import numpy as np In [1]: import matplotlib.pyplot as plt import os import sys import pandas as pd from mpltools import annotation # Import the .txt-file generated by 'autotest.sh' In [2]: txtname = "tree\_test.txt" data = np.loadtxt(txtname) data.shape Out[2]: (24, 7) # Map boolean input of quadrupoles to strings. In [3]: def convert(x): **if** x == 0: return "Mono" else: return "Quad" = "N" N In [4]: theta = "\$\\theta\_c\$" = "Multipole expansion" quad tree = "Tree code t(s)" dirsum = "Direct t(s)' relerr = "Relative error" intact = "Avg interactions"  $data_ls = []$ for i in range(len(data)): data\_ls.append({ # Number of particles int(data[i,0]), N: theta: data[i,1], # Opening angle threshold quad: convert(int(data[i,2])), # Using quadrupoles (bool) # Execution time tree code (sec) data[i,3], tree: dirsum: data[i,4], # Execution time direct summation (sec) relerr: data[i,5], # Mean relative force error # Mean number of particle-node interactions intact: data[i,6] }) df = pd.DataFrame(data\_ls) # Create dataframe # Retain an unaltered copy of the dataframe. df\_copy = df.copy() df\_copy Multipole expansion Tree code t(s) Direct t(s) Relative error Avg interactions Out[4]: 5000 0.2 Mono 0.604802 1.78396 0.000279 1735.480 5000 0.2 1.066630 1.77675 0.000022 1735.480 Quad 1 5000 0.4 Mono 0.163367 1.77674 0.002177 501.845 2 5000 0.4 0.297304 1.77734 0.000384 501.845 3 Quad 5000 0.8 Mono 0.034382 1.78090 0.014926 109.968 5 5000 0.8 Quad 0.064255 1.78542 0.006886 109.968 10000 0.2 2.109320 7.51979 0.000245 2403.890 Mono **7** 10000 0.2 2403.890 Quad 3.418570 8.09434 0.000020 621.909 8 10000 0.4 Mono 0.485566 8.28296 0.001814 10000 0.4 0.951817 8.40635 0.000332 621.909 Quad **10** 10000 0.8 8.13719 Mono 0.094773 0.012776 127.757 **11** 10000 0.8 0.184383 8.12671 0.006108 127.757 Quad 31.64360 **12** 20000 0.2 Mono 5.444850 0.000225 3120.220 **13** 20000 0.2 9.589140 31.52720 0.000019 3120.220 Quad 0.001527 **14** 20000 0.4 Mono 1.129170 29.20930 737.402 15 20000 0.4 1.927480 29.28240 0.000294 737.402 Quad 16 20000 0.8 0.225642 30.22290 0.010606 144.197 Mono 29 02400 20000 0.8 17 0.420236 0.005340 144.197 Quad 18 40000 0.2 16.810400 116.58500 0.000195 3976.430 Mono 19 40000 0.2 3976.430 Quad 25.686300 117.13400 0.000016 20 40000 0.4 Mono 3.272910 117.49900 0.001299 866.969 **21** 40000 0.4 4.933170 119.81100 0.000248 Quad 866.969 0.008915 22 40000 0.8 0.587553 117.86300 161.829 Mono 40000 0.8 0.888337 118.21700 0.004823 161.829 23 Quad # Group by N, theta and quadrupole moment In [5]: df = df.groupby([N, theta, quad]).first().unstack() Out[5]: Tree code t(s) Direct t(s) Relative error Avg interactions Multipole expansion Mono Mono Mono Quad Quad Quad Mono Quad N 5000 0.2 0.604802 1.066630 1.78396 1.77675 0.000279 0.000022 1735.480 1735.480 0.163367 0.297304 1.77674 1.77734 0.002177 0.000384 501.845 501.845 0.4 0.8 0.034382 0.064255 1.78090 1.78542 0.014926 0.006886 109.968 109.968 10000 0.2 2.109320 3.418570 7.51979 8.09434 0.000245 0.000020 2403.890 2403.890 0.4 0.485566 0.951817 8.28296 8.40635 0.001814 0.000332 621.909 621.909 0.8 0.094773 0.184383 8.13719 8.12671 0.012776 0.006108 127.757 127.757 20000 0.2 5.444850 9.589140 31.64360 31.52720 0.000225 0.000019 3120.220 3120.220 1.129170 1.927480 29.20930 29.28240 0.001527 0.000294 737.402 737.402 0.4 0.8 0.225642 0.420236 30.22290 29.02400 0.010606 0.005340 144.197 144.197 40000 **0.2** 16.810400 25.686300 116.58500 117.13400 0.000195 0.000016 3976.430 3976.430 0.4 3.272910 4.933170 117.49900 119.81100 0.001299 0.000248 866.969 866,969 0.8 0.587553 0.888337 117.86300 118.21700 0.008915 0.004823 161.829 161.829 Need to fix the direct summation column, misleading to have 'Mono' and 'Quad' labels there. We also calculate the mean execution time of the direct summation method as the average of the two elements at each row. # Calculate the mean of direct summation columns, insert back into the dataframe. In [6]: df[('Direct t(s)','-')] = df.iloc[:,2:4].mean(axis=1)# Drop the two old direct summation columns. df = df.drop(df.columns[2:4], axis=1) # Rearange order of columns, such that the direct summation (avg) # column retakes its old position. col\_list = list(df.columns) col\_list.insert(2,col\_list[-1]) col\_list.pop(-1) df = df[col\_list] In [7]: # Specify formatting of floating point numbers for each column. format\_1 = "{0:.3f}".format # Timings, tree code and direct summation format\_2 = "{0:.2e}".format # Relative errors format\_3 = "{0:.1f}".format # Average interactions df[list(df.columns)[0:3]]= df[list(df.columns)[0:3]].applymap(format\_1) df[list(df.columns)[3:5]]= df[list(df.columns)[3:5]].applymap(format\_2) df[list(df.columns)[5:]] = df[list(df.columns)[5:]].applymap(format\_3) df Tree code t(s) Direct t(s) Out[7]: Relative error Avg interactions Multipole expansion Mono Quad Mono Quad Mono Quad  $\theta_c$ Ν 5000 0.605 1.067 1.780 2.79e-04 2.21e-05 1735.5 1735.5 0.2 0.163 0.297 1.777 2.18e-03 3.84e-04 501.8 501.8 0.4 0.034 0.064 1.783 1.49e-02 6.89e-03 110.0 110.0 0.8 10000 7.807 2.45e-04 2.01e-05 0.2 2.109 3.419 2403.9 2403.9 0.486 0.952 8.345 1.81e-03 3.32e-04 621.9 621.9 0.095 0.184 8.132 1.28e-02 6.11e-03 127.8 127.8 0.8 20000 3120.2 31.585 2.25e-04 1.90e-05 3120.2 5.445 9.589 1.129 1.927 29.246 1.53e-03 2.94e-04 737.4 737.4 0.4 0.226 0.420 29.623 1.06e-02 5.34e-03 144.2 0.8 144.2 40000 16.810 25.686 116.859 1.95e-04 1.59e-05 3976.4 0.2 3976.4 4.933 118.655 1.30e-03 2.48e-04 3.273 867.0 867.0 0.8 0.588 0.888 118.040 8.91e-03 4.82e-03 161.8 161.8 In [8]: # Export table to latex print(df.to\_latex()) \begin{tabular}{lllllllll} \toprule &  $\{\}$  & \multicolumn $\{2\}\{1\}\{Tree\ code\ t(s)\}$  & Direct t(s) & \multicolumn $\{2\}\{1\}\{Relative\ error\}$  & \multicolumn $\{2\}\{Relative\ error\}$  & \multicolumn $\{2\}\{Relative\$ olumn{2}{l}{Avg interactions} \\ & Multipole expansion & Mono & Quad & - & Mono & Quad & Quad \\ Mono & N &  $\$  \\$\textbackslash theta\\_c\\$ & & & & // \midrule 1.780 & 5000 & 0.2 & 0.605 & 1.067 & 2.79e-04 & 2.21e-05 & 1735.5 & 173 5.5 \\ 1.777 & & 0.4 & 0.163 & 0.297 &2.18e-03 & 3.84e-04 & 501.8 & 50 1.8 \\ 0.034 & 1.783 & 110.0 & & 0.8 & 0.064 & 1.49e-02 & 6.89e-03 & 11 0.0 \\ 10000 & 0.2 & 2.109 & 3.419 & 7.807 & 2.45e-04 & 2.01e-05 & 2403.9 & 240 3.9 \\ 8.345 & & 0.4 & 0.486 & 0.952 & 1.81e-03 & 3.32e-04 & 621.9 & 62 1.9 \\ & 0.8 & 0.095 & 0.184 & 8.132 & 1.28e-02 & 6.11e-03 & 127.8 & 12 7.8 \\ 20000 & 0.2 & 5.445 & 9.589 & 31.585 & 2.25e-04 & 1.90e-05 & 3120.2 & 312 0.2 \\ & 0.4 & 1.129 & 1.927 & 29.246 & 1.53e-03 & 2.94e-04 & 737.4 & 73 7.4 \\ & 0.8 & 0.226 & 0.420 & 29.623 & 1.06e-02 & 5.34e-03 & 144.2 & 14 4.2 \\ 40000 & 0.2 & 16.810 & 25.686 & 116.859 & 1.95e-04 & 1.59e-05 & 3976.4 & 397 6.4 \\ & 0.4 & 3.273 & 4.933 & 118.655 & 1.30e-03 & 2.48e-04 & 867.0 & 86 7.0 \\ & 0.8 & 0.588 & 0.888 & 118.040 & 8.91e-03 & 4.82e-03 & 161.8 & 16 1.8 \\ **\bottomrule** \end{tabular} In [9]: # Mask specific threshold mask\_theta = df\_copy[theta] == 0.4 # Mask mono-/quadrupoles mask\_mono = df\_copy[quad] == 'Mono' mask\_quad = mask\_mono == False # Extract timing vectors. df\_copy[mask\_theta & mask\_mono] = df\_copy[mask\_theta & mask\_mono][N].to\_numpy() t\_tree\_mono = df\_copy[mask\_theta & mask\_mono][tree].to\_numpy() t\_dirsum\_m = df\_copy[mask\_theta & mask\_mono][dirsum].to\_numpy() t\_tree\_quad = df\_copy[mask\_theta & mask\_quad][tree].to\_numpy()  $t_dirsum_q = df_copy[mask_theta & mask_quad][dirsum].to_numpy()$ # Get average of direct summation vectors.  $t_dirsum = (t_dirsum_m + t_dirsum_q)/2$ # Simple linear regression in loglog. coef\_mono = np.polyfit(np.log(N\_vec), np.log(t\_tree\_mono), 1) coef\_quad = np.polyfit(np.log(N\_vec), np.log(t\_tree\_quad), 1) coef\_dirsum = np.polyfit(np.log(N\_vec), np.log(t\_dirsum), 1)  $N_{lin} = np.linspace(4500, 45000)$ # Positions of slope markers  $mono\_slope\_pos = (25000, 1.4)$  $quad\_slope\_pos = (32000, 4.5)$  $dirsum_slope_pos = (25000, 40)$ # Plotting plt.figure() # Obtained data points plt.loglog(N\_vec, t\_tree\_mono, 'o', label='Monopole') plt.loglog(N\_vec, t\_tree\_quad, 'o', label='Quadrupole') plt.loglog(N\_vec, t\_dirsum, 'o', label='Direct summation') # Linear regressions plt.loglog(N\_lin, N\_lin\*\*coef\_mono[0]\*np.exp(coef\_mono[1]),'--', color='tab:blue') plt.loglog(N\_lin, N\_lin\*\*coef\_quad[0]\*np.exp(coef\_quad[1]),'--', color='tab:orange') plt.loglog(N\_lin, N\_lin\*\*coef\_dirsum[0]\*np.exp(coef\_dirsum[1]),'--', color='tab:green') # Slope indicators annotation.slope\_marker(mono\_slope\_pos, np.around(coef\_mono[0], 3), poly\_kwargs={'facecolor': 'tab:blue', 'alpha': 0.8}) annotation.slope\_marker(quad\_slope\_pos, np.around(coef\_quad[0], 3), invert=True, poly\_kwargs={'facecolor': 'tab:orange', 'alpha': 0.8}) annotation.slope\_marker(dirsum\_slope\_pos, np.around(coef\_dirsum[0], 3), poly\_kwargs={'facecolor': 'tab:green', 'alpha': 0.8}) # Plot settings plt.ylabel('Execution time \$t(s)\$') plt.xlabel('Number of particles \$N\$') plt.title('Execution time t(s) as a function of number of \nparticles N, for opening angle the shold t(s)plt.legend() # Manual xticks  $x_list = ['\$\mathdefault{5 \cdot 10^{3}}$', '$\mathdefault{10^{4}}$',$ plt.xticks(N\_vec, x\_list) # Save figure plt.savefig('loglog\_TofN.pdf') Execution time t(s) as a function of number of particles N, for opening angle the shold  $\theta_c = 0.4$ Monopole  $10^{2}$ Ouadrupole Direct summation Execution time t(s) 10° 10°  $5 \cdot 10^{3}$  $10^{4}$  $2 \cdot 10^{4}$  $4 \cdot 10^{4}$ Number of particles N Extrapolate execution time for all methods, for a system with  $N=10^{10}$  particles. # Using the previous linear regression for the extrapolation. In [10]:  $N_{extr} = 10**10$ coef\_list = [coef\_mono, coef\_quad, coef\_dirsum]  $t_{estimate} = np.zeros(3)$ for i in range(3): t\_estimate[i] = N\_extr\*\*coef\_list[i][0]\*np.exp(coef\_list[i][1]) # Print results  $print(f'Tree\ code,\ monopoles:\ Est.\ time\ for\ N = \{N_extr:.2e\},\ t = \{t_estimate[0]:.2\}\ sec\ or\ \{t_estimate[0]:.2\}$  $print(f'Tree\ code,\ quadrupoles:\ Est.\ time\ for\ N = \{N_extr:.2e\},\ t = \{t_estimate[1]:.2\}\ sec\ or\ \{t_estimate[1]:.2\}$ print(f'Direct summation: Est. time for  $N = \{N_{extr}: 2e\}$ ,  $t = \{t_{estimate}[2]: 2\}$  sec or  $\{t_{estimate}[2]\}$ Tree code, monopoles: Est. time for N = 1.00e+10, t = 1.5e+08 sec or 4.7 years. Tree code, quadrupoles: Est. time for N = 1.00e+10, t = 6.5e+07 sec or 2.1 years. Est. time for N = 1.00e+10, t = 7.5e+12 sec or 2.4e+05 years.

Direct summation: