

```
In [1]: import numpy as np
import matplotlib.pyplot as plt
from scipy import integrate
```

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In [2]: rng = np.random.default_rng()
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In [3]: #Let me write down all the functions first
def f_26(x):
    return (x**2 + x)/np.sqrt(x) * np.exp(-x)

def f_27_a(x):
    return np.exp(np.sin(2*x))

def f_27_b(x):
    return 1 / (2 + np.cos(x))

def f_27_c(x):
    return np.exp(-x**2 /2)

def f_28(x):
    return np.sqrt(x)*np.cos(x)
```

```
In [4]: #7.26
#Monte Carlo Integration as provided in class

x_samp = np.linspace(0, 10, 2**7)

def p(x):
    return np.exp(-x)

def cumdist(f,x):
    return np.cumsum(f(x)) / np.sum(f(x))

from scipy import interpolate
inverse_cumdist = interpolate.interp1d(cumdist(p,x_samp), x_samp)

N = 200_000
xs = rng.uniform(min(cumdist(p,x_samp)), max(cumdist(p,x_samp)), size = N )

def z(x):
    return (x**2 + x) / np.sqrt(x)

print("Monte Carlo Integration: ", np.mean(z(inverse_cumdist(xs)))) 

def F(x):
    return (x**2 + x) / np.sqrt(x) * np.exp(-x)

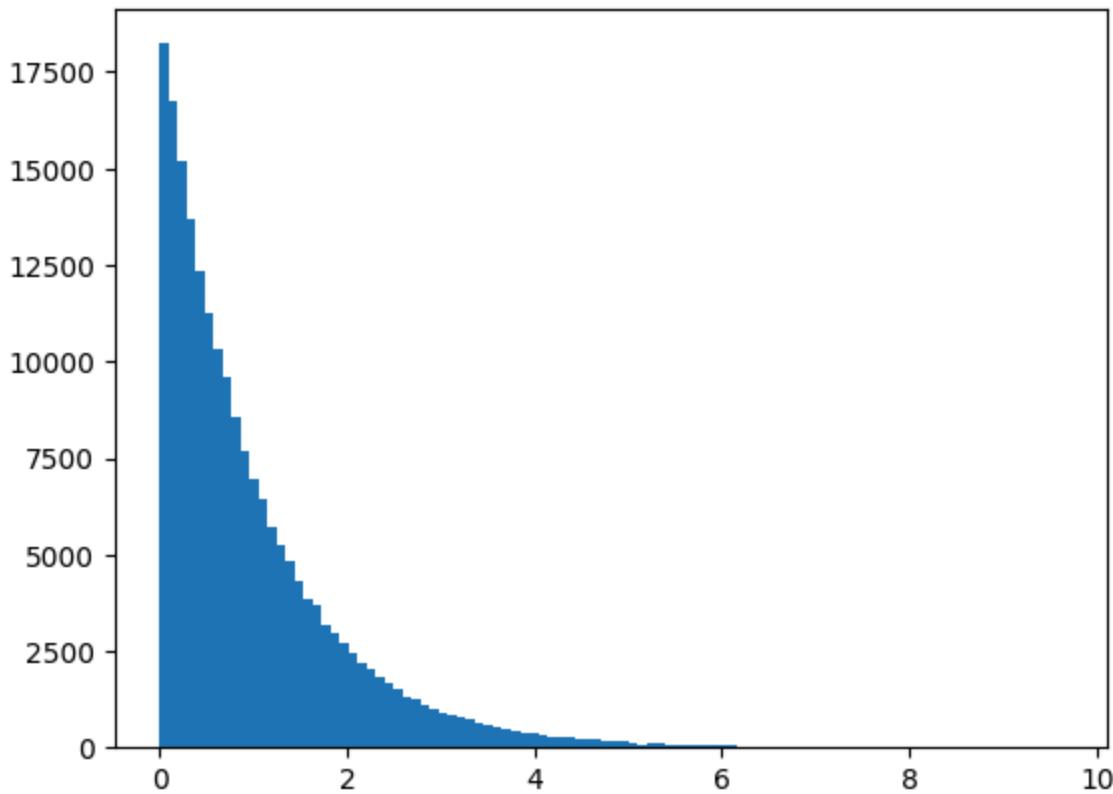
quad_out, err = integrate.quad(F, min(x_samp), max(x_samp))
print("Scipy Integration: ", quad_out)

#Visualization
fig, axs = plt.subplots()
plt.hist(inverse_cumdist(xs), bins = 100)
```

Monte Carlo Integration: 2.216456096594415

Scipy Integration: 2.2137555660034107

```
Out[4]: (array([1.8235e+04, 1.6731e+04, 1.5186e+04, 1.3686e+04, 1.2328e+04,
   1.1237e+04, 1.0310e+04, 9.5840e+03, 8.5870e+03, 7.6750e+03,
   6.9520e+03, 6.4140e+03, 5.7130e+03, 5.2680e+03, 4.8580e+03,
   4.2950e+03, 3.8420e+03, 3.6760e+03, 3.1760e+03, 2.9750e+03,
   2.7160e+03, 2.4320e+03, 2.1850e+03, 2.0210e+03, 1.8370e+03,
   1.6730e+03, 1.4990e+03, 1.3100e+03, 1.2410e+03, 1.1240e+03,
   1.0150e+03, 9.0200e+02, 8.2300e+02, 7.9500e+02, 7.4900e+02,
   6.4300e+02, 5.9000e+02, 5.1500e+02, 4.7500e+02, 4.3500e+02,
   3.7600e+02, 3.7000e+02, 3.2200e+02, 2.8100e+02, 2.7600e+02,
   2.4800e+02, 2.3600e+02, 2.1100e+02, 1.9800e+02, 1.7800e+02,
   1.4600e+02, 1.5000e+02, 1.3600e+02, 8.1000e+01, 9.3000e+01,
   1.0600e+02, 6.8000e+01, 7.9000e+01, 7.9000e+01, 6.3000e+01,
   4.9000e+01, 5.3000e+01, 4.2000e+01, 4.8000e+01, 3.8000e+01,
   3.1000e+01, 2.6000e+01, 2.5000e+01, 2.1000e+01, 3.2000e+01,
   2.0000e+01, 2.1000e+01, 2.9000e+01, 1.2000e+01, 1.4000e+01,
   1.7000e+01, 1.4000e+01, 8.0000e+00, 7.0000e+00, 7.0000e+00,
   9.0000e+00, 6.0000e+00, 5.0000e+00, 7.0000e+00, 5.0000e+00,
   4.0000e+00, 4.0000e+00, 6.0000e+00, 9.0000e+00, 2.0000e+00,
   3.0000e+00, 5.0000e+00, 0.0000e+00, 2.0000e+00, 6.0000e+00,
   3.0000e+00, 1.0000e+00, 2.0000e+00, 0.0000e+00, 2.0000e+00]),
 array([2.80700880e-07, 9.61587156e-02, 1.92317150e-01, 2.88475585e-01,
   3.84634020e-01, 4.80792455e-01, 5.76950890e-01, 6.73109325e-01,
   7.69267760e-01, 8.65426194e-01, 9.61584629e-01, 1.05774306e+00,
   1.15390150e+00, 1.25005993e+00, 1.34621837e+00, 1.44237680e+00,
   1.53853524e+00, 1.63469367e+00, 1.73085211e+00, 1.82701054e+00,
   1.92316898e+00, 2.01932741e+00, 2.11548585e+00, 2.21164428e+00,
   2.30780272e+00, 2.40396115e+00, 2.50011959e+00, 2.59627802e+00,
   2.69243646e+00, 2.78859489e+00, 2.88475333e+00, 2.98091176e+00,
   3.07707020e+00, 3.17322863e+00, 3.26938707e+00, 3.36554550e+00,
   3.46170394e+00, 3.55786237e+00, 3.65402081e+00, 3.75017924e+00,
   3.84633768e+00, 3.94249611e+00, 4.03865454e+00, 4.13481298e+00,
   4.23097141e+00, 4.32712985e+00, 4.42328828e+00, 4.51944672e+00,
   4.61560515e+00, 4.71176359e+00, 4.80792202e+00, 4.90408046e+00,
   5.00023889e+00, 5.09639733e+00, 5.19255576e+00, 5.28871420e+00,
   5.38487263e+00, 5.48103107e+00, 5.57718950e+00, 5.67334794e+00,
   5.76950637e+00, 5.86566481e+00, 5.96182324e+00, 6.05798168e+00,
   6.15414011e+00, 6.25029855e+00, 6.34645698e+00, 6.44261542e+00,
   6.53877385e+00, 6.63493229e+00, 6.73109072e+00, 6.82724916e+00,
   6.92340759e+00, 7.01956603e+00, 7.11572446e+00, 7.21188290e+00,
   7.30804133e+00, 7.40419977e+00, 7.50035820e+00, 7.59651663e+00,
   7.69267507e+00, 7.78883350e+00, 7.88499194e+00, 7.98115037e+00,
   8.07730881e+00, 8.17346724e+00, 8.26962568e+00, 8.36578411e+00,
   8.46194255e+00, 8.55810098e+00, 8.65425942e+00, 8.75041785e+00,
   8.84657629e+00, 8.94273472e+00, 9.03889316e+00, 9.13505159e+00,
   9.23121003e+00, 9.32736846e+00, 9.42352690e+00, 9.51968533e+00,
   9.61584377e+00]),
 <BarContainer object of 100 artists>)
```



```
In [5]: # 7.27
# Monte Carlo Integration function

#Visualization
x_27_a = np.linspace(0, 2*np.pi, 100)
x_27_c = np.linspace(-1, 1, 100)

x_rand_points = rng.uniform(0, 2*np.pi, 200_000)
x_rand_points_c = rng.uniform(-1, 1, 200_000)
y_rand_points = rng.uniform(0, 3, 200_000)

is_inside_a = y_rand_points < f_27_a(x_rand_points)
is_inside_b = y_rand_points < f_27_b(x_rand_points)
is_inside_c = y_rand_points < f_27_c(x_rand_points_c)

box_area = (2*np.pi - 0)*(3 - 0)
box_area_c = (1 - (-1))*(3 - 0)

integral_a = box_area*np.sum(is_inside_a) / N
integral_b = box_area*np.sum(is_inside_b) / N
integral_c = box_area_c*np.sum(is_inside_c) / N

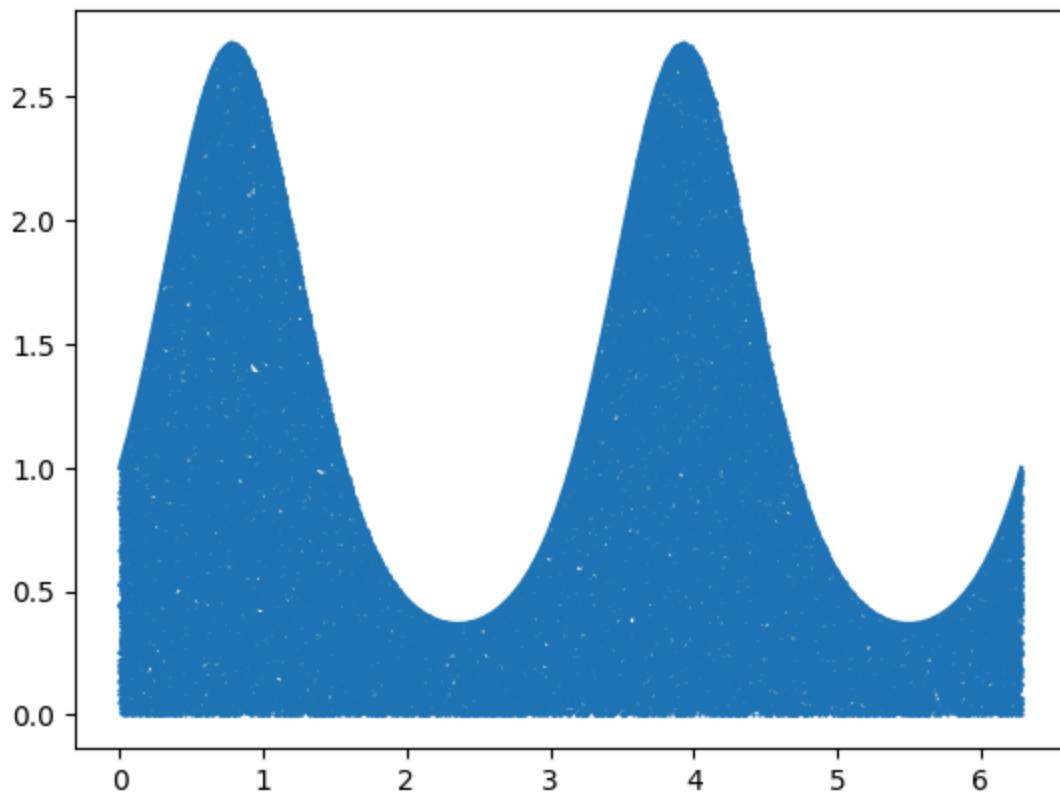
print("7.27 (a): ", "Monte Carlo Integration: ", integral_a, "Scipy Integration: ",
print("7.27 (b): ", "Monte Carlo Integration: ", integral_b, "Scipy Integration: ",
print("7.27 (c): ", "Monte Carlo Integration: ", integral_c, "Scipy Integration: ",

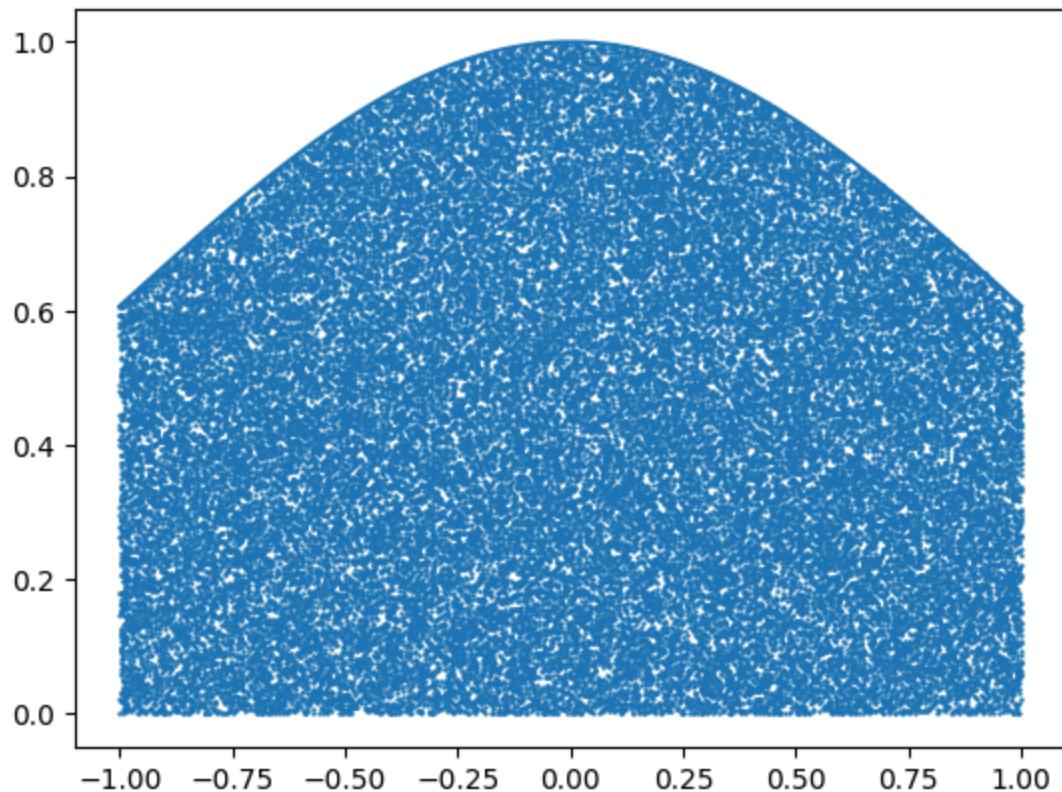
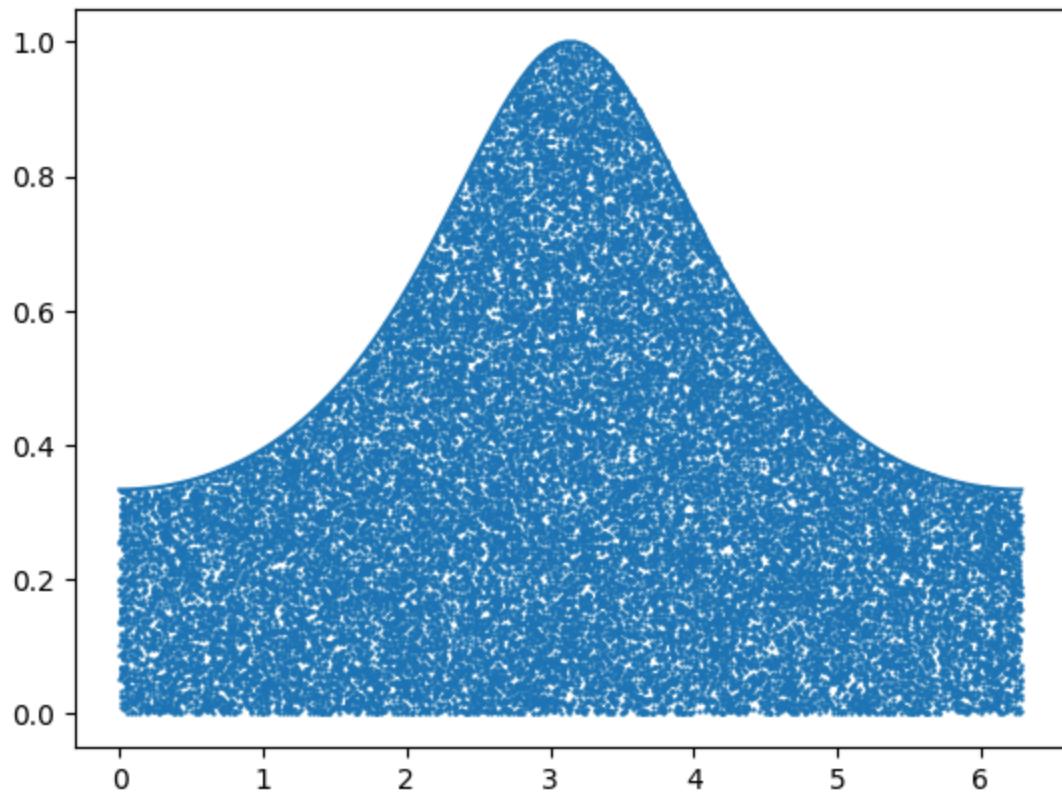
plt.scatter(x_rand_points[is_inside_a], y_rand_points[is_inside_a], s=0.5)
plt.plot(x_27_a, f_27_a(x_27_a))
plt.show()
```

```
plt.scatter(x_rand_points[is_inside_b], y_rand_points[is_inside_b], s=0.5)
plt.plot(x_27_a, f_27_b(x_27_a))
plt.show()

plt.scatter(x_rand_points_c[is_inside_c], y_rand_points[is_inside_c], s=0.5)
plt.plot(x_27_c, f_27_c(x_27_c))
plt.show()
```

7.27 (a): Monte Carlo Integration: 7.945276316487802 Scipy Integration: (7.954926
521012845, 4.052187353026371e-10)
7.27 (b): Monte Carlo Integration: 3.603940844418603 Scipy Integration: (3.627598
7284684352, 7.282921932194597e-09)
7.27 (c): Monte Carlo Integration: 1.6989 Scipy Integration: (1.7112487837842973,
1.8998678006191232e-14)





```
In [6]: x = np.linspace(0, np.pi, 1000)
plt.plot(x, f_28(x))

Nsamp = 200_000
x_i = rng.uniform(0, np.pi, Nsamp)
```

```
y_i = rng.uniform(-2, 1, Nsamp)

is_inside = y_i < f_28(x_i)
is_pos = y_i > 0
is_neg = y_i < 0

plt.scatter(x_i[is_inside*is_pos], y_i[is_inside*is_pos], c="blue", s=0.5)
plt.scatter(x_i[~is_inside*is_neg], y_i[~is_inside*is_neg], c="red", s=0.5)

plt.axhline(0, linestyle="--")

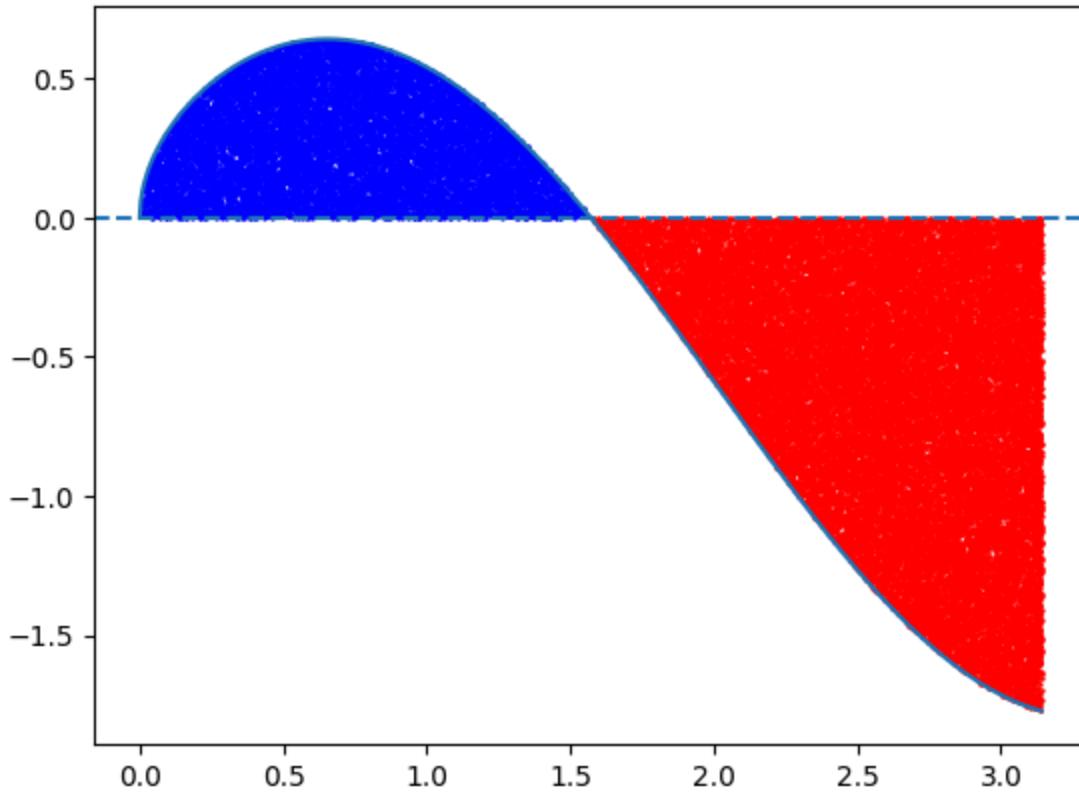
total_area = (np.pi - 0) * (1 - (-2))
print("Scipy Integration: ", integrate.quad(f_28, 0, np.pi))

# number of positive and negative samples that are under their respective parts
N_pos = np.sum(is_inside & is_pos)
N_neg = np.sum(~is_inside) & is_neg

integral = total_area * (N_pos - N_neg) / Nsamp
print("Monte Carlo Integration: ", integral)
```

Scipy Integration: (-0.8948314694841447, 5.742084585591556e-11)

Monte Carlo Integration: -0.9044959408950374



In []: