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## M5-L2 Problem 1

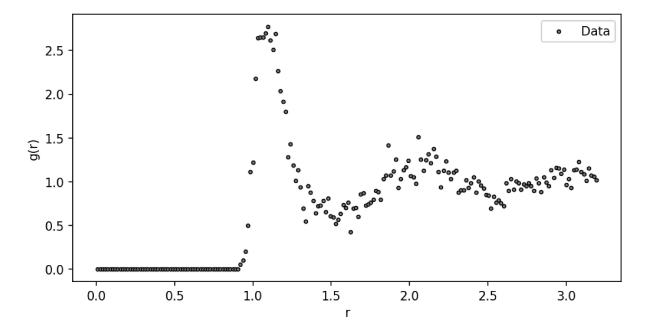
256 particles of liquid argon are simulated at 100K. A radial distribution function g(r) describes the density of particles a distance of r from each particle in the system. When an g(r) is computed in a simulation, it is done by creating a histogram of particle distances for a single simulation frame, resulting in a noisy function that is most often averaged over several frames.

Given g(r) vs. r data for a single frame, you will train a decision tree regressor to represent the underlying function.

First, run the cell below to load the data, etc.:

```
In [1]: import numpy as np
 import matplotlib.pyplot as plt
from sklearn.tree import DecisionTreeRegressor,plot tree
 r = np.array([0.008, 0.024, 0.04, 0.056, 0.072, 0.088, 0.104, 0.12, 0.136, 0.152, 0.168, 0.184]
 def plot(r, g, dt = None):
    if dt is not None:
        plt.figure(figsize=(12,3),dpi=150)
        plt.subplot(121)
        rs = np.linspace(0,4,1000)
        gs = dt.predict(rs.reshape(-1,1))
        plt.plot(rs,gs,color="red",label="Regression Tree",alpha=0.7)
    else:
        plt.figure(figsize=(8,4),dpi=150)
    plt.scatter(r,g,s=8,c="gray", label="Data", edgecolors="black",linewidths=.8)
    plt.legend(loc="upper right")
    plt.xlabel("r")
    plt.ylabel("g(r)")
    if dt is not None:
        plt.subplot(122)
        plot tree(dt)
        plt.title(f"Tree max. depth: {dt.max_depth}",y=-.2)
    plt.show()
 plot(r,g)
```

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## Training regression trees

For input r and output g, train a DecisionTreeRegressor() to perform the regression with max\_depth values of 1, 2, 6, 10.

Complete the code below, which will plot your decision tree results and visualize the tree. Name each decision tree within the loop dt.

Note: you may need to resize the input r as r.reshape(-1,1) before passing it as input into the fitting function.

```
In [4]: for max_depth in [1, 2, 6, 10]:
       # YOUR CODE GOES HERE
       # Create and fit `dt`
       dt = DecisionTreeRegressor(max_depth=max_depth)
       r = r.reshape(-1,1)
       dt.fit(r,g)
       dt.predict(r)
       plot(r,g,dt)
                                    Regression Tree
                                                                        X[0] \le 0.976
   2.5
                                    Data
                                                                     squared error = 0.41
                                                                        samples = 200
   2.0
                                                                        value = 0.772
 £ 1.5
   1.0
                                                          squared_error = 0.005
                                                                               squared error = 0.225
                                                              samples = 61
                                                                                   samples = 139
   0.5
                                                              value = 0.014
                                                                                   value = 1.104
   0.0
       0.0
                                              4.0
                                                                       Tree max. depth: 1
```

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