```
In [3]:
        diabetes
Out[3]: {'data': array([[ 0.03807591, 0.05068012, 0.06169621, ..., -0.00259226,
                  0.01990749, -0.01764613],
                [-0.00188202, -0.04464164, -0.05147406, ..., -0.03949338,
                 -0.06833155, -0.092204051,
                [0.08529891, 0.05068012, 0.04445121, ..., -0.00259226,
                  0.00286131, -0.025930341,
                [0.04170844, 0.05068012, -0.01590626, ..., -0.01107952,
                 -0.04688253, 0.01549073],
                [-0.04547248, -0.04464164, 0.03906215, ..., 0.02655962,
                  0.04452873, -0.02593034],
                [-0.04547248, -0.04464164, -0.0730303, ..., -0.03949338,
                 -0.00422151, 0.00306441]]),
         'target': array([151., 75., 141., 206., 135., 97., 138., 63., 110., 3
        10., 101.,
                 69., 179., 185., 118., 171., 166., 144., 97., 168.,
                                                                       68., 49.,
                 68., 245., 184., 202., 137., 85., 131., 283., 129.,
                                                                       59., 341.,
                       65., 102., 265., 276., 252., 90., 100., 55.,
                 87.,
                                                                       61., 92.,
                       53., 190., 142., 75., 142., 155., 225., 59., 104., 182.,
                259.,
In [1]: from sklearn import datasets
In [2]: | diabetes = datasets.load diabetes()
```

In [4]: |print(diabetes.DESCR)

.. _diabetes_dataset:

Diabetes dataset

Ten baseline variables, age, sex, body mass index, average blood pressure, and six blood serum measurements were obtained for each of n=442 diabetes patients, as well as the response of interest, a quantitative measure of disease progression one year after baseline.

Data Set Characteristics:

:Number of Instances: 442

:Number of Attributes: First 10 columns are numeric predictive values

:Target: Column 11 is a quantitative measure of disease progression one year after baseline

:Attribute Information:

- age age in years
- sex
- bmi body mass index
- bp average blood pressure
- s1 tc, total serum cholesterol
- s2 ldl, low-density lipoproteins
- s3 hdl, high-density lipoproteins
- s4 tch, total cholesterol / HDL
- s5 ltg, possibly log of serum triglycerides level
- s6 glu, blood sugar level

Note: Each of these 10 feature variables have been mean centered and scal ed by the standard deviation times the square root of `n_samples` (i.e. the sum of squares of each column totals 1).

Source URL:

https://www4.stat.ncsu.edu/~boos/var.select/diabetes.html (https://www4.stat.ncsu.edu/~boos/var.select/diabetes.html)

For more information see:

Bradley Efron, Trevor Hastie, Iain Johnstone and Robert Tibshirani (2004) "Least Angle Regression," Annals of Statistics (with discussion), 407-49 9.

(https://web.stanford.edu/~hastie/Papers/LARS/LeastAngle 2002.pdf)

In [6]: print(diabetes.feature_names) #print features name once again

```
['age', 'sex', 'bmi', 'bp', 's1', 's2', 's3', 's4', 's5', 's6']
```

```
In [7]: X = diabetes.data
         Y = diabetes.target
 In [8]: X.shape, Y.shape
 Out[8]: ((442, 10), (442,))
 In [9]: from sklearn.model_selection import train_test_split
         #import necessary library for data split
In [10]: X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2)
         #data split 20% goes to the test set
In [11]: X train.shape, Y train.shape
         #data dimension i.e 80% training data
Out[11]: ((353, 10), (353,))
In [13]: from sklearn import linear model
         #import library for model
         from sklearn.metrics import mean_squared_error, r2_score
          #import library for computing model
In [15]: model = linear model.LinearRegression()
         #Defines the regression model
In [17]: model.fit(X train, Y train)
         # Build actual training model
Out[17]: LinearRegression()
         In a Jupyter environment, please rerun this cell to show the HTML representation or trust the
         notebook.
         On GitHub, the HTML representation is unable to render, please try loading this page with
         nbviewer.org.
In [19]: Y pred = model.predict(X test)
         # Apply trained model to make prediction (on test set)
```

```
In [21]: # **Print model performance**
         print('Coefficients:', model.coef )
         #coeffiecient is stored in model.coef
         print('Intercept:', model.intercept_)
         #intercept stored in model.intercept
         print('Mean squared error (MSE): %.2f'
                % mean squared error(Y test, Y pred))
         #mean square stored in mean squared error, parameter passed
         #Y test is the actual value, Y pred is the predicted value.
         print('Coefficient of determination (R^2): %.2f'
               % r2_score(Y_test, Y_pred))
         Coefficients: [ -14.26812489 -294.66323133
                                                     556.98765558
                                                                   284.78133515 -9
         73.00174526
           591.72330419 240.24164297 310.43219074
                                                     766.60619727
                                                                    81.88776627]
         Intercept: 152.50905379010226
         Mean squared error (MSE): 2829.91
         Coefficient of determination (R^2): 0.49
In [22]: #equation of linear regression model is y=-14.26812489(age) -294.66323133 (
         ## **String formatting**
         r2 score(Y test, Y pred)
Out[22]: 0.49021677588018786
In [23]: r2 score(Y test, Y pred).dtype
Out[23]: dtype('float64')
         '%.2f' %0.49021677588018786
In [24]:
Out[24]: '0.49'
In [25]: #Now make scatterplot
         # **Import library**
         import seaborn as sns
In [26]: Y test
         #look at data
Out[26]: array([202., 220., 199., 121., 109., 249., 252., 142., 252., 52., 179.,
                272., 308.,
                            48., 163., 268., 219., 83., 131., 70., 217., 206.,
                 49., 89.,
                             57., 90., 104., 142., 258., 150., 91., 181., 310.,
                       78., 171., 71., 75., 185., 142., 225., 200., 152., 212.,
                174.,
                101., 296., 87., 60., 236., 37., 102., 263., 44., 111., 202.,
                 53., 131., 121., 183., 189., 166., 168., 137., 197., 297., 244.,
                132., 128., 122., 74., 232., 116., 84., 42., 97., 144.,
                 70., 45., 59., 40., 265., 65., 72., 198., 206., 196., 115.,
                186.])
```

In [27]: Y_pred

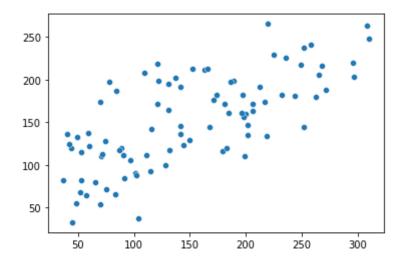
```
Out[27]: array([135.46158784, 265.41399678, 110.37504865, 171.90965885,
                208.05689615, 217.01382941, 145.06972288, 136.34426482,
                               68.3983279 , 116.68512593, 187.88530374,
                237.98372468,
                               55.15170817, 211.44458988, 216.47475712,
                263.56553173,
                               65.55012555, 164.50100619, 174.43876211,
                134.06499871,
                173.35929596, 171.58167701, 132.5712416 , 120.15964196,
                 63.93857131, 111.54262507, 36.83380647, 191.09431101,
                241.10242247, 129.08523114, 84.23977999, 171.90441026,
                            , 182.31523389, 198.01618097, 176.21876071,
                247.996688
                               72.08634669, 160.4785526, 145.5239425,
                229.2654962 , 160.22285002, 212.8514772 , 191.5221397 ,
                 89.96589371, 219.79560145, 117.4723048 , 122.71198415,
                226.2219284 , 82.17995131, 88.48149813, 179.61498644,
                120.15456018, 111.05451485, 146.31540844,
                                                           82.56015461,
                195.63156153, 218.99097859, 120.07273776, 199.164119
                213.32176127, 144.2632872 , 202.16360781, 182.29432575,
                203.18881599, 180.72013835, 117.07614913, 100.2325263 ,
                199.0939836 , 127.87902882, 182.43881278, 142.324929
                186.35730407, 124.36997395, 105.73926758, 123.46424494,
                115.2077598 , 54.34426743 , 32.79476599 , 137.94889864 ,
                135.78005204, 205.14189993, 79.90002847, 112.95278348,
                156.00285675, 163.6377009 , 161.3666516 ,
                197.01004328])
```

In [28]: sns.scatterplot(Y test, Y pred)

/Library/Frameworks/Python.framework/Versions/3.9/lib/python3.9/site-pack ages/seaborn/_decorators.py:36: FutureWarning: Pass the following variabl es as keyword args: x, y. From version 0.12, the only valid positional ar gument will be `data`, and passing other arguments without an explicit ke yword will result in an error or misinterpretation.

warnings.warn(

Out[28]: <AxesSubplot:>

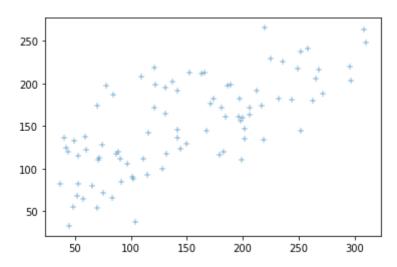


In [30]: sns.scatterplot(Y_test, Y_pred, marker="+")

/Library/Frameworks/Python.framework/Versions/3.9/lib/python3.9/site-pack ages/seaborn/_decorators.py:36: FutureWarning: Pass the following variabl es as keyword args: x, y. From version 0.12, the only valid positional ar gument will be `data`, and passing other arguments without an explicit ke yword will result in an error or misinterpretation.

warnings.warn(

Out[30]: <AxesSubplot:>

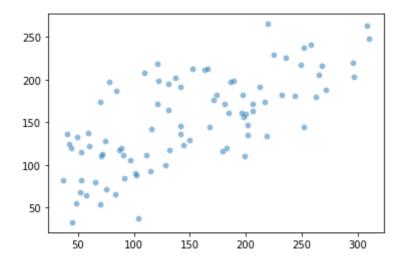


In [34]: sns.scatterplot(Y test, Y pred, alpha=0.5)

/Library/Frameworks/Python.framework/Versions/3.9/lib/python3.9/site-pack ages/seaborn/_decorators.py:36: FutureWarning: Pass the following variabl es as keyword args: x, y. From version 0.12, the only valid positional ar gument will be `data`, and passing other arguments without an explicit ke yword will result in an error or misinterpretation.

warnings.warn(

Out[34]: <AxesSubplot:>



In []: