

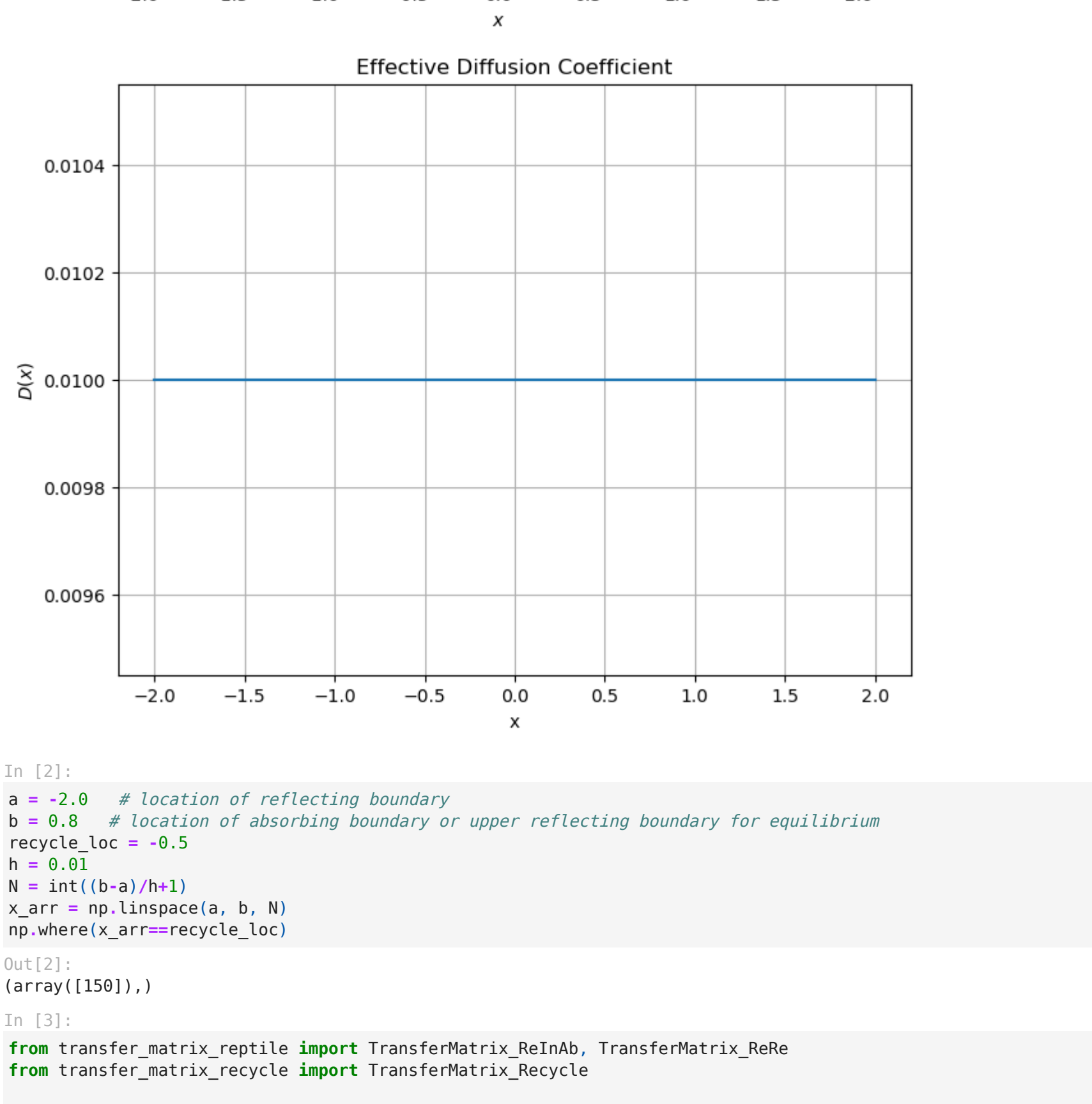
```
In [1]:
import numpy as np
import matplotlib.pyplot as plt
from scipy.sparse.linalg import spsolve
from scipy.integrate import quad
from scipy.interpolate import interp1d, PchipInterpolator

# Define the double-well potential using two Gaussian functions
def double_gaussian_potential(x, A1=3, mu1=-1, sigma1=0.5, A2=4, mu2=1, sigma2=0.6):
    V1 = A1 * np.exp(-((x - mu1)**2) / (2 * sigma1**2))
    V2 = A2 * np.exp(-((x - mu2)**2) / (2 * sigma2**2))
    return V1 + V2
beta_U = double_gaussian_potential

D0 = 0.01
def D(x):
    return D0*x**(2/3)
    return D0*x**0
x = np.linspace(-2, 2, 400)

# Plot the potential
plt.figure(figsize=(8, 6))
plt.plot(x, beta_U(x), label='Potential Energy')
plt.title('Double-Well Potential with Two Gaussian Functions')
plt.xlabel('$x$')
plt.ylabel('$\beta U(x)$')
# plt.axhline(0, color='black', linewidth=0.5)
plt.axvline(0.8, color='red', linewidth=1)
plt.grid(color='gray', linestyle='--', linewidth=0.5)
plt.legend()
plt.show()

plt.figure(figsize=(8, 6))
plt.plot(x, D(x))
plt.xlabel('$x$')
plt.ylabel('$D(x)$')
plt.title('Effective Diffusion Coefficient')
plt.grid(True)
plt.show()
```



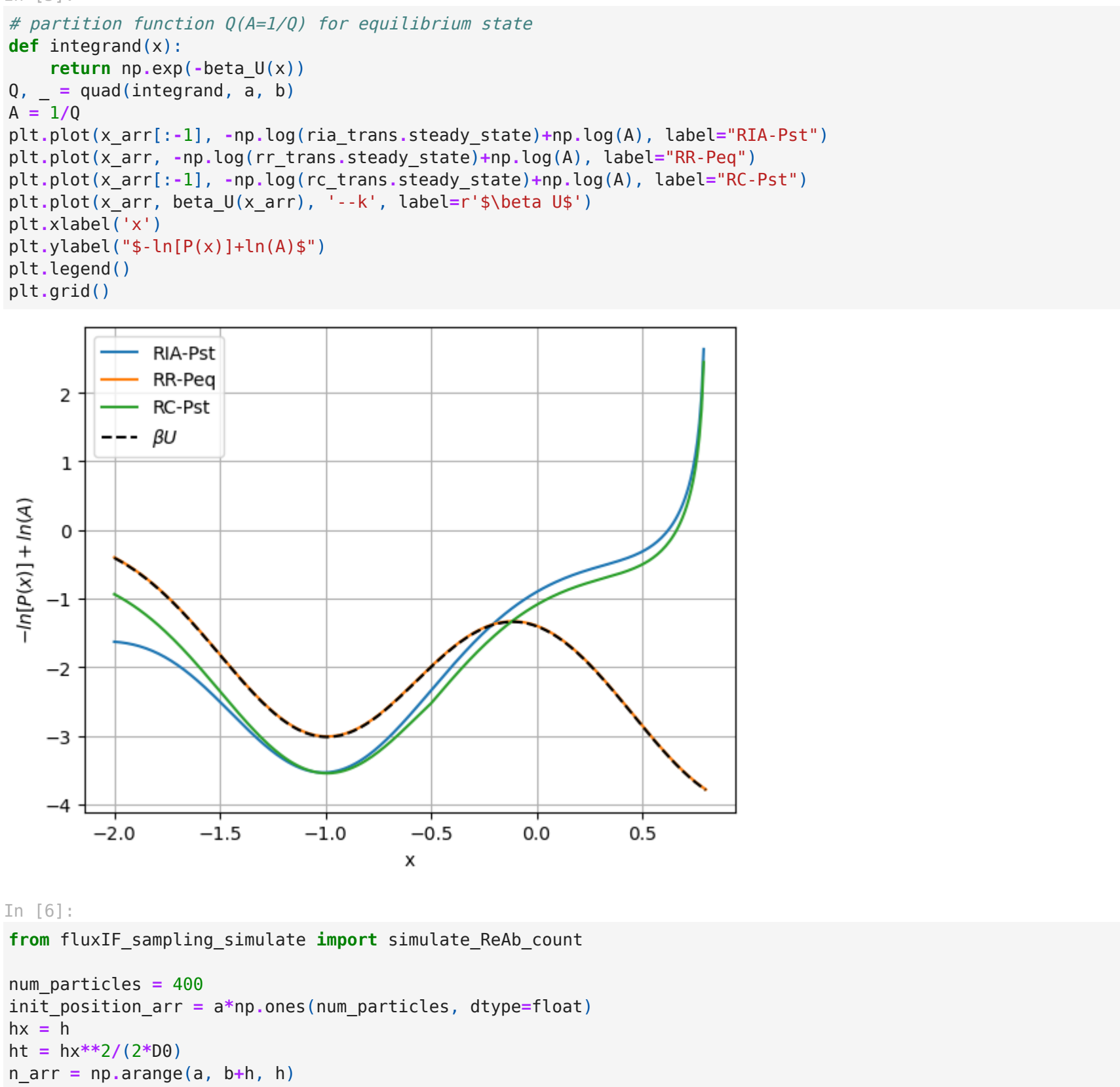
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In [2]:
a = -2.0 # location of reflecting boundary
b = 0.8 # location of absorbing boundary or upper reflecting boundary for equilibrium
recycle_loc = -0.5
n = 0.01
N = int((b-a)/h+1)
x_arr = np.linspace(a, b, N)
np.where(x_arr==recycle_loc)

Out[2]:
(array([150]),)
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In [3]:
from transfer_matrix_reptile import TransferMatrix_ReInAb, TransferMatrix_ReRe
from transfer_matrix_recycle import TransferMatrix_Recycle

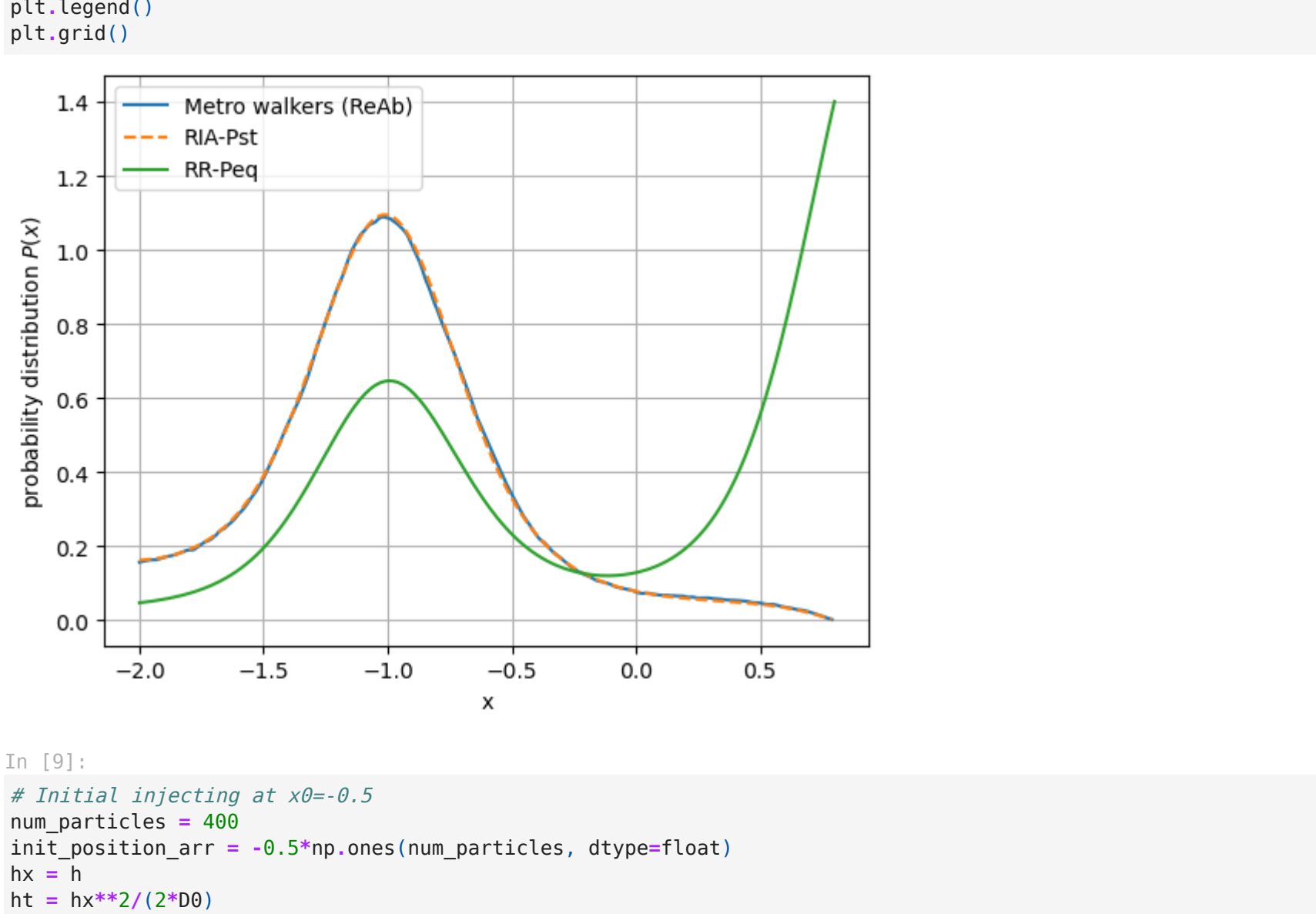
rr_trans = TransferMatrix_ReInAb(h, x_arr, beta_U, 0)
rr_trans = TransferMatrix_ReRe(h, x_arr, beta_U, 0)
rc_trans = TransferMatrix_Recycle(h, x_arr, beta_U, np.where(x_arr==recycle_loc, 0))

In [4]:
plt.plot(x_arr[:-1], ria_trans.steady_state, label='RIA-Pst')
plt.plot(x_arr, rr_trans.steady_state, label='RR-Peq')
# plt.plot(x_arr, 1.0/(h*np.sum(rr_trans.eig6_v[:, 1]))*rr_trans.eig6_v[:, 1], label='RR-Peq')
plt.plot(x_arr[:-1], rc_trans.steady_state, label='RC-Pst')
plt.xlabel('$x$')
plt.ylabel('probability distribution $P(x)$')
plt.legend()
plt.grid()
print(ria_trans.eig6_w)
print(rr_trans.eig6_w)
print(rc_trans.eig6_w)
```



```
In [5]:
# partition function Q(A=1/Q) for equilibrium state
def integrand(x):
    return np.exp(-beta_U(x))
Q, _ = quad(integrand, a, b)
A = 1/Q

plt.plot(x_arr[:-1], -np.log(ria_trans.steady_state)+np.log(A), label='RIA-Pst')
plt.plot(x_arr, -np.log(rr_trans.steady_state)+np.log(A), label='RR-Peq')
plt.plot(x_arr[:-1], -np.log(rc_trans.steady_state)+np.log(A), label='RC-Pst')
plt.plot(x_arr, beta_U(x), '--k', label='$\beta U(x)$')
plt.xlabel('$x$')
plt.ylabel('$-\ln(P(x))+\ln(A)$')
plt.legend()
plt.grid()
```

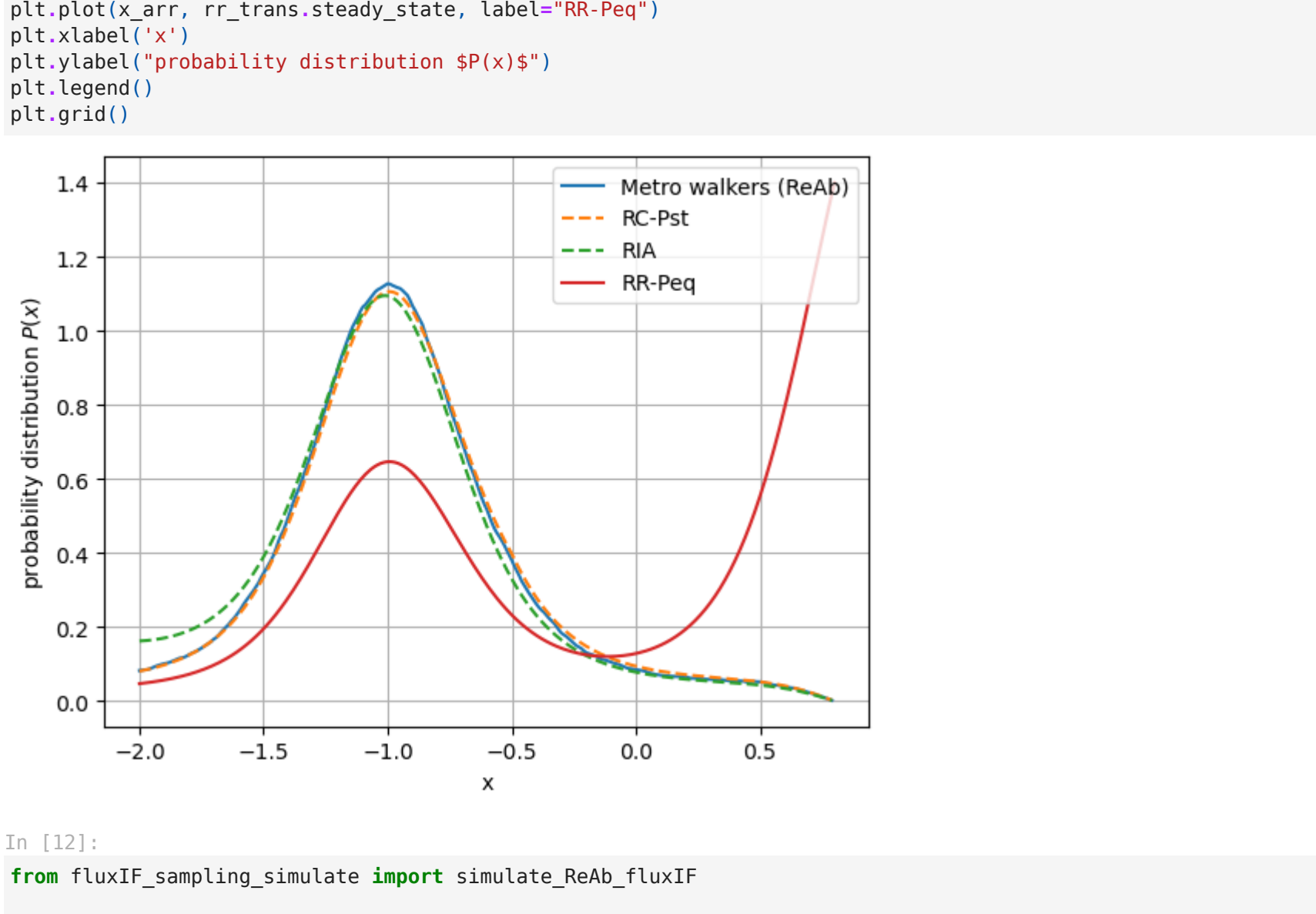


```
In [6]:
from fluxIF_sampling_simulate import simulate_ReAb_count

num_particles = 400
init_position_arr = a*np.ones(num_particles, dtype=float)
hx = h
ht = hx**2/(2*D0)
n_arr = np.arange(a, b+h, h)
n_arr = np.round(n_arr, decimals=5)
count_n = simulate_ReAb_count(init_position_arr=init_position_arr, beta_U=beta_U, n_arr=n_arr, a=a, b=b, hx=h)

Pst_n = count_n/(h*np.sum(count_n))
Pst_n = np.log(Pst_n[:-1]), label='Metro walkers (ReAb)')
plt.plot(x_arr[:-1], -np.log(ria_trans.steady_state), '--', label='RIA')

# Plot formatting
plt.xlabel('$n$')
plt.ylabel('$-\ln[P_{st}(n)]$')
plt.title('steady state distribution')
plt.legend()
plt.grid()
```



```
In [8]:
plt.plot(n_arr[:-1], Pst_n[:-1], label='Metro walkers (ReAb)')
plt.plot(x_arr[:-1], ria_trans.steady_state, '--', label='RIA-Pst')
plt.plot(x_arr, rr_trans.steady_state, '--', label='RR-Peq')
plt.xlabel('$x$')
plt.ylabel('probability distribution $P(x)$')
plt.legend()
plt.grid()
```



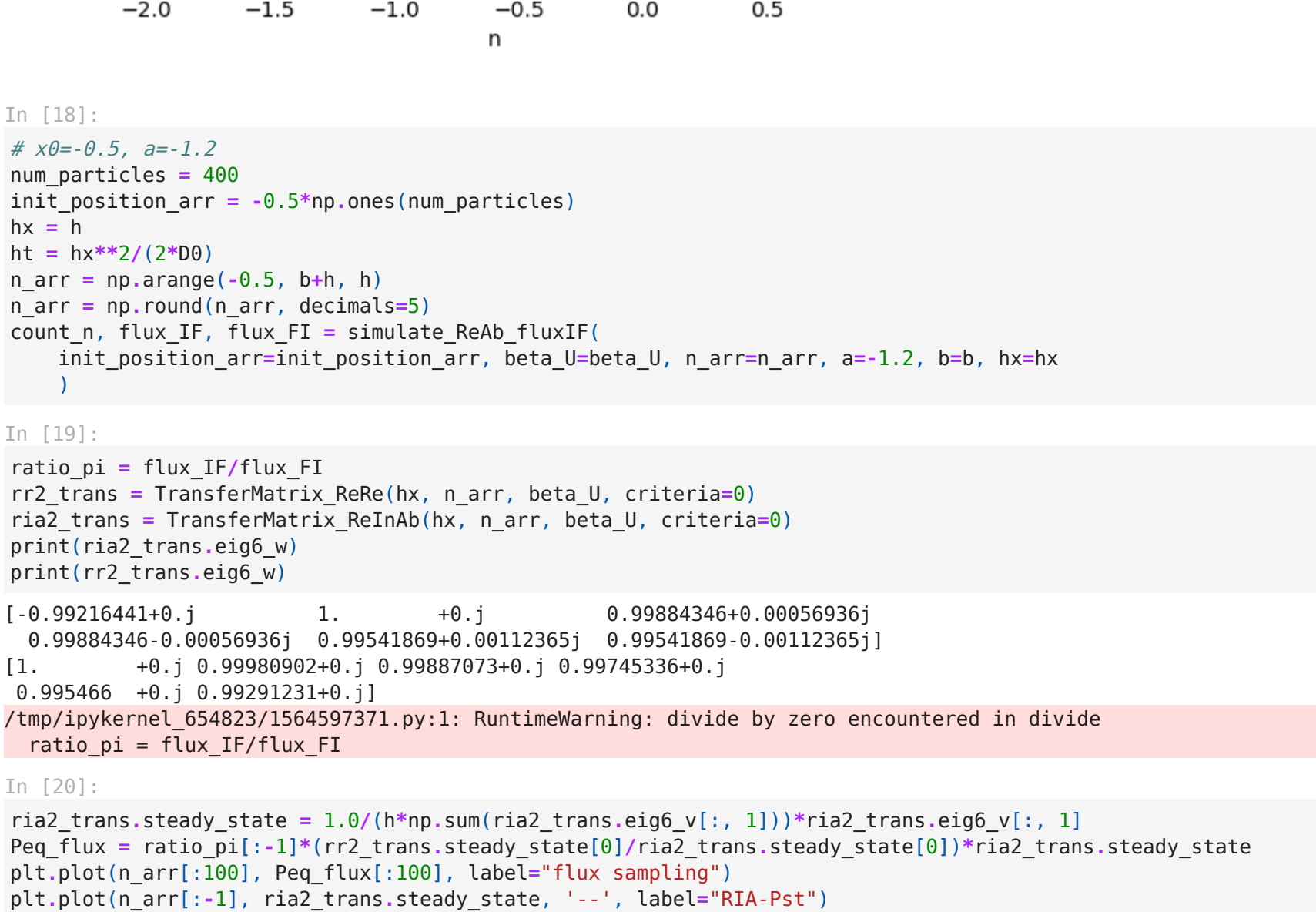
```
In [9]:
# Initial injecting at x0=-0.5
num_particles = 400
init_position_arr = -0.5*np.ones(num_particles, dtype=float)
hx = h
ht = hx**2/(2*D0)
n_arr = np.arange(a, b+h, h)
n_arr = np.round(n_arr, decimals=5)
count_n = simulate_ReAb_count(init_position_arr=init_position_arr, beta_U=beta_U, n_arr=n_arr, a=a, b=b, hx=h)

Pst_n = count_n/(h*np.sum(count_n))
Pst_n = np.log(Pst_n[:-1]), label='Metro walkers (ReAb)')
Pst_n = np.log(Pst_n[:-1]), label='RC-Pst')
Pst_n = np.log(Pst_n[:-1]), label='RR-Peq')
Pst_n = np.log(Pst_n[:-1]), label='RIA-Pst')

# Plot formatting
plt.xlabel('$n$')
plt.ylabel('$-\ln[P_{st}(n)]$')
plt.title('steady state distribution')
plt.legend()
plt.grid()
```



```
In [11]:
plt.plot(n_arr[:-1], Pst_n[:-1], label='Metro walkers (ReAb)')
plt.plot(x_arr[:-1], rc_trans.steady_state, '--', label='RC-Pst')
plt.plot(x_arr, rr_trans.steady_state, '--', label='RR-Peq')
plt.plot(x_arr[:-1], ria_trans.steady_state, '--', label='RIA-Pst')
plt.xlabel('$x$')
plt.ylabel('probability distribution $P(x)$')
plt.legend()
plt.grid()
```



```
In [12]:
from fluxIF_sampling_simulate import simulate_ReAb_fluxIF

num_particles = 400
init_position_arr = a*np.ones(num_particles, dtype=float)
hx = h
ht = hx**2/(2*D0)
n_arr = np.arange(a, b+h, h)
n_arr = np.round(n_arr, decimals=5)
count_n = simulate_ReAb_count(init_position_arr=init_position_arr, beta_U=beta_U, n_arr=n_arr, a=a, b=b, hx=h)

Pst_n = count_n/(h*np.sum(count_n))
Pst_n = np.log(Pst_n[:-1]), label='Metro walkers (ReAb)')
Pst_n = np.log(Pst_n[:-1]), label='RC-Pst')
Pst_n = np.log(Pst_n[:-1]), label='RR-Peq')
Pst_n = np.log(Pst_n[:-1]), label='RIA-Pst')

# Plot formatting
plt.xlabel('$n$')
plt.ylabel('$-\ln[P_{st}(n)]$')
plt.title('steady state distribution')
plt.legend()
plt.grid()
```





