Predicting the sales price of bulldozers using Machine learning algorithm

1. Definition

How well can we predict the future sale price of a bulldozer, given its characteristics and previous examples of how much similar bulldozers have been sold for?

2. Data

The data we have taken from kaggle: https://www.kaggle.com/c/bluebook-for-bulldozers/data (https://www.kaggle.com/c/bluebook-for-bulldozers/data)

There are 3 main datasets:

- Train.csv is the training set, which contains data through the end of 2011.
- Valid.csv is the validation set, which contains data from January 1, 2012 April 30, 2012 You
 make predictions on this set throughout the majority of the competition. Your score on this set
 is used to create the public leaderboard.
- Test.csv is the test set, which won't be released until the last week of the competition. It
 contains data from May 1, 2012 November 2012. Your score on the test set determines your
 final rank for the competition.

3. Evaluation

The evaluation metric for this competition is the RMSLE (root mean squared log error) between the actual and predicted auction prices.

For more on the evaluation of this project check: https://www.kaggle.com/c/bluebook-for-bulldozers/overview/evaluation (https://www.kaggle.com/c/bluebook-for-bulldozers/overview/evaluation)

Note. The goal for most regression evaluation metrics is to minimises the the error. For example, our goal for this project is to build a machine learning model which minimises RMSLE.

4. Features

Kaggle provides a data dictionary detailing all the features of the dataset.

https://www.kaggle.com/c/bluebook-for-bulldozers/data?select=Data+Dictionary.xlsx)

(https://www.kaggle.com/c/bluebook-for-bulldozers/data?select=Data+Dictionary.xlsx)

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

```
In [120]: # Importing training and validation sets
df=pd.read_csv("data/TrainAndValid.csv", low_memory=False)
```

In [121]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 412698 entries, 0 to 412697
Data columns (total 53 columns):

рата # 	Columns (total 53 columns	Non-Null Count	Dtype
0	SalesID	412698 non-null	int64
1	SalePrice	412698 non-null	float64
2	MachineID	412698 non-null	int64
3	ModelID	412698 non-null	int64
4	datasource	412698 non-null	int64
5	auctioneerID	392562 non-null	float64
6	YearMade	412698 non-null	int64
7	MachineHoursCurrentMeter	147504 non-null	float64
8	UsageBand	73670 non-null	object
9	saledate	412698 non-null	object
10	fiModelDesc	412698 non-null	object
11	fiBaseModel	412698 non-null	object
12	fiSecondaryDesc	271971 non-null	object
13	fiModelSeries	58667 non-null	object
14	fiModelDescriptor	74816 non-null	object
15	ProductSize	196093 non-null	object
16	fiProductClassDesc	412698 non-null	object
17	state	412698 non-null	object
18	ProductGroup	412698 non-null	object
19	ProductGroupDesc	412698 non-null	object
20	Drive_System	107087 non-null	object
21	Enclosure	412364 non-null	object
22	Forks	197715 non-null	object
23	Pad_Type	81096 non-null	object
24	Ride_Control	152728 non-null	object
25	Stick	81096 non-null	object
26	Transmission	188007 non-null	object
27	Turbocharged	81096 non-null	object
28	Blade_Extension	25983 non-null	object
29	Blade_Width	25983 non-null	object
30	Enclosure_Type	25983 non-null	object
31	Engine_Horsepower	25983 non-null	object
32	Hydraulics	330133 non-null	object
33	Pushblock	25983 non-null	object
34 25	Ripper	106945 non-null	object
35	Scarifier	25994 non-null	3
36	Tip_Control	25983 non-null 97638 non-null	object
37 38	Tire_Size	220679 non-null	object
39	Coupler System	44974 non-null	object
40	Coupler_System Grouser_Tracks	44875 non-null	object object
41	Hydraulics_Flow	44875 non-null	object
42	Track_Type	102193 non-null	object
43	Undercarriage_Pad_Width	102193 Non-Null	object
44	Stick_Length	102310 Non-Null	object
45	Thumb	102332 non-null	object
46	Pattern_Changer	102332 Non-Null	object
47	Grouser_Type	102101 non-null	object
48	Backhoe_Mounting	80712 non-null	object
49	Blade_Type	81875 non-null	object
. –	~~~_·>r)

50 Travel_Controls 81877 non-null object 51 Differential_Type 71564 non-null object 52 Steering_Controls 71522 non-null object

dtypes: float64(3), int64(5), object(45)

memory usage: 166.9+ MB

In [122]: df.isna().sum()

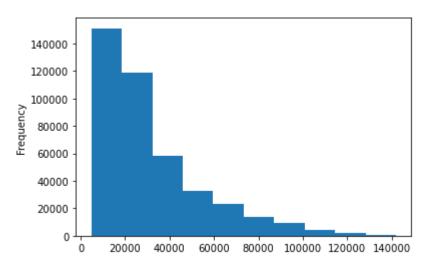
In [122]:	dT.1Sna().Sum()	
Out[122]:	SalesID	0
	SalePrice	0
	MachineID	0
	ModelID	0
	datasource	0
	auctioneerID	20136
	YearMade	0
	MachineHoursCurrentMeter	265194
	UsageBand	339028
	saledate	0
	fiModelDesc	0
	fiBaseModel	0
	fiSecondaryDesc	140727
	fiModelSeries	354031
	fiModelDescriptor	337882
	ProductSize	216605
	fiProductClassDesc	0
	state	0
	ProductGroup	0
	ProductGroupDesc	0
	Drive_System	305611
	Enclosure	334
	Forks	214983
	Pad_Type	331602
	Ride_Control	259970
	Stick	331602
	Transmission	224691
	Turbocharged	331602
	Blade_Extension	386715
	Blade_Width	386715
	Enclosure_Type	386715
	Engine_Horsepower	386715
	Hydraulics	82565
	Pushblock	386715
	Ripper	305753
	Scarifier	386704
	Tip Control	386715
	Tire_Size	315060
	Coupler	192019
	Coupler_System	367724
	Grouser_Tracks	367823
	Hydraulics_Flow	367823
	Track_Type	310505
	Undercarriage_Pad_Width	309782
	Stick_Length	310437
	Thumb	310366
	Pattern_Changer	310437
	Grouser_Type	310505
	Backhoe_Mounting	331986
	Blade_Type	330823
	Travel_Controls	330823
	Differential_Type	341134
	Steering_Controls	341176
	dtype: int64	5411/0
	acype. Incom	

```
In [123]: df.saledate
Out[123]: 0
                     11/16/2006 0:00
                      3/26/2004 0:00
          2
                      2/26/2004 0:00
          3
                      5/19/2011 0:00
                      7/23/2009 0:00
          412693
                       3/7/2012 0:00
          412694
                      1/28/2012 0:00
          412695
                      1/28/2012 0:00
          412696
                      3/7/2012 0:00
          412697
                      1/28/2012 0:00
          Name: saledate, Length: 412698, dtype: object
In [124]: | df.saledate.dtype
```

Out[124]: dtype('0')

```
In [125]: df.SalePrice.plot.hist()
```

Out[125]: <matplotlib.axes._subplots.AxesSubplot at 0x1816ebcf1c8>



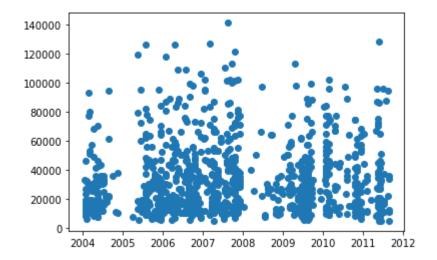
Parsing dates

When we work with time series data, we want to enrich the time & date component as much as possible.

We can do that by telling which of our columns has dates in it using the parse dates parameter.

```
In [127]: df.saledate.dtype
Out[127]: dtype('<M8[ns]')</pre>
In [128]: df.saledate[:1000]
Out[128]: 0
                 2006-11-16
                 2004-03-26
           2
                 2004-02-26
           3
                 2011-05-19
           4
                 2009-07-23
          995
                 2009-07-16
          996
                 2007-06-14
          997
                 2005-09-22
                 2005-07-28
           998
          999
                 2011-06-16
          Name: saledate, Length: 1000, dtype: datetime64[ns]
In [129]: fig, ax =plt.subplots()
          ax.scatter(df["saledate"][:1000],df["SalePrice"][:1000])
```

Out[129]: <matplotlib.collections.PathCollection at 0x18137af4308>



In [130]: df.head()

Out[130]:

	SalesID	SalePrice	MachinelD	ModelID	datasource	auctioneerID	YearMade	MachineHoursCuri
0	1139246	66000.0	999089	3157	121	3.0	2004	
1	1139248	57000.0	117657	77	121	3.0	1996	
2	1139249	10000.0	434808	7009	121	3.0	2001	
3	1139251	38500.0	1026470	332	121	3.0	2001	
4	1139253	11000.0	1057373	17311	121	3.0	2007	

5 rows × 53 columns

4

In [131]: df.head().T

Out[131]:		0	1	2	3	4
	SalesID	1139246	1139248	1139249	1139251	1139253
	SalePrice	66000	57000	10000	38500	11000
	MachinelD	999089	117657	434808	1026470	1057373
	ModelID	3157	77	7009	332	17311
	datasource	121	121	121	121	121
	auctioneerID	3	3	3	3	3
	YearMade	2004	1996	2001	2001	2007
	MachineHoursCurrentMeter	68	4640	2838	3486	722
	UsageBand	Low	Low	High	High	Medium
	saledate	2006-11-16 00:00:00	2004-03-26 00:00:00	2004-02-26 00:00:00	2011-05-19 00:00:00	2009-07-23 00:00:00
	fiModelDesc	521D	950FII	226	PC120-6E	S175
	fiBaseModel	521	950	226	PC120	S175
	fiSecondaryDesc	D	F	NaN	NaN	NaN
	fiModelSeries	NaN	II	NaN	-6E	NaN
	fiModelDescriptor	NaN	NaN	NaN	NaN	NaN
	ProductSize	NaN	Medium	NaN	Small	NaN
	fiProductClassDesc	Wheel Loader - 110.0 to 120.0 Horsepower	Wheel Loader - 150.0 to 175.0 Horsepower	Skid Steer Loader - 1351.0 to 1601.0 Lb Operat	Hydraulic Excavator, Track - 12.0 to 14.0 Metr	Skid Steer Loader - 1601.0 to 1751.0 Lb Operat
	state	Alabama	North Carolina	New York	Texas	New York
	ProductGroup	WL	WL	SSL	TEX	SSL
	ProductGroupDesc	Wheel Loader	Wheel Loader	Skid Steer Loaders	Track Excavators	Skid Steer Loaders
	Drive_System	NaN	NaN	NaN	NaN	NaN
	Enclosure	EROPS w AC	EROPS w AC	OROPS	EROPS w AC	EROPS
	Forks	None or Unspecified	None or Unspecified	None or Unspecified	NaN	None or Unspecified
	Pad_Type	NaN	NaN	NaN	NaN	NaN
	Ride_Control	None or Unspecified	None or Unspecified	NaN	NaN	NaN
	Stick	NaN	NaN	NaN	NaN	NaN
	Transmission	NaN	NaN	NaN	NaN	NaN
	Turbocharged	NaN	NaN	NaN	NaN	NaN
	Blade_Extension	NaN	NaN	NaN	NaN	NaN

	0	1	2	3	4
Blade_Width	NaN	NaN	NaN	NaN	NaN
Enclosure_Type	NaN	NaN	NaN	NaN	NaN
Engine_Horsepower	NaN	NaN	NaN	NaN	NaN
Hydraulics	2 Valve	2 Valve	Auxiliary	2 Valve	Auxiliary
Pushblock	NaN	NaN	NaN	NaN	NaN
Ripper	NaN	NaN	NaN	NaN	NaN
Scarifier	NaN	NaN	NaN	NaN	NaN
Tip_Control	NaN	NaN	NaN	NaN	NaN
Tire_Size	None or Unspecified	23.5	NaN	NaN	NaN
Coupler	None or Unspecified				
Coupler_System	NaN	NaN	None or Unspecified	NaN	None or Unspecified
Grouser_Tracks	NaN	NaN	None or Unspecified	NaN	None or Unspecified
Hydraulics_Flow	NaN	NaN	Standard	NaN	Standard
Track_Type	NaN	NaN	NaN	NaN	NaN
Undercarriage_Pad_Width	NaN	NaN	NaN	NaN	NaN
Stick_Length	NaN	NaN	NaN	NaN	NaN
Thumb	NaN	NaN	NaN	NaN	NaN
Pattern_Changer	NaN	NaN	NaN	NaN	NaN
Grouser_Type	NaN	NaN	NaN	NaN	NaN
Backhoe_Mounting	NaN	NaN	NaN	NaN	NaN
Blade_Type	NaN	NaN	NaN	NaN	NaN
Travel_Controls	NaN	NaN	NaN	NaN	NaN
Differential_Type	Standard	Standard	NaN	NaN	NaN
Steering_Controls	Conventional	Conventional	NaN	NaN	NaN

```
In [132]: df.saledate[:20]
Out[132]: 0
                2006-11-16
                2004-03-26
           2
                2004-02-26
           3
                2011-05-19
           4
                2009-07-23
           5
                2008-12-18
                2004-08-26
           7
                2005-11-17
           8
                2009-08-27
           9
                2007-08-09
           10
                2008-08-21
           11
                2006-08-24
           12
                2005-10-20
           13
                2006-01-26
           14
                2006-01-03
           15
                2006-11-16
           16
                2007-06-14
           17
                2010-01-28
           18
                2006-03-09
           19
                2005-11-17
           Name: saledate, dtype: datetime64[ns]
```

Sort DataFrame by saledate

When working with time series data, its a good practice to sort the data in ascending order of date.

```
In [133]: | df.sort_values(by=["saledate"],inplace=True,ascending=True)
           df.saledate.head(20)
Out[133]: 205615
                    1989-01-17
           274835
                    1989-01-31
           141296
                    1989-01-31
           212552
                    1989-01-31
           62755
                    1989-01-31
           54653
                    1989-01-31
           81383
                    1989-01-31
           204924
                    1989-01-31
           135376
                    1989-01-31
           113390
                    1989-01-31
           113394
                    1989-01-31
           116419
                    1989-01-31
           32138
                    1989-01-31
           127610
                    1989-01-31
           76171
                    1989-01-31
           127000
                    1989-01-31
           128130
                    1989-01-31
           127626
                    1989-01-31
           55455
                    1989-01-31
           55454
                    1989-01-31
           Name: saledate, dtype: datetime64[ns]
```

In [134]: df.head().T

$\wedge \cdot \cdot + \mid$	[12/1	
UU L I	1 1 3 4 1	

	205615	274835	141296	212552	62755
SalesID	1646770	1821514	1505138	1671174	1329056
SalePrice	9500	14000	50000	16000	22000
MachinelD	1126363	1194089	1473654	1327630	1336053
ModelID	8434	10150	4139	8591	4089
datasource	132	132	132	132	132
auctioneerID	18	99	99	99	99
YearMade	1974	1980	1978	1980	1984
MachineHoursCurrentMeter	NaN	NaN	NaN	NaN	NaN
UsageBand	NaN	NaN	NaN	NaN	NaN
saledate	1989-01-17 00:00:00	1989-01-31 00:00:00	1989-01-31 00:00:00	1989-01-31 00:00:00	1989-01-31 00:00:00
fiModelDesc	TD20	A66	D7G	A62	D3B
fiBaseModel	TD20	A66	D7	A62	D3
fiSecondaryDesc	NaN	NaN	G	NaN	В
fiModelSeries	NaN	NaN	NaN	NaN	NaN
fiModelDescriptor	NaN	NaN	NaN	NaN	NaN
ProductSize	Medium	NaN	Large	NaN	NaN
fiProductClassDesc	Track Type Tractor, Dozer - 105.0 to 130.0 Hor	Wheel Loader - 120.0 to 135.0 Horsepower	Track Type Tractor, Dozer - 190.0 to 260.0 Hor	Wheel Loader - Unidentified	Track Type Tractor, Dozer - 20.0 to 75.0 Horse
state	Texas	Florida	Florida	Florida	Florida
ProductGroup	TTT	WL	TTT	WL	TTT
ProductGroupDesc	Track Type Tractors	Wheel Loader	Track Type Tractors	Wheel Loader	Track Type Tractors
Drive_System	NaN	NaN	NaN	NaN	NaN
Enclosure	OROPS	OROPS	OROPS	EROPS	OROPS
Forks	NaN	None or Unspecified	NaN	None or Unspecified	NaN
Pad_Type	NaN	NaN	NaN	NaN	NaN
Ride_Control	NaN	None or Unspecified	NaN	None or Unspecified	NaN
Stick	NaN	NaN	NaN	NaN	NaN
Transmission	Direct Drive	NaN	Standard	NaN	Standard
Turbocharged	NaN	NaN	NaN	NaN	NaN
Blade_Extension	NaN	NaN	NaN	NaN	NaN
Blade_Width	NaN	NaN	NaN	NaN	NaN

	205615	274835	141296	212552	62755
Enclosure_Type	NaN	NaN	NaN	NaN	NaN
Engine_Horsepower	NaN	NaN	NaN	NaN	NaN
Hydraulics	2 Valve				
Pushblock	NaN	NaN	NaN	NaN	NaN
Ripper	None or Unspecified	NaN	None or Unspecified	NaN	None or Unspecified
Scarifier	NaN	NaN	NaN	NaN	NaN
Tip_Control	NaN	NaN	NaN	NaN	NaN
Tire_Size	NaN	None or Unspecified	NaN	None or Unspecified	NaN
Coupler	NaN	None or Unspecified	NaN	None or Unspecified	NaN
Coupler_System	NaN	NaN	NaN	NaN	NaN
Grouser_Tracks	NaN	NaN	NaN	NaN	NaN
Hydraulics_Flow	NaN	NaN	NaN	NaN	NaN
Track_Type	NaN	NaN	NaN	NaN	NaN
Undercarriage_Pad_Width	NaN	NaN	NaN	NaN	NaN
Stick_Length	NaN	NaN	NaN	NaN	NaN
Thumb	NaN	NaN	NaN	NaN	NaN
Pattern_Changer	NaN	NaN	NaN	NaN	NaN
Grouser_Type	NaN	NaN	NaN	NaN	NaN
Backhoe_Mounting	None or Unspecified	NaN	None or Unspecified	NaN	None or Unspecified
Blade_Type	Straight	NaN	Straight	NaN	PAT
Travel_Controls	None or Unspecified	NaN	None or Unspecified	NaN	Lever
Differential_Type	NaN	Standard	NaN	Standard	NaN
Steering_Controls	NaN	Conventional	NaN	Conventional	NaN

In [135]: ## Make a copy of original dataframe, so that when we manipulate the copy we stil
df_tmp=df.copy()

In [136]: df_tmp.head()

Out[136]:

	SalesID	SalePrice	MachinelD	ModelID	datasource	auctioneerID	YearMade	MachineHou
205615	1646770	9500.0	1126363	8434	132	18.0	1974	
274835	1821514	14000.0	1194089	10150	132	99.0	1980	
141296	1505138	50000.0	1473654	4139	132	99.0	1978	
212552	1671174	16000.0	1327630	8591	132	99.0	1980	
62755	1329056	22000.0	1336053	4089	132	99.0	1984	

5 rows × 53 columns

```
In [137]: ## We have done some feature engineering which required in timeSeries kind of pro
    df_tmp["saleYear"]=df_tmp.saledate.dt.year
    df_tmp["saleMonth"]=df_tmp.saledate.dt.month
    df_tmp["saleDay"]=df_tmp.saledate.dt.day
    df_tmp["saleDayOfWeek"]=df_tmp.saledate.dt.dayofweek
    df_tmp["saleDayOfYear"]=df_tmp.saledate.dt.dayofyear
```

In [138]: df_tmp.head().T

Out[138]:

	205615	274835	141296	212552	62755
SalesID	1646770	1821514	1505138	1671174	1329056
SalePrice	9500	14000	50000	16000	22000
MachinelD	1126363	1194089	1473654	1327630	1336053
ModelID	8434	10150	4139	8591	4089
datasource	132	132	132	132	132
auctioneerID	18	99	99	99	99
YearMade	1974	1980	1978	1980	1984
MachineHoursCurrentMeter	NaN	NaN	NaN	NaN	NaN
UsageBand	NaN	NaN	NaN	NaN	NaN
saledate	1989-01-17 00:00:00	1989-01-31 00:00:00	1989-01-31 00:00:00	1989-01-31 00:00:00	1989-01-31 00:00:00
fiModelDesc	TD20	A66	D7G	A62	D3B
fiBaseModel	TD20	A66	D7	A62	D3
fiSecondaryDesc	NaN	NaN	G	NaN	В
fiModelSeries	NaN	NaN	NaN	NaN	NaN
fiModelDescriptor	NaN	NaN	NaN	NaN	NaN
ProductSize	Medium	NaN	Large	NaN	NaN
fiProductClassDesc	Track Type Tractor, Dozer - 105.0 to 130.0 Hor	Wheel Loader - 120.0 to 135.0 Horsepower	Track Type Tractor, Dozer - 190.0 to 260.0 Hor	Wheel Loader - Unidentified	Track Type Tractor, Dozer - 20.0 to 75.0 Horse
state	Texas	Florida	Florida	Florida	Florida
ProductGroup	TTT	WL	TTT	WL	TTT
ProductGroupDesc	Track Type Tractors	Wheel Loader	Track Type Tractors	Wheel Loader	Track Type Tractors
Drive_System	NaN	NaN	NaN	NaN	NaN
Enclosure	OROPS	OROPS	OROPS	EROPS	OROPS
Forks	NaN	None or Unspecified	NaN	None or Unspecified	NaN
Pad_Type	NaN	NaN	NaN	NaN	NaN
Ride_Control	NaN	None or Unspecified	NaN	None or Unspecified	NaN
Stick	NaN	NaN	NaN	NaN	NaN
Transmission	Direct Drive	NaN	Standard	NaN	Standard
Turbocharged	NaN	NaN	NaN	NaN	NaN
Blade_Extension	NaN	NaN	NaN	NaN	NaN

	205615	274835	141296	212552	62755
Enclosure_Type	NaN	NaN	NaN	NaN	NaN
Engine_Horsepower	NaN	NaN	NaN	NaN	NaN
Hydraulics	2 Valve				
Pushblock	NaN	NaN	NaN	NaN	NaN
Ripper	None or Unspecified	NaN	None or Unspecified	NaN	None or Unspecified
Scarifier	NaN	NaN	NaN	NaN	NaN
Tip_Control	NaN	NaN	NaN	NaN	NaN
Tire_Size	NaN	None or Unspecified	NaN	None or Unspecified	NaN
Coupler	NaN	None or Unspecified	NaN	None or Unspecified	NaN
Coupler_System	NaN	NaN	NaN	NaN	NaN
Grouser_Tracks	NaN	NaN	NaN	NaN	NaN
Hydraulics_Flow	NaN	NaN	NaN	NaN	NaN
Track_Type	NaN	NaN	NaN	NaN	NaN
Undercarriage_Pad_Width	NaN	NaN	NaN	NaN	NaN
Stick_Length	NaN	NaN	NaN	NaN	NaN
Thumb	NaN	NaN	NaN	NaN	NaN
Pattern_Changer	NaN	NaN	NaN	NaN	NaN
Grouser_Type	NaN	NaN	NaN	NaN	NaN
Backhoe_Mounting	None or Unspecified	NaN	None or Unspecified	NaN	None or Unspecified
Blade_Type	Straight	NaN	Straight	NaN	PAT
Travel_Controls	None or Unspecified	NaN	None or Unspecified	NaN	Lever
Differential_Type	NaN	Standard	NaN	Standard	NaN
Steering_Controls	NaN	Conventional	NaN	Conventional	NaN
saleYear	1989	1989	1989	1989	1989
saleMonth	1	1	1	1	1
saleDay	17	31	31	31	31
saleDayOfWeek	1	1	1	1	1
saleDayOfYear	17	31	31	31	31

```
In [140]: df tmp.head(20).T
Out[140]:
                                            205615
                                                         274835
                                                                    141296
                                                                                  212552
                                                                                              62755
                                                                                                          5
                               SalesID
                                           1646770
                                                        1821514
                                                                   1505138
                                                                                 1671174
                                                                                            1329056
                                                                                                        130
                                                                                                          2
                              SalePrice
                                              9500
                                                                      50000
                                                                                   16000
                                                                                              22000
                                                          14000
                             MachineID
                                           1126363
                                                                    1473654
                                                                                 1327630
                                                                                            1336053
                                                                                                        118
                                                        1194089
                               ModelID
                                              8434
                                                          10150
                                                                       4139
                                                                                    8591
                                                                                               4089
                            datasource
                                               132
                                                            132
                                                                        132
                                                                                     132
                                                                                                132
                           auctioneerID
                                                18
                                                             99
                                                                         99
                                                                                      99
                                                                                                 99
                             YearMade
                                              1974
                                                           1980
                                                                       1978
                                                                                    1980
                                                                                                1984
             MachineHoursCurrentMeter
                                                                                    NaN
                                                                                                NaN
                                              NaN
                                                           NaN
                                                                       NaN
                            UsageBand
                                              NaN
                                                           NaN
                                                                       NaN
                                                                                    NaN
                                                                                                NaN
                           fiModelDesc
                                                                                                D3B
                                             TD20
                                                            A66
                                                                       D7G
                                                                                     A62
                           fiBaseModel
                                             TD20
                                                            A66
                                                                        D7
                                                                                     A62
                                                                                                 D3
```

```
In [141]: df_tmp.saledate
          AttributeError
                                                     Traceback (most recent call last)
          <ipython-input-141-0b8254724953> in <module>
           ----> 1 df_tmp.saledate
          ~\AppData\Roaming\Python\Python37\site-packages\pandas\core\generic.py in __get
          attr__(self, name)
                               if self._info_axis._can_hold_identifiers_and_holds_name(nam
             5139
          e):
             5140
                                   return self[name]
           -> 5141
                               return object.__getattribute__(self, name)
             5142
             5143
                       def __setattr__(self, name: str, value) -> None:
          AttributeError: 'DataFrame' object has no attribute 'saledate'
```

In [144]: df_tmp.state.value_counts()

In [144]:	df_tmp.state.value	e_counts(
Out[144]:		67320
	Texas	53110
	California	29761
	Washington	16222
	Georgia	14633
	Maryland	13322
	Mississippi	13240
	Ohio	12369
	Illinois	11540
	Colorado	11529
	New Jersey	11156
	North Carolina	10636
	Tennessee	10298
	Alabama	10292
	Pennsylvania	10234
	South Carolina	9951
	Arizona	9364
	New York	8639
	Connecticut	8276
	Minnesota	7885
	Missouri	7178
	Nevada	6932
	Louisiana	6627
	Kentucky	5351
	Maine	5096
	Indiana	4124
	Arkansas	3933
	New Mexico	3631
	Utah	3046
	Unspecified	2801
	Wisconsin	2745
	New Hampshire	2738
	Virginia	2353
	Idaho	2025
	Oregon	1911
	Michigan	1831
	Wyoming	1672
	Montana	1336
	Iowa	1336
	Oklahoma	1326
	Nebraska	866
	West Virginia	840
	Kansas	667
	Delaware	510
	North Dakota	480
	Alaska	430
	Massachusetts	347
	Vermont	300
	South Dakota	244
	Hawaii	118
	Rhode Island	83
	Puerto Rico	42
	Washington DC	2
	Name: state, dtype	e: int64

```
In [145]: len(df_tmp)
```

Out[145]: 412698

5. Modelling

We have done enough EDA (we could always do more) but let's start to do some model-driven EDA

```
In [146]: ## Let's build a machine learning model
          from sklearn.ensemble import RandomForestRegressor
          model=RandomForestRegressor(n jobs=1,
                                       random state=42) # same as np.seed(42)
          model.fit(df_tmp.drop("SalePrice", axis=1), df_tmp["SalePrice"])
          ValueError
                                                      Traceback (most recent call last)
          <ipython-input-146-b3ff0195a76f> in <module>
                3 model=RandomForestRegressor(n jobs=1,
                                               random_state=42) # same as np.seed(42)
                4
          ----> 5 model.fit(df tmp.drop("SalePrice", axis=1), df tmp["SalePrice"])
          ~\AppData\Roaming\Python\Python37\site-packages\sklearn\ensemble\_forest.py in
          fit(self, X, y, sample_weight)
              303
                               )
              304
                           X, y = self._validate_data(X, y, multi_output=True,
          --> 305
                                                      accept sparse="csc", dtype=DTYPE)
              306
                           if sample weight is not None:
              307
                               sample_weight = _check_sample_weight(sample_weight, X)
          ~\AppData\Roaming\Python\Python37\site-packages\sklearn\base.py in validate da
          ta(self, X, y, reset, validate_separately, **check_params)
                                   y = check_array(y, **check_y_params)
              431
              432
          --> 433
                                   X, y = \text{check}_X_y(X, y, **\text{check}_params)
              434
                               out = X, y
              435
          ~\AppData\Roaming\Python\Python37\site-packages\sklearn\utils\validation.py in
          inner_f(*args, **kwargs)
               61
                               extra_args = len(args) - len(all_args)
               62
                               if extra args <= 0:</pre>
                                   return f(*args, **kwargs)
          ---> 63
               64
               65
                               # extra_args > 0
          ~\AppData\Roaming\Python\Python37\site-packages\sklearn\utils\validation.py in
          check_X_y(X, y, accept_sparse, accept_large_sparse, dtype, order, copy, force_a
          11 finite, ensure 2d, allow nd, multi output, ensure min samples, ensure min fe
          atures, y numeric, estimator)
              876
                                       ensure_min_samples=ensure_min_samples,
              877
                                       ensure min features=ensure min features,
          --> 878
                                       estimator=estimator)
              879
                       if multi output:
              880
                           y = check array(y, accept sparse='csr', force all finite=True,
          ~\AppData\Roaming\Python\Python37\site-packages\sklearn\utils\validation.py in
          inner_f(*args, **kwargs)
               61
                               extra args = len(args) - len(all args)
               62
                               if extra args <= 0:</pre>
          ---> 63
                                   return f(*args, **kwargs)
               64
               65
                               # extra_args > 0
```

~\AppData\Roaming\Python\Python37\site-packages\sklearn\utils\validation.py in check_array(array, accept_sparse, accept_large_sparse, dtype, order, copy, forc

```
e_all_finite, ensure_2d, allow_nd, ensure_min_samples, ensure_min_features, est
imator)
   671
                            array = array.astype(dtype, casting="unsafe", copy=
False)
    672
                        else:
                            array = np.asarray(array, order=order, dtype=dtype)
--> 673
                    except ComplexWarning as complex warning:
    674
                        raise ValueError("Complex data not supported\n"
    675
~\AppData\Roaming\Python\Python37\site-packages\numpy\core\ asarray.py in asarr
ay(a, dtype, order)
     81
     82
---> 83
            return array(a, dtype, copy=False, order=order)
     84
     85
~\AppData\Roaming\Python\Python37\site-packages\pandas\core\generic.py in __arr
ay (self, dtype)
   1779
            def __array__(self, dtype=None) -> np.ndarray:
   1780
-> 1781
                return np.asarray(self. values, dtype=dtype)
   1782
   1783
            def __array_wrap__(self, result, context=None):
~\AppData\Roaming\Python\Python37\site-packages\numpy\core\ asarray.py in asarr
ay(a, dtype, order)
     81
            .....
     82
            return array(a, dtype, copy=False, order=order)
---> 83
     84
     85
ValueError: could not convert string to float: 'Low'
```

In [147]: df_tmp.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 412698 entries, 205615 to 409203
Data columns (total 57 columns):

# 	Columns (total 5/ columns	Non-Null Count	Dtype
0	SalesID	412698 non-null	int64
1	SalePrice	412698 non-null	float64
2	MachineID	412698 non-null	int64
3	ModelID	412698 non-null	int64
4	datasource	412698 non-null	int64
5	auctioneerID	392562 non-null	float64
6	YearMade	412698 non-null	int64
7	MachineHoursCurrentMeter	147504 non-null	float64
8	UsageBand	73670 non-null	object
9	fiModelDesc	412698 non-null	object
10	fiBaseModel	412698 non-null	object
11	fiSecondaryDesc	271971 non-null	object
12	fiModelSeries	58667 non-null	object
13	fiModelDescriptor	74816 non-null	object
14	ProductSize	196093 non-null	object
15	fiProductClassDesc	412698 non-null	object
16	state	412698 non-null	object
17	ProductGroup	412698 non-null	object
18	ProductGroupDesc	412698 non-null	object
19	Drive_System	107087 non-null	object
20	Enclosure	412364 non-null	object
21	Forks	197715 non-null	object
22	Pad_Type	81096 non-null	object
23	Ride_Control	152728 non-null	object
24	Stick	81096 non-null	object
25	Transmission	188007 non-null	object
26	Turbocharged	81096 non-null	object
27	Blade_Extension	25983 non-null	object
28	Blade_Width	25983 non-null	object
29	Enclosure_Type	25983 non-null	object
30	Engine_Horsepower	25983 non-null	3
31	Hydraulics	330133 non-null	object
32	Pushblock	25983 non-null	object
33	Ripper	106945 non-null	object
34	Scarifier	25994 non-null	object
35	Tip_Control	25983 non-null	object
36	Tire_Size	97638 non-null	object
37	Coupler Cystem	220679 non-null	object
38	Coupler_System	44974 non-null	object
39 40	Grouser_Tracks	44875 non-null 44875 non-null	object object
41	Hydraulics_Flow	102193 non-null	•
41 42	Track_Type Undercarriage_Pad_Width	102193 Non-Null	object object
42 43		102916 Non-Null	•
43 44	Stick_Length Thumb	102332 non-null	object object
44 45	Pattern_Changer	102332 Non-Null	object
45 46	Grouser_Type	102201 non-null	object
47	Backhoe_Mounting	80712 non-null	object
48	Blade_Type	81875 non-null	object
49	Travel_Controls	81877 non-null	object
-FJ	4461_661161 013	OLO,, HOH HULL	

```
50 Differential_Type
                              71564 non-null
                                              object
 51 Steering_Controls
                                              object
                              71522 non-null
 52 saleYear
                              412698 non-null
                                              int64
 53 saleMonth
                              412698 non-null
                                              int64
 54 saleDay
                              412698 non-null
                                              int64
 55 saleDayOfWeek
                              412698 non-null
                                              int64
 56 saleDayOfYear
                              412698 non-null int64
dtypes: float64(3), int64(10), object(44)
memory usage: 182.6+ MB
```

```
In [148]: df_tmp["UsageBand"].dtype
```

Out[148]: dtype('0')

In [149]: df_tmp.isna().sum()

TH [T45].	u	
Out[149]:	SalesID	0
	SalePrice	0
	MachineID	0
	ModelID	0
	datasource	0
	auctioneerID	20136
	YearMade	0
	MachineHoursCurrentMeter	265194
	UsageBand	339028
	fiModelDesc	0
	fiBaseModel	0
	fiSecondaryDesc	140727
	fiModelSeries	354031
	fiModelDescriptor	337882
	ProductSize	216605
	fiProductClassDesc	0
	state	0
	ProductGroup	0
	ProductGroupDesc	0
	Drive_System	305611
	Enclosure	334
	Forks	214983
	Pad_Type	331602
	Ride_Control	259970
	Stick	331602
	Transmission	224691
	Turbocharged	331602
	Blade_Extension	386715
	Blade_Width	386715
	Enclosure_Type	386715
	Engine_Horsepower	386715
	Hydraulics	82565
	Pushblock	386715
	Ripper	305753
	Scarifier	386704
	Tip_Control	386715
	Tire_Size	315060
	Coupler	192019
	Coupler_System	367724
	Grouser_Tracks	367823
	Hydraulics_Flow	367823
	Track_Type	310505
	Undercarriage_Pad_Width	309782
	Stick_Length	310437
	Thumb	310366
	Pattern_Changer	310437
	Grouser_Type	310505
	Backhoe_Mounting	331986
	Blade_Type	330823
	Travel_Controls	330821
	Differential_Type	341134
	Steering_Controls	341176
	saleYear	0
	saleMonth	0
	saleDay	0

saleDayOfWeek	e
saleDayOfYear	e
dtvpe: int64	

Convert String to Categorical data

One way we can turn all our data into numbers is by converting them into pandas categories.

```
In [150]: pd.api.types.is_string_dtype(df_tmp["UsageBand"])
Out[150]: True
```

UsageBand fiModelDesc fiBaseModel fiSecondaryDesc fiModelSeries fiModelDescriptor ProductSize fiProductClassDesc state ProductGroup ProductGroupDesc Drive_System Enclosure Forks Pad Type Ride Control Stick Transmission Turbocharged Blade_Extension Blade_Width Enclosure Type Engine_Horsepower Hydraulics Pushblock Ripper Scarifier Tip Control Tire_Size Coupler Coupler_System **Grouser Tracks** Hydraulics_Flow Track Type Undercarriage_Pad_Width Stick_Length Thumb Pattern_Changer Grouser Type Backhoe Mounting Blade_Type Travel_Controls

Differential_Type Steering_Controls

```
In [152]: # This will turn all the string values into categorical values

for label, content in df_tmp.items():
    if pd.api.types.is_string_dtype(content):
        df_tmp[label]=content.astype("category").cat.as_ordered()
```

In [153]: df_tmp.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 412698 entries, 205615 to 409203
Data columns (total 57 columns):

# 	Columns (total 5/ columns	Non-Null Count	Dtype
0	SalesID	412698 non-null	int64
1	SalePrice	412698 non-null	float64
2	MachineID	412698 non-null	int64
3	ModelID	412698 non-null	int64
4	datasource	412698 non-null	int64
5	auctioneerID	392562 non-null	float64
6	YearMade	412698 non-null	int64
7	MachineHoursCurrentMeter	147504 non-null	float64
8	UsageBand	73670 non-null	category
9	fiModelDesc	412698 non-null	category
10	fiBaseModel	412698 non-null	category
11	fiSecondaryDesc	271971 non-null	category
12	fiModelSeries	58667 non-null	category
13	fiModelDescriptor	74816 non-null	category
14	ProductSize	196093 non-null	category
15	fiProductClassDesc	412698 non-null	category
16	state	412698 non-null	category
17	ProductGroup	412698 non-null	category
18	ProductGroupDesc	412698 non-null	category
19	Drive_System	107087 non-null	category
20	Enclosure	412364 non-null	category
21	Forks	197715 non-null	category
22	Pad_Type	81096 non-null	category
23	Ride_Control	152728 non-null	category
24	Stick	81096 non-null	category
25	Transmission	188007 non-null	category
26	Turbocharged	81096 non-null	category
27	Blade_Extension	25983 non-null	category
28	Blade_Width	25983 non-null	category
29	Enclosure_Type	25983 non-null	category
30	Engine_Horsepower	25983 non-null	category
31	Hydraulics	330133 non-null	category
32	Pushblock	25983 non-null	category
33	Ripper	106945 non-null	category
34	Scarifier	25994 non-null	category
35	Tip_Control	25983 non-null	category
36	Tire_Size	97638 non-null	category
37	Coupler	220679 non-null	category
38	Coupler_System	44974 non-null	category
39	Grouser_Tracks	44875 non-null	category
40	Hydraulics_Flow	44875 non-null	category
41 42	Track_Type	102193 non-null	category
42 43	Undercarriage_Pad_Width	102916 non-null 102261 non-null	category
	Stick_Length		category
44 45	Thumb	102332 non-null 102261 non-null	category
45 46	Pattern_Changer	102261 non-null	category
46 47	Grouser_Type Backhoe_Mounting	80712 non-null	category
47 48	Blade_Type	81875 non-null	category category
49	Travel_Controls	81877 non-null	category
77	11 4461 601161 013	OTO// HOH-HUTT	cacegor y

```
51 Steering_Controls
                                          71522 non-null
                                                           category
           52 saleYear
                                          412698 non-null
                                                           int64
           53 saleMonth
                                          412698 non-null
                                                           int64
           54 saleDay
                                          412698 non-null
                                                           int64
           55 saleDayOfWeek
                                         412698 non-null
                                                           int64
           56 saleDayOfYear
                                         412698 non-null
                                                           int64
          dtypes: category(44), float64(3), int64(10)
          memory usage: 63.3 MB
In [154]: | df tmp.state.cat.categories
Out[154]: Index(['Alabama', 'Alaska', 'Arizona', 'Arkansas', 'California', 'Colorado',
                  'Connecticut', 'Delaware', 'Florida', 'Georgia', 'Hawaii', 'Idaho',
                  'Illinois', 'Indiana', 'Iowa', 'Kansas', 'Kentucky', 'Louisiana',
                  'Maine', 'Maryland', 'Massachusetts', 'Michigan', 'Minnesota',
                  'Mississippi', 'Missouri', 'Montana', 'Nebraska', 'Nevada',
                  'New Hampshire', 'New Jersey', 'New Mexico', 'New York',
                  'North Carolina', 'North Dakota', 'Ohio', 'Oklahoma', 'Oregon',
                 'Pennsylvania', 'Puerto Rico', 'Rhode Island', 'South Carolina',
                 'South Dakota', 'Tennessee', 'Texas', 'Unspecified', 'Utah', 'Vermont',
                  'Virginia', 'Washington', 'Washington DC', 'West Virginia', 'Wisconsin',
                  'Wyoming'],
                dtype='object')
In [155]: | df_tmp.state.cat.codes
Out[155]: 205615
                    43
          274835
                     8
          141296
                     8
          212552
                     8
          62755
                     8
          410879
                     4
          412476
                     4
          411927
                     4
          407124
                     4
                     4
          409203
          Length: 412698, dtype: int8
          Save preprocessed data
In [156]: # Export current tmp dataFrame
          df_tmp.to_csv("data/train_tmp.csv",index=False)
In [157]: # Import preprocessed data
          df tmp=pd.read csv("data/train tmp.csv",
                              low memory=False)
```

71564 non-null

category

50 Differential Type

In [158]: df_tmp.head().T

Out[158]:		0	1	2	3	4
	SalesID	1646770	1821514	1505138	1671174	1329056
	SalePrice	9500	14000	50000	16000	22000
	MachinelD	1126363	1194089	1473654	1327630	1336053
	ModelID	8434	10150	4139	8591	4089
	datasource	132	132	132	132	132
	auctioneerID	18	99	99	99	99
	YearMade	1974	1980	1978	1980	1984
	MachineHoursCurrentMeter	NaN	NaN	NaN	NaN	NaN
	UsageBand	NaN	NaN	NaN	NaN	NaN
	fiModelDesc	TD20	A66	D7G	A62	D3B
	fiBaseModel	TD20	A66	D7	A62	D3
	fiSecondaryDesc	NaN	NaN	G	NaN	В
	fiModelSeries	NaN	NaN	NaN	NaN	NaN
	fiModelDescriptor	NaN	NaN	NaN	NaN	NaN
	ProductSize	Medium	NaN	Large	NaN	NaN
	fiProductClassDesc	Track Type Tractor, Dozer - 105.0 to 130.0 Hor	Wheel Loader - 120.0 to 135.0 Horsepower	Track Type Tractor, Dozer - 190.0 to 260.0 Hor	Wheel Loader - Unidentified	Track Type Tractor, Dozer - 20.0 to 75.0 Horse
	state	Texas	Florida	Florida	Florida	Florida
	ProductGroup	TTT	WL	TTT	WL	TTT
	ProductGroupDesc	Track Type Tractors	Wheel Loader	Track Type Tractors	Wheel Loader	Track Type Tractors
	Drive_System	NaN	NaN	NaN	NaN	NaN
	Enclosure	OROPS	OROPS	OROPS	EROPS	OROPS
	Forks	NaN	None or Unspecified	NaN	None or Unspecified	NaN
	Pad_Type	NaN	NaN	NaN	NaN	NaN
	Ride_Control	NaN	None or Unspecified	NaN	None or Unspecified	NaN
	Stick	NaN	NaN	NaN	NaN	NaN
	Transmission	Direct Drive	NaN	Standard	NaN	Standard
	Turbocharged	NaN	NaN	NaN	NaN	NaN
	Blade_Extension	NaN	NaN	NaN	NaN	NaN
	Blade_Width	NaN	NaN	NaN	NaN	NaN
	Enclosure_Type	NaN	NaN	NaN	NaN	NaN

	0	1	2	3	4
Engine_Horsepower	NaN	NaN	NaN	NaN	NaN
Hydraulics	2 Valve				
Pushblock	NaN	NaN	NaN	NaN	NaN
Ripper	None or Unspecified	NaN	None or Unspecified	NaN	None or Unspecified
Scarifier	NaN	NaN	NaN	NaN	NaN
Tip_Control	NaN	NaN	NaN	NaN	NaN
Tire_Size	NaN	None or Unspecified	NaN	None or Unspecified	NaN
Coupler	NaN	None or Unspecified	NaN	None or Unspecified	NaN
Coupler_System	NaN	NaN	NaN	NaN	NaN
Grouser_Tracks	NaN	NaN	NaN	NaN	NaN
Hydraulics_Flow	NaN	NaN	NaN	NaN	NaN
Track_Type	NaN	NaN	NaN	NaN	NaN
ndercarriage_Pad_Width	NaN	NaN	NaN	NaN	NaN
Stick_Length	NaN	NaN	NaN	NaN	NaN
Thumb	NaN	NaN	NaN	NaN	NaN
Pattern_Changer	NaN	NaN	NaN	NaN	NaN
Grouser_Type	NaN	NaN	NaN	NaN	NaN
Backhoe_Mounting	None or Unspecified	NaN	None or Unspecified	NaN	None or Unspecified
Blade_Type	Straight	NaN	Straight	NaN	PAT
Travel_Controls	None or Unspecified	NaN	None or Unspecified	NaN	Lever
Differential_Type	NaN	Standard	NaN	Standard	NaN
Steering_Controls	NaN	Conventional	NaN	Conventional	NaN
saleYear	1989	1989	1989	1989	1989
saleMonth	1	1	1	1	1
saleDay	17	31	31	31	31
saleDayOfWeek	1	1	1	1	1
saleDayOfYear	17	31	31	31	31

In [159]: df_tmp.isnull().sum()

TH [T33].	d1_cmp:13Hd11():3dm()	
Out[159]:	SalesID	0
	SalePrice	0
	MachineID	0
	ModelID	0
	datasource	0
	auctioneerID	20136
	YearMade	0
	MachineHoursCurrentMeter	265194
	UsageBand	339028
	fiModelDesc	0
	fiBaseModel	0
	fiSecondaryDesc	140727
	fiModelSeries	354031
	fiModelDescriptor	337882
	ProductSize	216605
	fiProductClassDesc	0
	state	0
	ProductGroup	0
	ProductGroupDesc	0
	Drive_System	305611
	Enclosure	334
	Forks	214983
	Pad_Type	331602
	Ride_Control	259970
	Stick	331602
	Transmission	224691
	Turbocharged	331602
	Blade_Extension	386715
	Blade_Width	386715
	Enclosure_Type	386715
	Engine_Horsepower	386715
	Hydraulics	82565
	Pushblock	386715
	Ripper	305753
	Scarifier	386704
	Tip_Control	386715
	Tire_Size	315060
	Coupler	192019
	Coupler_System	367724
	Grouser_Tracks	367823
	Hydraulics_Flow	367823
	Track_Type	310505
	Undercarriage_Pad_Width	309782
	Stick_Length	310437
	Thumb	310366
	Pattern_Changer	310437
	Grouser_Type	310505
	Backhoe_Mounting	331986
	Blade_Type	330823
	Travel_Controls	330821
	Differential_Type	341134
	Steering_Controls	341176
	saleYear	0
	saleMonth	0
	saleDay	0

Fill the missing values

In [160]: # First check the numerical columns

First fill the numeric missing values

```
for label,content in df tmp.items():
              if pd.api.types.is numeric dtype(content):
                  print(label)
          SalesID
          SalePrice
          MachineID
          ModelID
          datasource
          auctioneerID
          YearMade
          MachineHoursCurrentMeter
          saleYear
          saleMonth
          saleDay
          saleDayOfWeek
          saleDayOfYear
In [161]: # Check for which numeric columns have null values
          for label, content in df_tmp.items():
              if pd.api.types.is numeric dtype(content):
                  if pd.isnull(content).sum():
                      print(label)
          auctioneerID
          MachineHoursCurrentMeter
In [162]: # Fill numeric rows with median
          for label, content in df_tmp.items():
              if pd.api.types.is_numeric_dtype(content):
                  if pd.isnull(content).sum():
                       # Add a binary column which tells us if the data was missing or not
                      df_tmp[label+"_is_missing"]=pd.isnull(content)
                      # Fill missing numeric values with median
                      df_tmp[label]=content.fillna(content.median())
In [164]: # Check if there is null numeric value
          for label, content in df_tmp.items():
              if pd.api.types.is_numeric_dtype(content):
                  if pd.isnull(content).sum():
                       print(label)
```

In [165]: # Checkk to see how many examples were missing
df_tmp.auctioneerID_is_missing.value_counts()

Out[165]: False 392562 True 20136

Name: auctioneerID_is_missing, dtype: int64

```
In [166]: df_tmp.isna().sum()
Out[166]: SalesID
                                                         0
                                                         0
           SalePrice
           MachineID
                                                          0
           ModelID
                                                          0
           datasource
                                                          0
           auctioneerID
                                                          0
                                                          0
           YearMade
                                                         0
           MachineHoursCurrentMeter
                                                    339028
           UsageBand
           fiModelDesc
                                                         0
           fiBaseModel
                                                          0
           fiSecondaryDesc
                                                    140727
           fiModelSeries
                                                    354031
           fiModelDescriptor
                                                    337882
           ProductSize
                                                    216605
           fiProductClassDesc
                                                         0
                                                         0
           state
           ProductGroup
                                                         0
                                                         0
           ProductGroupDesc
           Drive System
                                                    305611
           Enclosure
                                                        334
                                                    214983
           Forks
           Pad_Type
                                                    331602
           Ride_Control
                                                    259970
           Stick
                                                    331602
           Transmission
                                                    224691
           Turbocharged
                                                    331602
           Blade_Extension
                                                    386715
           Blade Width
                                                    386715
           Enclosure_Type
                                                    386715
           Engine_Horsepower
                                                    386715
           Hydraulics
                                                     82565
           Pushblock
                                                    386715
           Ripper
                                                    305753
           Scarifier
                                                    386704
           Tip_Control
                                                    386715
           Tire_Size
                                                    315060
           Coupler
                                                    192019
           Coupler_System
                                                    367724
           Grouser_Tracks
                                                    367823
           Hydraulics_Flow
                                                    367823
           Track_Type
                                                    310505
           Undercarriage_Pad_Width
                                                    309782
           Stick Length
                                                    310437
           Thumb
                                                    310366
           Pattern_Changer
                                                    310437
           Grouser_Type
                                                    310505
           Backhoe Mounting
                                                    331986
           Blade_Type
                                                    330823
           Travel_Controls
                                                    330821
           Differential_Type
                                                    341134
           Steering_Controls
                                                    341176
           saleYear
                                                         0
           saleMonth
                                                         0
           saleDay
                                                          0
```

Filling and turning categorical values with numerical values

```
In [167]: # Find the categorical column with null values
          for label, content in df_tmp.items():
              if not pd.api.types.is numeric dtype(content):
                  if pd.isnull(content).sum():
                       print(label)
          UsageBand
          fiSecondaryDesc
          fiModelSeries
          fiModelDescriptor
          ProductSize
          Drive_System
          Enclosure
          Forks
          Pad_Type
          Ride_Control
          Stick
          Transmission
          Turbocharged
          Blade Extension
          Blade_Width
          Enclosure_Type
          Engine Horsepower
          Hydraulics
          Pushblock
          Ripper
          Scarifier
          Tip_Control
          Tire_Size
          Coupler
          Coupler_System
          Grouser_Tracks
          Hydraulics Flow
          Track_Type
          Undercarriage_Pad_Width
          Stick_Length
          Thumb
          Pattern_Changer
          Grouser_Type
          Backhoe_Mounting
          Blade_Type
          Travel Controls
          Differential Type
```

Steering_Controls

```
In [168]: pd.Categorical(df tmp["UsageBand"]).codes+1
Out[168]: array([0, 0, 0, ..., 0, 0, 0], dtype=int8)
In [169]: # Turn categorical values into number and fill missing values
           for label, content in df tmp.items():
               if not pd.api.types.is_numeric_dtype(content):
                    # Add binary columns to indicate whether sample had missing values or not
                    df_tmp[label+"_is_missing"]=pd.isnull(content)
                    # Turn categories into numbers and add +1
                    df tmp[label]=pd.Categorical(content).codes+1
In [170]: df_tmp.info()
           <class 'pandas.core.frame.DataFrame'>
           RangeIndex: 412698 entries, 0 to 412697
           Columns: 103 entries, SalesID to Steering_Controls_is_missing
           dtypes: bool(46), float64(3), int16(4), int64(10), int8(40)
           memory usage: 77.9 MB
In [171]: df tmp.head().T
Out[171]:
                                             0
                                                     1
                                                              2
                               SalesID 1646770 1821514 1505138 1671174 1329056
                              SalePrice
                                          9500
                                                 14000
                                                          50000
                                                                           22000
                                                                  16000
                                      1126363
                             MachinelD
                                               1194089 1473654 1327630 1336053
                               ModelID
                                          8434
                                                 10150
                                                           4139
                                                                   8591
                                                                           4089
                            datasource
                                           132
                                                   132
                                                            132
                                                                    132
                                                                            132
                                                    ...
                                                             ...
            Backhoe_Mounting_is_missing
                                                   True
                                                          False
                                                                   True
                                                                           False
                                          False
                                                   True
                                                          False
                                                                           False
                  Blade_Type_is_missing
                                          False
                                                                   True
               Travel_Controls_is_missing
                                          False
                                                   True
                                                          False
                                                                   True
                                                                           False
              Differential_Type_is_missing
                                          True
                                                  False
                                                           True
                                                                   False
                                                                            True
             Steering_Controls_is_missing
                                                  False
                                                           True
                                                                   False
                                                                            True
                                          True
```

103 rows × 5 columns

```
In [172]: | df_tmp.isna().sum()
Out[172]: SalesID
                                             0
           SalePrice
                                             0
           MachineID
                                             0
           ModelID
                                             0
           datasource
                                             0
           Backhoe_Mounting_is_missing
           Blade_Type_is_missing
           Travel_Controls_is_missing
                                             0
           Differential_Type_is_missing
                                             0
           Steering_Controls_is_missing
                                             0
           Length: 103, dtype: int64
           Now let's build the Machine Learning model
In [173]: len(df_tmp)
Out[173]: 412698
In [174]: | df_tmp.head()
Out[174]:
                      SalePrice
               SalesID
                               MachinelD
                                          ModelID
                                                  datasource auctioneerID YearMade
                                                                                  MachineHoursCur
            0 1646770
                         9500.0
                                  1126363
                                             8434
                                                         132
                                                                    18.0
                                                                             1974
            1 1821514
                        14000.0
                                  1194089
                                            10150
                                                         132
                                                                    99.0
                                                                             1980
            2 1505138
                        50000.0
                                  1473654
                                                        132
                                                                    99.0
                                                                             1978
                                             4139
              1671174
                        16000.0
                                  1327630
                                             8591
                                                         132
                                                                    99.0
                                                                             1980
              1329056
                        22000.0
                                  1336053
                                             4089
                                                         132
                                                                    99.0
                                                                             1984
           5 rows × 103 columns
In [175]: | %%time
           # Instantiate model
           model = RandomForestRegressor(n_jobs=-1,
                                           random_state=42)
           # Fit the model
           model.fit(df_tmp.drop("SalePrice", axis=1), df_tmp["SalePrice"])
           Wall time: 7min 44s
Out[175]: RandomForestRegressor(n_jobs=-1, random_state=42)
In [176]: # Score the model
           model.score(df_tmp.drop("SalePrice", axis=1), df_tmp["SalePrice"])
Out[176]: 0.9875468079970562
```

```
In [177]: | df_tmp.saleYear
Out[177]: 0
                     1989
           1
                     1989
           2
                     1989
           3
                     1989
           4
                     1989
                      . . .
           412693
                     2012
           412694
                     2012
           412695
                     2012
           412696
                     2012
           412697
                     2012
           Name: saleYear, Length: 412698, dtype: int64
In [178]: | df_tmp.saleYear.value_counts()
Out[178]: 2009
                   43849
           2008
                   39767
           2011
                   35197
           2010
                   33390
           2007
                   32208
           2006
                   21685
           2005
                   20463
           2004
                   19879
           2001
                   17594
           2000
                   17415
           2002
                   17246
           2003
                   15254
           1998
                   13046
           1999
                   12793
           2012
                   11573
           1997
                    9785
           1996
                    8829
           1995
                    8530
                    7929
           1994
           1993
                    6303
           1992
                    5519
           1991
                    5109
           1989
                    4806
           1990
                    4529
           Name: saleYear, dtype: int64
In [179]: | df_valid = df_tmp[df_tmp.saleYear==2012]
           df_train = df_tmp[df_tmp.saleYear!=2012]
           len(df_valid),len(df_train)
Out[179]: (11573, 401125)
```

```
In [180]: # Split the train and valid data in X and y
          X_train, y_train = df_train.drop("SalePrice", axis=1), df_train["SalePrice"]
          X_valid, y_valid = df_valid.drop("SalePrice", axis=1), df_valid["SalePrice"]
          X train.shape, y train.shape, X valid.shape, y valid.shape
Out[180]: ((401125, 102), (401125,), (11573, 102), (11573,))
In [181]: y train
Out[181]: 0
                     9500.0
          1
                    14000.0
          2
                    50000.0
          3
                    16000.0
          4
                    22000.0
                     . . .
          401120
                    29000.0
          401121
                    11000.0
          401122
                    11000.0
          401123
                    18000.0
          401124
                    13500.0
          Name: SalePrice, Length: 401125, dtype: float64
```

Build an Evaluation Metrics

```
In [182]: # Create a evaluatio metrics i.e RMSLE
          from sklearn.metrics import mean_squared_log_error, mean_absolute_error, r2_score
          def rmsle(y_test, y_preds):
              Calculate root mean squured log error between prediction and true labels.
              return np.sqrt(mean_squared_log_error(y_test, y_preds))
          # Create function to evaluate model on a few different levels.
          def show score(model):
              train preds=model.predict(X train)
              valid preds=model.predict(X valid)
              scores={"Training MAE":mean_absolute_error(y_train,train_preds),
                      "Valid MAE":mean_absolute_error(y_valid,valid_preds),
                      "Training RMSLE":rmsle(y_train,train_preds),
                      "Valid RMSLE":rmsle(y_valid,valid_preds),
                      "Training R^2":r2 score(y train, train preds),
                      "Valid R^2":r2 score(y valid, valid preds)}
              return scores
```

Testing our model on Subset (to tune the hyperparameters)

```
In [183]: # This takes far too long .... for experimenting
          # %%time
          # model=RandomForestRegressor(n jobs=-1,
                                        random state=42)
          # model.fit(X_train,y_train)
In [184]: len(X_train)
Out[184]: 401125
In [185]: # Change max samples value
          from sklearn.ensemble import RandomForestRegressor
          model=RandomForestRegressor(n_jobs=-1,
                                       random state=42,
                                       max samples=10000)
In [186]: | %%time
          # Cutting down on the max number of samples each estimator can see improve train
          model.fit(X_train, y_train)
          Wall time: 25.4 s
Out[186]: RandomForestRegressor(max_samples=10000, n_jobs=-1, random_state=42)
In [187]: show_score(model)
Out[187]: {'Training MAE': 5561.2988092240585,
            'Valid MAE': 7177.26365505919,
            'Training RMSLE': 0.257745378256977,
            'Valid RMSLE': 0.29362638671089003,
            'Training R^2': 0.8606658995199189,
            'Valid R^2': 0.8320374995090507}
```

Hyperparameter tuning with RandomizedSearchCV

```
In [188]: | %%time
          from sklearn.model selection import RandomizedSearchCV
          # Different RandomForestRegressor hyperparameters
          rf_grid={"n_estimators":np.arange(10,100,10),
                    "max_depth":[None,3,5,10],
                    "min_samples_split":np.arange(2,20,2),
                   "min_samples_leaf":np.arange(1,20,2),
                    "max_features":[0.5,1,"sqrt","auto"],
                    "max_samples":[10000]}
          # Instantiate RandomizedSearchCV model
          rs_model = RandomizedSearchCV(RandomForestRegressor(n_jobs=-1,
                                                                random state=42),
                                        param distributions=rf grid,
                                        n_iter=2,
                                        cv=5,
                                        verbose=True)
          # Fit the model
          rs model.fit(X train,y train)
          Fitting 5 folds for each of 2 candidates, totalling 10 fits
          Wall time: 3min 25s
Out[188]: RandomizedSearchCV(cv=5,
                              estimator=RandomForestRegressor(n jobs=-1, random state=42),
                              n iter=2,
                              param_distributions={'max_depth': [None, 3, 5, 10],
                                                    'max features': [0.5, 1, 'sqrt',
                                                                     'auto'],
                                                    'max_samples': [10000],
                                                    'min samples leaf': array([ 1, 3, 5,
          7, 9, 11, 13, 15, 17, 19]),
                                                    'min_samples_split': array([ 2, 4, 6,
          8, 10, 12, 14, 16, 18]),
                                                    'n estimators': array([10, 20, 30, 40,
          50, 60, 70, 80, 90])},
                              verbose=True)
In [189]: # Find the best model hyperparameters
          rs model.best params
Out[189]: {'n_estimators': 40,
            'min_samples_split': 12,
            'min_samples_leaf': 9,
            'max samples': 10000,
            'max_features': 'auto',
            'max depth': None}
```

```
In [190]: # Evaluate the RandomizedSearchCV model
show_score(rs_model)

Out[190]: {'Training MAE': 6086.99131166347,
    'Valid MAE': 7674.949511960607,
    'Training RMSLE': 0.27648998624118665,
    'Valid RMSLE': 0.3072434602008128,
    'Training R^2': 0.8312379190695804,
    'Valid R^2': 0.8002384558444482}
```

Train a model with the best hyperparameters

Note: These were found after 100 iterations of RandomizedSearchCV

```
In [191]: | %%time
          # Most ideal hyperparameter
          ideal model = RandomForestRegressor(n estimators=40,
                                               min_samples_leaf=1,
                                               min samples split=14,
                                               max features=0.5,
                                               n jobs=-1,
                                               max samples=None,
                                               random state=42)
          # Fit the ideal model
          ideal_model.fit(X_train,y_train)
          Wall time: 1min 34s
Out[191]: RandomForestRegressor(max_features=0.5, min_samples_split=14, n_estimators=40,
                                 n jobs=-1, random state=42)
In [192]: # Scores for ideal model (trained on all the data)
          show score(ideal model)
Out[192]: {'Training MAE': 2953.8161137163484,
            'Valid MAE': 5951.247761444453,
            'Training RMSLE': 0.14469006962371858,
            'Valid RMSLE': 0.24524163989538328,
            'Training R^2': 0.9588145522577225,
            'Valid R^2': 0.8818019502450094}
In [193]: # Scores for rs_model (only trained on 10,000 examples)
          show score(rs model)
Out[193]: {'Training MAE': 6086.99131166347,
            'Valid MAE': 7674.949511960607,
            'Training RMSLE': 0.27648998624118665,
            'Valid RMSLE': 0.3072434602008128,
            'Training R^2': 0.8312379190695804,
            'Valid R^2': 0.8002384558444482}
```

Test our model using test data

	df_test.head().T	sc_uuccs=[surcuate])			
Out[203]:		0	1	2	3	4
	SalesiD	1227829	1227844	1227847	1227848	1227863
	MachinelD	1006309	1022817	1031560	56204	1053887
	ModelID	3168	7271	22805	1269	22312
	datasource	121	121	121	121	121
	auctioneerID	3	3	3	3	3
	YearMade	1999	1000	2004	2006	2005
	MachineHoursCurrentMeter	3688	28555	6038	8940	2286
	UsageBand	Low	High	Medium	High	Low
	saledate	2012-05-03 00:00:00	2012-05-10 00:00:00	2012-05-10 00:00:00	2012-05-10 00:00:00	2012-05-10 00:00:00
	fiModelDesc	580G	936	EC210BLC	330CL	650K
	fiBaseModel	580	936	EC210	330	650
	fiSecondaryDesc	G	NaN	В	С	K
	fiModelSeries	NaN	NaN	NaN	NaN	NaN
	fiModelDescriptor	NaN	NaN	LC	L	NaN
	ProductSize	NaN	Medium	Large / Medium	Large / Medium	NaN
	fiProductClassDesc	Backhoe Loader - 14.0 to 15.0 Ft Standard Digg	Wheel Loader - 135.0 to 150.0 Horsepower	Hydraulic Excavator, Track - 21.0 to 24.0 Metr	Hydraulic Excavator, Track - 33.0 to 40.0 Metr	Track Type Tractor, Dozer - 20.0 to 75.0 Horse
	state	Wyoming	Virginia	New Jersey	New Jersey	Florida
	ProductGroup	BL	WL	TEX	TEX	TTT
	ProductGroupDesc	Backhoe Loaders	Wheel Loader	Track Excavators	Track Excavators	Track Type Tractors
	Drive_System	Two Wheel Drive	NaN	NaN	NaN	NaN
	Enclosure	OROPS	EROPS	EROPS w AC	EROPS w AC	OROPS
	Forks	Yes	Yes	NaN	NaN	NaN
	Pad_Type	None or Unspecified	NaN	NaN	NaN	NaN
	Ride_Control	No	None or Unspecified	NaN	NaN	NaN
	Stick	Standard	NaN	NaN	NaN	NaN
	Transmission	Standard	NaN	NaN	NaN	Hydrostatic

	0	1	2	3	4
Turbocharged	None or Unspecified	NaN	NaN	NaN	NaN
Blade_Extension	NaN	NaN	NaN	NaN	NaN
Blade_Width	NaN	NaN	NaN	NaN	NaN
Enclosure_Type	NaN	NaN	NaN	NaN	NaN
Engine_Horsepower	NaN	NaN	NaN	NaN	NaN
Hydraulics	NaN	2 Valve	Auxiliary	Standard	2 Valve
Pushblock	NaN	NaN	NaN	NaN	NaN
Ripper	NaN	NaN	NaN	NaN	None or Unspecified
Scarifier	NaN	NaN	NaN	NaN	NaN
Tip_Control	NaN	NaN	NaN	NaN	NaN
Tire_Size	NaN	20.5	NaN	NaN	NaN
Coupler	NaN	None or Unspecified	None or Unspecified	None or Unspecified	NaN
Coupler_System	NaN	NaN	NaN	NaN	NaN
Grouser_Tracks	NaN	NaN	NaN	NaN	NaN
Hydraulics_Flow	NaN	NaN	NaN	NaN	NaN
Track_Type	NaN	NaN	Steel	Steel	NaN
Undercarriage_Pad_Width	NaN	NaN	None or Unspecified	None or Unspecified	NaN
Stick_Length	NaN	NaN	9' 6"	None or Unspecified	NaN
Thumb	NaN	NaN	Manual	Manual	NaN
Pattern_Changer	NaN	NaN	None or Unspecified	Yes	NaN
Grouser_Type	NaN	NaN	Double	Triple	NaN
Backhoe_Mounting	NaN	NaN	NaN	NaN	None or Unspecified
Blade_Type	NaN	NaN	NaN	NaN	PAT
Travel_Controls	NaN	NaN	NaN	NaN	None or Unspecified
Differential_Type	NaN	Standard	NaN	NaN	NaN
Steering_Controls	NaN	Conventional	NaN	NaN	NaN

```
In [204]: df_test.saledate[:1000]
Out[204]: 0
                 2012-05-03
                 2012-05-10
          1
          2
                 2012-05-10
          3
                 2012-05-10
          4
                 2012-05-10
                    . . .
          995
                 2012-06-12
          996
                 2012-06-12
          997
                 2012-05-01
          998
                 2012-05-01
          999
                 2012-06-05
          Name: saledate, Length: 1000, dtype: datetime64[ns]
```

In [205]: df_test.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 12457 entries, 0 to 12456
Data columns (total 52 columns):

νаτа	columns (total 52 columns	•	
#	Column	Non-Null Count	Dtype
0	SalesID	12457 non-null	 int64
1	MachineID	12457 non-null	int64
2	ModelID	12457 non-null	int64
3	datasource	12457 non-null	int64
4	auctioneerID	12457 non-null	int64
5	YearMade	12457 non-null	int64
6	MachineHoursCurrentMeter	2129 non-null	float64
7	UsageBand	1834 non-null	object
8	saledate	12457 non-null	datetime64[ns]
9	fiModelDesc	12457 non-null	object
10	fiBaseModel	12457 non-null	object
11	fiSecondaryDesc	8482 non-null	object
12	fiModelSeries	2006 non-null	object
13	fiModelDescriptor	3024 non-null	object
14	ProductSize	6048 non-null	object
15	fiProductClassDesc	12457 non-null	object
16	state	12457 non-null	object
17	ProductGroup	12457 non-null	object
18	ProductGroupDesc	12457 non-null	object
19	Drive_System	2759 non-null	object
20	Enclosure	12455 non-null	object
21	Forks	6308 non-null	object
22	Pad_Type	2108 non-null	object
23	Ride_Control	4241 non-null	object
24	Stick	2108 non-null	object
25	Transmission	4818 non-null	object
26	Turbocharged	2108 non-null	object
27	Blade_Extension	651 non-null	object
28	Blade_Width	651 non-null	object
29	Enclosure_Type	651 non-null	object
30	Engine_Horsepower	651 non-null	object
31	Hydraulics	10315 non-null	
32	Pushblock	651 non-null	object
33	Ripper	2704 non-null	object
34	Scarifier	651 non-null	object
35	Tip_Control	651 non-null	object
36	Tire Size	2778 non-null	object
37	Coupler	7601 non-null	object
38	Coupler_System	2066 non-null	object
39	Grouser_Tracks	2066 non-null	object
40	Hydraulics_Flow	2066 non-null	object
41	Track_Type	3394 non-null	object
42	Undercarriage_Pad_Width	3398 non-null	object
43	Stick_Length	3394 non-null	object
44	Thumb	3395 non-null	object
45	Pattern_Changer	3394 non-null	object
46	Grouser_Type	3394 non-null	object
47	Backhoe_Mounting	2051 non-null	object
48	Blade_Type	2058 non-null	object
49	Travel_Controls	2058 non-null	object

50 Differential_Type 2129 non-null object 51 Steering_Controls 2129 non-null object

dtypes: datetime64[ns](1), float64(1), int64(6), object(44)

memory usage: 4.9+ MB

In [206]: df_test.isna().sum() Out[206]: SalesID 0 0 MachineID ModelID 0 0 datasource auctioneerID 0 YearMade 0 MachineHoursCurrentMeter 10328 10623 UsageBand saledate 0 fiModelDesc 0 fiBaseModel 0 fiSecondaryDesc 3975 fiModelSeries 10451 fiModelDescriptor 9433 ProductSize 6409 0 fiProductClassDesc 0 state ProductGroup 0 0 ProductGroupDesc 9698 Drive System Enclosure 2 6149 **Forks** Pad_Type 10349 Ride_Control 8216 Stick 10349 Transmission 7639 Turbocharged 10349 Blade_Extension 11806 Blade_Width 11806 Enclosure_Type 11806 Engine_Horsepower 11806 Hydraulics 2142 Pushblock 11806 Ripper 9753 Scarifier 11806 Tip_Control 11806 Tire_Size 9679 Coupler 4856 Coupler_System 10391 Grouser_Tracks 10391 Hydraulics_Flow 10391 Track_Type 9063 Undercarriage_Pad_Width 9059 Stick Length 9063 Thumb 9062 Pattern_Changer 9063 Grouser_Type 9063 Backhoe_Mounting 10406

dtype: int64

Travel_Controls

Differential_Type

Steering_Controls

Blade_Type

10399

10399

10328

10328

Preprocessing the data (getting the test dataset in the same format as our training dataset)

```
In [207]: | def preprocess_data(df):
              Performs some transformation odf and returns transformed df.
              df["saleYear"]=df.saledate.dt.year
              df["saleMonth"]=df.saledate.dt.month
              df["saleDay"]=df.saledate.dt.day
              df["saleDayOfWeek"]=df.saledate.dt.dayofweek
              df["saleDayOfYear"]=df.saledate.dt.dayofyear
              df.drop(labels="saledate",axis=1,inplace=True)
              # Fill the numeric rows with median
              for label,content in df.items():
                  if pd.api.types.is_numeric_dtype(content):
                      if pd.isnull(content).sum():
                          # Add a binary column which tells us if the data was missing or r
                          df[label+"_is_missing"]=pd.isnull(content)
                          # Fill missing numeric values with median
                          df[label]=content.fillna(content.median())
              # Fill the categorical missing data and turn categories into numbers
                  if not pd.api.types.is_numeric_dtype(content):
                      df[label+"_is_missing"]=pd.isnull(content)
                      df[label]=pd.Categorical(content).codes+1
              return df
```

```
In [208]: # Preprocess the test data
df_test=preprocess_data(df_test)
df_test.head()
```

\sim			$\Gamma \sim$	$\overline{}$	\sim	т.
()	u.	Τ.	2	и	×	
$^{\circ}$	u	_	_	v	o	

	SalesID	MachinelD	ModelID	datasource	auctioneerID	YearMade	MachineHoursCurrentMeter
0	1227829	1006309	3168	121	3	1999	3688.0
1	1227844	1022817	7271	121	3	1000	28555.0
2	1227847	1031560	22805	121	3	2004	6038.0
3	1227848	56204	1269	121	3	2006	8940.0
4	1227863	1053887	22312	121	3	2005	2286.0

5 rows × 101 columns

```
In [209]: # Make predicytions on update test data
          test preds=ideal model.predict(df test)
          ValueError
                                                     Traceback (most recent call last)
          <ipython-input-209-2cc34114db25> in <module>
                1 # Make predicytions on update test data
          ----> 2 test_preds=ideal_model.predict(df_test)
          ~\AppData\Roaming\Python\Python37\site-packages\sklearn\ensemble\ forest.py in
          predict(self, X)
              782
                          check is fitted(self)
              783
                          # Check data
          --> 784
                          X = self._validate_X_predict(X)
              785
                          # Assign chunk of trees to jobs
              786
          ~\AppData\Roaming\Python\Python37\site-packages\sklearn\ensemble\ forest.py in
          _validate_X_predict(self, X)
              420
                          check_is_fitted(self)
              421
          --> 422
                          return self.estimators [0]. validate X predict(X, check input=T
          rue)
              423
              424
                      @property
          ~\AppData\Roaming\Python\Python37\site-packages\sklearn\tree\ classes.py in va
          lidate X predict(self, X, check input)
              406
                          if check_input:
              407
                              X = self._validate_data(X, dtype=DTYPE, accept_sparse="cs
          r",
          --> 408
                                                       reset=False)
                               if issparse(X) and (X.indices.dtype != np.intc or
              409
                                                   X.indptr.dtype != np.intc):
              410
          ~\AppData\Roaming\Python\Python37\site-packages\sklearn\base.py in _validate_da
          ta(self, X, y, reset, validate_separately, **check_params)
              435
                          if check params.get('ensure 2d', True):
              436
                               self. check n features(X, reset=reset)
          --> 437
              438
              439
                          return out
          ~\AppData\Roaming\Python\Python37\site-packages\sklearn\base.py in check n fea
          tures(self, X, reset)
                          if n features != self.n features in :
              364
              365
                              raise ValueError(
          --> 366
                                   f"X has {n_features} features, but {self.__class__.__na
          me__} "
              367
                                   f"is expecting {self.n features in } features as inpu
          t.")
              368
          ValueError: X has 101 features, but DecisionTreeRegressor is expecting 102 feat
```

ures as input.

```
In [210]: X train.head()
Out[210]:
               SalesID
                       MachinelD ModelID datasource auctioneerID YearMade
                                                                            MachineHoursCurrentMeter
               1646770
                          1126363
                                     8434
                                                 132
                                                             18.0
                                                                      1974
                                                                                                0.0
              1821514
                         1194089
                                    10150
                                                 132
                                                             99.0
                                                                      1980
                                                                                                0.0
              1505138
                         1473654
                                                             99.0
                                                                      1978
                                     4139
                                                 132
                                                                                                0.0
               1671174
                                                             99.0
                                                                      1980
                         1327630
                                     8591
                                                 132
                                                                                                0.0
               1329056
                         1336053
                                     4089
                                                 132
                                                             99.0
                                                                      1984
                                                                                                0.0
           5 rows × 102 columns
In [211]: # we can find how the columns differ using sets
           set(X train.columns)-set(df test.columns)
Out[211]: {'auctioneerID_is_missing'}
In [212]: # Manually adjust df test to have auctioneerID is missing column
           df_test["auctioneerID_is_missing"]=False
           df_test.head()
Out[212]:
               SalesID
                       MachinelD ModelID
                                          datasource auctioneerID
                                                                  YearMade
                                                                            MachineHoursCurrentMeter
            0 1227829
                         1006309
                                     3168
                                                 121
                                                               3
                                                                      1999
                                                                                             3688.0
            1 1227844
                         1022817
                                     7271
                                                 121
                                                               3
                                                                      1000
                                                                                             28555.0
                                                                      2004
              1227847
                         1031560
                                    22805
                                                 121
                                                               3
                                                                                             6038.0
               1227848
                           56204
                                     1269
                                                 121
                                                               3
                                                                      2006
                                                                                             8940.0
               1227863
                         1053887
                                                               3
                                                                      2005
                                                                                             2286.0
                                    22312
                                                 121
           5 rows × 102 columns
In [213]: # Make predicitions on the test data
           test preds=ideal model.predict(df test)
In [214]: test preds
Out[214]: array([20614.36780887, 19897.80170658, 44852.21959446, ...,
                   14296.98620472, 22164.85757662, 31683.80063427])
```

```
In [215]: # Format predicitons into the same format kaggle is asking
    df_preds=pd.DataFrame()
    df_preds["SalesID"]=df_test["SalesID"]
    df_preds["SalesPrice"]=test_preds
    df_preds
```

Out[215]:

	SalesID	SalesPrice
0	1227829	20614.367809
1	1227844	19897.801707
2	1227847	44852.219594
3	1227848	68346.325323
4	1227863	39487.349708
12452	6643171	46466.092910
12453	6643173	17500.493352
12454	6643184	14296.986205
12455	6643186	22164.857577
12456	6643196	31683.800634

12457 rows × 2 columns

```
In [216]: # Export predictive data
df_preds.to_csv("data/test_predicitions.csv", index=False)
```

In []: