```
import pandas as pd
import numpy as np
import seaborn as sns
from scipy.stats import skew
%matplotlib inline
import matplotlib.pyplot as plt
plt.style.use("ggplot")
plt.rcParams['figure.figsize'] = (12, 8)
df=pd.read_csv('Advertising.csv')
```

df.head(10)

_		TV	radio	newspaper	sales	
	0	230.1	37.8	69.2	22.1	ıl.
	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	
	5	8.7	48.9	75.0	7.2	
	6	57.5	32.8	23.5	11.8	
	7	120.2	19.6	11.6	13.2	
	8	8.6	2.1	1.0	4.8	
	9	199.8	2.6	21.2	10.6	

Generate code with df Next steps:

View recommended plots

New interactive sheet

df.info()

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<class 'pandas.core.frame.DataFrame'> RangeIndex: 200 entries, 0 to 199

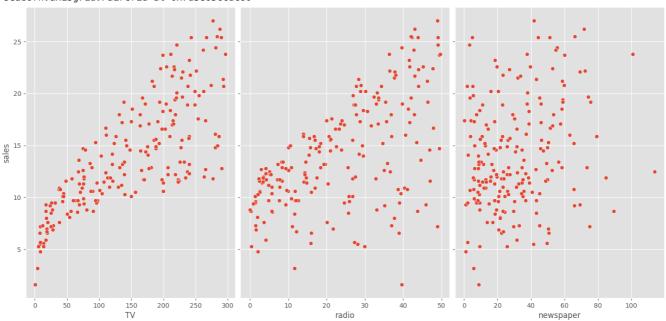
Data columns (total 4 columns):
Column Non-Null Count Dtype ___ TV float64 0 200 non-null 1 radio 200 non-null float64 200 non-null float64 newspaper

sales 200 non-null dtypes: float64(4) memory usage: 6.4 KB

sns.pairplot(df,x_vars=['TV','radio','newspaper'],y_vars='sales',height=7,aspect=0.7)

float64

<seaborn.axisgrid.PairGrid at 0x7d3e65ccdc60>



sns.heatmap(df.corr(),annot=True)

 $https://colab.research.google.com/drive/1aScafdUKE4gDhfJInafezhn3_SrrJson\#scrollTo=9pRPsA5E0ijP\&printMode=true$

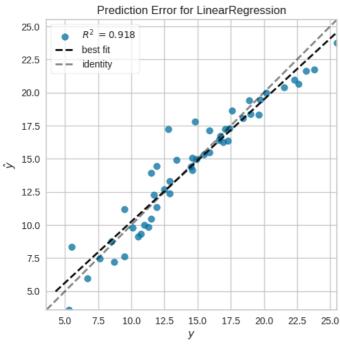
→ <Axes: >



```
from sklearn.metrics import r2_score
lm2=LinearRegression()
lm2.fit(x[['TV','radio']],y)
lm2_pred=lm2.predict(x[['TV','radio']])
print(r2_score(y,lm2_pred))
0.8971942610828957
lm3=LinearRegression()
lm3.fit(x[['TV','radio','newspaper']],y)
lm3_pred=lm3.predict(x[['TV','radio','newspaper']])
print(r2_score(y,lm3_pred))
→ 0.8972106381789522
from sklearn.model_selection import train_test_split
from sklearn.metrics import mean_squared_error
X=df.drop('sales',axis=1)
y=df['sales']
x\_train, x\_test, y\_train, y\_test=train\_test\_split(X, y, random\_state=1)
lm4=LinearRegression().fit(x_train,y_train)
lm4_pred=lm4.predict(x_test)
print("RMSE:",np.sqrt(mean_squared_error(y_test,lm4_pred)))
print("R*2:",r2_score(y_test,lm4_pred))
RMSE: 1.404651423032895
    R*2: 0.9156213613792232
X=df.drop(['sales','newspaper'],axis=1)
y=df['sales']
x\_train, x\_test, y\_train, y\_test=train\_test\_split(X, y, random\_state=1)
lm5=LinearRegression().fit(x_train,y_train)
lm5_pred=lm5.predict(x_test)
print("RMSE:",np.sqrt(mean_squared_error(y_test,lm5_pred)))
print("R*2:",r2_score(y_test,lm5_pred))
    RMSE: 1.3879034699382888
\overline{2}
    R*2: 0.9176214942248907
```

from yellowbrick.regressor import PredictionError,ResidualsPlot v=PredictionError(lm5).fit(x_train,y_train) v.score(x_test,y_test) v.poof()

/usr/local/lib/python3.10/dist-packages/sklearn/base.py:465: UserWarning: X does not have valid feature names, but L:
 warnings.warn(



<Axes: title={'center': 'Prediction Error for LinearRegression'}, xlabel='\$y\$', ylabel='\$\\hat{y}\$'>

df['interaction']=df['TV']*df['radio']
X=df[['TV','radio','interaction']]
y=df['sales']
x_train,x_test,y_train,y_test=train_test_split(X,y,random_state=1)
lm6=LinearRegression().fit(x_train,y_train)
lm6_pred=lm6.predict(x_test)
print("RMSE:",np.sqrt(mean_squared_error(y_test,lm6_pred)))
print("R*2:",r2_score(y_test,lm6_pred))

RMSE: 0.7011871137164328 R*2: 0.978973681468126