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

In [10]:
```
df=pd.read_csv('diabetes.csv')
```

In [11]:
```
df.head()
```

Out[11]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	Pedigree	Age	Outcome
0	6	148	72	35	0	33.6	0.627	50	1
1	1	85	66	29	0	26.6	0.351	31	0
2	8	183	64	0	0	23.3	0.672	32	1
3	1	89	66	23	94	28.1	0.167	21	0
4	0	137	40	35	168	43.1	2.288	33	1

In [12]:
```
df.tail()
```

Out[12]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	Pedigree	Age	Outcome
763	10	101	76	48	180	32.9	0.171	63	0
764	2	122	70	27	0	36.8	0.340	27	0
765	5	121	72	23	112	26.2	0.245	30	0
766	1	126	60	0	0	30.1	0.349	47	1
767	1	93	70	31	0	30.4	0.315	23	0

In [13]:
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 9 columns):
 #   Column         Non-Null Count  Dtype  
---  ------         --------------  -----  
 0   Pregnancies    768 non-null    int64  
 1   Glucose        768 non-null    int64  
 2   BloodPressure  768 non-null    int64  
 3   SkinThickness  768 non-null    int64  
 4   Insulin        768 non-null    int64  
 5   BMI            768 non-null    float64
 6   Pedigree       768 non-null    float64
 7   Age            768 non-null    int64  
 8   Outcome        768 non-null    int64  
dtypes: float64(2), int64(7)
memory usage: 54.1 KB
```

In [14]:
```
df.isnull().sum()
```

Out[14]:
```
Pregnancies      0
Glucose          0
BloodPressure    0
SkinThickness    0
Insulin          0
BMI              0
Pedigree         0
Age              0
Outcome          0
dtype: int64
```

In [15]:
```
df.shape
```

Out[15]:
```
(768, 9)
```

In [16]:
```
df.describe()
```

Out[16]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	Pedigree	
count	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	76
mean	3.845052	120.894531	69.105469	20.536458	79.799479	31.992578	0.471876	3
std	3.369578	31.972618	19.355807	15.952218	115.244002	7.884160	0.331329	1
min	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.078000	2
25%	1.000000	99.000000	62.000000	0.000000	0.000000	27.300000	0.243750	2
50%	3.000000	117.000000	72.000000	23.000000	30.500000	32.000000	0.372500	2
75%	6.000000	140.250000	80.000000	32.000000	127.250000	36.600000	0.626250	4
max	17.000000	199.000000	122.000000	99.000000	846.000000	67.100000	2.420000	8

In [17]:
```
df.head()
```

Out[17]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	Pedigree	Age	Outcome
0	6	148	72	35	0	33.6	0.627	50	1
1	1	85	66	29	0	26.6	0.351	31	0
2	8	183	64	0	0	23.3	0.672	32	1
3	1	89	66	23	94	28.1	0.167	21	0
4	0	137	40	35	168	43.1	2.288	33	1

In [18]:
```
df.tail()
```

Out[18]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	Pedigree	Age	Outcome
763	10	101	76	48	180	32.9	0.171	63	0
764	2	122	70	27	0	36.8	0.340	27	0
765	5	121	72	23	112	26.2	0.245	30	0
766	1	126	60	0	0	30.1	0.349	47	1
767	1	93	70	31	0	30.4	0.315	23	0

In [19]:
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 9 columns):
 #   Column         Non-Null Count  Dtype  
---  ------         --------------  -----  
 0   Pregnancies    768 non-null    int64  
 1   Glucose        768 non-null    int64  
 2   BloodPressure  768 non-null    int64  
 3   SkinThickness  768 non-null    int64  
 4   Insulin        768 non-null    int64  
 5   BMI            768 non-null    float64
 6   Pedigree       768 non-null    float64
 7   Age            768 non-null    int64  
 8   Outcome        768 non-null    int64  
dtypes: float64(2), int64(7)
memory usage: 54.1 KB
```

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

Out[20]:
```
Pregnancies      0
Glucose          0
BloodPressure    0
SkinThickness    0
Insulin          0
BMI              0
Pedigree         0
Age              0
Outcome          0
dtype: int64
```

In [21]:
```
df.shape
```

Out[21]:
```
(768, 9)
```

In [22]:
```
df.describe()
```

Out[22]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	Pedigree	
count	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	76
mean	3.845052	120.894531	69.105469	20.536458	79.799479	31.992578	0.471876	3
std	3.369578	31.972618	19.355807	15.952218	115.244002	7.884160	0.331329	1
min	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.078000	2
25%	1.000000	99.000000	62.000000	0.000000	0.000000	27.300000	0.243750	2
50%	3.000000	117.000000	72.000000	23.000000	30.500000	32.000000	0.372500	2
75%	6.000000	140.250000	80.000000	32.000000	127.250000	36.600000	0.626250	4
max	17.000000	199.000000	122.000000	99.000000	846.000000	67.100000	2.420000	8

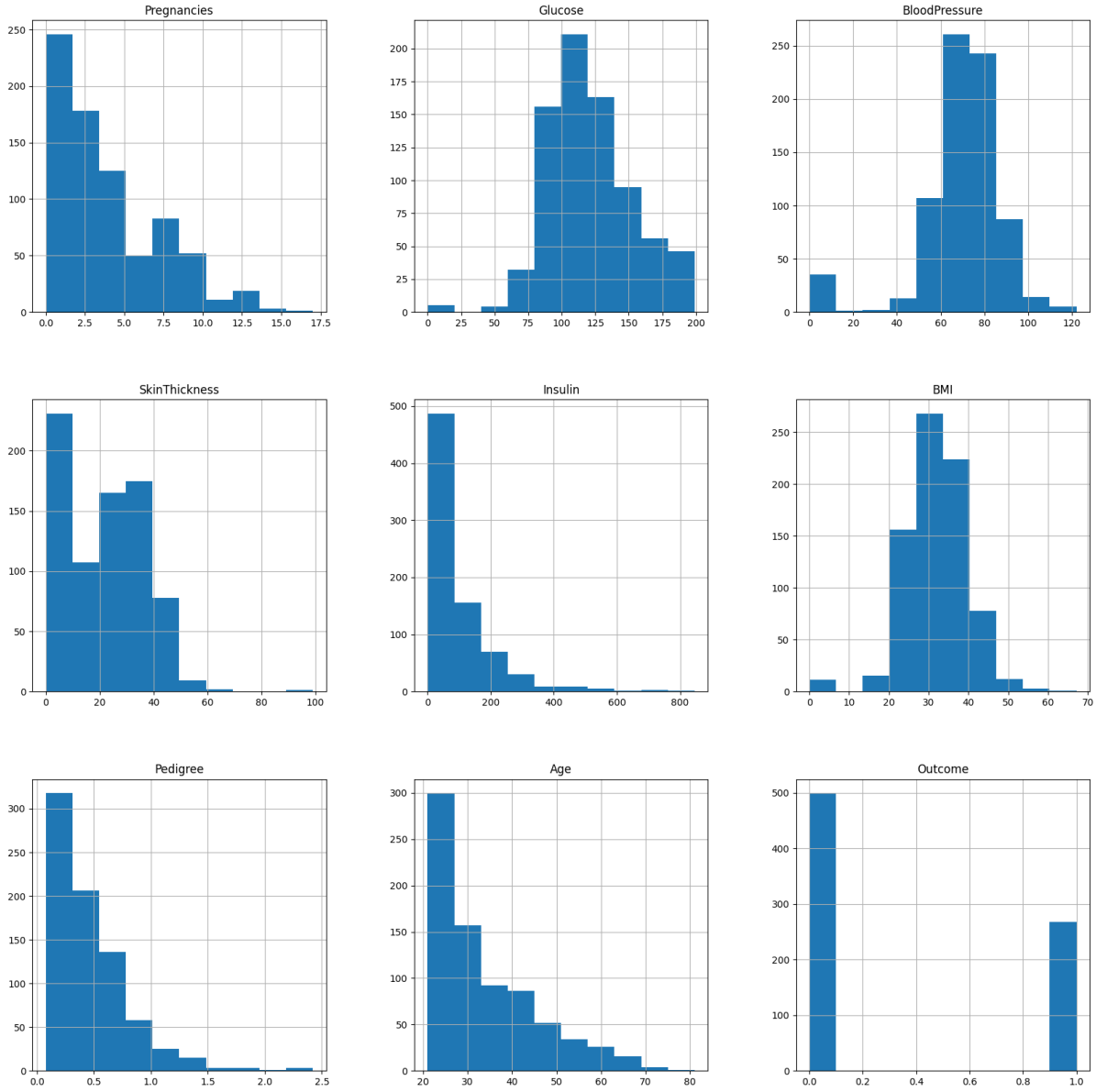
In [23]:
```
df.corr()
```

Out[23]:

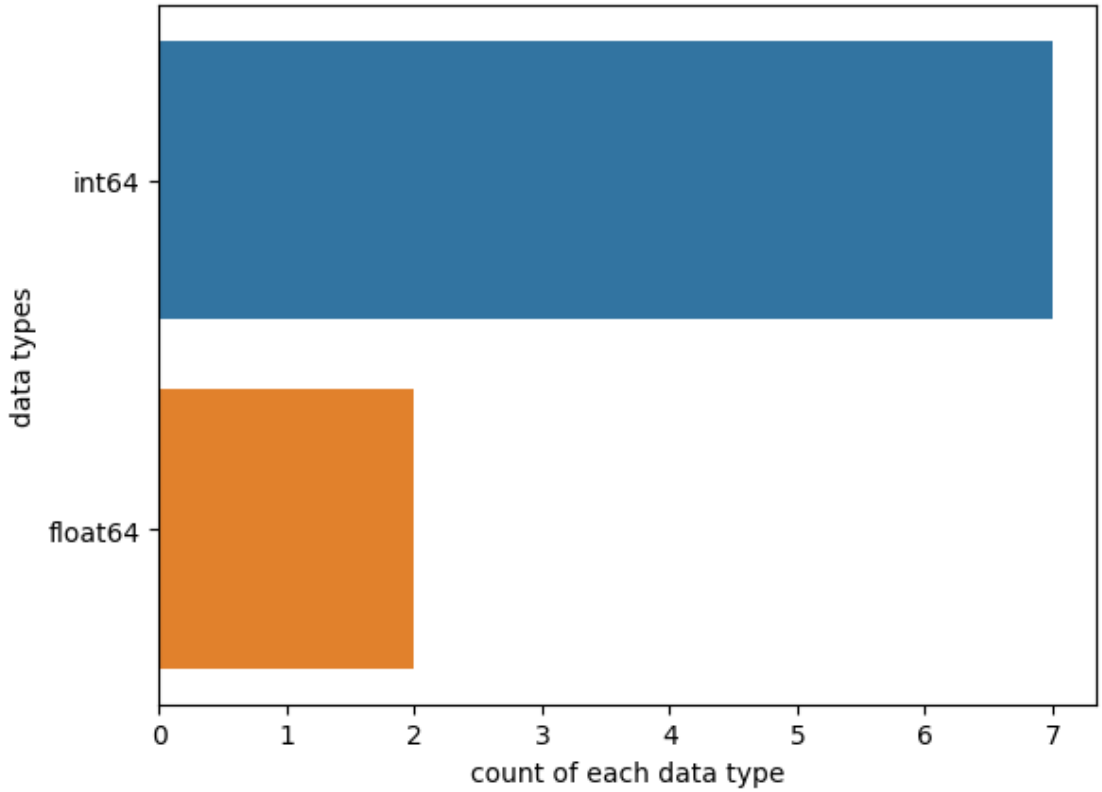
	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	Pedigree
Pregnancies	1.000000	0.129459	0.141282	-0.081672	-0.073535	0.017683	-0.033523
Glucose	0.129459	1.000000	0.152590	0.057328	0.331357	0.221071	0.137337
BloodPressure	0.141282	0.152590	1.000000	0.207371	0.088933	0.281805	0.041265
SkinThickness	-0.081672	0.057328	0.207371	1.000000	0.436783	0.392573	0.183928
Insulin	-0.073535	0.331357	0.088933	0.436783	1.000000	0.197859	0.185071
BMI	0.017683	0.221071	0.281805	0.392573	0.197859	1.000000	0.140647
Pedigree	-0.033523	0.137337	0.041265	0.183928	0.185071	0.140647	1.000000
Age	0.544341	0.263514	0.239528	-0.113970	-0.042163	0.036242	0.033561
Outcome	0.221898	0.466581	0.065068	0.074752	0.130548	0.292695	0.173844

In [24]:
```
df['Glucose'].fillna(df['Glucose'].mean(), inplace = True)
df['BloodPressure'].fillna(df['BloodPressure'].mean(), inplace = True)
df['SkinThickness'].fillna(df['SkinThickness'].median(), inplace = True)
df['Insulin'].fillna(df['Insulin'].median(), inplace = True)
df['BMI'].fillna(df['BMI'].median(), inplace = True)
```

In [25]:
```
p = df.hist(figsize = (20,20))
```

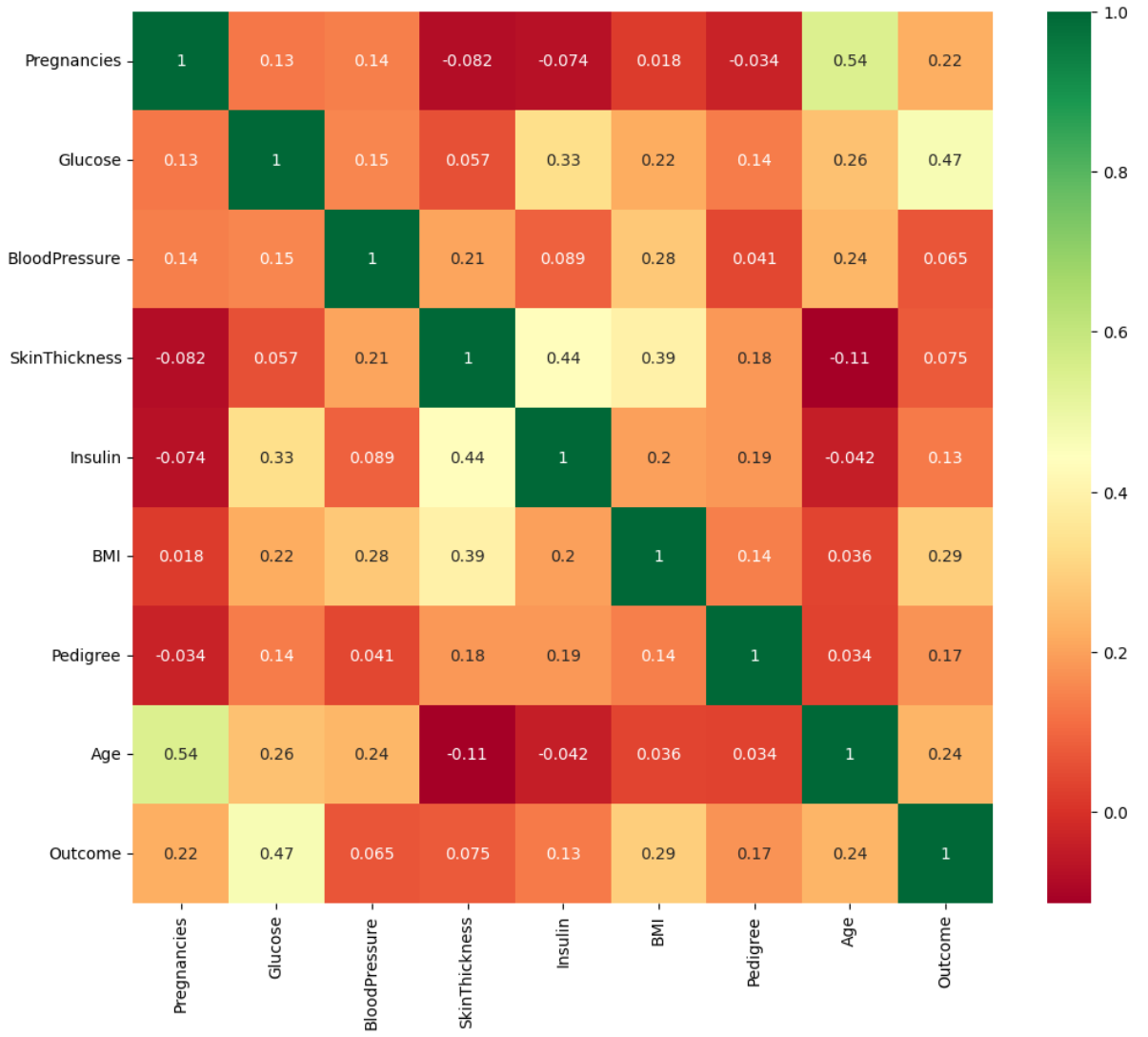


In [26]:
```
sns.countplot(y=df.dtypes ,data=df)
plt.xlabel("count of each data type")
plt.ylabel("data types")
plt.show()
```



In [27]:
```
p=sns.pairplot(df)
```

In [28]:
```
plt.figure(figsize=(12,10)) # on this line I just set the size of figure to 12 by 1
p=sns.heatmap(df.corr(), annot=True,cmap ='RdYlGn')
```



In [29]:
```
from sklearn.preprocessing import StandardScaler
sc_X = StandardScaler()
X = pd.DataFrame(sc_X.fit_transform(df.drop(["Outcome"],axis = 1),),
 columns=['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
 'BMI', 'DiabetesPedigreeFunction', 'Age'])
```

In [30]:
```
y = df.Outcome
```

In [31]:
```
X.head()
```

Out[31]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunctic
0	0.639947	0.848324	0.149641	0.907270	-0.692891	0.204013	0.46849
1	-0.844885	-1.123396	-0.160546	0.530902	-0.692891	-0.684422	-0.36506
2	1.233880	1.943724	-0.263941	-1.288212	-0.692891	-1.103255	0.60439
3	-0.844885	-0.998208	-0.160546	0.154533	0.123302	-0.494043	-0.92076
4	-1.141852	0.504055	-1.504687	0.907270	0.765836	1.409746	5.48490

In [32]:
```
from sklearn.model_selection import train_test_split

# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=1/3, random_sta
```

In [33]:
```
from sklearn.neighbors import KNeighborsClassifier
test_scores = []
train_scores = []
for i in range(1,15):
 knn = KNeighborsClassifier(i)
 knn.fit(X_train,y_train)

 train_scores.append(knn.score(X_train,y_train))
 test_scores.append(knn.score(X_test,y_test))
```

In [36]:
```
max_train_score = max(train_scores)
train_scores_ind = [i for i, v in enumerate(train_scores) if v == max_train_score]
k_values_for_max_score = list(map(lambda x: x + 1, train_scores_ind))

print('Max train score {:.2f}% and k = {}'.format(max_train_score * 100, k_values_f
```

```
Max train score 100.00% and k = [1]
```

In [38]:
```
max_test_score = max(test_scores)
test_scores_ind = [i for i, v in enumerate(test_scores) if v == max_test_score]
k_values_for_max_test_score = list(map(lambda x: x + 1, test_scores_ind))

print('Max test score {:.2f}% and k = {}'.format(max_test_score * 100, k_values_for
```

```
Max test score 76.95% and k = [11]
```

In [39]:
```
knn = KNeighborsClassifier(11)
knn.fit(X_train,y_train)
knn.score(X_test,y_test)
```

Out[39]:
```
0.76953125
```

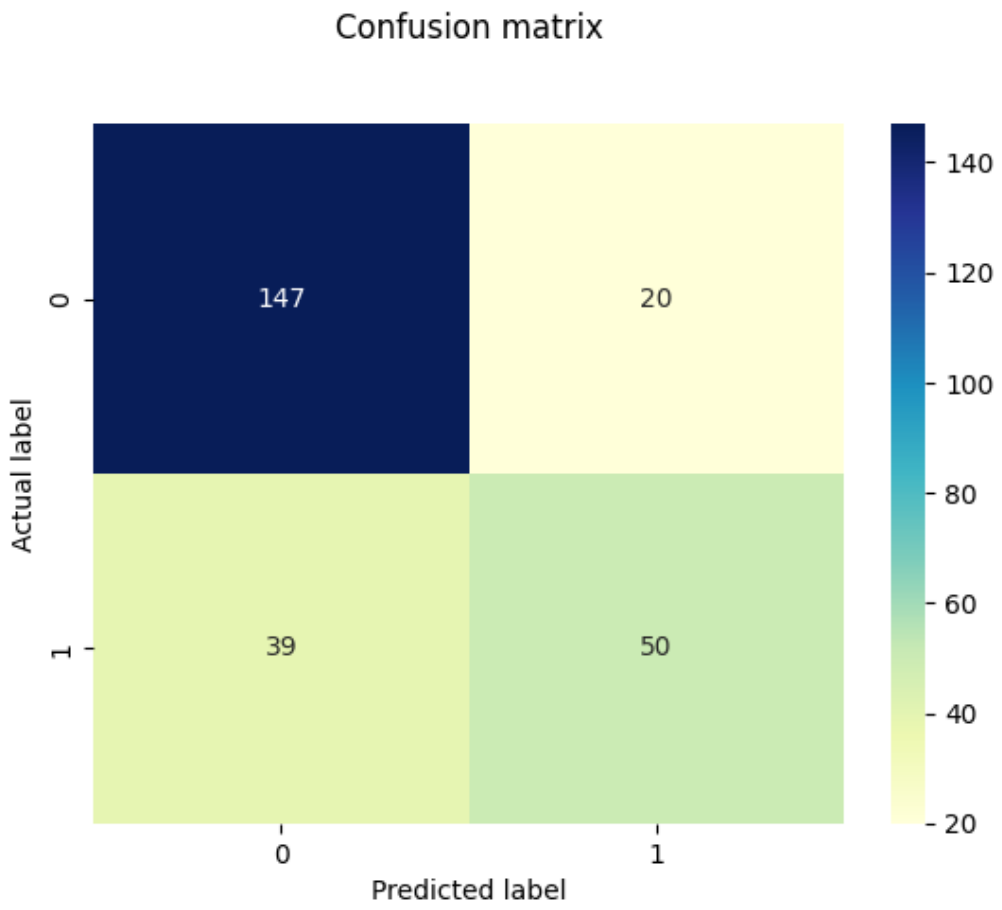
In [40]:
```
from sklearn.metrics import confusion_matrix
#let us get the predictions using the classifier we had fit above
y_pred = knn.predict(X_test)
confusion_matrix(y_test,y_pred)
pd.crosstab(y_test, y_pred, rownames=['True'], colnames=['Predicted'], margins=True
```

Out[40]:

Predicted	0	1	All
True			
0	147	20	167
1	39	50	89
All	186	70	256

In [41]:
```
y_pred = knn.predict(X_test)
from sklearn import metrics
cnf_matrix = metrics.confusion_matrix(y_test, y_pred)
p = sns.heatmap(pd.DataFrame(cnf_matrix), annot=True, cmap="YlGnBu" ,fmt='g')
plt.title('Confusion matrix', y=1.1)
plt.ylabel('Actual label')
plt.xlabel('Predicted label')
```

Out[41]:
```
Text(0.5, 23.52222222222222, 'Predicted label')
```



In [42]:
```
from sklearn.metrics import classification_report
print(classification_report(y_test,y_pred))
```

```
              precision    recall  f1-score   support

           0       0.79      0.88      0.83       167
           1       0.71      0.56      0.63        89

    accuracy                           0.77       256
   macro avg       0.75      0.72      0.73       256
weighted avg       0.76      0.77      0.76       256
```

In [ ]:
```

```