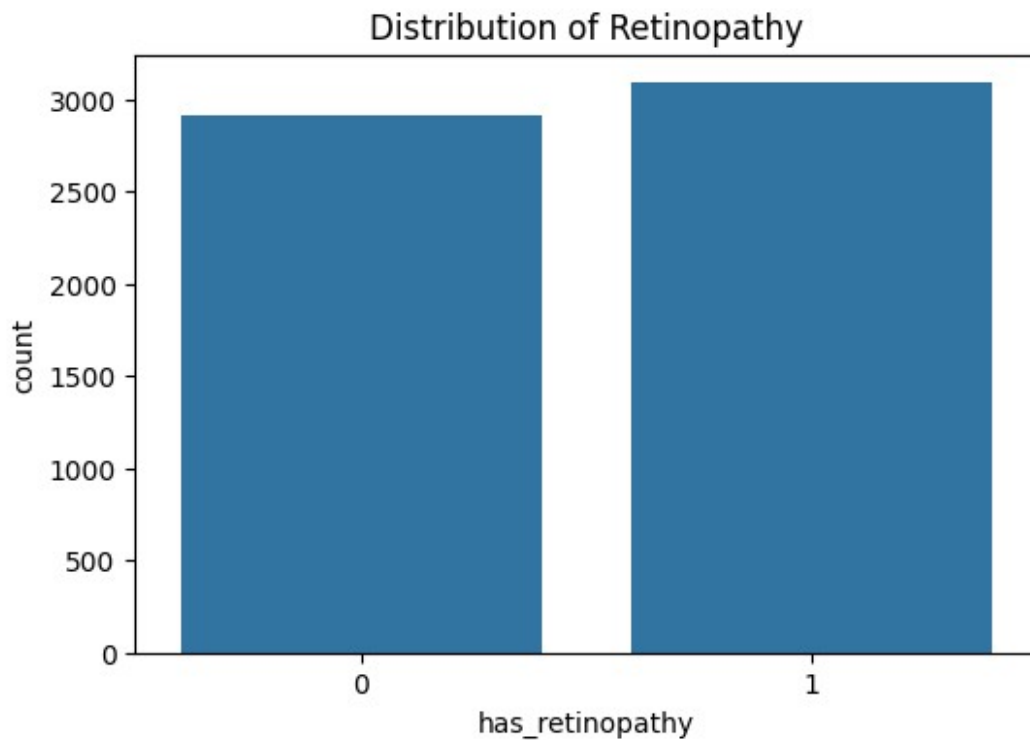


```
axes[1, 1].set_title('Systolic vs Diastolic Blood Pressure')
```

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



```
plt.figure(figsize=(6, 4))  
sns.countplot(x='has_retinopathy', data=df)  
plt.title('Distribution of Retinopathy')  
plt.show()
```

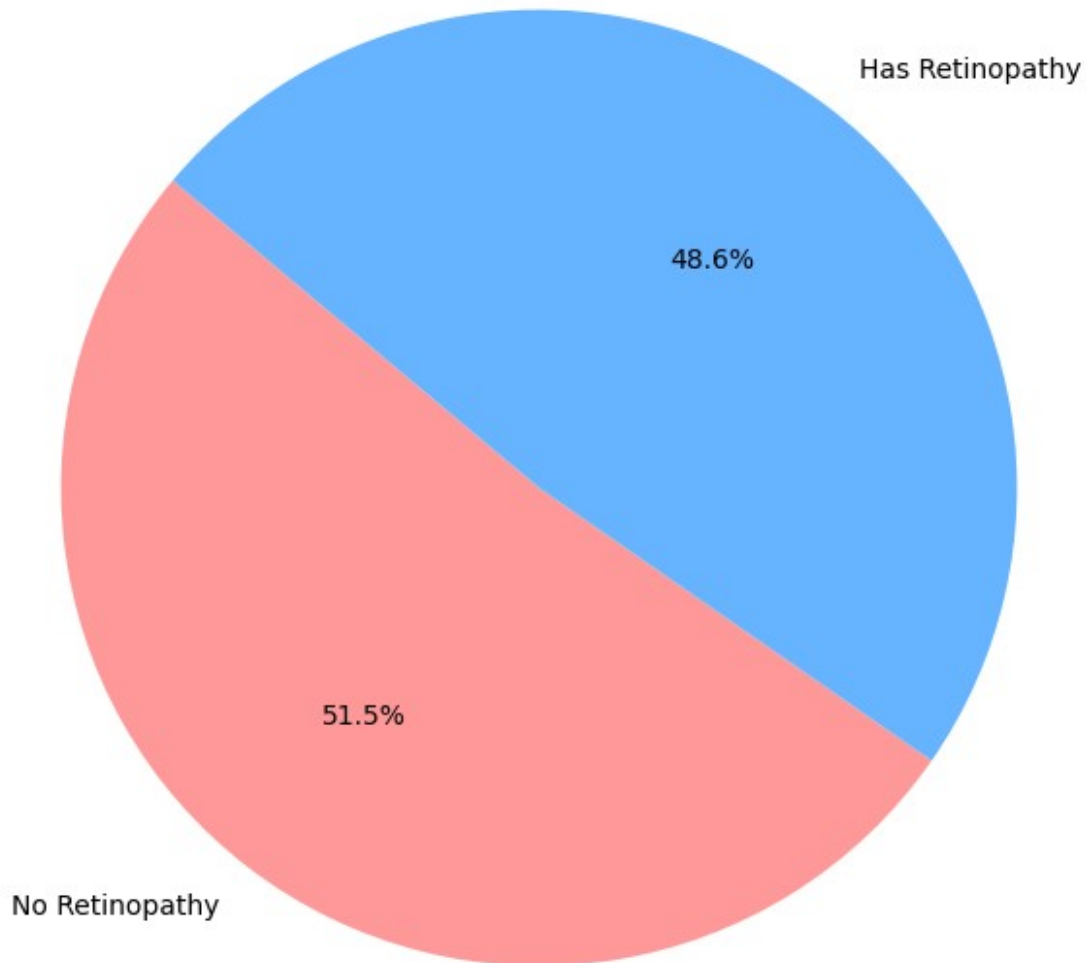



```
retinopathy_counts = df['has_retinopathy'].value_counts()

labels = ['No Retinopathy', 'Has Retinopathy']
colors = ['#ff9999', '#66b3ff']

plt.figure(figsize=(8, 8))
plt.pie(retinopathy_counts, labels=labels, colors=colors,
        autopct='%1.1f%%', startangle=140)
plt.title('Distribution of Retinopathy')
plt.show()
```

Distribution of Retinopathy



```
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.linear_model import LogisticRegression
from sklearn.svm import SVC
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score

X = df[['age', 'systolic_blood_pressure', 'diastolic_blood_pressure',
        'cholesterol']]
y = df['has_retinopathy']
```

```

X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.3, random_state=42)

scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)

log_reg = LogisticRegression()
log_reg.fit(X_train_scaled, y_train)
y_pred_log_reg = log_reg.predict(X_test_scaled)
accuracy_log_reg = accuracy_score(y_test, y_pred_log_reg)

svc = SVC()
svc.fit(X_train_scaled, y_train)
y_pred_svc = svc.predict(X_test_scaled)
accuracy_svc = accuracy_score(y_test, y_pred_svc)

dt = DecisionTreeClassifier()
dt.fit(X_train, y_train)
y_pred_dt = dt.predict(X_test)
accuracy_dt = accuracy_score(y_test, y_pred_dt)

rf = RandomForestClassifier()
rf.fit(X_train, y_train)
y_pred_rf = rf.predict(X_test)
accuracy_rf = accuracy_score(y_test, y_pred_rf)

print(f"Logistic Regression Accuracy: {accuracy_log_reg:.4f}")
print(f"SVC Accuracy: {accuracy_svc:.4f}")
print(f"Decision Tree Accuracy: {accuracy_dt:.4f}")
print(f"Random Forest Accuracy: {accuracy_rf:.4f}")

Logistic Regression Accuracy: 0.7444
SVC Accuracy: 0.7411
Decision Tree Accuracy: 0.6528
Random Forest Accuracy: 0.7322

# Define the accuracy scores
accuracies = {
    'Logistic Regression': accuracy_log_reg,
    'SVC': accuracy_svc,
    'Decision Tree': accuracy_dt,
    'Random Forest': accuracy_rf
}

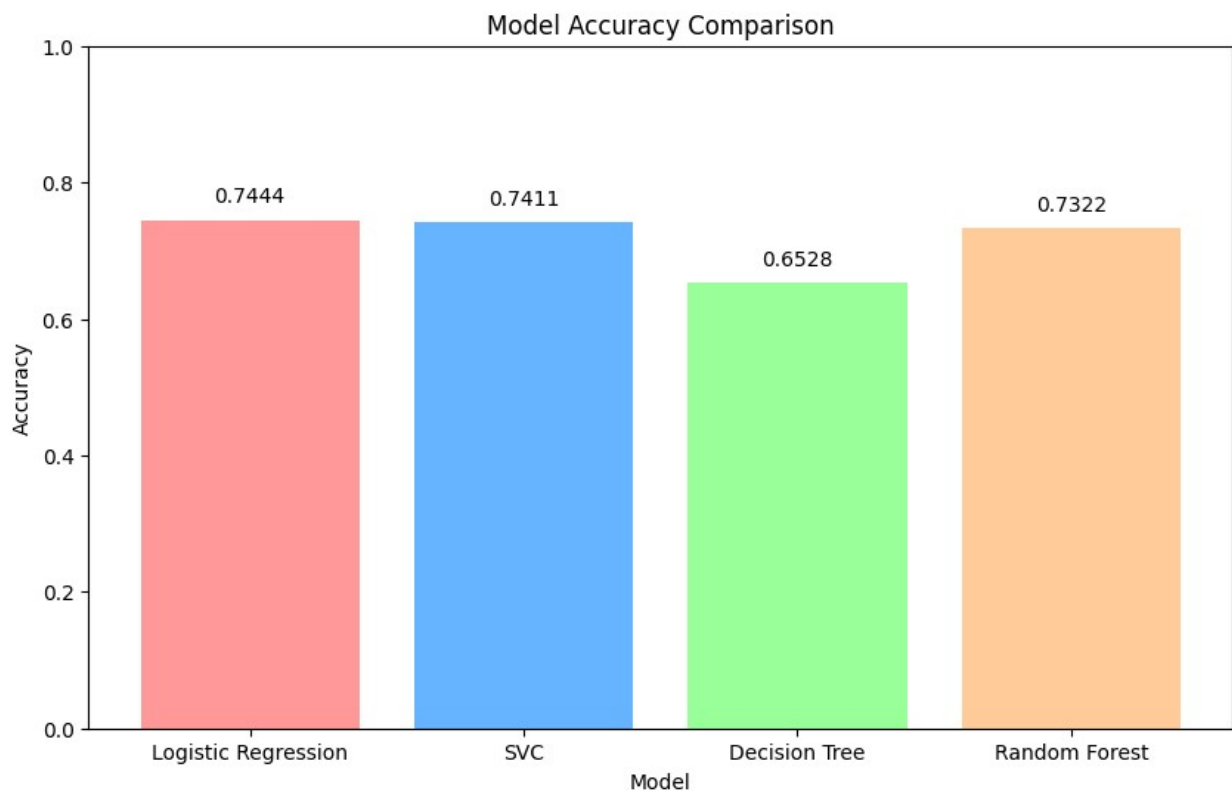
plt.figure(figsize=(10, 6))
bars = plt.bar(accuracies.keys(), accuracies.values(),
color=['#ff9999', '#66b3ff', '#99ff99', '#ffcc99'])
plt.xlabel('Model')
plt.ylabel('Accuracy')
plt.title('Model Accuracy Comparison')

```

```
plt.ylim(0, 1)

for bar in bars:
    yval = bar.get_height()
    plt.text(bar.get_x() + bar.get_width()/2, yval + 0.02,
f'{yval:.4f}', ha='center', va='bottom')

plt.show()
```



Thanks !!!