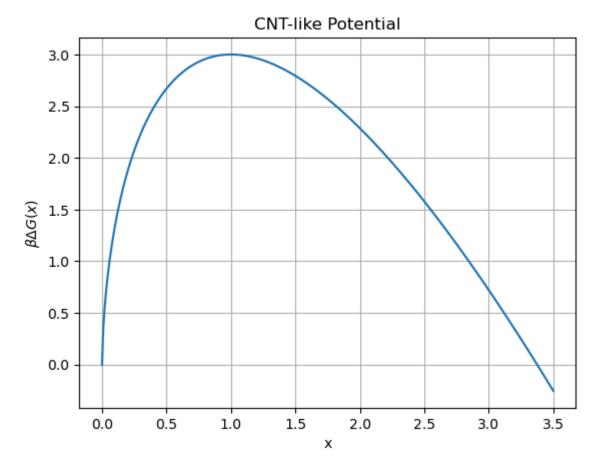
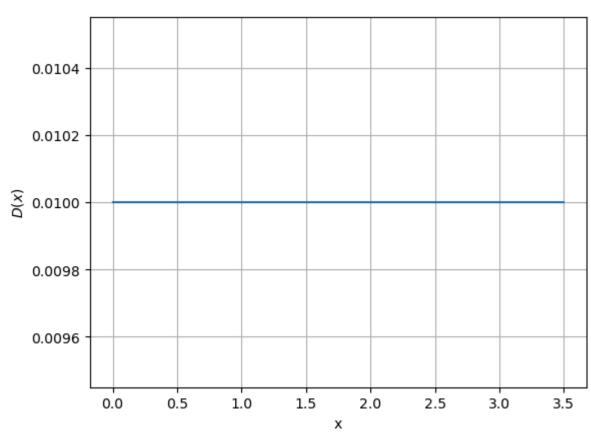
```
In [1]: import numpy as np
        import matplotlib.pyplot as plt
        from scipy.sparse.linalg import spsolve
        from scipy.integrate import quad
        from scipy.interpolate import interpld, PchipInterpolator
In [2]: beta Gstar = 3 \# kT
        D0 = 0.01 # rescaled D/D0 so it won't be used
        def beta Gcnt(x):
            return 2*beta_Gstar*(-x+3.0/2*x**(2/3))
        def Dcnt(x):
            # return D0*x**(2/3)
            return D0*x**0
        x_{arr} = np.linspace(0, 3.5, 351)
In [3]: plt.plot(x_arr, beta_Gcnt(x_arr))
        plt.grid()
        plt.xlabel('x')
        plt.ylabel("$ \\beta \\Delta G(x)$")
        plt.title('CNT-like Potential')
        plt.show()
        plt.plot(x_arr, Dcnt(x_arr))
        plt.grid()
        plt.xlabel('x')
```

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



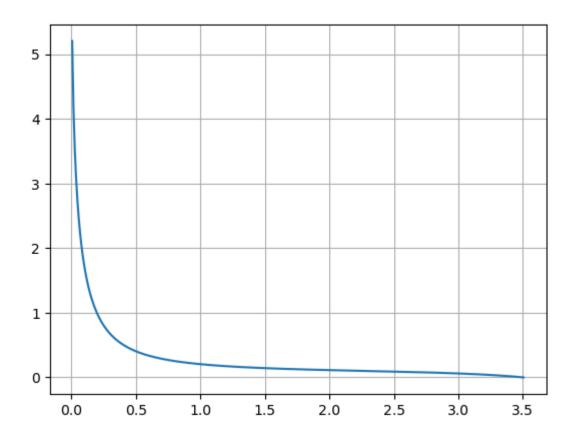
Out[3]: Text(0, 0.5, '\$ D(x)\$')



Numerical solve ODE with injecting boundary condition to find steady state solution

```
In [4]: def kappa(x):
            return Dcnt(x)*np.exp(-beta Gcnt(x))
        N = x arr.size
        h = x arr[1] - x arr[0]
        x minus half = x arr[:]
        x plus half = x arr+h
        b arr = x arr + h/2.0
        u_arr = np.zeros(b_arr.size)
In [5]: 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 \text{ vect}[0] = 1
        u arr[0:-1] = spsolve(A, f vect)
        Pst arr = u arr*np.exp(-beta Gcnt(b arr))
        Pst arr /= (np.sum(Pst arr)*h)
        inject rate = Dcnt(b arr[-1])*np.exp(-beta Gcnt(b arr[-1])+beta Gcnt(b arr[-
        print(inject rate)
        plt.plot(b arr, Pst arr)
        plt.grid()
       0.0009524325195362317
```

```
/tmp/ipykernel_787626/3470158452.py:11: SparseEfficiencyWarning: spsolve req
uires A be CSC or CSR matrix format
  u_arr[0:-1] = spsolve(A, f_vect)
```

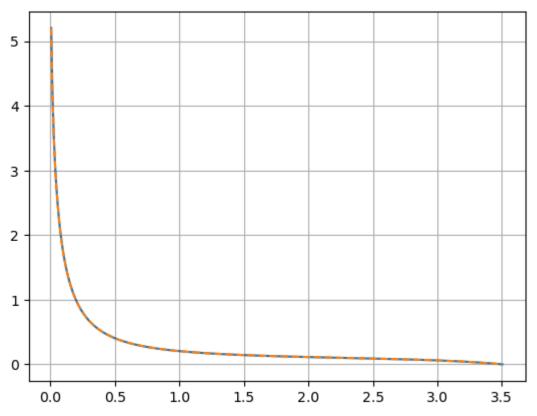


Numerically Nest integrate for Steady-State Flux and Probability distribution function

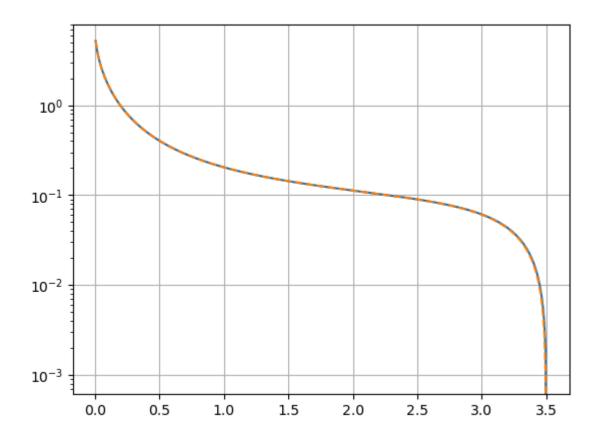
```
In [6]: a = 0 # location of reflecting boundary
        b = 3.5 # location of absorbing boundary
        # Define the inner function to integrate as a function of y
        def inner integrand(y):
            return 1.0/Dcnt(y)*np.exp(beta Gcnt(y))
        # Define the inner integral as a function 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 Gcnt(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
        print(st_flux)
```

```
In [7]: def st P func(x):
            def integrand(y):
                return 1.0/Dcnt(y)*np.exp(beta_Gcnt(y))
            # Perform the integration
            y lower = x
            y upper = b
            result, error = quad(integrand, y lower, y upper)
            result *= st flux*np.exp(-beta Gcnt(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(b_arr[i])
        plt.plot(b arr, Pst arr)
        plt.plot(b arr, st P arr, '--')
        plt.grid()
        print(Pst_arr[:5], st_P_arr[:5])
```

[5.20375997 4.15675355 3.533006 3.08971019 2.75050347] [5.19959192 4.15342 731 3.53017949 3.08723841 2.74830295]



```
In [8]: plt.semilogy(b_arr, Pst_arr)
  plt.semilogy(b_arr, st_P_arr, '--')
  plt.grid()
```



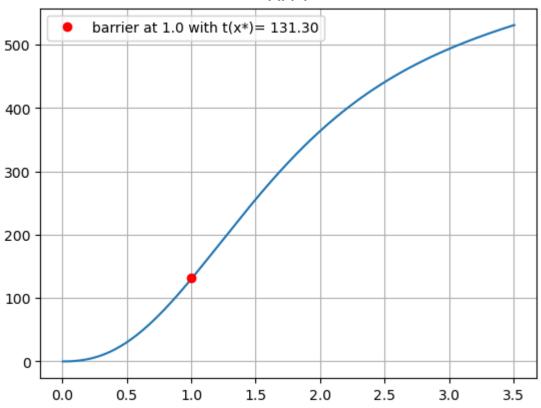
Numerically Nest integrate for MFPT

```
In [9]: \# a = 0.0 \# location of reflecting boundary
        x0 = a # Regura's method, overlap the starting point and reflecting boundary
        # b = 3.5 # location of absorbing boundary
        # Define the inner function to integrate as a function of z
        def inner_integrand(z):
            return np.exp(-beta_Gcnt(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 Gcnt(y))*inner integral(y)/Dcnt(y)
        # Perform the outer integration (from x0 to b)
        result, error = quad(outer integrand, y lower, y upper)
```

```
In [10]: mfpt arr = np.zeros(b arr.size)
          for i in np.arange(b arr.size):
              mfpt_arr[i], _ = quad(outer_integrand, y_lower, b_arr[i])
          idx_barrier = np.where((b_arr > 1.0) & (b_arr < 1.0 + h))
          t star = mfpt arr[idx barrier][0]
          print(f''\{2*t star\} = 2*t(x*)")
          print(f"{mfpt arr[-1]} = t(b)")
          print(f"{invert st flux} = 1/J")
          print(f"{1/inject rate/2} = 1/inject rate")
          plt.plot(b arr, mfpt arr)
          plt.plot(1.0, t star , 'ro', label=f"barrier at 1.0 with t(x^*)=\{t \text{ star: } .2f\}
          plt.title('MFPT')
         plt.legend()
         plt.grid()
        262.6019205305602 = 2*t(x*)
        531.063861616151 = t(b)
        530.7418784987556 = 1/J
```

MFPT

524.9715751447307 = 1/inject rate

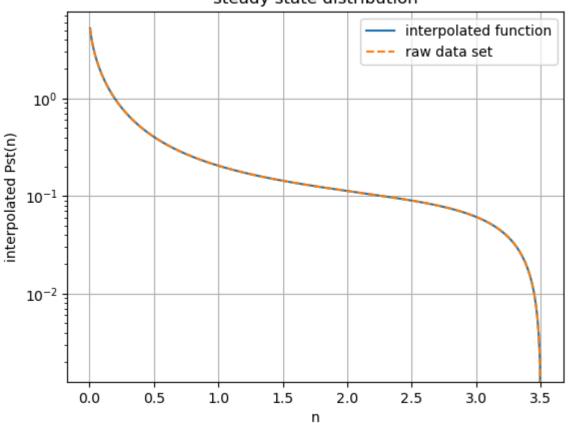


Reconstruct Free Energy from Numerical Solution of Pst_arr and Numerical Integration of mfpt

```
In [11]: interp_Pst_func = interpld(b_arr, Pst_arr, kind='cubic', fill_value="extrapo"
# interp_Pst_func = PchipInterpolator(b_arr, Pst_arr)
```

```
plt.semilogy(b_arr, interp_Pst_func(b_arr), label="interpolated function")
plt.semilogy(b_arr, Pst_arr, '--', label="raw data set")
# Plot formatting
plt.xlabel('n')
plt.ylabel('interpolated Pst(n)')
plt.title('steady state distribution')
plt.legend()
plt.grid()
```

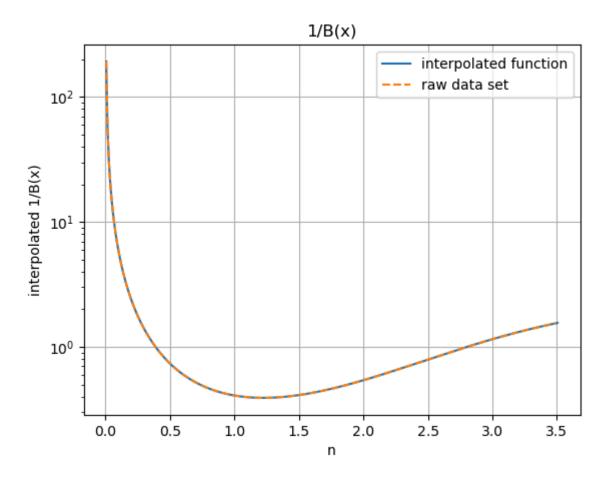
steady state distribution



```
In [12]: # Pst(b) = 0, and also avoid D(x)=0
integral_Pst_arr = np.zeros(b_arr.size-1)
Bx_arr = np.zeros(b_arr.size-1):
    integral_Pst_arr[i], _ = quad(interp_Pst_func, b_arr[i], b_arr[-1])
    Bx_arr[i] = -1.0/Pst_arr[i]*(integral_Pst_arr[i]-(mfpt_arr[-1]-mfpt_arr[

interp_invertBx_func = interpld(b_arr[:-1], 1.0/Bx_arr, kind='cubic', fill_v

plt.semilogy(b_arr, interp_invertBx_func(b_arr), label="interpolated functic plt.semilogy(b_arr[:-1], 1.0/Bx_arr, '--', label="raw data set")
# Plot formatting
plt.xlabel('n')
plt.ylabel('interpolated 1/B(x)')
plt.title('1/B(x)')
plt.legend()
plt.grid()
```



```
In [13]: integral_invertBx_arr = np.zeros(b_arr.size-1)
    beta_G_arr = np.zeros(b_arr.size-1):
        # Here x0 is b_arr[0]
        integral_invertBx_arr[i], _ = quad(interp_invertBx_func, b_arr[0], b_arr
        beta_G_arr[i] = beta_Gcnt(b_arr[0])+np.log(Bx_arr[i]/Bx_arr[0])-integral

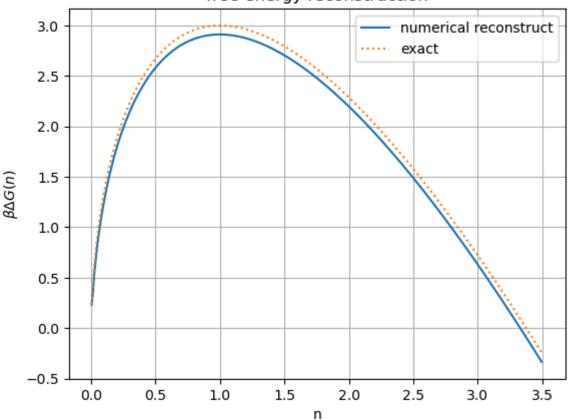
print(_)

plt.plot(b_arr[:-1], beta_G_arr, label="numerical reconstruct")
    plt.plot(b_arr, beta_Gcnt(b_arr), ':', label="exact")

# Plot formatting
    plt.xlabel('n')
    plt.ylabel('$ \\beta \Delta G(n) $')
    plt.title('free energy reconstruction')
    plt.legend()
    plt.grid()
```

6.84755899320907e-08

free energy reconstruction

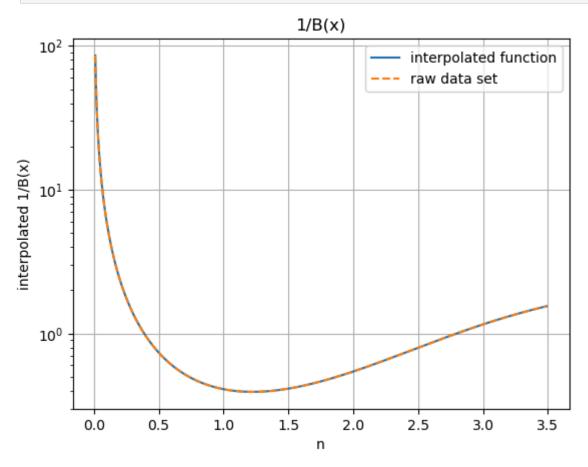


Reconstruct Free Energy from Numerical Integration of mfpt and Pst (st_P_func) - assume to be exact version

```
In [15]: # Except for the absorbing boundary Pst(b) = 0, D(x)=0, also avoid at reflect
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]-st_flux*(mfp))
interp_invertBx_func = interpld(x_arr[1:-1], 1.0/Bx_arr, kind='cubic') # fil

plt.semilogy(x_arr[1:-1], interp_invertBx_func(x_arr[1:-1]), label="interpol plt.semilogy(x_arr[1:-1], 1.0/Bx_arr, '--', label="raw data set")
# Plot formatting
plt.xlabel('n')
plt.ylabel('interpolated 1/B(x)')
plt.title('1/B(x)')
```

```
plt.legend()
plt.grid()
```



```
In [16]: 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
    beta_Grec2_arr[i] = beta_Gcnt(x_arr[1])+np.log(Bx_arr[i]/Bx_arr[0])-inte

print(_)

plt.plot(x_arr[1:-1], beta_Grec2_arr, label="exact reconstruct")
plt.plot(x_arr[1:-1], beta_Gcnt(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()
```

1.5 1.0 0.5 0.0

1.5

n

2.0

2.5

3.0

3.5

0.5

0.0

1.0

Verify Hill Relation: plot the $1/flux,\ \tau(b),\ and\ 2\tau(x*)$ v.s different location of absorbing boundary

```
In [18]:
    def exact_flux(a, b):
        # Define the inner function to integrate as a function of y
        def inner_integrand(y):
            return 1.0/Dcnt(y)*np.exp(beta_Gcnt(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_Gcnt(x))*inner_integral(x)

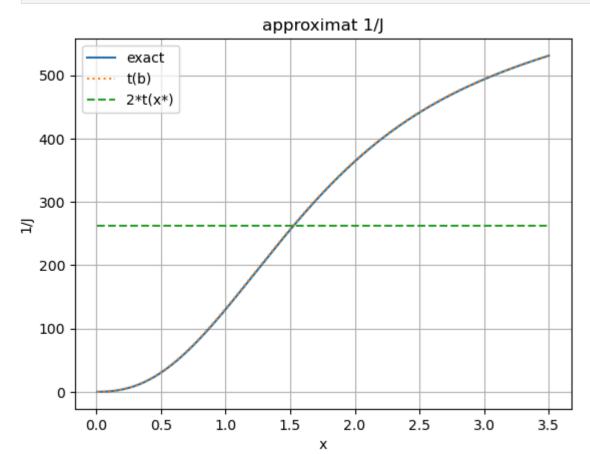
# Perform the outer integration
invert_Jst, error = quad(outer_integrand, x_lower, x_upper)
return l/invert_Jst
```

```
In [19]: # for valid value, exclude the refleting boundary
    Jst_arr = np.zeros(N-1)

for i in np.arange(N-1):
        Jst_arr[i] = exact_flux(x_arr[0], x_arr[1+i])
```

```
In [20]: plt.plot(x_arr[1:], 1/Jst_arr, label="exact")
   plt.plot(x_arr[1:], mfpt_arr[1:], ':', label="t(b)")
   plt.plot(x_arr[1:], 2*t_star*np.ones(x_arr[1:].size), '---', label="2*t(x*)")

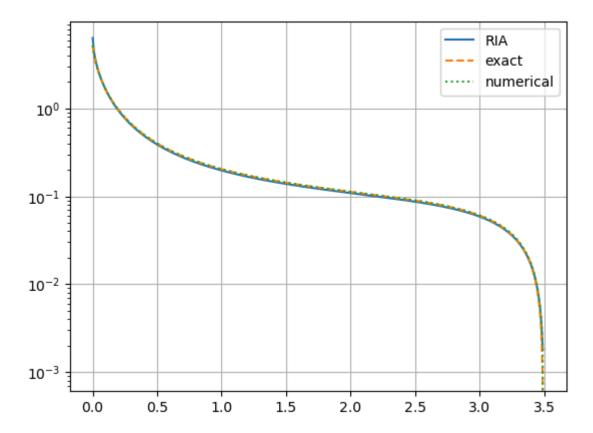
# Plot formatting
   plt.xlabel('x')
   plt.ylabel('1/J')
   plt.title('approximat 1/J')
   plt.legend()
   plt.grid()
```



Transfer Matrix with Recycling Boundary Condition

```
In [21]: from transfer matrix reptile import TransferMatrix ReInAb
         ria trans = TransferMatrix ReInAb(hx=h, x arr=x arr, beta U=beta Gcnt, crite
In [22]: # plt.plot(x arr[:-1], ria trans.steady state, label="RIA")
         # # plt.plot(x arr[:-1], 1.0/(h*np.sum(ria trans.eig6 v[:, 1]))*ria trans.ei
         # plt.plot(x_arr, st_P_arr, '--', label="exact")
         # plt.plot(b_arr, Pst_arr, ':', label="numerical")
         plt.semilogy(x_arr[:-1], ria_trans.steady state, label="RIA")
         # plt.semilogy(x arr[:-1], 1.0/(h*np.sum(ria trans.eig6 v[:, 1]))*ria trans.
         plt.semilogy(x_arr, st_P_arr, '--', label="exact")
         plt.semilogy(x_arr, Pst_arr, ':', label="numerical")
         plt.legend()
         plt.grid()
         print(ria_trans.relax timescale)
        (5659.746400019532-1606.5616490138746j)
        /home/yjiang23/anaconda3/lib/python3.11/site-packages/matplotlib/cbook.py:16
        99: ComplexWarning: Casting complex values to real discards the imaginary pa
          return math.isfinite(val)
        /home/yjiang23/anaconda3/lib/python3.11/site-packages/matplotlib/cbook.py:13
        45: ComplexWarning: Casting complex values to real discards the imaginary pa
```

return np.asarray(x, float)

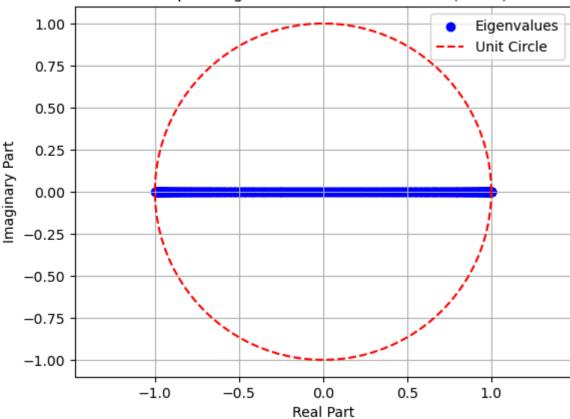


In [23]: ria_trans.plot_eigenvalues()
print(ria_trans.eig6_w)

/home/yjiang23/Desktop/research/MFPT/MFPT_reconst_experiment/transfer_matrix _reptile.py:547: RuntimeWarning: k >= N - 1 for N * N square matrix. Attempt ing to use scipy.linalg.eig instead.

eigenvalues, eigenvectors_mat = eigs(self.trans_mat, k=num_eigenvalues)

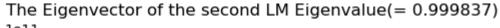
Complex Eigenvalues and Unit Circle (ReAb)

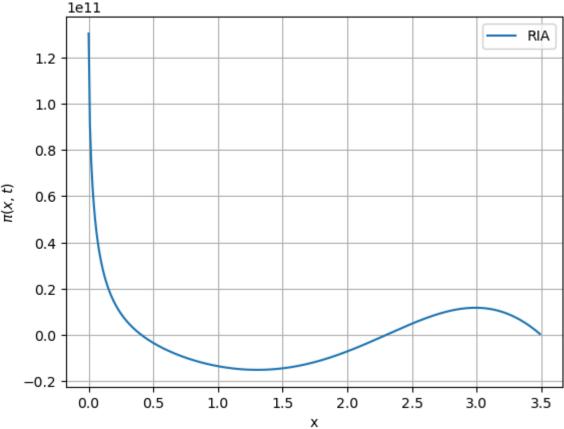


```
[1. +0.00000000e+00j 0.9998365 -4.64063578e-05j 0.9998365 +4.64063578e-05j 0.9993665 -8.80166794e-05j 0.9993665 +8.80166794e-05j 0.99857693-1.24414070e-04j]
```

```
In [24]: plt.plot(x_arr[:-1], 1.0/(h*np.sum(ria_trans.eig6_v[:, 1]))*ria_trans.eig6_v
# Plot formatting
plt.xlabel('x')
plt.ylabel('$ \pi(x,t) $')
plt.title(f'The Eigenvector of the second LM Eigenvalue(= {ria_trans.eig6_w[
plt.legend()
plt.grid(True)
print(1.0/(h*np.sum(ria_trans.eig6_v[:, 1]))*ria_trans.eig6_v[:, 1][:10])

[1.30286356e+11+3.99873573e+11j 9.06163097e+10+2.81056666e+11j
7.47924582e+10+2.34841677e+11j 6.41514601e+10+2.04237752e+11j
5.61917605e+10+1.81659750e+11j 4.99003188e+10+1.64048196e+11j
4.47503754e+10+1.49817877e+11j 4.04293689e+10+1.38031297e+11j
3.67356718e+10+1.28085591e+11j 3.35314183e+10+1.19569776e+11j]
```





In []:

Extract Steady State Distribution and MFPT from Simulation

```
In [25]: from mfpt_Pst_RW_simulate import simulate_ReAb, simulate_ReAb_accelerate

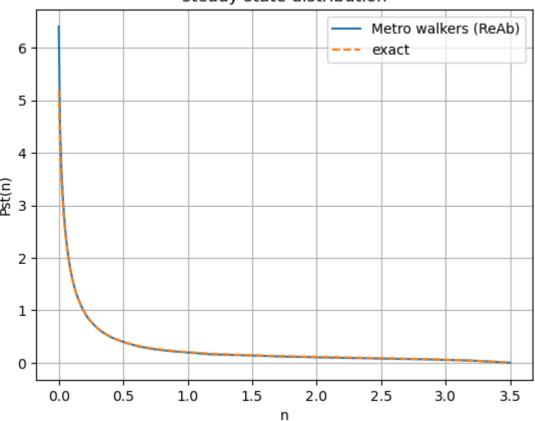
num_particles = 600
init_position_arr = np.zeros(num_particles, dtype=float)
simu_x_arr = np.linspace(a, b, 351)
hx = simu_x_arr[1]-simu_x_arr[0]
ht = hx**2/(2*D0)
n_arr = np.arange(a, b+0.01, 0.01)
n_arr = np.round(n_arr, decimals=5)
hx == 0.01
```

Out[25]: True

```
In [26]: count_n, ti_n = simulate_ReAb_accelerate(init_position_arr=init_position_arr
In [27]: Pst_n = count_n/(hx*np.sum(count_n))
    plt.plot(n_arr, Pst_n, label="Metro walkers (ReAb)")
    plt.plot(x_arr, st_P_arr, '--', label="exact")
    # plt.plot(x_arr[:-1], ria_trans.steady_state, label="RIA")
    # plt.plot(x_arr[:-1], 1.0/(h*np.sum(ria_trans.eig6_v[:, 1]))*ria_trans.eig6
```

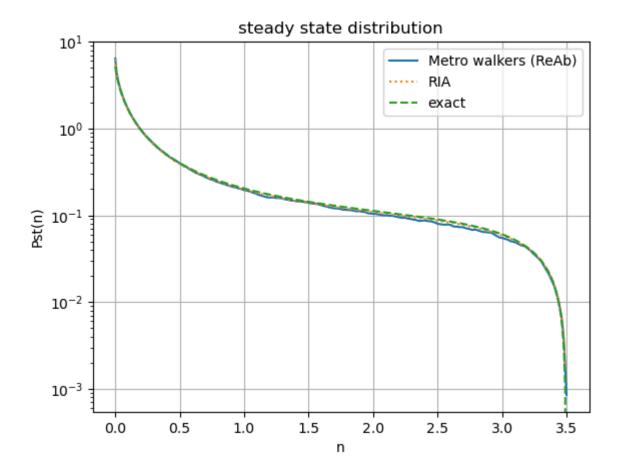
```
# Plot formatting
plt.xlabel('n')
plt.ylabel('Pst(n)')
plt.title('steady state distribution')
plt.legend()
plt.grid()
```

steady state distribution



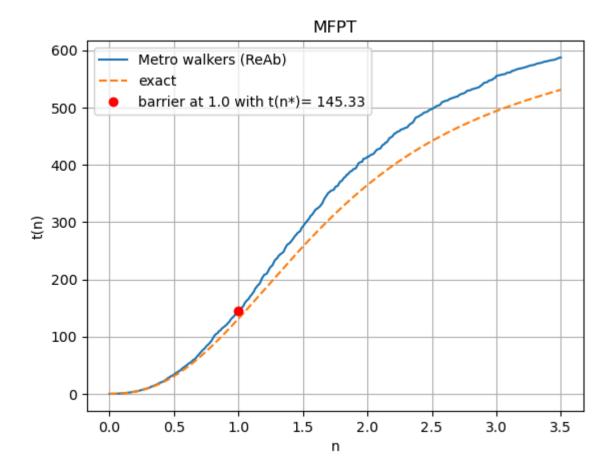
```
In [28]: plt.semilogy(n_arr, Pst_n, label="Metro walkers (ReAb)")
    plt.semilogy(x_arr[:-1], ria_trans.steady_state, ':', label="RIA")
    # plt.semilogy(x_arr[:-1], 1.0/(h*np.sum(ria_trans.eig6_v[:, 1]))*ria_trans.
    plt.semilogy(x_arr, st_P_arr, '--', label="exact")

# Plot formatting
    plt.xlabel('n')
    plt.ylabel('Pst(n)')
    plt.title('steady state distribution')
    plt.legend()
    plt.grid()
```



```
In [29]: mfpt_simu_arr = np.mean(ti_n, axis=0)
    idx_barrier = np.where(n_arr == 1.0)[0][0]
    t_star = mfpt_simu_arr[idx_barrier]

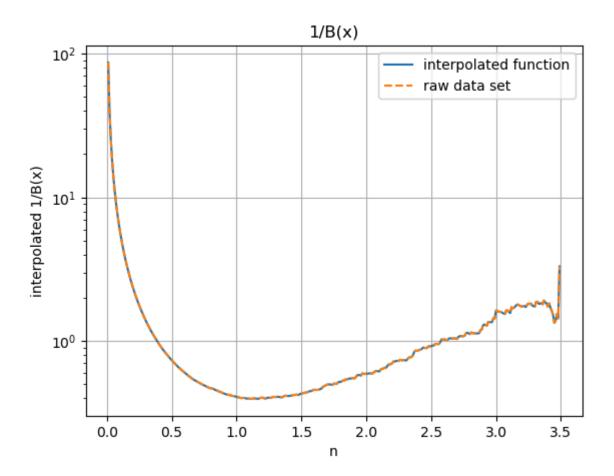
plt.plot(n_arr, mfpt_simu_arr, label="Metro walkers (ReAb)")
    plt.plot(x_arr, mfpt_arr, '--', label="exact")
    plt.plot(1.0, t_star , 'ro', label=f'barrier at 1.0 with t(n*)={t_star: .2f}
    # Plot formatting
    plt.xlabel('n')
    plt.ylabel('t(n)')
    plt.title('MFPT')
    plt.legend()
    plt.grid()
```



Reconstruct Free Energy Using mfpt (mfpt_simu_arr) and Pst (Pst_n) extracted from Simulation

(1) Exact Steady State Distribution with simulated MFPT

```
# Except for the absorbing boundary Pst(b) = 0, D(x)=0, also avoid at reflect
In [30]:
         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_a
         interp invertBx func = interpld(x arr[1:-1], 1.0/Bx arr, kind='cubic') # fil
         plt.semilogy(x arr[1:-1], interp invertBx func(x arr[1:-1]), label="interpol
         plt.semilogy(x arr[1:-1], 1.0/Bx arr, '--', label="raw data set")
         # Plot formatting
         plt.xlabel('n')
         plt.ylabel('interpolated 1/B(x)')
         plt.title('1/B(x)')
         plt.legend()
         plt.grid()
```



```
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 beta_Grec2_arr[i] = beta_Gcnt(x_arr[1])+np.log(Bx_arr[i]/Bx_arr[0])-inte

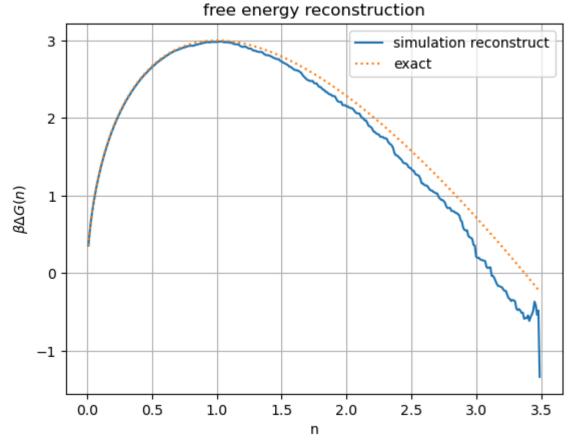
print(_)

plt.plot(x_arr[1:-1], beta_Grec2_arr, label="simulation reconstruct")
plt.plot(x_arr[1:-1], beta_Gcnt(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()
```

/tmp/ipykernel_787626/2563123347.py:6: IntegrationWarning: The maximum numbe
r of subdivisions (50) has been achieved.
 If increasing the limit yields no improvement it is advised to analyze
 the integrand in order to determine the difficulties. If the position of
a
 local difficulty can be determined (singularity, discontinuity) one will
 probably gain from splitting up the interval and calling the integrator
 on the subranges. Perhaps a special-purpose integrator should be used.
 integral_invertBx_arr[i], _ = quad(interp_invertBx_func, x_arr[1], x_arr[1+i])

0.005085227757393399



(2) Transfer Matrix Steady State Distribution with exact MFPT and then with simulated MFPT

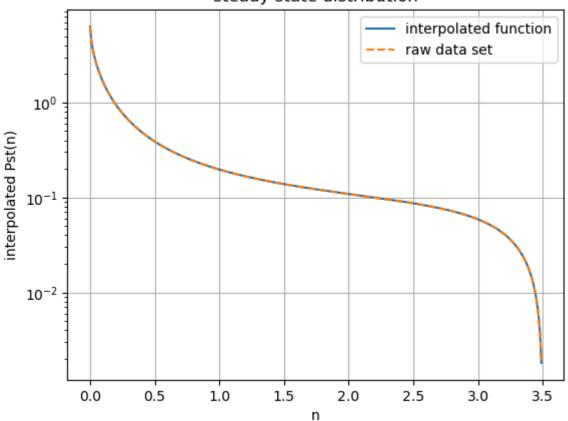
```
In [32]: interp_Pst_func = interpld(x_arr[:-1], np.array(ria_trans.steady_state, dtyp
# interp_Pst_func = interpld(x_arr[:-1], np.array(1.0/(h*np.sum(ria_trans.ei)

plt.semilogy(x_arr[:-1], interp_Pst_func(x_arr[:-1]), label="interpolated fuplt.semilogy(x_arr[:-1], ria_trans.steady_state, '--', label="raw data set")
# plt.semilogy(x_arr[:-1], np.array(1.0/(h*np.sum(ria_trans.eig6_v[:, 1]))*r

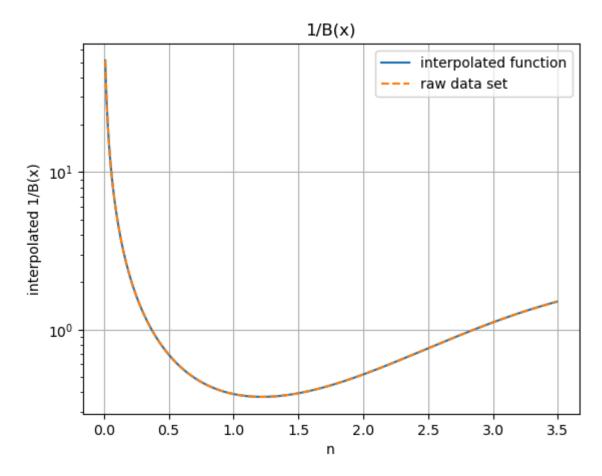
# Plot formatting
plt.xlabel('n')
plt.ylabel('interpolated Pst(n)')
plt.title('steady state distribution')
plt.legend()
plt.grid()
```

```
/tmp/ipykernel_787626/2788019905.py:1: ComplexWarning: Casting complex value
s to real discards the imaginary part
  interp_Pst_func = interpld(x_arr[:-1], np.array(ria_trans.steady_state, dt
ype=float), kind='cubic', fill value="extrapolate")
```

steady state distribution



```
# Except for the absorbing boundary Pst(b) = 0, D(x) = 0, also avoid at reflect
In [33]:
         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
             # Bx arr[i] = -1.0/interp Pst func(x arr[1+i])*(integral Pst arr[i]-st 1
         interp invertBx func = interpld(x arr[1:-1], 1.0/Bx arr, kind='cubic') # fil
         plt.semilogy(x_arr[1:-1], interp_invertBx_func(x_arr[1:-1]), label="interpol")
         plt.semilogy(x_arr[1:-1], 1.0/Bx_arr, '--', label="raw data set")
         # Plot formatting
         plt.xlabel('n')
         plt.ylabel('interpolated 1/B(x)')
         plt.title('1/B(x)')
         plt.legend()
         plt.grid()
```



```
In [34]: 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
    beta_Grec2_arr[i] = beta_Gcnt(x_arr[1])+np.log(Bx_arr[i]/Bx_arr[0])-inte

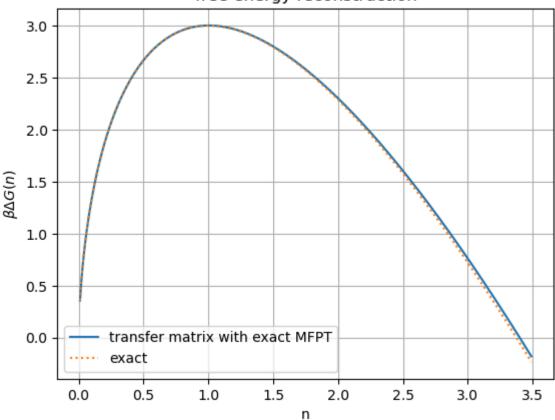
print(_)

plt.plot(x_arr[1:-1], beta_Grec2_arr, label="transfer matrix with exact MFPT
plt.plot(x_arr[1:-1], beta_Gcnt(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()
```

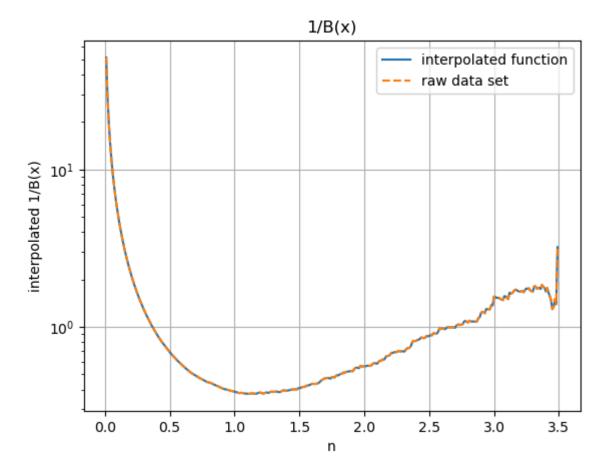
3.24027710664547e-08

free energy reconstruction



```
In [35]: # Except for the absorbing boundary Pst(b) = 0, D(x)=0, also avoid at reflect
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_interp_invertBx_func = interpld(x_arr[1:-1], 1.0/Bx_arr, kind='cubic') # filt

plt.semilogy(x_arr[1:-1], interp_invertBx_func(x_arr[1:-1]), label="interpoling plt.semilogy(x_arr[1:-1], 1.0/Bx_arr, '--', label="raw data set")
# Plot formatting
plt.xlabel('n')
plt.ylabel('interpolated 1/B(x)')
plt.title('1/B(x)')
plt.legend()
plt.grid()
```



```
In [36]: integral_invertBx_arr = np.zeros(N-2)
beta_Grec2_arr = np.zeros(N-2)

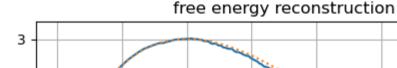
for i in range(N-2):
    # Here x0 is x_arr[1]
    integral_invertBx_arr[i], _ = quad(interp_invertBx_func, x_arr[1], x_arr
    beta_Grec2_arr[i] = beta_Gcnt(x_arr[1])+np.log(Bx_arr[i]/Bx_arr[0])-inte

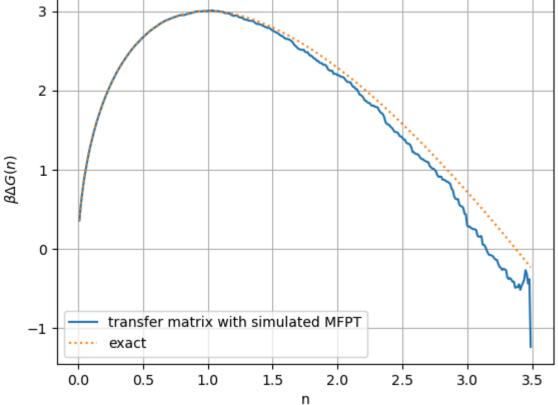
print(_)

plt.plot(x_arr[1:-1], beta_Grec2_arr, label="transfer matrix with simulated
plt.plot(x_arr[1:-1], beta_Gcnt(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()
```

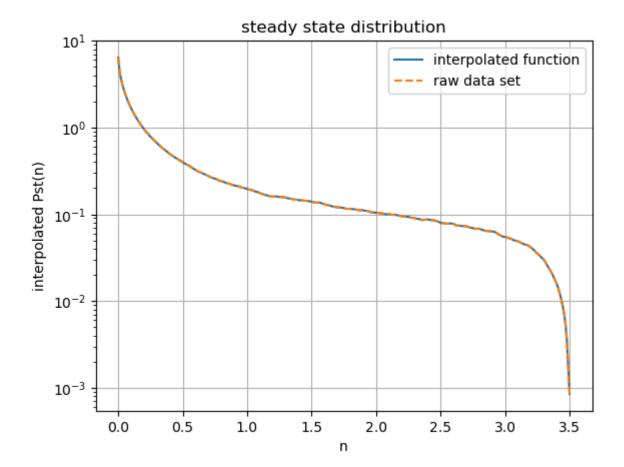
```
/tmp/ipykernel 787626/2741261074.py:6: IntegrationWarning: The maximum numbe
r of subdivisions (50) has been achieved.
  If increasing the limit yields no improvement it is advised to analyze
  the integrand in order to determine the difficulties. If the position of
  local difficulty can be determined (singularity, discontinuity) one will
  probably gain from splitting up the interval and calling the integrator
  on the subranges. Perhaps a special-purpose integrator should be used.
  integral_invertBx_arr[i], _ = quad(interp_invertBx_func, x_arr[1], x_arr[1]
0.004713660730942138
```



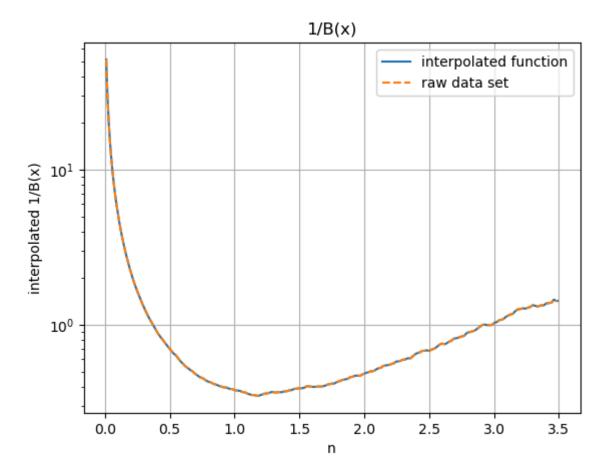


(3) Simulated Steady State Distribution with exact MFPT and then with simulated MFPT

```
In [37]: interp_simu_Pst_func = interpld(simu_x_arr, Pst_n, kind='cubic', fill_value=
         # interp Pst func = PchipInterpolator(b arr, Pst arr)
         plt.semilogy(simu x arr, interp simu Pst func(simu x arr), label="interpolat")
         plt.semilogy(simu x arr, Pst n, '--', label="raw data set")
         # Plot formatting
         plt.xlabel('n')
         plt.ylabel('interpolated Pst(n)')
         plt.title('steady state distribution')
         plt.legend()
         plt.grid()
```



```
In [38]:
         # Except for the absorbing boundary Pst(b) = 0, D(x) = 0, also avoid at reflect
         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, simu_x_arr[1+i], sim
             simu Bx arr[i] = -1.0/Pst n[1+i]*(integral Pst <math>arr[i]-(mfpt arr[-1]-mfpt
         interp simu invertBx func = interp1d(x arr[1:-1], 1.0/simu Bx arr, kind='cut
         plt.semilogy(simu x arr[1:-1], interp simu invertBx func(x arr[1:-1]), label
         plt.semilogy(simu x arr[1:-1], 1.0/simu Bx arr, '--', label="raw data set")
         # Plot formatting
         plt.xlabel('n')
         plt.ylabel('interpolated 1/B(x)')
         plt.title('1/B(x)')
         plt.legend()
         plt.grid()
        /tmp/ipykernel 787626/3951444635.py:5: IntegrationWarning: The maximum numbe
        r of subdivisions (50) has been achieved.
          If increasing the limit yields no improvement it is advised to analyze
          the integrand in order to determine the difficulties. If the position of
        а
          local difficulty can be determined (singularity, discontinuity) one will
          probably gain from splitting up the interval and calling the integrator
          on the subranges. Perhaps a special-purpose integrator should be used.
          integral Pst arr[i], = quad(interp simu Pst func, simu x arr[1+i], simu
        x arr[-1])
```



```
In [39]: 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, simu_x_arr
    beta_Grec2_arr[i] = beta_Gcnt(simu_x_arr[1])+np.log(simu_Bx_arr[i]/simu_print(_)

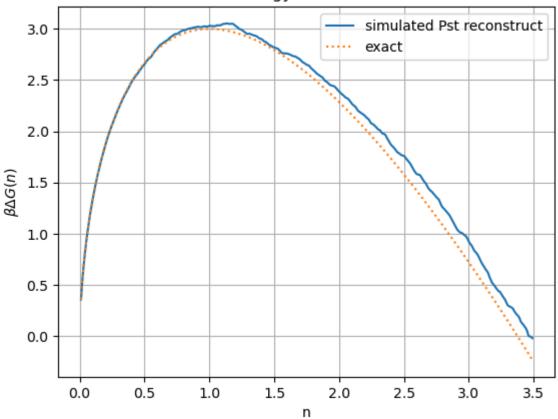
plt.plot(simu_x_arr[1:-1], beta_Grec2_arr, label="simulated Pst reconstruct"
plt.plot(simu_x_arr[1:-1], beta_Gcnt(x_arr[1:-1]), ':', label="exact")

# Plot formatting
plt.xlabel('n')
plt.ylabel('$ \\beta \Delta G(n) $')
plt.title('free energy reconstruction')
plt.legend()
plt.grid()
```

```
/tmp/ipykernel_787626/4156853672.py:6: IntegrationWarning: The maximum numbe
r of subdivisions (50) has been achieved.
   If increasing the limit yields no improvement it is advised to analyze
   the integrand in order to determine the difficulties. If the position of
a
   local difficulty can be determined (singularity, discontinuity) one will
   probably gain from splitting up the interval and calling the integrator
   on the subranges. Perhaps a special-purpose integrator should be used.
   integral_invertBx_arr[i], _ = quad(interp_simu_invertBx_func, simu_x_arr
[1], simu_x_arr[1+i])
```

0.0015712234963261276

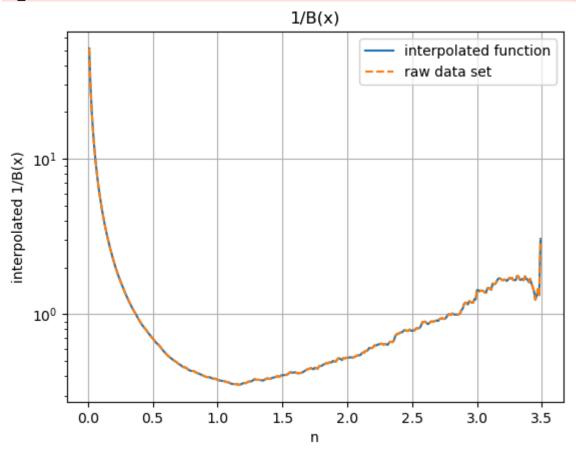




```
In [40]: # Except for the absorbing boundary Pst(b) = 0, D(x)=0, also avoid at reflect
simu_Bx_arr = np.zeros(N-2)
integral_Pst_arr = np.zeros(N-2)
for i in range(N-2):
    integral_Pst_arr[i], _ = quad(interp_simu_Pst_func, simu_x_arr[1+i], sim
        simu_Bx_arr[i] = -1.0/Pst_n[1+i]*(integral_Pst_arr[i]-(mfpt_simu_arr[-1])
interp_simu_invertBx_func = interpld(simu_x_arr[1:-1], 1.0/simu_Bx_arr, kinc

plt.semilogy(simu_x_arr[1:-1], interp_simu_invertBx_func(x_arr[1:-1]), label
plt.semilogy(simu_x_arr[1:-1], 1.0/simu_Bx_arr, '--', label="raw data set")
# Plot formatting
plt.xlabel('n')
plt.ylabel('interpolated 1/B(x)')
plt.title('1/B(x)')
plt.legend()
plt.grid()
```

```
/tmp/ipykernel_787626/1053423012.py:5: IntegrationWarning: The maximum numbe
r of subdivisions (50) has been achieved.
   If increasing the limit yields no improvement it is advised to analyze
   the integrand in order to determine the difficulties. If the position of
a
   local difficulty can be determined (singularity, discontinuity) one will
   probably gain from splitting up the interval and calling the integrator
   on the subranges. Perhaps a special-purpose integrator should be used.
   integral_Pst_arr[i], _ = quad(interp_simu_Pst_func, simu_x_arr[1+i], simu_x_arr[-1])
```



```
In [41]: 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, simu_x_arr
    beta_Grec2_arr[i] = beta_Gcnt(simu_x_arr[1])+np.log(simu_Bx_arr[i]/simu_print(_)

plt.plot(simu_x_arr[1:-1], beta_Grec2_arr, label="simulated Pst&MFPT reconst
plt.plot(simu_x_arr[1:-1], beta_Gcnt(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()
```

/tmp/ipykernel_787626/1290495388.py:6: IntegrationWarning: The maximum numbe r of subdivisions (50) has been achieved.

If increasing the limit yields no improvement it is advised to analyze the integrand in order to determine the difficulties. If the position of

local difficulty can be determined (singularity, discontinuity) one will
probably gain from splitting up the interval and calling the integrator
on the subranges. Perhaps a special-purpose integrator should be used.
integral_invertBx_arr[i], _ = quad(interp_simu_invertBx_func, simu_x_arr
[1], simu x arr[1+i])

0.004144138969086164

free energy reconstruction

