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In [11]: import numpy as np # Importing NumPy library
import pandas as pd # Importing Pandas library
import matplotlib.pyplot as plt # Importing Matplotlib library's "pyplot" module
import seaborn as sns # Importing Seaborn library
import os
from sklearn.model_selection import train_test_split

In [12]: #importing SVM Model
from sklearn.svm import SVR
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In [13]: data = pd.read_csv("/Users/catherinebetancourt-lee/BMEN 415/Volumetric_features.csv")

data
```

Out[13]:

| | S.No | Left-Lateral-Ventricle | Left-Inf-Lat-Vent | Left-Cerebellum-White-Matter | Left-Cerebellum-Cortex | Left-Thalamus | Left-Caudate | Left-Putamen | Left-Pallidum | 3rd-Ventricle | ... | rh_supramarginal_thickness | rh_frontalpole_thickness | rh_temporalpole_thickness | rh_trans |
|--|------|------------------------|-------------------|------------------------------|------------------------|---------------|--------------|--------------|---------------|---------------|--------|----------------------------|--------------------------|---------------------------|----------|
| | 0 | 1 | 22916.9 | 982.7 | 15196.7 | 55796.4 | 6855.5 | 2956.4 | 4240.7 | 2223.9 | 2034.4 | ... | 2.408 | 2.629 | 3.519 |
| | 1 | 2 | 22953.2 | 984.5 | 15289.7 | 55778.6 | 6835.1 | 3064.2 | 4498.6 | 2354.1 | 1927.1 | ... | 2.417 | 2.640 | 3.488 |
| | 2 | 3 | 23320.4 | 1062.1 | 15382.1 | 55551.2 | 7566.0 | 3231.7 | 4456.2 | 1995.4 | 2064.7 | ... | 2.374 | 2.601 | 3.342 |
| | 3 | 4 | 24360.0 | 1000.5 | 14805.4 | 54041.8 | 8004.6 | 3137.3 | 4262.2 | 1983.4 | 2017.7 | ... | 2.366 | 2.639 | 3.361 |
| | 4 | 5 | 25769.4 | 1124.4 | 16331.1 | 54108.6 | 6677.4 | 2964.4 | 4204.6 | 2409.7 | 2251.8 | ... | 2.381 | 2.555 | 3.450 |
| | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| | 4221 | 4222 | 27065.6 | 532.4 | 12425.1 | 51042.9 | 6354.8 | 3822.6 | 4490.5 | 2019.4 | 1256.2 | ... | 2.505 | 2.666 | 2.915 |
| | 4222 | 4223 | 28408.8 | 912.7 | 14024.8 | 43103.5 | 6060.7 | 3114.2 | 3731.0 | 1937.4 | 1669.9 | ... | 2.385 | 3.008 | 3.572 |
| | 4223 | 4224 | 34467.9 | 1659.6 | 12744.5 | 54924.8 | 6256.7 | 3573.4 | 3526.6 | 2189.9 | 3063.1 | ... | 2.028 | 2.995 | 3.706 |
| | 4224 | 4225 | 31627.5 | 1334.4 | 15883.2 | 57148.2 | 6982.4 | 4475.8 | 4464.4 | 2317.8 | 3809.0 | ... | 2.491 | 2.865 | 3.456 |
| | 4225 | 4226 | 14879.4 | 704.2 | 11346.6 | 50468.5 | 6935.4 | 3258.5 | 3751.5 | 2226.5 | 1898.4 | ... | 2.474 | 3.150 | 3.691 |

4226 rows × 141 columns

```
In [14]: #Separating target variable and features
y = data['rh_supramarginal_thickness']
x = data.drop(['rh_supramarginal_thickness'], axis = 1)
x
```

Out[14]:

| | S.No | Left-Lateral-Ventricle | Left-Inf-Lat-Vent | Left-Cerebellum-White-Matter | Left-Cerebellum-Cortex | Left-Thalamus | Left-Caudate | Left-Putamen | Left-Pallidum | 3rd-Ventricle | ... | rh_superiortemporal_thickness | rh_frontalpole_thickness | rh_temporalpole_thickness | rh_tra |
|--|------|------------------------|-------------------|------------------------------|------------------------|---------------|--------------|--------------|---------------|---------------|--------|-------------------------------|--------------------------|---------------------------|--------|
| | 0 | 1 | 22916.9 | 982.7 | 15196.7 | 55796.4 | 6855.5 | 2956.4 | 4240.7 | 2223.9 | 2034.4 | ... | 2.648 | 2.629 | 3.519 |
| | 1 | 2 | 22953.2 | 984.5 | 15289.7 | 55778.6 | 6835.1 | 3064.2 | 4498.6 | 2354.1 | 1927.1 | ... | 2.660 | 2.640 | 3.488 |
| | 2 | 3 | 23320.4 | 1062.1 | 15382.1 | 55551.2 | 7566.0 | 3231.7 | 4456.2 | 1995.4 | 2064.7 | ... | 2.597 | 2.601 | 3.342 |
| | 3 | 4 | 24360.0 | 1000.5 | 14805.4 | 54041.8 | 8004.6 | 3137.3 | 4262.2 | 1983.4 | 2017.7 | ... | 2.604 | 2.639 | 3.361 |
| | 4 | 5 | 25769.4 | 1124.4 | 16331.1 | 54108.6 | 6677.4 | 2964.4 | 4204.6 | 2409.7 | 2251.8 | ... | 2.597 | 2.555 | 3.450 |
| | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| | 4221 | 4222 | 27065.6 | 532.4 | 12425.1 | 51042.9 | 6354.8 | 3822.6 | 4490.5 | 2019.4 | 1256.2 | ... | 2.457 | 2.666 | 2.915 |
| | 4222 | 4223 | 28408.8 | 912.7 | 14024.8 | 43103.5 | 6060.7 | 3114.2 | 3731.0 | 1937.4 | 1669.9 | ... | 2.497 | 3.008 | 3.572 |
| | 4223 | 4224 | 34467.9 | 1659.6 | 12744.5 | 54924.8 | 6256.7 | 3573.4 | 3526.6 | 2189.9 | 3063.1 | ... | 2.407 | 2.995 | 3.706 |
| | 4224 | 4225 | 31627.5 | 1334.4 | 15883.2 | 57148.2 | 6982.4 | 4475.8 | 4464.4 | 2317.8 | 3809.0 | ... | 2.700 | 2.865 | 3.456 |
| | 4225 | 4226 | 14879.4 | 704.2 | 11346.6 | 50468.5 | 6935.4 | 3258.5 | 3751.5 | 2226.5 | 1898.4 | ... | 2.746 | 3.150 | 3.691 |

4226 rows × 140 columns

```
In [15]: #training and testing datasets
X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.2, random_state=42)
```

```
In [16]: #Create and fit in a SVM regression model
svr_model = SVR(kernel='rbf', C=5.0)
svr_model.fit(X_train, y_train)
```

Out[16]: SVR(C=5.0)

```
In [17]: #predict on the dataset
y_pred = svr_model.predict(X_test)
```

```
In [18]: from sklearn.model_selection import cross_val_predict # For K-Fold Cross Validation
from sklearn.metrics import r2_score # For find accuracy with R2 Score
from sklearn.metrics import mean_squared_error # For MSE
from math import sqrt # For squareroot operation

y_pred_train = svr_model.predict(X_train)
y_pred_test = svr_model.predict(X_test)

accuracy_train = r2_score(y_train, y_pred_train)
print("Training R2 for Multiple Linear Regression Model: ", accuracy_train)

accuracy_test = r2_score(y_test, y_pred_test)
print("Testing R2 for Multiple Linear Regression Model: ", accuracy_test)

RMSE_train = sqrt(mean_squared_error(y_train, y_pred_train))
print("RMSE for Training Data: ", RMSE_train)

RMSE_test = sqrt(mean_squared_error(y_test, y_pred_test))
print("RMSE for Testing Data: ", RMSE_test)

Training R2 for Multiple Linear Regression Model:  0.6290578198420312
Testing R2 for Multiple Linear Regression Model:  0.6415445394915742
RMSE for Training Data:  0.11167245415649324
RMSE for Testing Data:  0.1160629552873467
```

```
In [19]: true_val = y_train
pred_val = y_pred_train
```

```
In [20]: plt.figure(figsize=(8,8))
plt.scatter(true_val, pred_val, c='crimson')
plt.yscale('log')
plt.xscale('log')

p1 = max(max(pred_val), max(true_val))
p2 = min(min(pred_val), min(true_val))
plt.plot([p1, p2], [p1, p2], 'b-')
plt.xlabel('True Values', fontsize=15)
plt.ylabel('Predicted Values', fontsize=15)
plt.title("SVM R: True Values vs Predicted Values")
plt.axis('equal')
plt.show()
```

