Importing Libraries

```
In [1]:
```

```
import random
import pandas as pd
import numpy as np
from scipy.integrate import quad
from sklearn.linear_model import LinearRegression
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_squared_error
from sklearn.metrics import mean_absolute_error
from sklearn.model_selection import cross_val_score, KFold
```

Question 1

```
In [2]:
```

```
np.random.seed(1)
x1_train = np.random.gamma(shape=1, scale=1, size=1000)
x2_train = np.random.gamma(shape=1, scale=1, size=1000)
x3_train = np.random.gamma(shape=1, scale=1, size=1000)
w_train = np.random.normal(loc=0, scale=2, size=1000)
y_train = (0.5*x1_train) + (3*x2_train) + (5*x3_train) + (5*x2_train*x3_train) + (2*x1_train*x2_train*x3_train) + w_train

np.random.seed(2)
x1_test = np.random.gamma(shape=1, scale=1, size=1000)
x2_test = np.random.gamma(shape=1, scale=1, size=1000)
x3_test = np.random.gamma(shape=1, scale=1, size=1000)
w_test = np.random.normal(loc=0, scale=2, size=1000)
y_test = (0.5*x1_test) + (3*x2_test) + (5*x3_test) + (5*x2_test*x3_test) + (2*x1_test*x2_test*x3_test) + w_test
```

In [3]:

```
train_df = pd.DataFrame({"x1": x1_train, "x2": x2_train, "x3": x3_train, "y": y_train})
x_train = train_df[["x1", "x2", "x3"]]
y_train = train_df["y"]

test_df = pd.DataFrame({"x1": x1_test, "x2": x2_test, "x3": x3_test, "y": y_test})
x_test = test_df[["x1", "x2", "x3"]]
y_test = test_df["y"]
```

In [4]:

```
lr_model = LinearRegression()
lr_model.fit(x_train, y_train)
y_lr_pred = lr_model.predict(x_test)
lr_mse = mean_squared_error(y_test, y_lr_pred)
print("Linear Regression MSE:", lr_mse)
print("Linear Regression Coefficients:", lr_model.coef_)
```

Linear Regression MSE: 74.56753771040981 Linear Regression Coefficients: [2.19957962 8.38793456 12.19228899]

In [5]:

```
rf_model = RandomForestRegressor(n_estimators = 500)
rf_model.fit(x_train, y_train)
y_rf_pred = rf_model.predict(x_test)
rf_mse = mean_squared_error(y_test, y_rf_pred)
print("Random Forest Regression MSE:", rf_mse)
```

Random Forest Regression MSE: 27.467311656382822

Question 2

```
In [6]:
def function(x):
   return 3 + (x**2) - (2*np.sin(x))
a, b = 1, 8
true value, error = quad(function, a, b)
print("True Value:", true value, u"\u00B1", error)
True Value: 189.9617286539798 ± 2.1089988494924473e-12
In [15]:
samples = np.random.uniform(low=a, high=b, size=10000)
function samples = [function(sample) for sample in samples]
estimated value = np.mean(function samples) * (b-a)
print("Estimated Value: ", estimated value)
Estimated Value: 190.97185353936558
In [8]:
sd = 0
sd = sum([(result - estimated value)**2 for result in samples])
sd = (sd/(10000 - 1))**0.5
lower = estimated value - 1.96*sd/(10000**0.5)
upper = estimated value + 1.96*sd/(10000**0.5)
confidence interval = [lower, upper]
print("Confidence Interval: ", confidence interval)
Confidence Interval: [184.94448784143083, 192.16113042048005]
Question 6
In [9]:
df = pd.DataFrame({"x1": [0, 1, 2, 3, 4], "y": [1, 2, 3, 2, 1]})
In [10]:
lr model 1 = LinearRegression()
lr model 1.fit(df[["x1"]], df["y"])
y pred 1 = lr model 1.predict(df[["x1"]])
print("model 1 intercept:", lr_model_1.intercept_)
lr_model_2 = LinearRegression(fit intercept=False)
lr_model_2.fit(df[["x1"]], df["y"])
y pred 2 = lr model 2.predict(df[["x1"]])
print("model 2 coefficients: ", lr_model_2.coef_)
model 1 intercept: 1.8
model 2 coefficients: [0.6]
In [11]:
squared_error_1 = ((y_pred_1 - df["y"])**2).mean()
squared error 2 = ((y \text{ pred } 2 - \text{df}["y"]) **2).mean()
print("Average Squared Error for model 1:", squared error 1, "model 2:", squared error 2)
absolute error 1 = (np.abs(y pred 1 - df["y"])).mean()
absolute error 2 = (np.abs(y pred 2 - df["y"])).mean()
print("Average Absolute Error for model_1:", absolute_error_1, "model_2:", absolute_error
_2)
```

```
1_15_model_1 = (np.abs(y_pred_1 - df["y"])**1.5).mean()
1_15_model_2 = (np.abs(y_pred_2 - df["y"])**1.5).mean()
print("L1.5 loss for model_1:", 1_15_model_1, "model_2:", 1_15_model_2)
```

Average Squared Error for model_1: 0.560000000000000 model_2: 1.64

Average Absolute Error for model_1: 0.64000000000001 model_2: 1.1600000000001

L1.5 loss for model 1: 0.5849006163624495 model 2: 1.36348016266711

Question 7

```
In [12]:
```

```
#Reading Data
df = pd.read csv(r"C:\Users\jaske\Desktop\studies\OneDrive\Statistical Methods for Data S
cience\Assignments\Assignment 1\data.csv")
df["x2"] = df["x2"].astype(int)
df["x2"].unique()
Out[12]:
array([1, 2, 3])
In [13]:
#Ordinal Data
ordinal df = df.copy()
ordinal model = LinearRegression()
kf = KFold(n splits=10, shuffle=True, random state=42)
mse list = -cross val score(ordinal model, ordinal df.drop(columns=["y"]), ordinal df["y"
], cv=kf, scoring="neg mean squared error")
mse = (mse list).mean()
print("Ordered mean squared error:", mse)
```

Ordered mean squared error: 1.2685717076212815

In [14]:

```
#Non Ordinal Data
non_ordinal_df = pd.get_dummies(df, columns=["x2"])
non_ordinal_model = LinearRegression()
kf = KFold(n_splits=10, shuffle=True, random_state=42)
mse_list = -cross_val_score(non_ordinal_model, non_ordinal_df.drop(columns=["y"]), non_ordinal_df["y"], cv=kf, scoring="neg_mean_squared_error")
mse = np.mean(mse_list)
print("UnOrdered mean squared error:", mse)
```

UnOrdered mean squared error: 3.757360926169872e-30

References

Monte Carlo integration in Python. (2022, March 10). GeeksforGeeks. https://www.geeksforgeeks.org/monte-carlo-integration-in-python/ An Easy Guide to K-Fold Cross-Validation. (2020. November 4). Statology. https://www.statology.org/k-fold-cross-validation/