

Day Objectives

Identifying and Eliminating Outliers

- Outliers are observations that are significantly different from other data points
- Outliers can adversely affect the training process of a machine learning algorithm, resulting in a loss of accuracy.
- Need to use the mathematical formula and retrieve the outlier data.

$$\text{interquartile range(IQR)} = Q3(\text{quantile}(0.75)) - Q1(\text{quantile}(0.25))$$

In [4]:

```
import pandas as pd
import matplotlib.pyplot as plt
```

In [2]:

```
adv = pd.read_csv("https://raw.githubusercontent.com/APSSDC-Data-Analysis/Data-Analysis-batch8/main/Datasets/Advertising.csv")
adv.head()
```

Out[2]:

	TV	radio	newspaper	sales
0	230.1	37.8	69.2	22.1
1	44.5	39.3	45.1	10.4
2	17.2	45.9	69.3	9.3
3	151.5	41.3	58.5	18.5
4	180.8	10.8	58.4	12.9

In [3]:

```
adv.shape
```

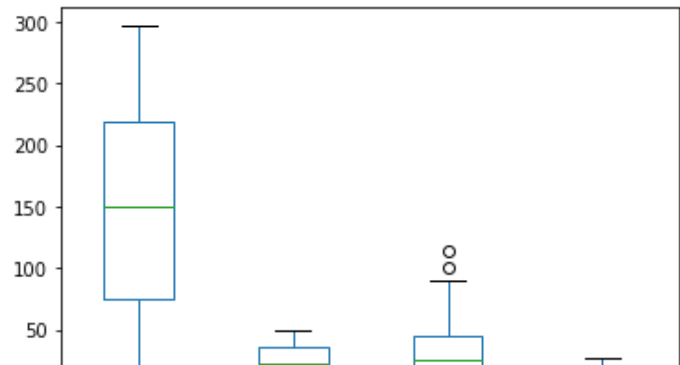
Out[3]:

(200, 4)

Identifing Outliers

In [6]:

```
adv.plot( kind = "box")
plt.show()
```





In [7]:

```
adv.describe()
```

Out[7]:

	TV	radio	newspaper	sales
count	200.000000	200.000000	200.000000	200.000000
mean	147.042500	23.264000	30.554000	14.022500
std	85.854236	14.846809	21.778621	5.217457
min	0.700000	0.000000	0.300000	1.600000
25%	74.375000	9.975000	12.750000	10.375000
50%	149.750000	22.900000	25.750000	12.900000
75%	218.825000	36.525000	45.100000	17.400000
max	296.400000	49.600000	114.000000	27.000000

In [8]:

```
# interquartile range(IQR) = Q3(quantile(0.75)) - Q1(quantile(0.25))
```

In [9]:

```
Q1 = adv.quantile(0.25)
Q1
```

Out[9]:

```
TV          74.375
radio       9.975
newspaper   12.750
sales      10.375
Name: 0.25, dtype: float64
```

In [10]:

```
Q3 = adv.quantile(0.75)
Q3
```

Out[10]:

```
TV          218.825
radio       36.525
newspaper   45.100
sales      17.400
Name: 0.75, dtype: float64
```

In [11]:

```
IQR = Q3 - Q1
IQR
```

Out[11]:

```
TV          144.450
radio       26.550
newspaper   32.350
sales       7.025
dtype: float64
```

In [12]:

```
min_value = (Q1 - 1.5*IQR)
max_value = (Q3 + 1.5*IQR)
```

In [15]:

```
outliers_df = adv[((adv < min_value ) | (adv > max_value)).any(axis = 1)]
outliers_df
```

Out[15]:

	TV	radio	newspaper	sales
16	67.8	36.6	114.0	12.5
101	296.4	36.3	100.9	23.8

In [20]:

```
filter_df = adv[~((adv < min_value ) | (adv > max_value)).any(axis = 1)]
filter_df
```

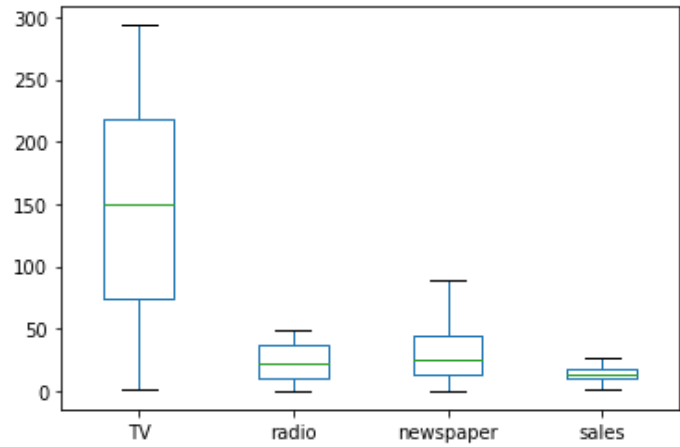
Out[20]:

	TV	radio	newspaper	sales
0	230.1	37.8	69.2	22.1
1	44.5	39.3	45.1	10.4
2	17.2	45.9	69.3	9.3
3	151.5	41.3	58.5	18.5
4	180.8	10.8	58.4	12.9
...
195	38.2	3.7	13.8	7.6
196	94.2	4.9	8.1	9.7
197	177.0	9.3	6.4	12.8
198	283.6	42.0	66.2	25.5
199	232.1	8.6	8.7	13.4

198 rows x 4 columns

In [18]:

```
filter_df.plot(kind = "box")
plt.show()
```



In []:

In []:

Preprocessing Techniques

- Data Preprocessing is a technique that is used to convert the raw data into a clean data set

Data preprocessing steps

- Loading data
- Exploring data
- Cleaning data
- Transforming data
 - will learn data preprocessing techniques with scikit-learn, one of the most popular frameworks used for industry data science
 - The scikit-learn library includes tools for data preprocessing and data mining. It is imported in Python via the statement `import sklearn`.



Data Imputation

- if the dataset is missing too many values, we just don't use it
- if only a few of the values are missing, we can perform data imputation to substitute the missing data with some other value(s).
- There are many different methods for data imputation
 - Using the mean value
 - Using the median value
 - Using the most frequent value
 - Filling in missing values with a constant

Feature Scaling

1. Standardizing Data

- Data scientists will convert the data into a standard format to make it easier to understand.
- The standard format refers to data that has 0 mean and unit variance (i.e. standard deviation = 1), and the process of converting data into this format is called data standardization.
- improve the performance of models
- it rescales the data to have mean = 0 and variance (statistical measure that provides indicator of data's dispersion) = 1
- Standardization rescales data so that it has a mean of 0 and a standard deviation of 1.
- The formula for this is: $(x - \mu) / \sigma$
 - We subtract the mean (μ) from each value (x) and then divide by the standard deviation (σ)



2. Data Range

- Scale data by compressing it into a fixed range
- One of the biggest use cases for this is compressing data into the range [0, 1]
- Classifier is `MinMaxScaler`



3. Normalizing Data

- Want to scale the individual data observations (i.e. rows)
- Used in classification Problems and data mining
- ex : columns : `salary,ex_yr,position_levels`
- when clustering data we need to apply L2 normalization to each row

- L2 normalization applied to a particular row of a data array
- L2 norm of a row is just the square root of the sum of squared values for the row

4. Robust Scaling

- Deal with is outliers (data point that is significantly further away from the other data points)
- Robustly scale the data, i.e. avoid being affected by outliers
- Scaling by using data's median and Interquartile Range (IQR)
- Here mean affected but median remains same
- Subtract the median from each data value then scale to the IQR

Data Imputation

- Handling Missing Data
 - There are many different methods for data imputation
 - Using the mean value
 - Using the median value
 - Using the most frequent value
 - Filling in missing values with a constant

In [27]:

```
titanic = pd.read_csv("https://raw.githubusercontent.com/APSSDC-Data-Analysis/Data-Analysis-batch8/main/Datasets/titanic.csv")
titanic.head()
```

Out[27]:

	survived	pclass	name	sex	age	sibsp	parch	ticket	fare	cabin	embarked
0	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1	0	PC 17599	71.2833	C85	C
2	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S
4	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S

In [28]:

```
titanic.shape # 1000 emp details
```

Out[28]:

(891, 11)

In [29]:

```
titanic.isnull().sum()
```

Out[29]:

```

survived      0
pclass        0
name          0
sex           0
age          177
sibsp         0
parch         0
ticket        0
fare          0
cabin        687

```

```
cabin      2
embarked    2
dtype: int64
```

In [30]:

```
titanic.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 11 columns):
#   Column      Non-Null Count  Dtype  
---  -
0   survived    891 non-null    int64  
1   pclass      891 non-null    int64  
2   name        891 non-null    object  
3   sex         891 non-null    object  
4   age         714 non-null    float64 
5   sibsp       891 non-null    int64  
6   parch       891 non-null    int64  
7   ticket      891 non-null    object  
8   fare        891 non-null    float64 
9   cabin       204 non-null    object  
10  embarked    889 non-null    object  
dtypes: float64(2), int64(4), object(5)
memory usage: 76.7+ KB
```

In [24]:

```
from sklearn.impute import SimpleImputer
```

In [25]:

```
si = SimpleImputer(strategy= "mean")
si
```

Out[25]:

```
SimpleImputer()
```

In [32]:

```
si.fit_transform(titanic[["age"]])
```

Out[32]:

```
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In [33]:

```
titanic[["age"]].mean()
```

Out[33]:

```
age      29.699118
dtype: float64
```

In [34]:

```
si = SimpleImputer(strategy= "median")
si.fit_transform(titanic[["age"]])
```

Out[34]:

```
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```

In [35]:

```
titanic[["age"]].median()
```

Out[35]:

```

age      28.0
dtype: float64

```

In [38]:

```

si = SimpleImputer(strategy= "most_frequent")
si.fit_transform(titanic[["age"]])

```

Out[38]:

```

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In [41]:

```
titanic[["age"]].mode()
```

Out[41]:

	age
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In [42]:

```
si = SimpleImputer(strategy= "constant", fill_value = -1)
si.fit_transform(titanic[["age"]])
```

Out[42]:

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```

1. Standardizing Data

scaling

example: 20m, 30mm, 75cm, 1km lengths ---- Total length -- 20+30+75+1 =

single unit m or km or mm --- adding --- accurate output

weights... kg, g, mg, tones,

In []:

In [43]:

```
adv.head()
```

Out[43]:

	TV	radio	newspaper	sales
0	230.1	37.8	69.2	22.1
1	44.5	39.3	45.1	10.4
2	17.2	45.9	69.3	9.3
3	151.5	41.3	58.5	18.5
4	180.8	10.8	58.4	12.9

In [44]:

```
adv.isnull().sum()
```

Out[44]:

```
TV          0
radio       0
newspaper   0
sales       0
dtype: int64
```

In [45]:

```
adv.mean()    # mean becomes 0
```

Out[45]:

```
TV          147.0425
radio       23.2640
newspaper   30.5540
sales       14.0225
dtype: float64
```

In [46]:

```
adv.std()    # std becomes 1    stande format
```

Out[46]:

```
TV          85.854236
radio       14.846809
newspaper   21.778621
sales       5.217457
dtype: float64
```

In [47]:

```
from sklearn.preprocessing import scale
```

In [48]:

```
adv_new = scale(adv)
adv_new
```

Out[48]:

```
array([[ 9.69852266e-01,  9.81522472e-01,  1.77894547e+00,
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       [-1.19737623e+00,  1.08280781e+00,  6.69578760e-01,
```

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-6.96046111e-01],
[-5.93681653e-01, -5.71519414e-01, 3.84181516e-01,
-5.03900877e-01],
[-7.87313476e-02, -1.44257334e+00, -9.92169710e-01,

-7.15260635e-01],
[1.08662104e+00, -1.07794612e+00, -1.00597925e+00,
-1.58039455e-01],
[1.12281936e+00, 1.73778635e+00, 6.32753309e-01,
2.18613240e+00],
[-1.27327593e+00, 1.15033137e+00, -8.58677450e-01,
-5.99973494e-01],
[-1.19504085e+00, 1.71239749e-01, -4.58200672e-01,
-7.53689682e-01],
[1.56070228e+00, -6.32290618e-01, 2.96721070e-01,
3.99181724e-01],
[-3.04095087e-01, -1.00367020e+00, 8.35293289e-01,
-4.65471830e-01],
[5.90353742e-01, 2.43084817e-03, -7.52804279e-01,
4.95254341e-01],
[2.83251860e-01, 1.10981724e+00, 3.28943340e-01,
9.56402904e-01],
[4.75920341e-01, -1.46120984e-01, -9.69153803e-01,
3.03109107e-01],
[-1.66912209e+00, -7.87594807e-01, -1.14407469e+00,
-2.07949180e+00],
[-6.20538471e-01, 1.36640677e+00, 9.18150553e-01,
2.45465537e-01],
[3.21989902e-02, -1.48308748e+00, -2.87882962e-01,
-7.53689682e-01],
[-1.58037782e+00, 9.20751268e-01, 6.74181942e-01,
-1.29169634e+00],
[-1.79152496e-01, -3.28434597e-01, 1.86244718e-01,
-2.15683025e-01],
[2.97264113e-01, -3.48691665e-01, 6.72064478e-03,
7.25348259e-02],
[-7.16288868e-01, 8.46475352e-01, 8.62912377e-01,
-1.38824932e-01],
[4.82926468e-01, -3.48691665e-01, -2.28041604e-01,
1.68607443e-01],
[1.92172214e-01, 9.13998912e-01, -1.06582061e+00,
7.64257669e-01],
[-3.48467222e-01, -5.78271770e-01, -1.15788424e+00,
-4.07828260e-01],
[1.02123053e+00, -1.34128800e+00, 2.49704176e+00,
-4.07828260e-01],
[-1.50798117e+00, 9.68017760e-01, -4.12168859e-01,
-1.15719467e+00],
[6.97781017e-01, -1.21974559e+00, -5.13438849e-01,
-3.50184689e-01],
[7.98202165e-01, 2.26879163e-02, 1.24497643e+00,
5.91326959e-01],
[1.60273904e+00, -8.55118367e-01, -1.11185242e+00,
1.87821967e-01],
[-1.13315340e+00, -7.87594807e-01, -5.59470662e-01,
-1.08033658e+00],
[2.03849092e-01, -1.59625696e-01, 7.75451931e-01,
9.17493494e-02],
[-1.48813048e+00, -2.13644545e-01, -6.23915201e-01,
-1.23405277e+00],
[2.49388915e-01, -1.09145083e+00, -8.17248818e-01,
-4.46257307e-01],
[8.79940308e-01, -1.34128800e+00, -8.03439274e-01,
-4.84686354e-01],
[1.51633014e+00, 1.73103399e+00, 5.17673775e-01,
2.49356478e+00],
[1.18353913e+00, 4.68343414e-01, -4.72010216e-01,
1.18697718e+00],
[2.70407294e-01, -1.04418434e+00, 2.13863806e-01,
-4.46257307e-01],
[1.51399477e+00, -1.41556392e+00, -3.15502050e-01,
-4.27042783e-01],
[2.16693657e-01, -8.95632503e-01, -5.96296113e-01,
-2.73326596e-01],
[1.11601758e-01, -1.39530685e+00, -1.02439198e+00,
-6.76831588e-01],
[8.34400486e-01, -1.20624088e+00, -1.45184340e-01,

```

-3.50184689e-01],
[-1.06075676e+00, -1.18598381e+00, -3.93111688e-02,
-1.02269301e+00],
[ 1.64127273e+00,  1.33264499e+00,  1.89862818e+00,
 2.33984859e+00],
[ 1.24659427e+00, -1.32616272e-01, -2.55016247e-02,
 6.87399576e-01],
[ 6.76762637e-01,  1.47444446e+00, -5.04232486e-01,
 1.64812575e+00],
[-8.80728498e-02, -1.42906863e+00, -1.82009791e-01,
-7.15260635e-01],
[ 5.14454038e-01,  3.67058074e-01, -5.68677025e-01,
 6.29756005e-01],
[ 1.62258973e+00, -6.32290618e-01, -1.23613832e+00,
 3.60752677e-01],
[-1.49863967e+00, -7.53833027e-01, -3.29311594e-01,
-1.40698348e+00],
[-1.25576062e+00,  1.20435022e+00, -1.13947151e+00,
-6.19188018e-01],
[-8.35393020e-01, -8.41613655e-01, -1.13026515e+00,
-7.92118728e-01],
[-1.51615499e+00, -1.29402151e+00,  4.81492770e-02,
-1.56069967e+00],
[ 2.30705910e-01,  1.26512143e+00, -1.24074150e+00,
 1.07169004e+00],
[ 3.10313024e-02,  8.32970639e-01, -1.13026515e+00,
 6.29756005e-01],
[-1.27094056e+00, -1.32103093e+00, -7.71217005e-01,
-1.23405277e+00],
[-6.17035408e-01, -1.24000266e+00, -1.03359834e+00,
-8.30547775e-01],
[ 3.49810063e-01, -9.42898996e-01, -1.11185242e+00,
-2.34897549e-01],
[ 1.59456522e+00,  1.26512143e+00,  1.64085003e+00,
 2.20534693e+00],
[ 9.93206022e-01, -9.90165488e-01, -1.00597925e+00,
-1.19610408e-01]]))

```

In [49]:

```

# 2D array into df
adv_new_df = pd.DataFrame(adv_new, columns = adv.columns)
adv_new_df.head()

```

Out[49]:

	TV	radio	newspaper	sales
0	0.969852	0.981522	1.778945	1.552053
1	-1.197376	1.082808	0.669579	-0.696046
2	-1.516155	1.528463	1.783549	-0.907406
3	0.052050	1.217855	1.286405	0.860330
4	0.394182	-0.841614	1.281802	-0.215683

In [51]:

```

adv_new_df.mean().round(2)

```

Out[51]:

```

TV          0.0
radio       -0.0
newspaper   0.0
sales       -0.0
dtype: float64

```

In [52]:

```

adv_new_df.std().round(2)

```

Out[52]:

```
TV          1.0
radio       1.0
newspaper   1.0
sales       1.0
dtype: float64
```

In [55]:

```
(adv["TV"][0] - adv["TV"].mean()) / (adv["TV"].std())
```

Out[55]:

```
0.9674245973763037
```

In [56]:

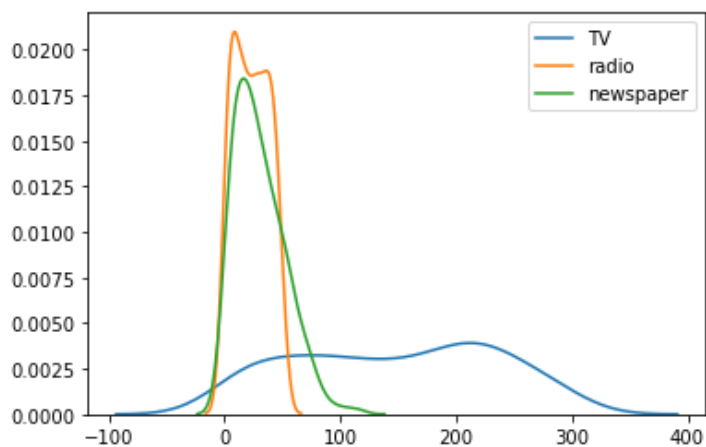
```
import seaborn as sns
```

In [57]:

```
sns.kdeplot(adv["TV"])
sns.kdeplot(adv["radio"])
sns.kdeplot(adv["newspaper"])
```

Out[57]:

<matplotlib.axes._subplots.AxesSubplot at 0x1939a9c96d0>

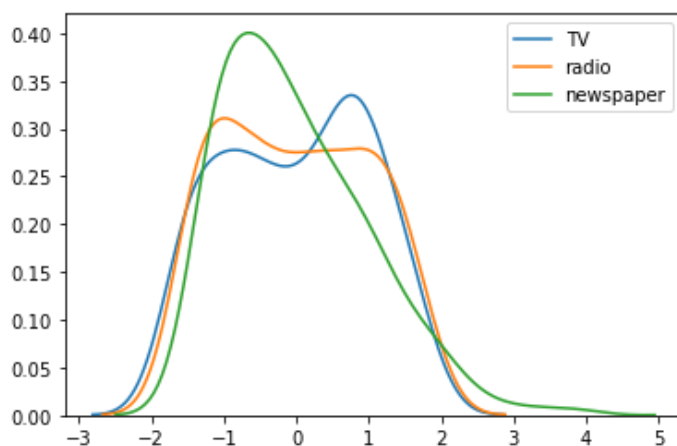


In [58]:

```
sns.kdeplot(adv_new_df["TV"])
sns.kdeplot(adv_new_df["radio"])
sns.kdeplot(adv_new_df["newspaper"])
```

Out[58]:

<matplotlib.axes._subplots.AxesSubplot at 0x1939aa54580>



2. Data Range

In [59]:

```
adv.head()
```

Out[59]:

	TV	radio	newspaper	sales
0	230.1	37.8	69.2	22.1
1	44.5	39.3	45.1	10.4
2	17.2	45.9	69.3	9.3
3	151.5	41.3	58.5	18.5
4	180.8	10.8	58.4	12.9

In [60]:

```
from sklearn.preprocessing import MinMaxScaler
```

In [61]:

```
mnscale = MinMaxScaler()  
mnscale
```

Out[61]:

```
MinMaxScaler()
```

In [63]:

```
mnscale = mnscale.fit_transform(adv)
```

In [64]:

```
mnscaled_data = pd.DataFrame(mnscale, columns= adv.columns)  
mnscaled_data
```

Out[64]:

	TV	radio	newspaper	sales
0	0.775786	0.762097	0.605981	0.807087
1	0.148123	0.792339	0.394019	0.346457
2	0.055800	0.925403	0.606860	0.303150
3	0.509976	0.832661	0.511873	0.665354
4	0.609063	0.217742	0.510994	0.444882
...
195	0.126818	0.074597	0.118734	0.236220
196	0.316199	0.098790	0.068602	0.318898
197	0.596212	0.187500	0.053650	0.440945
198	0.956713	0.846774	0.579595	0.940945
199	0.782550	0.173387	0.073879	0.464567

200 rows x 4 columns

In [67]:

```
adv.min(), mnscaled_data.min()
```

Out[67]:

```
(TV          0.7
 radio       0.0
 newspaper   0.3
 sales       1.6
 dtype: float64,
 TV          0.0
 radio       0.0
 newspaper   0.0
 sales       0.0
 dtype: float64)
```

In [68]:

```
adv.max(), mnscaled_data.max()
```

Out[68]:

```
(TV          296.4
 radio        49.6
 newspaper   114.0
 sales        27.0
 dtype: float64,
 TV          1.0
 radio        1.0
 newspaper    1.0
 sales        1.0
 dtype: float64)
```

In [70]:

```
(adv["TV"][0] - adv["TV"].min()) / (adv["TV"].max() - adv["TV"].min())
```

Out[70]:

```
0.7757862698681096
```

3. Normalizing Data

In [71]:

```
adv.head()
```

Out[71]:

	TV	radio	newspaper	sales
0	230.1	37.8	69.2	22.1
1	44.5	39.3	45.1	10.4
2	17.2	45.9	69.3	9.3
3	151.5	41.3	58.5	18.5
4	180.8	10.8	58.4	12.9

In [72]:

```
from sklearn.preprocessing import Normalizer
```

In [73]:

```
norm_data = Normalizer()
```

In [75]:

```
norm_data = norm_data.fit_transform(adv)
```

In [76]:

```
# 2d into df
```

```
norm_data = pd.DataFrame(norm_data, columns = adv.columns)
norm_data
```

Out[76]:

	TV	radio	newspaper	sales
0	0.942116	0.154767	0.283331	0.090486
1	0.591135	0.522059	0.599106	0.138153
2	0.201426	0.537527	0.811561	0.108911
3	0.898632	0.244974	0.346997	0.109734
4	0.947881	0.056621	0.306174	0.067631
...
195	0.920781	0.089186	0.332638	0.183192
196	0.989807	0.051487	0.085111	0.101923
197	0.995383	0.052300	0.035991	0.071983
198	0.960250	0.142209	0.224149	0.086341
199	0.996958	0.036940	0.037370	0.057558

200 rows x 4 columns

4. Robust Scaling

In [77]:

```
adv.head()
```

Out[77]:

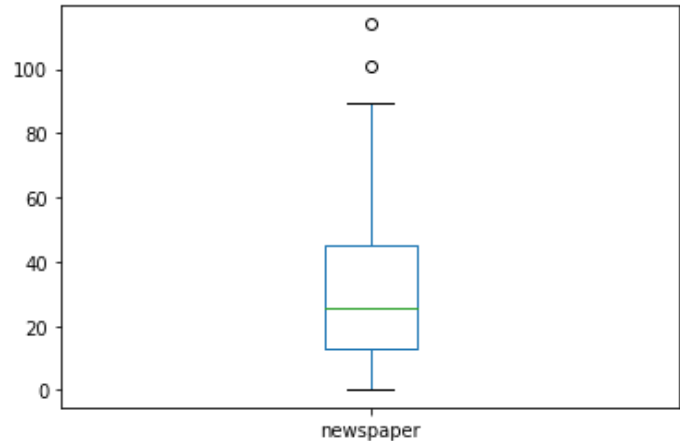
	TV	radio	newspaper	sales
0	230.1	37.8	69.2	22.1
1	44.5	39.3	45.1	10.4
2	17.2	45.9	69.3	9.3
3	151.5	41.3	58.5	18.5
4	180.8	10.8	58.4	12.9

In [78]:

```
adv["newspaper"].plot(kind = "box")
```

Out[78]:

<matplotlib.axes._subplots.AxesSubplot at 0x1939ac6ad90>



In [79]:

```
In [79]:
```

```
from sklearn.preprocessing import RobustScaler
```

```
In [80]:
```

```
rscale = RobustScaler()  
rscale.fit(adv)
```

```
Out[80]:
```

```
RobustScaler()
```

```
In [82]:
```

```
rscale = rscale.transform(adv)
```

```
In [83]:
```

```
rscale = pd.DataFrame(rscale, columns = adv.columns)  
rscale
```

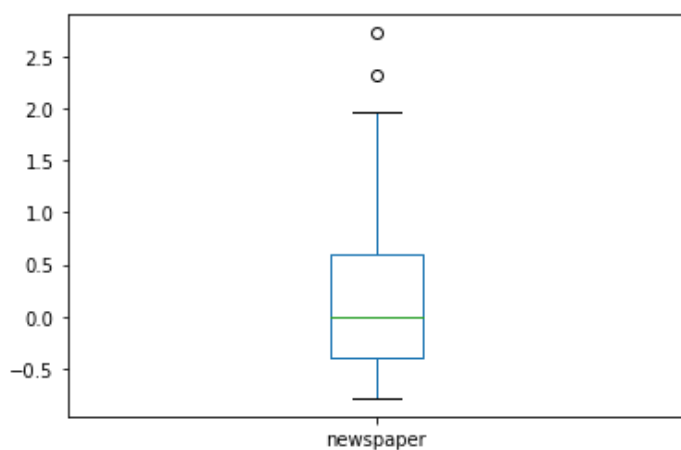
```
Out[83]:
```

	TV	radio	newspaper	sales
0	0.556248	0.561205	1.343122	1.309609
1	-0.728626	0.617702	0.598145	-0.355872
2	-0.917619	0.866290	1.346213	-0.512456
3	0.012115	0.693032	1.012365	0.797153
4	0.214953	-0.455744	1.009274	0.000000
...
195	-0.772240	-0.723164	-0.369397	-0.754448
196	-0.384562	-0.677966	-0.545595	-0.455516
197	0.188647	-0.512241	-0.598145	-0.014235
198	0.926618	0.719397	1.250386	1.793594
199	0.570093	-0.538606	-0.527048	0.071174

200 rows x 4 columns

```
In [85]:
```

```
rscale["newspaper"].plot(kind = "box")  
plt.show()
```

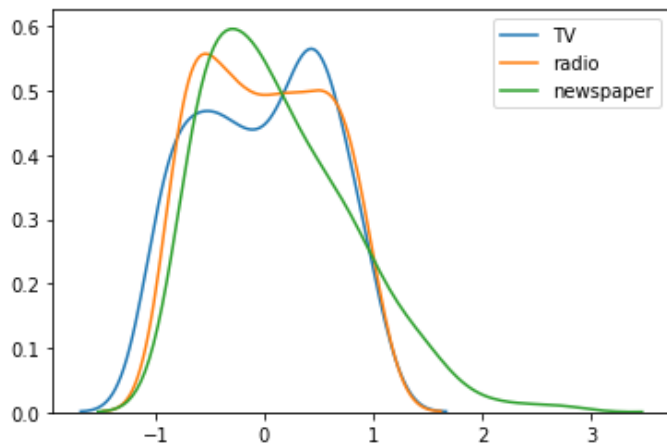


```
In [86]:
```

```
sns.kdeplot(rscale["TV"])  
sns.kdeplot(rscale["radio"])  
sns.kdeplot(rscale["newspaper"])
```

Out[86]:

<matplotlib.axes._subplots.AxesSubplot at 0x1939a6cf160>



Label Encoder

In []:

```
# True and False
# 1 or 0
# used to convert categorical data into numeric type data
```

In [87]:

```
emp.head()
```

Out[87]:

	First Name	Gender	Start Date	Last Login Time	Salary	Bonus %	Senior Management	Team
0	Douglas	Male	8/6/1993	12:42 PM	97308	6.945	True	Marketing
1	Thomas	Male	3/31/1996	6:53 AM	61933	4.170	True	NaN
2	Maria	Female	4/23/1993	11:17 AM	130590	11.858	False	Finance
3	Jerry	Male	3/4/2005	1:00 PM	138705	9.340	True	Finance
4	Larry	Male	1/24/1998	4:47 PM	101004	1.389	True	Client Services

In [88]:

```
emp["Gender"].value_counts()
```

Out[88]:

```
Female    431
Male      424
Name: Gender, dtype: int64
```

In [89]:

```
from sklearn.preprocessing import LabelEncoder
```

In [90]:

```
Le_data = LabelEncoder()
```

In [92]:

```
Le_data.fit_transform(emp[["Gender"]])
```

```
C:\Users\lavan\anaconda3\lib\site-packages\sklearn\utils\validation.py:73: DataConversion
Warning: A column-vector y was passed when a 1d array was expected. Please change the sha
pe of y to (n_samples, ), for example using ravel().
return f(**kwargs)
```

```
return 1(**kwargs)
```

```
-----
TypeError                                Traceback (most recent call last)
~\anaconda3\lib\site-packages\sklearn\preprocessing\_label.py in _encode(values, uniques,
encode, check_unknown)
    112         try:
--> 113             res = _encode_python(values, uniques, encode)
    114         except TypeError:

~\anaconda3\lib\site-packages\sklearn\preprocessing\_label.py in _encode_python(values, u
niques, encode)
    60     if uniques is None:
---> 61         uniques = sorted(set(values))
    62         uniques = np.array(uniques, dtype=values.dtype)
```

TypeError: '<' not supported between instances of 'str' and 'float'

During handling of the above exception, another exception occurred:

```
TypeError                                Traceback (most recent call last)
<ipython-input-92-c5b8c3f778d3> in <module>
----> 1 Le_data.fit_transform(emp[["Gender"]])

~\anaconda3\lib\site-packages\sklearn\preprocessing\_label.py in fit_transform(self, y)
    254         """
    255         y = column_or_1d(y, warn=True)
--> 256         self.classes_, y = _encode(y, encode=True)
    257         return y
    258

~\anaconda3\lib\site-packages\sklearn\preprocessing\_label.py in _encode(values, uniques,
encode, check_unknown)
    115         types = sorted(t.__qualname__
    116                         for t in set(type(v) for v in values))
--> 117         raise TypeError("Encoders require their input to be uniformly "
    118                         f"strings or numbers. Got {types}")
    119         return res
```

TypeError: Encoders require their input to be uniformly strings or numbers. Got ['float', 'str']

In []:

```
# Got ['float', 'str']

# 100/100
```

In [93]:

```
emp["Gender"].isnull().sum()
# NAN -- not a number --- Special type of floating value
```

Out[93]:

145

In [94]:

```
emp.dropna(inplace = True)
```

In [95]:

```
emp.shape
```

Out[95]:

(764, 8)

In [96]:

```
emp.isnull().sum()
```

Out[96]:

```
First Name      0
Gender          0
Start Date      0
Last Login Time 0
Salary          0
Bonus %         0
Senior Management 0
Team            0
dtype: int64
```

In [97]:

```
Le_data.fit_transform(emp[["Gender"]])
```

Out[97]:

```
array([[1, 0, 1, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1, 1, 1,
        0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 0,
        0, 0, 1, 1, 0, 0, 1, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 0, 1, 0,
        0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 1, 0, 0, 0,
        1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0,
        1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1,
        1, 1, 0, 0, 1, 1, 0, 1, 0, 1, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0,
        1, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1,
        0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0,
        0, 1, 0, 0, 0, 0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 1,
        0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1,
        1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0,
        0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0,
        0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1,
        0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1,
        1, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 0, 1, 0, 1,
        0, 1, 0, 0, 1, 1, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1,
        1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 1, 1, 1, 0, 1, 1,
        1, 0, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1,
        0, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0,
        1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 0,
        0, 1, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0,
        0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0, 0, 0,
        0, 0, 0, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0,
        1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 0, 1, 0, 1, 0, 0,
        0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1,
        1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0,
        1, 0, 1, 1, 1, 0, 1, 0, 1, 1, 0, 0, 0, 1, 1, 1, 0, 0, 0, 1, 1, 0,
        1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1])
```

In [98]:

```
emp["Gender"]
```

Out[98]:

```
0      Male
2      Female
3      Male
4      Male
5      Male
...
994    Male
996    Male
997    Male
998    Male
999    Male
Name: Gender, Length: 764, dtype: object
```

In []: