In [15]:

```
# -*- coding: utf-8 -*-
"""
Created on Sun Sep 23 14:30:30 2018

@author: filipe.luz
"""

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.ensemble import RandomForestClassifier
from sklearn.model_selection import train_test_split
from sklearn.metrics import classification_report,confusion_matrix, roc_curve

df = pd.read_csv('C:\\Users\\filipe.luz\\Desktop\\Desafio_BB\\census.csv')

df.isnull().sum()

df_new = df.dropna()
df_new.info()
<class 'pandas.core.frame.DataFrame'>
```

```
Int64Index: 5532 entries, 0 to 5531
Data columns (total 14 columns):
age
                   5532 non-null int64
workclass
                   5532 non-null object
education level
                   5532 non-null object
education-num
                   5532 non-null float64
                   5532 non-null object
marital-status
occupation
                   5532 non-null object
relationship
                   5532 non-null object
race
                   5532 non-null object
                   5532 non-null object
sex
capital-gain
                   5532 non-null float64
capital-loss
                   5532 non-null float64
                   5532 non-null float64
hours-per-week
native-country
                   5532 non-null object
income
                   5532 non-null object
dtypes: float64(4), int64(1), object(9)
memory usage: 648.3+ KB
```

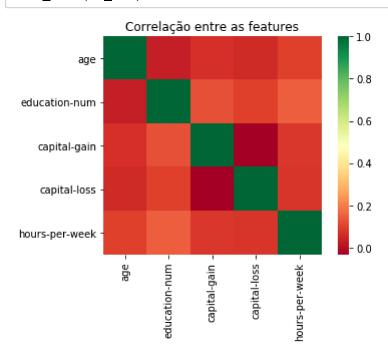
In [16]:

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 5532 entries, 0 to 5531
Data columns (total 14 columns):
                   5532 non-null int64
age
                   5532 non-null category
workclass
education level
                   5532 non-null category
education-num
                   5532 non-null float64
marital-status
                   5532 non-null category
occupation
                   5532 non-null category
relationship
                   5532 non-null category
                   5532 non-null category
race
                   5532 non-null category
sex
                   5532 non-null float64
capital-gain
capital-loss
                   5532 non-null float64
hours-per-week
                   5532 non-null float64
native-country
                   5532 non-null category
income
                   5532 non-null category
dtypes: category(9), float64(4), int64(1)
memory usage: 312.3 KB
```

In [17]:

```
def heat_corr (data):
    '''Analisando a correlação entre as features
        do dataset
    '''
    sns.heatmap(data.corr(),square=True,cmap='RdYlGn')
    plt.title('Correlação entre as features')
    plt.show()
```

#Chamando função para analisar a correlação entre as features originais do dataset heat_corr(df_new)



In [18]:

In [19]:

```
#Generating the X and y datasets

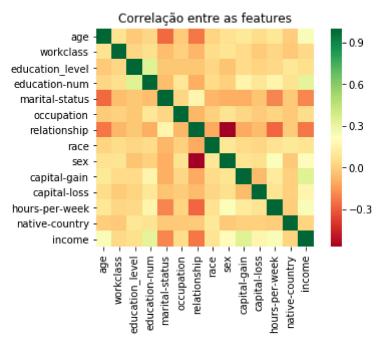
X = (df_new.drop(['income'], axis =1) )
y = np.where(df_new.iloc[:,-1] == '>50K',1,0)

#One Hot Encoding fuel rail column
from sklearn.preprocessing import LabelEncoder, OneHotEncoder
labelEncoder= LabelEncoder()
for i in range(0,len(X.columns)):
    X.iloc[:,i] = labelEncoder.fit_transform(X.iloc[:,i])

hotEncoder = OneHotEncoder()
for i in range(0,len(X.columns)):
    X.iloc[:,i] = labelEncoder.fit_transform(X.iloc[:,i])
```

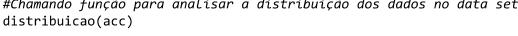
In [20]:

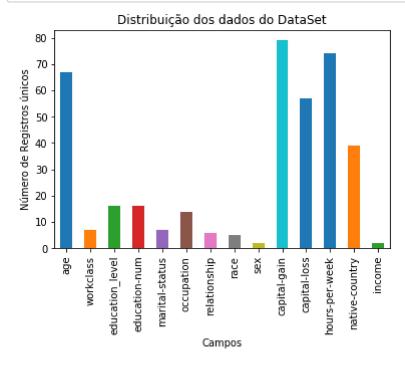
```
#Creating a new df to can analyse the correlation after OneHotEncoder
acc = X.copy()
acc['income'] = y.copy()
heat_corr(acc)
```



In [21]:

```
def distribuicao (data):
    Esta função exibirá a quantidade de registros únicos para cada coluna
   existente no dataset
    dataframe -> Histogram
    # Calculando valores unicos para cada label: num_unique_labels
   num_unique_labels = data.apply(pd.Series.nunique)
    # plotando valores
    num_unique_labels.plot( kind='bar')
    # Nomeando os eixos
    plt.xlabel('Campos')
    plt.ylabel('Número de Registros únicos')
    plt.title('Distribuição dos dados do DataSet')
    # Exibindo gráfico
    plt.show()
#Chamando função para analisar a distribuição dos dados no data set
distribuicao(acc)
```





In [22]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y , test_size=0.4, random_state = 0)

# Feature Scaling
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)

classifier = RandomForestClassifier(n_estimators=40, criterion='entropy', random_state= 0)

classifier.fit(X_train, y_train)
y_pred_train = classifier.predict_proba(X_train)

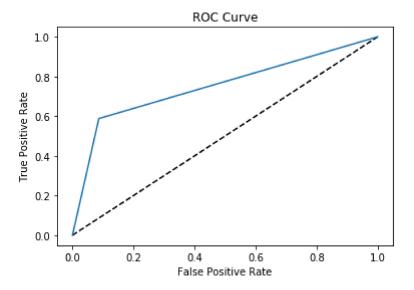
y_pred = classifier.predict_proba(X_test)

# Compute predicted probabilities: y_pred_prob
y_pred_prob = (y_pred)[:,1]
```

In [23]:

```
# Generate ROC curve values: fpr, tpr, thresholds
fpr, tpr,threshold = roc_curve(y_test,y_pred_prob.round())

# Plot ROC curve
plt.plot([0, 1], [0, 1], 'k--')
plt.plot(fpr,tpr)
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('ROC Curve')
plt.show()
```



In [24]:

```
# Making the Confusion Matrix
cm = confusion_matrix(y_test, y_pred[:,1].round())
ac_desemp(cm)
# Compute metrics
print(classification_report(y_test, y_pred[:,1].round()))
print("Log Loss Result: " + str( compute_log_loss(y_pred[:,1],y_test)))
```

```
Acuracia do modelo é : 0.8309986443741527
             precision
                          recall f1-score
                                              support
          0
                  0.87
                            0.91
                                       0.89
                                                 1652
          1
                  0.70
                            0.59
                                       0.64
                                                  561
avg / total
                  0.82
                            0.83
                                       0.83
                                                 2213
```

Log Loss Result: 0.474954156448957