1.5 -1.5-1.0-0.50.0 0.5 1.0 -2.02.0 Х Effective Diffusion Coefficient 0.0104 0.0102 S 0.0100 0.0098 0.0096 -2.0-1.5-1.0-0.51.0 1.5 2.0 0.0 0.5 Х In [2]: a = -2.0 # location of reflecting boundary b = 0.8 # location of absorbing boundary h = 0.01N = int((b-a)/h+1) $x_{arr} = np.linspace(a, b, N)$ Steady State Distribution - Numerical solve ODE with injecting boundary condition In [3]: def kappa(x): return D(x)*np.exp(-beta_U(x)) $x_{minus_half} = x_{arr} - h/2.0$ $x_plus_half = x_arr + h/2.0$ u_arr = np.zeros(x_arr.size) A = np.zeros((N-1, N-1))np.fill_diagonal(A, -(kappa(x_minus_half)+kappa(x_plus_half))) np.fill_diagonal(A[1:,], kappa(x_minus_half[1:-1])) # subdiagonal np.fill_diagonal(A[:, 1:], kappa(x_plus_half[:-2])) # superdiagonal A[0, 1] += kappa(x minus half[0])f_vect = np.zeros(N-1) $f_{vect[0]} = 1$ $u_arr[0:-1] = spsolve(A, f_vect)$ Pst_arr = u_arr*np.exp(-beta_U(x_arr)) Pst arr /= (np.sum(Pst arr)*h) Steady-State Flux and Probability Distribution Function - Numerically Nest integrate the expression (referred to be exact) In [4]: # Define the inner function to integrate as a function of y def inner integrand(y): return 1.0/D(y)*np.exp(beta_U(y)) # Define the inner integral as a function of x def inner_integral(x): $y_{lower} = x$ $y_{upper} = b$ result, error = quad(inner_integrand, y_lower, y_upper) return result # Define the outer integral x lower = a $x_upper = b$ # Define the outer function to integrate (also as a function of x) **def** outer integrand(x): return np.exp(-beta_U(x))*inner_integral(x) # Perform the outer integration invert_st_flux, error = quad(outer_integrand, x_lower, x_upper) st flux = 1.0/invert st flux In [5]: def st_P_func(x): def integrand(y): return 1.0/D(y)*np.exp(beta U(y)) # Perform the integration $y_{lower} = x$ $y_{upper} = b$ result, error = quad(integrand, y_lower, y_upper) result *= st_flux*np.exp(-beta_U(x)) return result st_P_arr = np.zeros(x_arr.size) for i in np.arange(x_arr.size): st_P_arr[i] = st_P_func(x_arr[i]) Steady State Distribution Plots In [6]: plt.plot(x arr, Pst arr, label='num ODE') plt.plot(x_arr, st_P_arr, '--', label='num_integration') plt.title('Steady-state Probability Distribution') plt.grid() plt.legend() print(Pst_arr[:5], st_P_arr[:5]) $\hbox{\tt [0.16254088 \ 0.16291158 \ 0.16336963 \ 0.16391799 \ 0.16455972] } \hbox{\tt [0.16267392 \ 0.16304497 \ 0.16350344 \ 0.16405229 \ 0.16469136963] } \\$ 458] Steady-state Probability Distribution num ODE num_integration 1.0 0.8 0.6 0.4 0.2 0.0 -2.0-1.5-1.0-0.50.0 0.5 In [7]: plt.plot(x arr, -np.log(Pst arr), label='num ODE') plt.plot(x_arr, -np.log(st_P_arr), '--', label='num_integration') plt.xlabel('x') plt.ylabel("\$-ln[Pst(x)]\$") plt.grid() plt.legend() <matplotlib.legend.Legend at 0x781098b2d5d0> num ODE num_integration 5 4 -ln[Pst(x)]2 1 0 -1.50.0 0.5 -2.0-1.0-0.5Х Numerically Nest integrate for MFPT - referred to be exact In [8]: # Regura's method, overlap the starting point and reflecting boundary # Define the inner function to integrate as a function of z def inner integrand(z): return np.exp(-beta U(z)) # Define the inner integral as a function of y def inner_integral(y): $z_{lower} = a$ $z_{upper} = y$ result, error = quad(inner_integrand, z_lower, z_upper) return result # Define the outer integral $y_lower = x0$ $y_{upper} = b$ # Define the outer function to integrate (also as a function of y) def outer integrand(y): return np.exp(beta_U(y))*inner_integral(y)/D(y) # Perform the outer integration (from x0 to b) result, error = quad(outer_integrand, y_lower, y_upper) mfpt_arr = np.zeros(x_arr.size) for i in np.arange(x_arr.size): mfpt_arr[i], _ = quad(outer_integrand, y_lower, x_arr[i]) plt.xlabel('x') plt.ylabel(r"\$\tau {MFPT} (x)\$") plt.plot(x_arr, mfpt_arr) plt.title('MFPT vs. Position of Absorbing Boundary') plt.grid() MFPT vs. Position of Absorbing Boundary 400 300 200 100 0 -1.5-1.00.0 -2.0-0.50.5 Х Reconstruct Free Energy from Numerical Solution of Pst_arr and Numerical Integration of MFPT In [9]: interp Pst func = interpld(x arr, Pst arr, kind='cubic', fill value="extrapolate") # Except for the absorbing boundary Pst(b) = 0, and also avoid D(x)=0integral_Pst_arr = np.zeros(N-1) $Bx_arr = np.zeros(N-1)$ for i in range(N-1): integral_Pst_arr[i], _ = quad(interp_Pst_func, x_arr[i], x_arr[-1]) $Bx_arr[i] = -1.0/Pst_arr[i]*(integral_Pst_arr[i]-(mfpt_arr[-1]-mfpt_arr[i])/(mfpt_arr[-1]))$ interp_invertBx_func = interpld(x_arr[:-1], 1.0/Bx_arr, kind='cubic', fill_value="extrapolate") integral_invertBx_arr = np.zeros(N-1) beta Urec1 arr = np.zeros(N-1) for i in range(x_arr.size-1): # Here x0 is b arr[0] integral_invertBx_arr[i], _ = quad(interp_invertBx_func, x_arr[0], x_arr[i]) beta_Urec1_arr[i] = beta_U(x_arr[0])+np.log(Bx_arr[i]/Bx_arr[0])-integral_invertBx_arr[i] print() plt.plot(x_arr[:-1], beta_Urec1_arr, label="numerical reconstruct") plt.plot(x arr, beta U(x arr), ":", label="exact") # Plot formatting plt.xlabel('x') plt.ylabel(r'\$ \beta U(x) \$') plt.title('free energy reconstruction') plt.legend() plt.grid() 1.0346551525309143e-07 free energy reconstruction numerical reconstruct -0.5····· exact -1.0-1.5-2.0-2.5-3.0-3.5-2.0-1.5-1.0-0.50.0 0.5 Х Pst and MFPT calculated by Transfer Matrix with Recycling Boundary Condition In [10]: from transfer_matrix_reptile import TransferMatrix_ReInAb ria_trans = TransferMatrix_ReInAb(h, x_arr, beta_U, 0) In [11]: # plt.plot(x_arr[:-1], ria_trans.steady_state, label="RIA") # plt.plot(x_arr, st_P_arr, '--', label="exact") plt.plot(x_arr[:-1], -np.log(ria_trans.steady_state), label="RIA") plt.plot(x_arr, -np.log(st_P_arr), '--', label="exact") print(ria_trans.eig6_w) plt.xlabel('x') plt.ylabel("\$-ln[Pst(x)]\$") plt.legend() plt.grid() +0.j 0.99963059-0.00017818j 0.99963059+0.00017818j 0.99891501-0.00038286j 0.99891501+0.00038286j 0.99768309-0.00056821j] RIA 6 exact 5 4 -ln[Pst(x)]3 2 1 0 -2.0-1.5-1.0-0.50.0 0.5 Х In [12]: ria_P = ria_trans.trans_mat.T vec_pi = ria_trans.eig6_v[:,0].real/np.sum(ria_trans.eig6_v[:,0].real) epsilon = np.ones(N-1) $mat_I = np.eye(N-1)$ $mat_E = np.ones((N-1, N-1))$ mat_Z = np.linalg.inv(mat_I-ria_P+np.outer(epsilon, vec_pi)) vec_Zdg = np.diag(mat_Z) mat_Zdg = np.diag(vec_Zdg) mat_Mdg = np.diag(1/vec_pi) mat_M = (mat_I - mat_Z + mat_E @ mat_Zdg) @ mat_Mdg mat_Mbar = mat_M - mat_Mdg $delt_t = h^{**2}/(2*D0)$ plt.xlabel('x') plt.ylabel(r"\$\tau_{MFPT} (x)\$") plt.plot(x_arr, mfpt_arr, '--') plt.plot(x_arr[:-1], delt_t*mat_Mbar[0]) plt.title('MFPT vs. Position of Absorbing Boundary') plt.grid() MFPT vs. Position of Absorbing Boundary 400 300 200 100 0.0 0.5 -2.0-1.5-1.0-0.5Extract Steady State Distribution and MFPT from Simulation In [13]: from mfpt_Pst_RW_simulate import simulate_ReAb, simulate_ReAb_accelerate num particles = 600init_position_arr = a*np.ones(num_particles, dtype=float) ht = hx**2/(2*D0) $n_arr = np.arange(a, b+h, h)$ n_arr = np.round(n_arr, decimals=5) In [14]: # count n, ti n = simulate ReAb accelerate(init position arr=init position arr, beta U=beta U, n arr=n arr, a count_n, ti_n = simulate_ReAb(init_position_arr=init_position_arr, beta_U=beta_U, n_arr=n_arr, a=a, b=b, hx=h In [15]: Pst n = count n/(h*np.sum(count n))plt.plot(n_arr[:-1], -np.log(Pst_n[:-1]), label="Metro walkers (ReAb)") # $plt.plot(x_arr[:-1], 1.0/(h*np.sum(ria_trans.eig6_v[:, 1]))*ria_trans.eig6_v[:, 1], label="RIA")$ plt.plot(x_arr[:-1], -np.log(ria_trans.steady_state), label="RIA") plt.plot(x_arr[:-1], -np.log(st_P_arr[:-1]), '--', label="exact") # Plot formatting plt.xlabel('n') plt.ylabel("\$-ln[Pst(n)]\$") plt.title('steady state distribution') plt.legend() plt.grid() steady state distribution Metro walkers (ReAb) 6 RIA exact 5 4 -ln[Pst(n)]2 1 0 -2.0-1.5-1.0-0.50.0 0.5 n In [16]: mfpt_simu_arr = np.mean(ti_n, axis=0) plt.plot(n_arr, mfpt_simu_arr, label="Metro walkers (ReAb)") plt.plot(x_arr, mfpt_arr, '--', label="exact") # Plot formatting plt.xlabel('n') plt.ylabel(r'\$ \tau_{MFPT} (n)\$') plt.title('MFPT vs. Location of Absorbing Boundary') plt.legend() plt.grid() MFPT vs. Location of Absorbing Boundary Metro walkers (ReAb) exact 400 300 $\tau_{MFPT}(n)$ 200 100 0 -2.0-1.5-1.0-0.50.0 0.5 n Reconstruct Free Energy Using data [mfpt (mfpt_simu_arr), Pst (Pst_n)] extracted from Simulation (1) Exact Steady State Distribution with simulated MFPT In [17]: # Except for the absorbing boundary Pst(b) = 0, D(x)=0, also avoid at reflecting boundary A(x)=0, B(x)=0, $Bx_arr = np.zeros(N-2)$ integral_Pst_arr = np.zeros(N-2) for i in range(N-2): integral Pst arr[i], = quad(st P func, x arr[1+i], x arr[-1]) $Bx_arr[i] = -1.0/st_P_func(x_arr[1+i])*(integral_Pst_arr[i]-(mfpt_simu_arr[-1]-mfpt_simu_arr[1+i])/mfpt_simu_arr[-1]-m$ interp_invertBx_func = interpld(x_arr[1:-1], 1.0/Bx_arr, kind='cubic') # fill_value="extrapolate") integral invertBx arr = np.zeros(N-2) beta Grec2 arr = np.zeros(N-2) for i in range(N-2): # Here x0 is x_arr[1] integral_invertBx_arr[i], _ = quad(interp_invertBx_func, x_arr[1], x_arr[1+i]) beta Grec2 arr[i] = beta U(x arr[1])+np.log(Bx arr[i]/Bx arr[0])-integral invertBx arr[i] print(_) plt.plot(x arr[1:-1], beta Grec2 arr, label="exact Pst with simulated MFPT") plt.plot(x_arr[1:-1], beta_U(x_arr[1:-1]), ':', label="exact") # Plot formatting plt.xlabel('n') plt.ylabel('\$ \\beta \Delta G(n) \$') plt.title('free energy reconstruction') plt.legend() plt.grid() 0.0005854517561246147 free energy reconstruction exact Pst with simulated MFPT -0.5exact -1.0-1.5-2.0-2.5-3.0-3.5-4.0-4.5-2.0-1.5-1.0-0.50.0 0.5 n (2) Simulated Steady State Distribution with exact MFPT In [18]: interp simu Pst func = interpld(n arr, Pst n, kind='cubic', fill value="extrapolate") # Except for the absorbing boundary Pst(b) = 0, D(x)=0, also avoid at reflecting boundary A(x)=0, B(x)=0, simu Bx arr = np.zeros(N-2)integral_Pst_arr = np.zeros(N-2) for i in range(N-2): integral_Pst_arr[i], _ = quad(interp_simu_Pst_func, n_arr[1+i], n_arr[-1]) $simu_Bx_arr[i] = -1.0/Pst_n[1+i]*(integral_Pst_arr[i]-(mfpt_arr[-1]-mfpt_arr[1+i])/mfpt_arr[-1])$ interp_simu_invertBx_func = interpld(x_arr[1:-1], 1.0/simu_Bx_arr, kind='cubic') # fill_value="extrapolate") integral invertBx arr = np.zeros(N-2) beta_Grec2_arr = np.zeros(N-2) for i in range(N-2): # Here x0 is x arr[1] integral_invertBx_arr[i], _ = quad(interp_simu_invertBx_func, n_arr[1], n_arr[1+i]) beta_Grec2_arr[i] = beta_U(n_arr[1])+np.log(simu_Bx_arr[i]/simu_Bx_arr[0])-integral_invertBx_arr[i] print(_) plt.plot(n_arr[1:-1], beta_Grec2_arr, label="simulated Pst with exact MFPT") plt.plot(n_arr[1:-1], beta_U(x_arr[1:-1]), ':', label="exact") # Plot formatting plt.xlabel('n') plt.ylabel('\$ \\beta \Delta G(n) \$') plt.title('free energy reconstruction') plt.legend() plt.grid() 0.00015107363757098464 free energy reconstruction simulated Pst with exact MFPT -0.5exact -1.0-1.5-2.0-2.5-3.0-3.5-0.5 -2.0-1.5-1.00.0 0.5 n (3) Simulated Steady State Distribution with simulated MFPT In [19]: # Except for the absorbing boundary Pst(b) = 0, D(x)=0, also avoid at reflecting boundary A(x)=0, B(x)=0, $simu_Bx_arr = np.zeros(N-2)$ integral_Pst_arr = np.zeros(N-2) for i in range(N-2): integral_Pst_arr[i], _ = quad(interp_simu_Pst_func, n_arr[1+i], n_arr[-1]) simu Bx arr[i] = -1.0/Pst n[1+i]*(integral Pst <math>arr[i]-(mfpt simu arr[-1]-mfpt simu arr[1+i])/mfpt simu arr[-1]interp_simu_invertBx_func = interpld(n_arr[1:-1], 1.0/simu_Bx_arr, kind='cubic') # fill_value="extrapolate") integral_invertBx_arr = np.zeros(N-2) beta_Grec2_arr = np.zeros(N-2) for i in range(N-2): # Here x0 is x_arr[1] integral_invertBx_arr[i], _ = quad(interp_simu_invertBx_func, n_arr[1], n_arr[1+i]) beta_Grec2_arr[i] = beta_U(n_arr[1])+np.log(simu_Bx_arr[i]/simu_Bx_arr[0])-integral_invertBx_arr[i] print(_) plt.plot(n_arr[1:-2], beta_Grec2_arr[:-1], label="simulatied Pst & MFPT") plt.plot(n_arr[1:-1], beta_U(x_arr[1:-1]), ':', label="exact") # Plot formatting plt.xlabel('n') plt.ylabel('\$ \\beta \Delta G(n) \$') plt.title('free energy reconstruction') plt.legend() plt.grid() 0.001023960112321978 free energy reconstruction simulatied Pst & MFPT -0.5····· exact -1.0-1.5 $\beta\Delta G(n)$ -2.0-2.5-3.0-3.5-2.0-1.5-1.0-0.50.0 0.5 Reconstruct Free Energy Using data [mfpt (mat_Mbar[0]), Pst (ria trans.steady state)] calculated by Transfer Matrix (1) Transfer Matrix Steady State Distribution with exact MFPT In [20]: # interp_Pst_func = interpld(x_arr[:-1], np.array(1.0/(h*np.sum(ria_trans.eig6_v[:, 1]))*ria_trans.eig6_v[:, 1])) interp_Pst_func = interpld(x_arr[:-1], np.array(ria_trans.steady_state, dtype=float), kind='cubic', fill_valu # Except for the absorbing boundary Pst(b) = 0, D(x)=0, also avoid at reflecting boundary A(x)=0, B(x)=0, $Bx_arr = np.zeros(N-2)$ integral_Pst_arr = np.zeros(N-2) for i in range(N-2): integral_Pst_arr[i], _ = quad(interp_Pst_func, x_arr[1+i], x_arr[-1]) $Bx_arr[i] = -1.0/interp_Pst_func(x_arr[1+i])*(integral_Pst_arr[i]-(mfpt_arr[-1]-mfpt_arr[1+i])/mfpt_arr[-1]-mfpt_arr[-1]$ interp invertBx func = interpld(x arr[1:-1], 1.0/Bx arr, kind='cubic') # fill value="extrapolate") integral invertBx arr = np.zeros(N-2) beta_Grec2_arr = np.zeros(N-2) for i in range(N-2): # Here x0 is x arr[1] integral_invertBx_arr[i], _ = quad(interp_invertBx_func, x_arr[1], x_arr[1+i]) beta_Grec2_arr[i] = beta_U(x_arr[1])+np.log(Bx_arr[i]/Bx_arr[0])-integral_invertBx_arr[i] print(_) plt.plot(x_arr[1:-1], beta_Grec2_arr, label="transfer matrix Pst with exact MFPT") plt.plot(x_arr[1:-1], beta_U(x_arr[1:-1]), ':', label="exact") # Plot formatting plt.xlabel('n') plt.ylabel('\$ \\beta \Delta G(n) \$') plt.title('free energy reconstruction') plt.legend() plt.grid() 8.408819427882198e-08 free energy reconstruction transfer matrix Pst with exact MFPT -0.5-1.0-1.5-2.0-2.5-3.0-3.5-2.0-1.5-1.00.0 0.5 -0.5n (2) Transfer Matrix Steady State Distribution with simulated MFPT

In [1]:

import numpy as np

import warnings

D0 = 0.01 **def** D(x):

Suppress all warnings

return -(V1 + V2)

return D0*x**(2/3)

plt.figure(figsize=(8, 6))

plt.ylabel(r'\$\beta U(x)\$')

plt.figure(figsize=(8, 6))

return D0*x**0 x = np.linspace(-2, 2, 400)

Plot the potential

plt.xlabel('\$x\$')

plt.plot(x, D(x))
plt.xlabel('x')

plt.grid(True)
plt.show()

-0.5

-1.0

-1.5

-2.0

-2.5

-3.0

-3.5

-4.0

plt.ylabel("\$D(x)\$")

plt.legend()
plt.show()

import matplotlib.pyplot as plt

from scipy.integrate import quad

warnings.filterwarnings("ignore")

beta_U = double_gaussian_potential

from scipy.sparse.linalg import spsolve

from scipy.interpolate import interpld, PchipInterpolator

Define the double-well potential using two Gaussian functions

V1 = A1 * np.exp(-((x - mu1)**2) / (2 * sigma1**2))V2 = A2 * np.exp(-((x - mu2)**2) / (2 * sigma2**2))

plt.plot(x, beta_U(x), label='Potential Energy')

plt.axhline(0, color='black',linewidth=0.5)
plt.axvline(0.8, color='red',linewidth=1)

plt.title('Effective Diffusion Coefficient')

plt.title('Double-Well Potential with Two Gaussian Functions')

plt.grid(color = 'gray', linestyle = '--', linewidth = 0.5)

def double_gaussian_potential(x, A1=3, mu1=-1, sigma1=0.5, A2=4, mu2=1, sigma2=0.6):

Double-Well Potential with Two Gaussian Functions

Potential Energy

In [21]: # Except for the absorbing boundary Pst(b) = 0, D(x)=0, also avoid at reflecting boundary A(x)=0, B(x)=0, $Bx_arr = np.zeros(N-2)$ integral_Pst_arr = np.zeros(N-2) for i in range(N-2): integral_Pst_arr[i], _ = quad(interp_Pst_func, x_arr[1+i], x_arr[-1]) $Bx_{arr[i]} = -1.0/interp_Pst_func(x_{arr[1+i]})*(integral_Pst_{arr[i]} - (mfpt_simu_arr[-1] - mfpt_simu_arr[1+i])/(mfpt_simu_arr[-1] - mfpt_simu_arr[-1] - (mfpt_simu_arr[-1] - mfpt_simu_arr[-1] - (mfpt_simu_arr[-1] - mfpt_simu_arr[-1] - (mfpt_simu_arr[-1] - mfpt_simu_arr[-1] - (mfpt_simu_arr[-1] - (mfpt_simu_arr[-1]$ $interp_invertBx_func = interpld(x_arr[1:-1], \ 1.0/Bx_arr, \ kind='cubic') \ \# \ fill_value="extrapolate")$ integral_invertBx_arr = np.zeros(N-2) beta_Grec2_arr = np.zeros(N-2) for i in range(N-2): # Here x0 is x_arr[1] integral_invertBx_arr[i], _ = quad(interp_invertBx_func, x_arr[1], x_arr[1+i]) beta_Grec2_arr[i] = beta_U(x_arr[1])+np.log(Bx_arr[i]/Bx_arr[0])-integral_invertBx_arr[i] print(_) $\label="transfer matrix Pst with simulated MFPT") \\ plt.plot(x_arr[1:-1], beta_U(x_arr[1:-1]), ':', label="exact") \\$ # Plot formatting plt.xlabel('n') plt.ylabel(r'\$ \beta U(n) \$') plt.title('free energy reconstruction') plt.legend() plt.grid() 0.0005756332315537067free energy reconstruction transfer matrix Pst with simulated MFPT -0.5····· exact -1.0-1.5-2.0 (u) ∂g −2.5 -3.0-3.5-4.0 -4.5-2.0-1.5-1.0-0.50.0 0.5 n (3) Transfer Matrix Steady State Distribution with MFPT matrix (mat_Mbar) In [22]: # Except for the absorbing boundary Pst(b) = 0, D(x)=0, also avoid at reflecting boundary A(x)=0, B(x)=0, $Bx_arr = np.zeros(N-2)$ integral_Pst_arr = np.zeros(N-2) for i in range(N-2): integral_Pst_arr[i], _ = quad(interp_Pst_func, x_arr[1+i], x_arr[-1]) $Bx_{arr[i]} = -1.0/interp_Pst_func(x_{arr[1+i]})*(integral_Pst_{arr[i]} - (mat_Mbar[0][-1] - mat_Mbar[0][1+i])/mat_mat_Mbar[0][-1] - mat_Mbar[0][-1] - mat$ interp_invertBx_func = interpld(x_arr[1:-1], 1.0/Bx_arr, kind='cubic') # fill_value="extrapolate") integral_invertBx_arr = np.zeros(N-2) beta_Grec2_arr = np.zeros(N-2) for i in range(N-2): # Here x0 is x_arr[1] integral_invertBx_arr[i], _ = quad(interp_invertBx_func, x_arr[1], x_arr[1+i])
beta_Grec2_arr[i] = beta_U(x_arr[1])+np.log(Bx_arr[i]/Bx_arr[0])-integral_invertBx_arr[i] print(_) plt.plot(x_arr[1:-1], beta_Grec2_arr, label="transfer matrix Pst & MFPT") plt.plot(x_arr[1:-1], beta_U(x_arr[1:-1]), ':', label="exact") # Plot formatting plt.xlabel('n') plt.ylabel(r'\$ \beta U(n) \$') plt.title('free energy reconstruction') plt.legend() plt.grid() 8.98490115805555e-08 free energy reconstruction transfer matrix Pst & MFPT ····· exact -1-2 -3 -4 -2.0 -1.5-1.0-0.50.0 0.5 n Obtain the diffusion coefficient from B(x) and the derivate of MFPT In [23]: dt_dx_trans = np.gradient(delt_t*mat_Mbar[0], h)
D0_trans = Bx_arr/dt_dx_trans[1:] plt.plot(x_arr[1:-1], D0_trans) plt.grid() 0.010 0.008 0.006 0.004 0.002 0.000 -1.5-2.0-1.0-0.50.0 0.5 In [24]: dt_dx_simu = np.gradient(mfpt_simu_arr, h) D0_simu = simu_Bx_arr/dt_dx_simu[1:-1] plt.plot(x_arr[1:-1], D0_simu) plt.grid() 0.05 0.04 0.03 0.02 0.01 0.00 -2.0-1.5-1.0-0.5 0.0 0.5 In []: