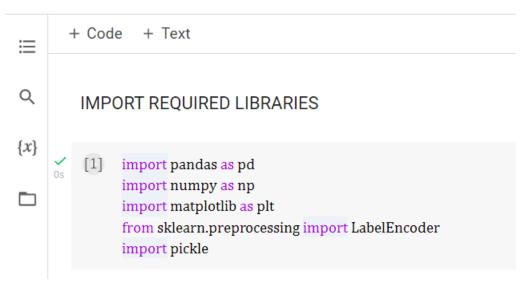
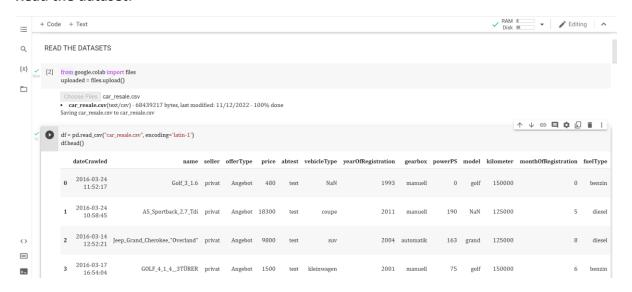
Development Phase Pre-process the data

Date	12 November 2022
Team ID	PNT2022TMID21553
Project Name	Project – Car Resale Value Prediction

Import the required libraries:

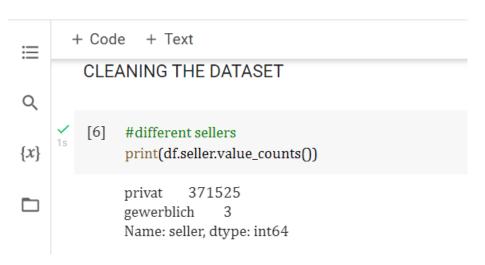


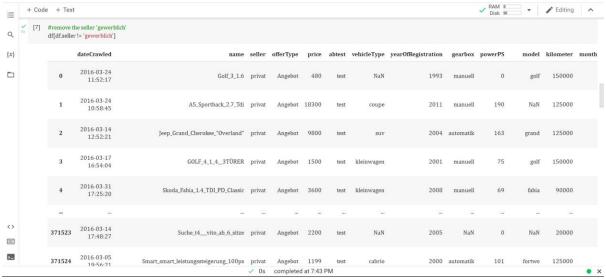
Read the dataset:

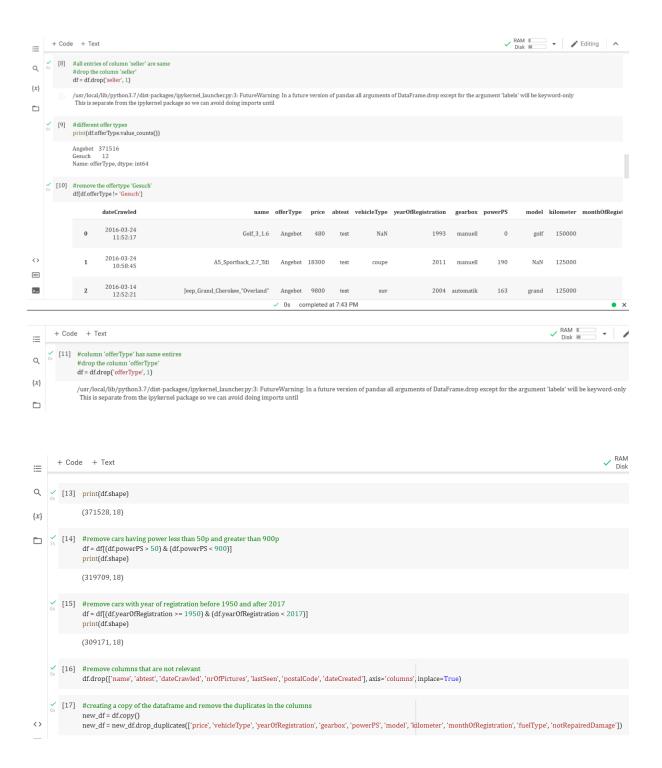




Cleaning the dataset:







```
+ Code + Text
\equiv
         [19] #clean the dataset of German words and replace with proper English words
Q
                new_df.gearbox.replace(('manuell', 'automatik'), ('manual', 'automatic'), inplace=True)
                new\_df.fuelType.replace \verb|('benzin', 'andere', 'elektro')|, ('petrol', 'others', 'electric')|, inplace = True|
{x}
                new\_df.vehicleType.replace(('kleinwagen', 'cabrio', 'kombi', 'andere'), ('small \ car', 'convertible', 'combination', 'others'), inplace=True)
                new\_df.notRepairedDamage.replace(('ja', 'nein'), ('Yes', 'No'), inplace=True)
[20] #Outlier Removal
                new\_df = new\_df[(new\_df.price >= 100) \& (new\_df.price <= 150000)]
         [21] #Fill the not declared values of the columns as NaN using fillna function
                new\_df['notRepairedDamage']. fill na(value='not-declared', in place=True)
                new_df['fuelType'].fillna(value='not-declared', inplace=True)
                new\_df['gearbox'].fillna(value='not-declared', inplace=True)
                new_df['vehicleType'].fillna(value='not-declared', inplace=True)
                new_df['model'].fillna(value='not-declared', inplace=True)
        [22] #save the dataframe as csv
                new\_df.to\_csv('car\_resale\_preprocessed.csv')
        + Code + Text
\equiv
                #label encode the categorical data
                labels = ['gearbox', 'notRepairedDamage', 'model', 'brand', 'fuelType', 'vehicleType']
\{x\}
                mapping = {}
                for i in labels:
                 mapping[i] = LabelEncoder()
                  mapping[i].fit(new_df[i])
                 tr = mapping[i].transform(new_df[i])
                  np.save(str('classes'+i+'.npy'), mapping[i].classes_)
                 print(i, ":", mapping[i])
                 new_df.loc[:, i+'_labels'] = pd.Series(tr, index=new_df.index)
                gearbox : LabelEncoder()
                notRepairedDamage : LabelEncoder()
                model : LabelEncoder()
                brand : LabelEncoder()
                fuelType : LabelEncoder()
                vehicleType : LabelEncoder()
      [24] #'labeled' dataframe contains the final data
                labelled = new_df[ ['price', 'yearOfRegistration', 'powerPS', 'kilometer', 'monthOfRegistration'] + [x+"_labels" for x in labels]]
                print(labelled.columns)
<>
                Index(['price', 'yearOfRegistration', 'powerPS', 'kilometer',
                    'month Of Registration', 'gearbox\_labels', 'not Repaired Damage\_labels',
\equiv
                    'model\_labels', 'brand\_labels', 'fuelType\_labels',
                    'vehicleType_labels'],
>_
                   dtype='object')
```

Splitting the data into independent and dependent variables

