In [3]: import numpy as np
 import pandas as pd
 import matplotlib.pyplot as plt
 import matplotlib.cm as cm

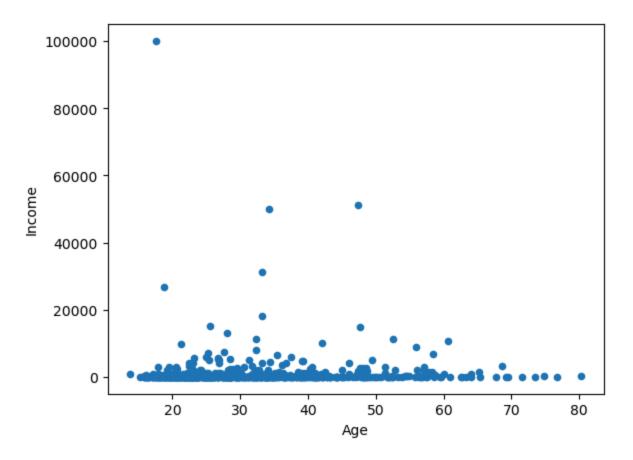
In [4]: credit_df = pd.read_csv("C:/Users/ankus/OneDrive/Desktop/credit_card.csv")
 credit_df

Out[4]:		Gender	Age	Debt	Married	BankCustomer	Industry	Ethnicity	YearsEmpl
	0	1	30.83	0.000	1	1	Industrials	White	
	1	0	58.67	4.460	1	1	Materials	Black	
	2	0	24.50	0.500	1	1	Materials	Black	
	3	1	27.83	1.540	1	1	Industrials	White	
	4	1	20.17	5.625	1	1	Industrials	White	
	•••	•••							
	685	1	21.08	10.085	0	0	Education	Black	
	686	0	22.67	0.750	1	1	Energy	White	
	687	0	25.25	13.500	0	0	Healthcare	Latino	
	688	1	17.92	0.205	1	1	ConsumerStaples	White	
	689	1	35.00	3.375	1	1	Energy	Black	

690 rows × 16 columns

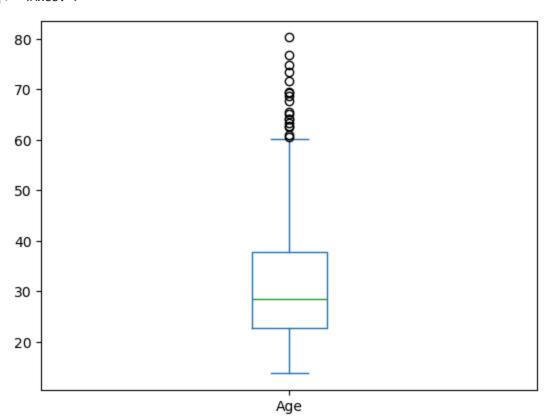
```
In [5]: credit_df.plot('Age','Income',kind='scatter',marker='o')
```

Out[5]: <Axes: xlabel='Age', ylabel='Income'>



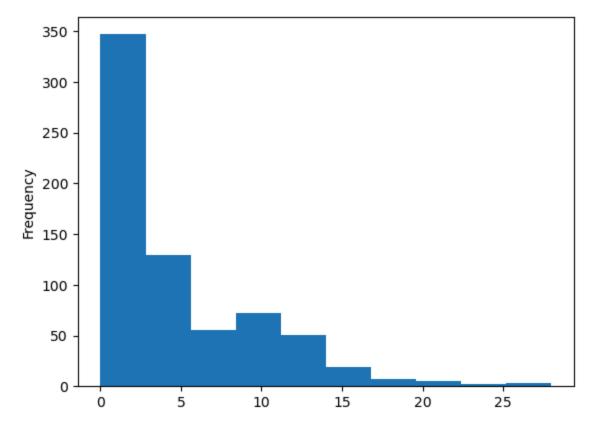
In [6]: credit_df['Age'].plot(kind='box')

Out[6]: <Axes: >



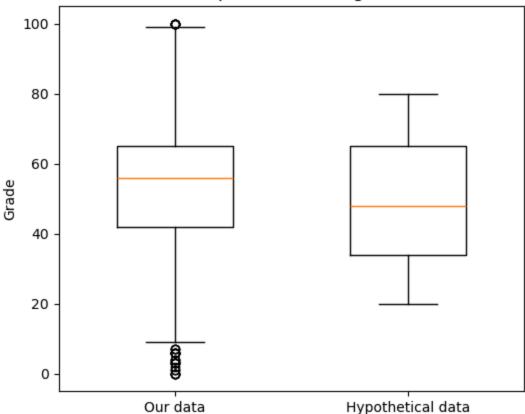
```
In [7]: credit_df['Debt'].plot(kind='hist')
```

Out[7]: <Axes: ylabel='Frequency'>



```
In [8]: C
                         3
6
                                                            3
6
                                                                      3 100 100 100 100
           0
4
        [
                             0
                                      2
                                           7
                                               6 100
                    3
                             6
                                               6 100 100
                                 6 100
                                                                 3
                                                                          1
                                                                                   0]
                                           7
                                                                     6
                                                                              6
```

Box plot of midterm grade



```
import numpy as np
data = [1, 2, 2, 2, 3, 1, 1, 15, 2, 2, 2, 3, 1, 1, 2]
mean = np.mean(data)
std = np.std(data)
print('mean of the dataset is', mean)
print('std. deviation is', std)
threshold = 3
outlier = []
for i in data:
    z = (i-mean)/std
    if z > threshold:
        outlier.append(i)
print('outlier in dataset of Z score is', outlier)
```

mean of the dataset is 2.666666666666655 std. deviation is 3.3598941782277745 outlier in dataset of Z score is [15]

```
In [10]: q1 = credit_df["Age"].quantile(0.25)
    q3 = credit_df['Age'].quantile(0.75)
    iqr = q3-q1
    upper_bound = q3+(1.5*iqr)
    lower_bound = q1-(1.5*iqr)
```

```
In [11]: upperIndex = credit_df[credit_df['Age']>upper_bound].index
    credit_df.drop(upperIndex,inplace=True)
    lowerIndex = credit_df[credit_df['Age']<lower_bound].index</pre>
```

```
credit_df.drop(lowerIndex,inplace=True)
          credit_df.info()
         <class 'pandas.core.frame.DataFrame'>
        Index: 672 entries, 0 to 689
        Data columns (total 16 columns):
             Column Non-Null Count Dtype
                              -----
             -----
         0 Gender 672 non-null int64
1 Age 672 non-null float64
2 Debt 672 non-null float64
3 Married 672 non-null int64
4 BankCustomer 672 non-null int64
         0 Gender
         5 Industry 672 non-null object
6 Ethnicity 672 non-null object
         7 YearsEmployed 672 non-null float64
         8 PriorDefault 672 non-null int64
         9 Employed 672 non-null int64
10 CreditScore 672 non-null int64
         11 DriversLicense 672 non-null int64
         12 Citizen 672 non-null object
13 ZipCode 672 non-null int64
14 Income 672 non-null int64
         14 Income
                             672 non-null int64
         15 Approved 672 non-null int64
        dtypes: float64(3), int64(10), object(3)
        memory usage: 89.2+ KB
In [12]: m = np.mean(credit_df['Age'])
          print('mean:',m)
          for i in credit_df['Age']:
              if i<lower_bound or i>upper_bound :
                 titanic_df['Age'] = titanic_df['Age'].replace(i,m)
        mean: 30.541651785714286
In [13]: m = credit_df['Age'].median()
          print("median",m)
          for i in credit_df['Age']:
              if i<lower_bound or i>upper_bound :
                  credit_df['Age'] = credit_df['Age'].replace(i,m)
        median 28.25
In [14]: for i in credit_df['Age']:
              if i<lower_bound or i>upper_bound :
                  credit_df['Age'] = credit_df['Age'].replace(i,0)
In [15]: import numpy as np
          import pandas as pd
          import matplotlib.pyplot as plt
          %matplotlib inline
          import seaborn as sns
          import math
In [16]: card_approval_df=pd.read_csv("C:/Users/ankus/OneDrive/Desktop/credit_card.csv")
          credit_df
```

```
print(card_approval_df.head())
          Gender
                   Age
                         Debt Married BankCustomer
                                                       Industry Ethnicity \
               1 30.83 0.000
                                1
                                                1 Industrials
                                                                   White
       1
               0 58.67 4.460
                                    1
                                                 1
                                                      Materials
                                                                   Black
       2
               0 24.50 0.500
                                    1
                                                 1
                                                      Materials
                                                                   Black
       3
               1 27.83 1.540
                                    1
                                                 1 Industrials
                                                                   White
       4
               1 20.17 5.625
                                    1
                                                 1 Industrials
                                                                   White
          YearsEmployed PriorDefault Employed CreditScore DriversLicense \
       0
                  1.25
                                  1
                                            1
                                                        1
                  3.04
                                                                       0
                                  1
                                            1
                                                        6
       1
       2
                  1.50
                                  1
                                            0
                                                        0
                                                                       0
                                                        5
       3
                  3.75
                                  1
                                            1
                                                                      1
       4
                  1.71
                                  1
                                            0
                                                        0
                                                                       0
               Citizen ZipCode Income Approved
       0
               ByBirth
                           202
                                    0
                                             1
               ByBirth
                           43
                                  560
                                             1
       1
       2
                                             1
               ByBirth
                           280
                                  824
       3
               ByBirth
                           100
                                    3
                                             1
       4 ByOtherMeans
                           120
                                    0
                                             1
In [17]: print(card_approval_df.info())
       <class 'pandas.core.frame.DataFrame'>
       RangeIndex: 690 entries, 0 to 689
       Data columns (total 16 columns):
            Column
                           Non-Null Count Dtype
            -----
                           -----
                                          ----
        0
            Gender
                           690 non-null
                                          int64
        1
            Age
                           690 non-null
                                       float64
            Debt
                         690 non-null float64
        3
           Married
                          690 non-null int64
            BankCustomer 690 non-null int64
        5
                         690 non-null object
            Industry
            Ethnicity 690 non-null object
        6
        7
            YearsEmployed 690 non-null float64
            PriorDefault
                           690 non-null
                                       int64
        9
            Employed
                           690 non-null
                                       int64
        10 CreditScore
                           690 non-null
                                       int64
                                          int64
        11 DriversLicense 690 non-null
        12 Citizen
                           690 non-null
                                          object
        13 ZipCode
                           690 non-null
                                          int64
        14 Income
                           690 non-null
                                          int64
        15 Approved
                           690 non-null
                                          int64
       dtypes: float64(3), int64(10), object(3)
       memory usage: 86.4+ KB
       None
In [18]: card_approval_df.duplicated().sum()
Out[18]: 0
In [19]: card_approval_df[['Age','Debt','YearsEmployed','CreditScore','Income']].describe()
```

		Age	Debt	YearsEmployed	CreditScore	Income
cou	unt	690.000000	690.000000	690.000000	690.00000	690.000000
me	ean	31.514116	4.758725	2.223406	2.40000	1017.385507
	std	11.860245	4.978163	3.346513	4.86294	5210.102598
n	nin	13.750000	0.000000	0.000000	0.00000	0.000000
2	5%	22.670000	1.000000	0.165000	0.00000	0.000000
50	0%	28.460000	2.750000	1.000000	0.00000	5.000000
7	5%	37.707500	7.207500	2.625000	3.00000	395.500000
m	nax	80.250000	28.000000	28.500000	67.00000	100000.000000

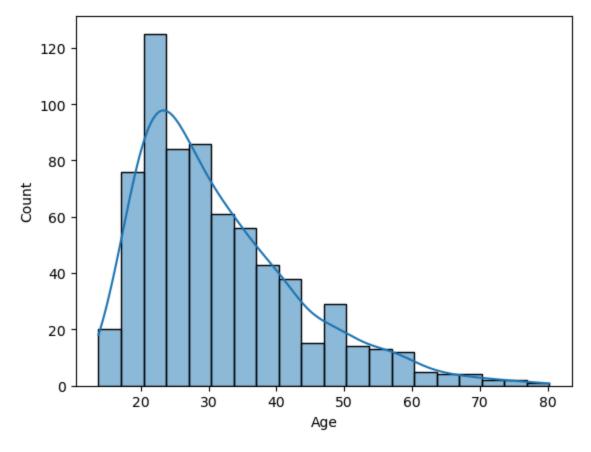
In [20]: sns.histplot(card_approval_df.Age,kde=True)

C:\ProgramData\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version. Convert inf values to NaN before operating instead.

with pd.option_context('mode.use_inf_as_na', True):

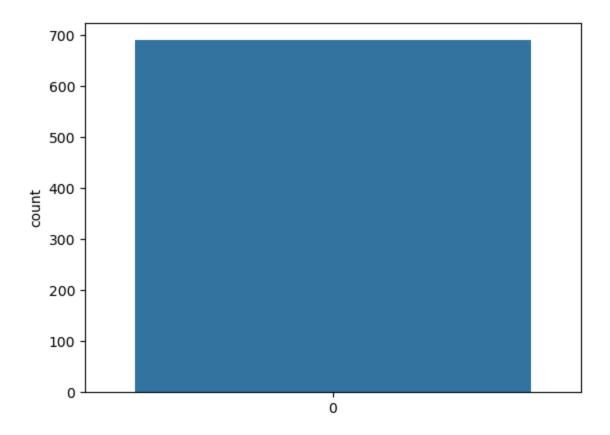
Out[20]: <Axes: xlabel='Age', ylabel='Count'>

Out[19]:



In [21]: sns.countplot(card_approval_df.Gender)

Out[21]: <Axes: ylabel='count'>



```
In [24]: import seaborn as sns
   import matplotlib.pyplot as plt

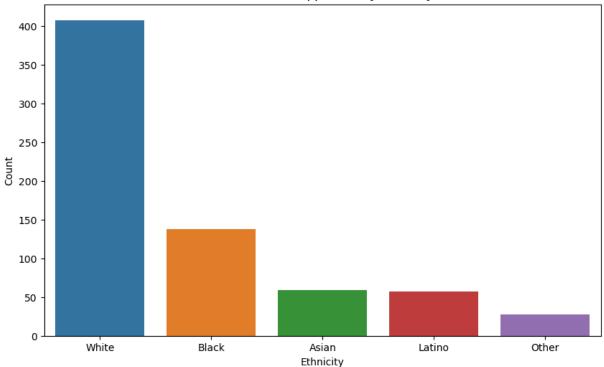
# Set the figure size for better visibility
   plt.figure(figsize=(10, 6))

# Create the count plot
   sns.countplot(data=card_approval_df, x='Ethnicity')

# Adding labels and title for clarity
   plt.xlabel('Ethnicity')
   plt.ylabel('Count')
   plt.title('Count of Card Approvals by Ethnicity')

# Show the plot
   plt.show()
```

Count of Card Approvals by Ethnicity



In [23]: card_approval_df[['Age','Debt','YearsEmployed','CreditScore','Income']].corr()

\cap		~ n n ~	١.
Uι	1 L		١.

	Age	Debt	YearsEmployed	CreditScore	Income
Age	1.000000	0.202177	0.391464	0.187327	0.018719
Debt	0.202177	1.000000	0.298902	0.271207	0.123121
YearsEmployed	0.391464	0.298902	1.000000	0.322330	0.051345
CreditScore	0.187327	0.271207	0.322330	1.000000	0.063692
Income	0.018719	0.123121	0.051345	0.063692	1.000000

```
In [25]: sns.scatterplot(card_approval_df.YearsEmployed,card_approval_df.Income)
   plt.ylim(0,20000)
```

```
In [30]: card_approval_df.groupby(by='Approved').agg('mean')[['Age','Debt','YearsEmployed']]
```

```
TypeError
                                          Traceback (most recent call last)
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\groupby\groupby.py:1874,
in GroupBy._agg_py_fallback(self, how, values, ndim, alt)
-> 1874
            res_values = self.grouper.agg_series(ser, alt, preserve_dtype=True)
  1875 except Exception as err:
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\groupby\ops.py:849, in B
aseGrouper.agg_series(self, obj, func, preserve_dtype)
    847
            preserve dtype = True
--> 849 result = self._aggregate_series_pure_python(obj, func)
    851 if len(obj) == 0 and len(result) == 0 and isinstance(obj.dtype, ExtensionDty
pe):
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\groupby\ops.py:877, in B
aseGrouper. aggregate series pure python(self, obj, func)
    876 for i, group in enumerate(splitter):
            res = func(group)
--> 877
   878
            res = extract_result(res)
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\groupby\groupby.py:2380,
in GroupBy.mean.<locals>.<lambda>(x)
  2377 else:
  2378
            result = self._cython_agg_general(
  2379
                "mean",
-> 2380
                alt=lambda x: Series(x).mean(numeric_only=numeric_only),
   2381
                numeric_only=numeric_only,
  2382
            return result.__finalize__(self.obj, method="groupby")
   2383
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\series.py:6225, in Serie
s.mean(self, axis, skipna, numeric_only, **kwargs)
  6217 @doc(make_doc("mean", ndim=1))
  6218 def mean(
  6219
            self,
   (…)
            **kwargs,
  6223
  6224 ):
-> 6225
            return NDFrame.mean(self, axis, skipna, numeric_only, **kwargs)
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\generic.py:11992, in NDF
rame.mean(self, axis, skipna, numeric_only, **kwargs)
 11985 def mean(
 11986
           self.
 11987
            axis: Axis | None = 0,
  (\ldots)
 11990
           **kwargs,
 11991 ) -> Series | float:
> 11992
            return self._stat_function(
 11993
                "mean", nanops.nanmean, axis, skipna, numeric_only, **kwargs
 11994
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\generic.py:11949, in NDF
rame._stat_function(self, name, func, axis, skipna, numeric_only, **kwargs)
  11947 validate_bool_kwarg(skipna, "skipna", none_allowed=False)
```

```
> 11949 return self._reduce(
            func, name=name, axis=axis, skipna=skipna, numeric_only=numeric_only
 11951 )
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\series.py:6133, in Serie
s._reduce(self, op, name, axis, skipna, numeric_only, filter_type, **kwds)
  6129
            raise TypeError(
   6130
                f"Series.{name} does not allow {kwd_name}={numeric_only} "
                "with non-numeric dtypes."
   6131
  6132
            )
-> 6133 return op(delegate, skipna=skipna, **kwds)
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\nanops.py:147, in bottle
neck_switch.__call__.<locals>.f(values, axis, skipna, **kwds)
    146 else:
            result = alt(values, axis=axis, skipna=skipna, **kwds)
--> 147
    149 return result
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\nanops.py:404, in _datet
imelike_compat.<locals>.new_func(values, axis, skipna, mask, **kwargs)
            mask = isna(values)
   402
--> 404 result = func(values, axis=axis, skipna=skipna, mask=mask, **kwargs)
    406 if datetimelike:
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\nanops.py:720, in nanmea
n(values, axis, skipna, mask)
    719 the_sum = values.sum(axis, dtype=dtype_sum)
--> 720 the_sum = _ensure_numeric(the_sum)
    722 if axis is not None and getattr(the_sum, "ndim", False):
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\nanops.py:1693, in _ensu
re numeric(x)
  1691 if isinstance(x, str):
            # GH#44008, GH#36703 avoid casting e.g. strings to numeric
  1692
            raise TypeError(f"Could not convert string '{x}' to numeric")
-> 1693
  1694 try:
```

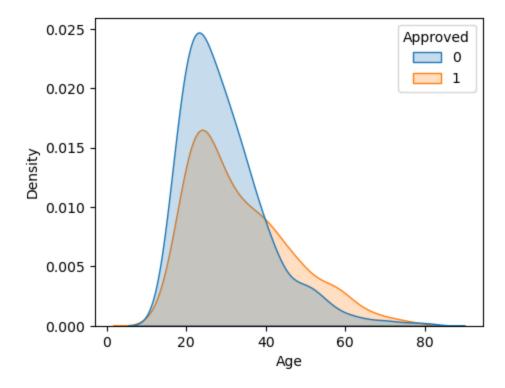
TypeError: Could not convert string 'EducationReal EstateInformationTechnologyCommun ication Services Energy Materials Consumer Staples Consumer Staples Financials Energy Health can be a supported by the consumer Staples of the consu $re Communication Services Communication Services Real\ Estate Consumer Staples Energy Real\ Estate Consumer Staples and Services Real\ Estate Consumer Staples Services Real\ Estate Consumer Servi$ ate Consumer Staples Consumer Staples Research Energy Energy Consumer Staples Energy Energy Consumer Staples Consumer Staplumer Discretion ary Consumer Staples Energy Materials Materials Energy Energy Materials Financial Staples Financial Stals Consumer Discretionary Education Financials Consumer Staples Utilities Energy IndustrialsHealthcareIndustrialsFinancialsEnergyMaterialsMaterialsFinancialsEnergyEnergyReal Es $tate Consumer Discretion ary Energy Real\ Estate Health care Consumer Discretion ary Utilities Health Consumer Discretion are Utilities Health Consumer Discretion and Utilities Health Consumer Discretion are Utilities Health Consumer Discretion and Utilities Health Consumer Discretion are Utilities Health Consumer Discretion and Utilities Health Consumer Discretion Are Utilities Health Consumer Discretion and Utilities Health Consumer Discretion Are Utilities Health Consumer Discre$ $althcare {\tt EnergyEducationReal\ Estate EnergyCommunicationServices Financials Information Terminal} \\$ $chnology Real\ Estate Consumer Discretion ary Materials Health care Communication Services Energy Consumer Consumer Discretion and Consumer Consumer Discretion are the Consumer Discretion and Consumer Discretion are the Consumer Discretion and Consumer Discretion are the Consumer Discretion and Consumer Discretion and Consumer Discretion are the Consumer Discretion and Consumer Discret$ gyInformationTechnologyEnergyMaterialsEnergyIndustrialsHealthcareIndustrialsEnergyRe search Communication Services Industrials Industrials Industrials Health care Health CarthcareConsumerStaplesConsumerStaplesEnergyHealthcareReal EstateEnergyMaterialsMateri als Health care Health care Financials Research Industrials Energy Consumer Staples Research Financials Research Financial Research Financia Financia Financia FinancierDiscretionaryIndustrialsConsumerDiscretionaryEnergyEnergyEnergyReal EstateInformat ion Technology Materials Energy Consumer Discretionary Consumer Staples Industrials Energy Consumer Staples (Staple Staples) and the Staples (Staples) and the Staples (StaplessumerStaples Consumer Discretion ary Consumer Discretion ary Financials Education Energy Energy Consumer Discretion are also also as a consumer Discretion are a consumer Discretion and a consumer Discretion are a consumer Discretion are a consumer Discretion are a consumer Discretion are a consumer Discretion and Discretion are a consumer Discretion are a consumer Discretion and Discretion are a consumer Discretion and Discretion are a consumer Discretion and Discretion are a consumer Discretion are a consumer Discretion and Discretion are a consumyEnergyIndustrialsResearchHealthcareCommunicationServicesReal EstateConsumerDiscreti onaryConsumerStaplesReal EstateMaterialsReal EstateHealthcareFinancialsHealthcareFin ancials Financials Education Communication Services Consumer Staples Communication Services Financials Financials Education Communication Services Financials Financial Fiin ancials Transport Industrials Education Financials Industrials Consumer Staples Education Research and the property of theal EstateEnergyConsumerDiscretionaryConsumerDiscretionaryConsumerDiscretionaryEnerg y Consumer Discretion ary Health care Financials Energy Consumer Staples Energy Consumer Discretion and the consion ary Communication Services Education Consumer Discretionary Consumer Staples Communication Consumer Discretion ConsumernServicesIndustrialsReal EstateUtilitiesHealthcareFinancialsEnergyResearchEnergyCons umer Discretion ary Materials Consumer Discretion ary Health care Financials Energy IndustrialsMaterialsHealthcareEnergyConsumerDiscretionaryConsumerDiscretionaryFinancialsHealthc are Financials Energy Energy Health care Consumer Discretionary Materials Industrials Information and the property of the prion Technology Energy Energy Industrials Education Health care Information Technology Health Care Information Technologre Energy Financials Consumer Staples Energy Health care Health care Health care Health care Health care Financials Consumer Staples Energy Health care Health ccials Financials Health care Communication Services Financials Research Health care Energy Health Communication Services Financials Research Health Care Energy Health Communication Services Financials Research Health Care Energy Health Care Financials Research Health Care Financial Research Financial R $th care {\tt Health care Energy Consumer Discretion} ary {\tt Energy Materials Industrials Real Estate Industrials} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer Discretion} and {\tt Entropy Consumer Discretion} are {\tt Entropy Consumer$ strialsEnergyEnergyConsumerDiscretionaryEnergyEnergyCommunicationServicesEnergyHealthcare Research Communication Services Energy Industrials Financials Consumer Discretionary Financials Consumer Discretionary Financials Consumer Discretionary Financials Consumer Discretion Financials Consumer Financial Fi $nancials Consumer Discretionary Consumer Discretionary Consumer Discretionary Energy Real \ Escape Consumer Discretionary Consumer Disc$ tate Consumer Discretion ary Consumer Discretion ary Consumer Discretion ary Energy Consumer Discretion are Consumer Discretcretion ary Financials Consumer Staples Energy Communication Services Health care Financials Consumer Staples Financial Financ $nsumer Staples Health care {\tt Materials Materials Energy Financials Consumer Discretion}$ $ary Financials {\tt Materials Energy Consumer Staples Consumer Discretionary Health care Consumer Staples {\tt Consumer Staples Consumer Discretionary Health Care Consumer Staples {\tt Consumer Staples Consumer Discretionary Health Care Consumer Staples {\tt Consumer Discretionary Health Care Consumer Discretion Health Care Consumer Discretion$ aples Health care Health care Utilities Industrials Energy Information Technology Consumer Discussion For the Scholar Consumer Discussion For the Scholar Consumer Discussion For the Consumer Discussion For the Scholar Consumer Discussion For the ConsumretionaryFinancialsIndustrialsConsumerDiscretionaryReal EstateIndustrialsEnergyFinan cialsMaterialsEnergyFinancialsFinancialsReal EstateUtilitiesFinancialsMaterialsIndus $trials {\tt EnergyEnergyEnergyConsumerDiscretionaryHealth} care {\tt CommunicationServicesConsumerDiscretionaryHealth} and {\tt ConsumerDiscretionaryHealth} and {\tt ConsumerDiscre$ Staples Consumer Discretionary Materials Information Technology Energy Consumer Discretion are the following the contraction of the contractionyHealthcareReal EstateIndustrialsMaterialsCommunicationServicesHealthcareFinancialsI $ndustrials {\tt EnergyCommunicationServicesConsumerStaplesConsumerStaplesFinancialsConsume}$ $\verb|rDiscretion| ary Information Technology Consumer Staples Consumer Staples Industrials Information Technology Consumer Staples Industrial Information Technology Consumer Staples Information Technolo$ ion Technology Energy Consumer Staples Financials Health care Consumer Discretionary MaterialsConsumerDiscretionaryConsumerDiscretionaryMaterialsHealthcareEnergyConsumerStaplesEn ergy Energy Consumer Staples Energy Industrials Information Technology Industrials Communication Technology Industrial Indus $ion Services Energy Energy Real\ Estate Health care Industrials Information Technology Consumer$ Staples Education Consumer Discretionary Consumer Staples Materials Communication Services Herrich Communication ServicesalthcareEnergyFinancialsReal EstateConsumerDiscretionaryReal EstateCommunicationServ $ices {\tt EducationEnergyHealthcareConsumerStaplesEnergy'}\ to\ numeric$

The above exception was the direct cause of the following exception:

```
TypeError
                                          Traceback (most recent call last)
Cell In[30], line 1
---> 1 card_approval_df.groupby(by='Approved').agg('mean')[['Age','Debt','YearsEmpl
oyed']]
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\groupby\generic.py:1445,
in DataFrameGroupBy.aggregate(self, func, engine, engine_kwargs, *args, **kwargs)
  1442
            kwargs["engine_kwargs"] = engine_kwargs
  1444 op = GroupByApply(self, func, args=args, kwargs=kwargs)
-> 1445 result = op.agg()
   1446 if not is_dict_like(func) and result is not None:
            # GH #52849
   1447
   1448
            if not self.as_index and is_list_like(func):
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\apply.py:172, in Apply.a
gg(self)
    169 kwargs = self.kwargs
```

```
171 if isinstance(func, str):
--> 172
           return self.apply_str()
   174 if is dict like(func):
   175
            return self.agg_dict_like()
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\apply.py:586, in Apply.a
pply_str(self)
    584
               else:
    585
                   self.kwargs["axis"] = self.axis
--> 586 return self._apply_str(obj, func, *self.args, **self.kwargs)
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\apply.py:669, in Apply._
apply_str(self, obj, func, *args, **kwargs)
    667 f = getattr(obj, func)
   668 if callable(f):
           return f(*args, **kwargs)
--> 669
   671 # people may aggregate on a non-callable attribute
    672 # but don't let them think they can pass args to it
   673 assert len(args) == 0
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\groupby\groupby.py:2378,
in GroupBy.mean(self, numeric_only, engine, engine_kwargs)
  2371
            return self._numba_agg_general(
  2372
               grouped_mean,
  2373
               executor.float_dtype_mapping,
  2374
               engine_kwargs,
  2375
               min_periods=0,
  2376 )
  2377 else:
-> 2378 result = self._cython_agg_general(
  2379
               "mean",
               alt=lambda x: Series(x).mean(numeric_only=numeric only),
  2380
  2381
               numeric_only=numeric_only,
  2382
  2383
           return result.__finalize__(self.obj, method="groupby")
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\groupby\groupby.py:1929,
in GroupBy. cython agg general(self, how, alt, numeric only, min count, **kwargs)
            result = self._agg_py_fallback(how, values, ndim=data.ndim, alt=alt)
  1926
  1927
            return result
-> 1929 new_mgr = data.grouped_reduce(array_func)
  1930 res = self._wrap_agged_manager(new_mgr)
  1931 out = self._wrap_aggregated_output(res)
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\internals\managers.py:14
28, in BlockManager.grouped_reduce(self, func)
  1424 if blk.is object:
  1425
          # split on object-dtype blocks bc some columns may raise
  1426
           # while others do not.
          for sb in blk. split():
  1427
               applied = sb.apply(func)
-> 1428
  1429
               result_blocks = extend_blocks(applied, result_blocks)
  1430 else:
File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\internals\blocks.py:366,
in Block.apply(self, func, **kwargs)
```

```
360 @final
            361 def apply(self, func, **kwargs) -> list[Block]:
            362
            363
                    apply the function to my values; return a block if we are not
            364
            365
        --> 366
                    result = func(self.values, **kwargs)
            368
                    result = maybe_coerce_values(result)
                    return self._split_op_result(result)
            369
        File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\groupby\groupby.py:1926,
        in GroupBy._cython_agg_general.<locals>.array_func(values)
           1923 else:
           1924
                   return result
        -> 1926 result = self._agg_py_fallback(how, values, ndim=data.ndim, alt=alt)
           1927 return result
        File C:\ProgramData\anaconda3\Lib\site-packages\pandas\core\groupby\groupby.py:1878,
        in GroupBy._agg_py_fallback(self, how, values, ndim, alt)
           1876
                    msg = f"agg function failed [how->{how},dtype->{ser.dtype}]"
           1877
                    # preserve the kind of exception that raised
        -> 1878
                    raise type(err)(msg) from err
           1880 if ser.dtype == object:
           1881
                  res_values = res_values.astype(object, copy=False)
       TypeError: agg function failed [how->mean,dtype->object]
In [27]: plt.figure(figsize=(5,4))
         sns.kdeplot(data=card_approval_df,x='Age',hue='Approved',fill=True)
        C:\ProgramData\anaconda3\Lib\site-packages\seaborn\_oldcore.py:1119: FutureWarning:
        use_inf_as_na option is deprecated and will be removed in a future version. Convert
        inf values to NaN before operating instead.
         with pd.option_context('mode.use_inf_as_na', True):
Out[27]: <Axes: xlabel='Age', ylabel='Density'>
```



In [28]: sns.countplot(data=card_approval_df,x='Approved',hue='Gender')

```
AttributeError
                                          Traceback (most recent call last)
Cell In[28], line 1
----> 1 sns.countplot(data=card approval df,x='Approved',hue='Gender')
File C:\ProgramData\anaconda3\Lib\site-packages\seaborn\categorical.py:2955, in coun
tplot(data, x, y, hue, order, hue_order, orient, color, palette, saturation, width,
dodge, ax, **kwargs)
   2952 if ax is None:
   2953
          ax = plt.gca()
-> 2955 plotter.plot(ax, kwargs)
  2956 return ax
File C:\ProgramData\anaconda3\Lib\site-packages\seaborn\categorical.py:1587, in _Bar
Plotter.plot(self, ax, bar_kws)
  1585 """Make the plot."""
  1586 self.draw bars(ax, bar kws)
-> 1587 self.annotate_axes(ax)
  1588 if self.orient == "h":
  1589
            ax.invert yaxis()
File C:\ProgramData\anaconda3\Lib\site-packages\seaborn\categorical.py:767, in _Cate
goricalPlotter.annotate axes(self, ax)
            ax.set_ylim(-.5, len(self.plot_data) - .5, auto=None)
    766 if self.hue_names is not None:
            ax.legend(loc="best", title=self.hue_title)
--> 767
File C:\ProgramData\anaconda3\Lib\site-packages\matplotlib\axes\_axes.py:322, in Axe
s.legend(self, *args, **kwargs)
    204 @_docstring.dedent_interpd
    205 def legend(self, *args, **kwargs):
   206
    207
            Place a legend on the Axes.
    208
   (\ldots)
    320
            .. plot:: gallery/text_labels_and_annotations/legend.py
   321
--> 322
            handles, labels, kwargs = mlegend._parse_legend_args([self], *args, **kw
args)
    323
            self.legend_ = mlegend.Legend(self, handles, labels, **kwargs)
    324
            self.legend_._remove_method = self._remove_legend
File C:\ProgramData\anaconda3\Lib\site-packages\matplotlib\legend.py:1361, in _parse
_legend_args(axs, handles, labels, *args, **kwargs)
  1357
            handles = [handle for handle, label
  1358
                       in zip(_get_legend_handles(axs, handlers), labels)]
  1360 elif len(args) == 0: # 0 args: automatically detect labels and handles.
-> 1361
            handles, labels = _get_legend_handles_labels(axs, handlers)
  1362
            if not handles:
  1363
                log.warning(
                    "No artists with labels found to put in legend. Note that "
  1364
  1365
                    "artists whose label start with an underscore are ignored "
  1366
                    "when legend() is called with no argument.")
File C:\ProgramData\anaconda3\Lib\site-packages\matplotlib\legend.py:1291, in _get_1
egend_handles_labels(axs, legend_handler_map)
```

```
1289 for handle in _get_legend_handles(axs, legend_handler_map):
1290    label = handle.get_label()
-> 1291    if label and not label.startswith('_'):
1292         handles.append(handle)
1293         labels.append(label)

AttributeError: 'numpy.int64' object has no attribute 'startswith'
```

