

102_hw2

October 4, 2022

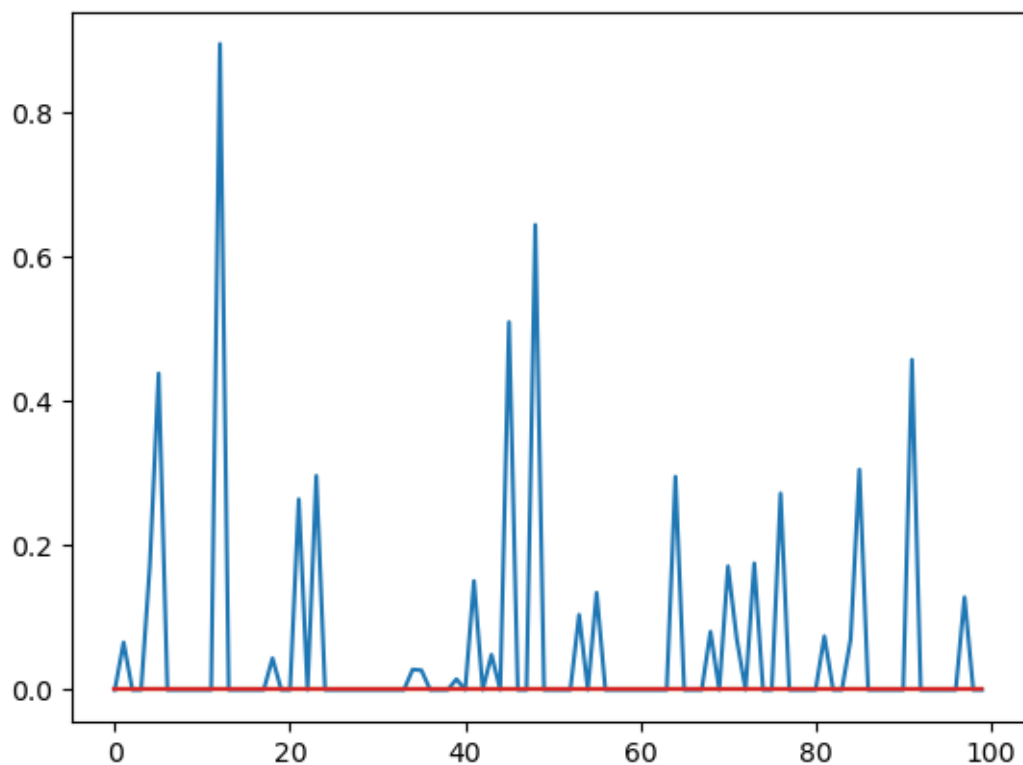
```
[155]: import numpy as np
      %matplotlib inline
      import matplotlib.pyplot as plt
      import numpy as np
      import pandas as pd
      import seaborn as sns
      from scipy.stats import pareto
      import pymc3 as pm
      from scipy import stats
```

0.1 Question 1 The One with all the Beetles

```
[156]: w = 15
      np.random.seed(10)

      y = np.random.uniform(low = 0, high = w , size = 100)
```

```
[157]: alpha = 1
      beta = 10
      days = [1, 10, 50, 100]
      for i in days:
          curr_x = np.random.uniform(low = 0, high = w, size = 100)
          plt.plot(pareto.pdf(x = curr_x, b = beta, scale = beta**alpha))
          beta = max(beta, max(curr_x))
          alpha = alpha + i
```



[]:

0.2 Question 2 Bayesian Fidget Spinners

$$z_i \sim \text{Bernoulli}(\pi)$$

$$q_k \sim \text{Beta}(\alpha_k, \beta_k)$$

$$x_i | z_i, q_0, q_1 \sim \text{Geometric}(q_{z_i})$$

0.2.1 b)

- i) $\pi < \frac{1}{2}$, indicating Nat thinks factory 0 produces more boxes than factory 1
- ii) Factory 0 has lower rate of defect comparing to factory 1

0.2.2 c)

```
[158]: import numpy as np
np.random.seed(5)

N = 100
pi = 0.45
```

```

q0 = 0.05
q1 = 0.18

n1 = int(N * pi)
n0 = N - n1

y_obs = np.zeros(N)
y_obs[:n0] = np.random.geometric(q0, size=n0)
y_obs[n0:] = np.random.geometric(q1, size=n1)

print(y_obs)

```

```

[ 5. 40.  5. 49. 14. 19. 29. 15.  7.  5.  2. 27. 12.  4. 42.  7. 11.  7.
 20. 17. 18.  7.  7.  6.  8.  4.  4. 65. 63.  5.  1.  5. 24. 30.  1. 17.
  1. 15. 20. 83.  6. 32. 40. 50.  1. 13. 78. 10. 33. 16. 29. 13.  1.  2.
  3.  2. 17.  6.  9.  5.  6.  9. 14. 13.  9.  1.  3.  1.  4.  9.  1.  2.
  3.  2.  8.  3.  6.  4.  5. 15. 15.  1. 10.  3.  1.  3.  4. 17.  3.  1.
 10.  1.  4. 10.  4.  3.  9.  1.  5.  3.]

```

```

[159]: alphas = [1, 5]
betas = [5, 1]
pi = 0.45

with pm.Model() as model:
    z = pm.Bernoulli(p = pi, shape = (100, 2))

    # Hint: you should use the shape= parameter here so that
    # q is a PyMC3 array with both q0 and q1.
    q = pm.Beta("q", alpha = alphas, beta = betas, shape = (100,2))

    # Hint: it may be useful to use "fancy indexing" like we did in class.
    # See below for an example
    X = pm.Geometric("X", p = q)

    trace = ....

# FANCY INDEXING

my_binary_array = np.array([0, 0, 1, 1, 0, 1])
my_real_array = np.array([0.27, 0.34])
print(my_real_array[my_binary_array])

```

Cell In [159], line 16

```
trace = ...
```

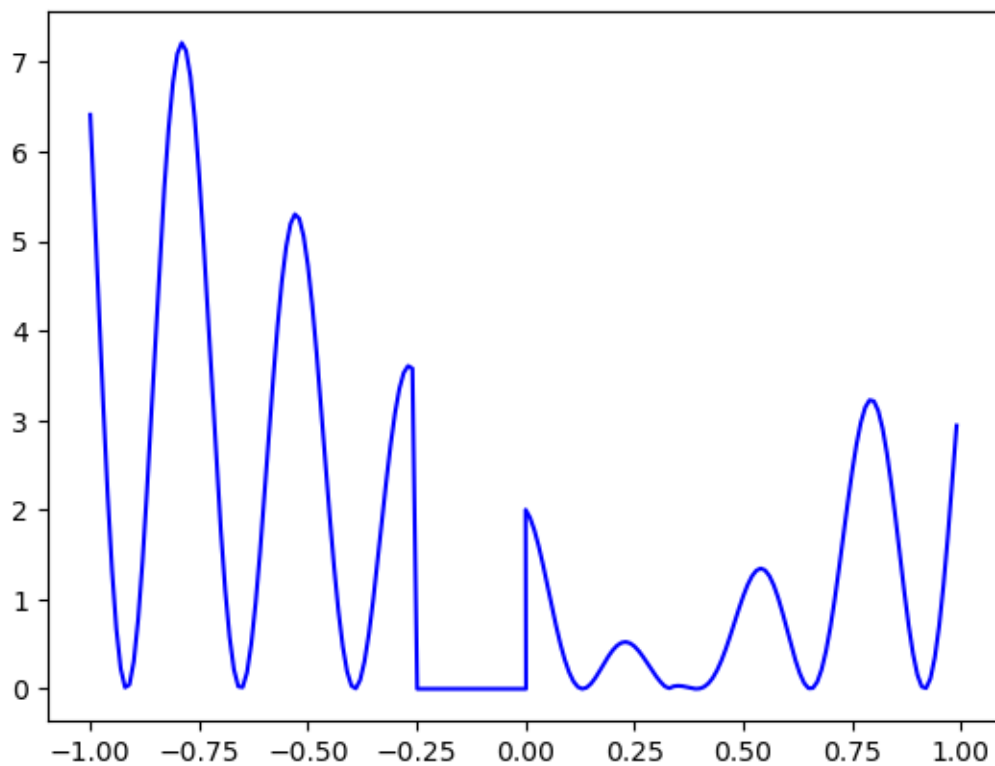
SyntaxError: invalid syntax

0.2.3 Question 3 Rejection Sampling

a) Plot $g(x)$

```
[160]: fig, ax = plt.subplots()
x_1 = np.arange(-1, -0.25, 0.01)
x_2 = np.arange(0, 1, 0.01)
x_3 = np.arange(-0.249, 0.01, 0.01)
g_1 = np.cos(12*x_1)**2 * abs(x_1**3 + 6*x_1 - 2)
g_2 = np.cos(12*x_2)**2 * abs(x_2**3 + 6*x_2 - 2)
g = np.concatenate([g_1, [0]*len(x_3), g_2])
x = np.concatenate([x_1, x_3, x_2])
ax.plot(x, g, 'b')
```

```
[160]: [<matplotlib.lines.Line2D at 0x7fbd82b88ca0>]
```



```
[161]: max(g)
```

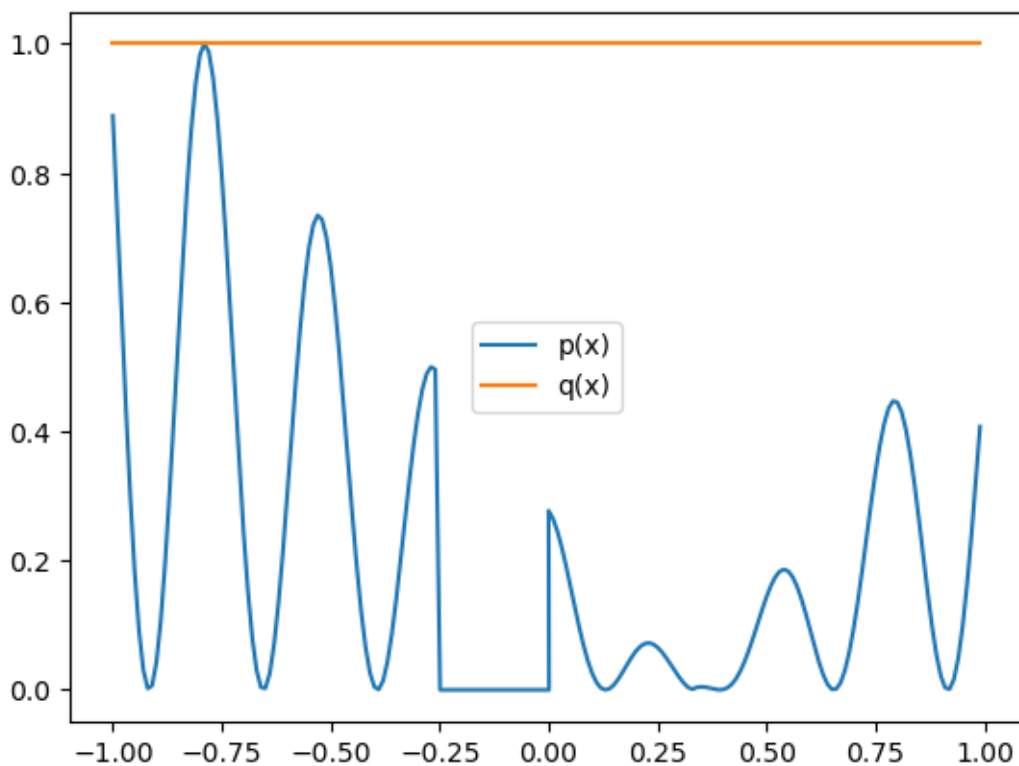
```
[161]: 7.2110044499935935
```

$$q \sim \text{Uniform}(-1, 1), M \leq \frac{1}{\max(g(x))} = \frac{1}{7.211}$$

b) plot proposal $q(x)$ and $p(x)$, estimate \hat{c}

```
[162]: fig, ax = plt.subplots()
plt.plot(x, (1/7.211)*g, label = 'p(x)')
plt.plot(np.arange(-1, 1, 0.01), len(np.arange(-1, 1, 0.01))*[1], label = 'q(x)')
plt.legend()
```

```
[162]: <matplotlib.legend.Legend at 0x7fbd848c7e50>
```



$$\frac{n}{N} \approx \frac{\int_{-1}^1 p(x) dx}{\int_{-1}^1 q(x) dx} = M \int_{-1}^1 g(x) dx = \frac{M}{c} \int f(x) dx = \frac{M}{c}$$

As a result, $\hat{c} = \frac{NM}{n}$

c) Rejection Sampling

```
[181]: def target (x):
return np.cos(12*x)**2 * abs(x**3 + 6*x -2)
```

```
[174]: def uniform_sampling_dist(t):
return stats.uniform.pdf(t, -1, 1)
```

```
[190]: def rejection_sample_uniform(num_samples=1000):
        proposals = np.random.uniform(low=-1, high=1, size=num_samples)
        target = [target(i)*(1/7.211) for i in proposals]

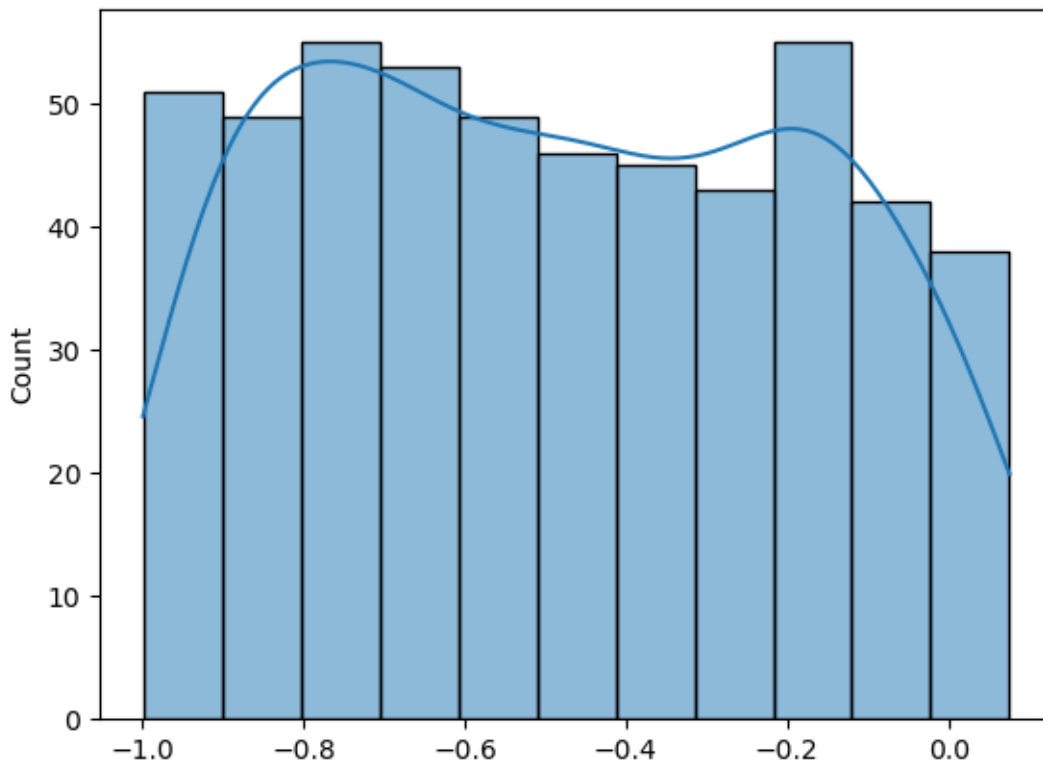
        num_accept = np.sum(accept)
        print('Accepted %d out of %d proposals' % (num_accept, num_samples))
        return proposals[accept]
```

```
[191]: samples = rejection_sample_uniform(num_samples=1000)

        # Plot a true histogram (comparable with density functions) using density=True
        sns.histplot(x = samples, kde=True)
```

Accepted 526 out of 1000 proposals

```
[191]: <AxesSubplot:ylabel='Count'>
```



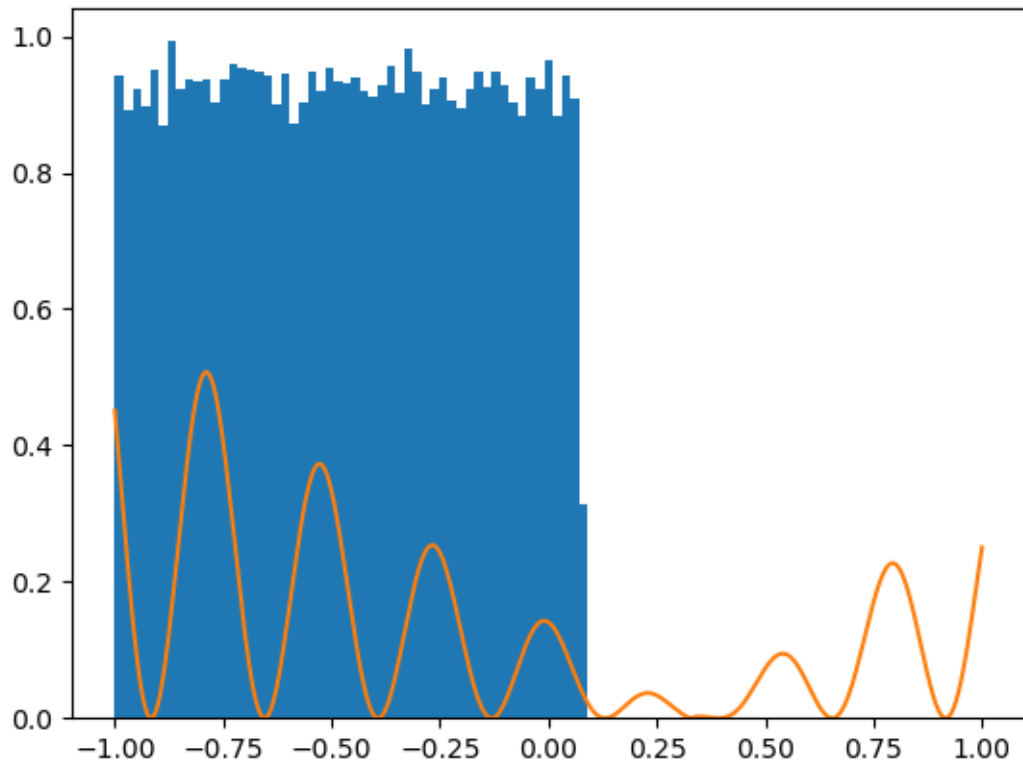
```
[194]: samples = rejection_sample_uniform(num_samples=100000)

        # Plot a true histogram (comparable with density functions) using density=True
        plt.hist(samples, bins=np.linspace(-1, 1, 100), density=True)
        t = np.linspace(-1, 1, 1000)
```

```
# Where did this magic number 0.36 come from? What happens if you change it?
plt.plot(t, target(t)*(1/14.211))
```

Accepted 54069 out of 100000 proposals

```
[194]: [<matplotlib.lines.Line2D at 0x7fbd80c46ac0>]
```



```
[ ]:
```

```
[189]: proposals = np.random.uniform(low=-1, high=1, size=1000)
target_proposals = target(proposals)*(1/7.211)
```

```
[ ]:
```

```
[ ]:
```

```
[109]: target(proposals)
```

ValueError

Traceback (most recent call last)

Cell In [109], line 1

```
----> 1 target(proposals)
```

HW2 Q1

a) Given $x \sim \text{Uniform}(0, w)$:

$$f_1(x|w) = \frac{1}{w} \mathbb{1}[x \leq w]$$

$$\begin{aligned} f_n(x_1, \dots, x_n | w) &= \frac{1}{w^n} \prod_{i=1}^n \mathbb{1}[x_i \leq w] \\ &= \frac{1}{w^n} \mathbb{1}[\max_i x_i \leq w] \end{aligned}$$

$$b) \text{Lik}(w) = \frac{1}{w^n} \mathbb{1}[\max_i x_i \leq w]$$

$$\max_i x_i > w \quad \text{Lik}(x_1, \dots, x_n; w) = 0$$

$$\max_i x_i \leq w \quad \text{Lik}(x_1, \dots, x_n; w) = \frac{1}{w^n} \Rightarrow \text{Mono-decreasing function}$$

So maximum $w = \max_i x_i$

$$c) p'(w | x_1, \dots, x_n) = \frac{f_n(x_1, \dots, x_n | w) p(w)}{\int_{w'} f_n(x_1, \dots, x_n | w') p(w') dw'}$$

$$\begin{aligned} \text{Numerator: } f_n(x_1, \dots, x_n | w) p(w) &= \frac{1}{w^n} \mathbb{1}[\max_i x_i \leq w] \cdot \frac{\alpha \beta^\alpha}{w^{\alpha+1}} \mathbb{1}(w > \beta) \\ &= \frac{\alpha \beta^\alpha}{w^{n+\alpha+1}} \mathbb{1}[\max_i x_i \leq w] \mathbb{1}(w > \beta) \\ &= \frac{\alpha \beta^\alpha}{w^{n+\alpha+1}} \mathbb{1}[\max_i w \geq \max(x_i, \beta)] \end{aligned}$$

$$\begin{aligned} \int_{w'} f_n(x_1, \dots, x_n | w') p(w') dw' &= \int_0^\infty \frac{\alpha \beta^\alpha}{w'^{n+\alpha+1}} \mathbb{1}[w' \geq \max[\max_i x_i, \beta]] dw' \\ &\stackrel{\alpha \beta^\alpha \text{ is constant}}{\Rightarrow} \alpha \beta^\alpha \int \frac{\mathbb{1}[w' \geq \max[\max_i x_i, \beta]]}{w'^{n+\alpha+1}} dw' \end{aligned}$$

$$\text{Let } m = \max(\max_i x_i, \beta) \Rightarrow \alpha \beta^\alpha \left[\frac{w'^{-\alpha-n}}{-\alpha-n} \right] \bigg|_m^\infty = \alpha \beta^\alpha \frac{m^{-\alpha-n}}{\alpha+n}$$

$$P'(w | x_1 \dots x_n) = \frac{\frac{\alpha \beta^\alpha}{w^{\alpha+1}} m}{\frac{\alpha \beta^\alpha}{w^{\alpha+1}} \frac{m^{-\alpha-n}}{\alpha+n}} = \frac{(\alpha+n) m^{\alpha+n}}{w^{\alpha+n+1}} \mathbb{1}[w > m]$$

set $\alpha+n = \delta \Rightarrow \frac{\delta m^\delta}{w^{\delta+1}} \mathbb{1}[w > m]$

$\hookrightarrow \text{Pareto}(\delta, m)$

d) prior: $w \sim \text{Pareto}(\alpha, \beta)$

posterior $\sim \text{Pareto}(\delta, m)$

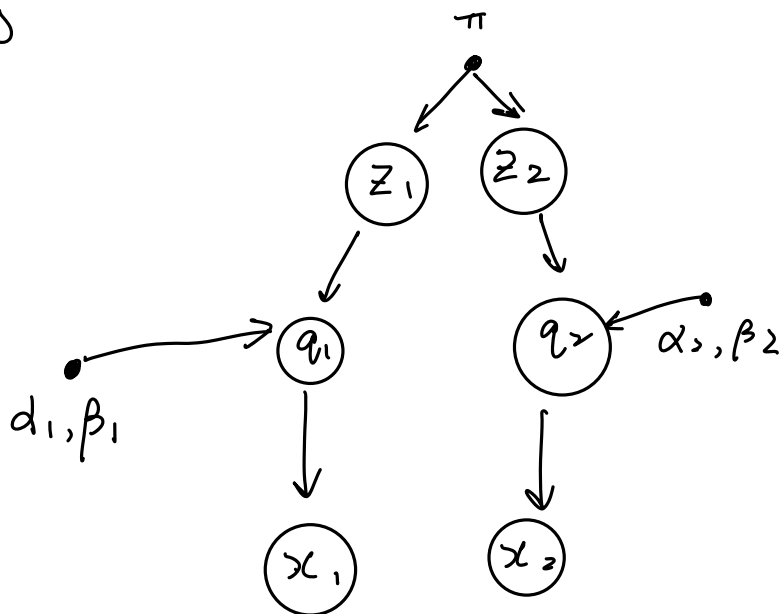
where $\delta = \alpha+n$ $m = \max(\max_i x_i, \beta)$

α : track # of samples seen

m : reward/update maximum length seen

Question 2

a)



c) i)

$$P[\text{Bin}(50, 0.45) \leq 50]$$

$$\Rightarrow \sum_{i=0}^{50} \binom{100}{i} (0.45)^i (0.55)^{100-i} = 0.87$$

d)

$$q_0 \sim \text{Beta}(\alpha_0 + n_0, \beta_0 - n_0 + \sum_{i: z_i=0} x_i)$$

$$q_1 \sim \text{Beta}(\alpha_1 + n_1, \beta_1 - n_1 + \sum_{i: z_i=1} x_i)$$

$$n_0 = \sum_{i=0}^N \mathbb{1}_{z_i=0}$$

$$n_1 = \sum_{i=0}^N \mathbb{1}_{z_i=1}$$