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In [ ]: import math
import matplotlib.pyplot as plt
import numpy as np
from datetime import datetime
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In [ ]: SAMPLE_MIN = 0
SAMPLE_MAX = 15
SAMPLE_RANGE = (SAMPLE_MIN, SAMPLE_MAX)
NORMAL_PARAMS = (5,4)
num_samples = 30
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In [ ]: def normal_dist(x: float, mean: float, deviation: float):

    return (1 / (math.sqrt(2 * math.pi) * deviation)) * math.exp(
        -0.5 * ((x - mean) ** 2) / (deviation**2))
    )

def p(x: int):
    return (
        0.3 * normal_dist(x, 2, 1)
        + 0.4 * normal_dist(x, 5, 2)
        + 0.3 * normal_dist(x, 9, 1)
    )

def q(x: int):
    return normal_dist(x,*NORMAL_PARAMS)
```

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In [ ]:
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In [ ]: def gen_normal_samples(k=num_samples, seed=0):
    rng = np.random.default_rng(seed)

    return rng.normal(*NORMAL_PARAMS,k)

def gen_uniform_samples(sample_max: int=SAMPLE_MAX, k=num_samples, seed=0):
    rng = np.random.default_rng(seed)

    return rng.random(k)*sample_max
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In [ ]: from numpy._typing import NDArray

def resample(samples: NDArray[np.float64], k=num_samples, p_func=p, q_func=q):
    target_density = np.array([p_func(i) for i in samples]) # np arrays allow division
    proposed_density = np.array([q_func(i) for i in samples]) # np arrays allow division
    weights = target_density / proposed_density
    normalized_weights = weights / sum(weights)

    return [samples[resampled_idx] for resampled_idx in np.random.choice(np.arange(k), k, p
```

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In [ ]: def print_distr(samples, normalize: bool = False):
    int_samples = [round(v) for v in samples]

    distr_dict = {v:0. for v in range(min(int_samples),max(int_samples)+1)}
    for sample in int_samples:
        distr_dict.setdefault(sample, 0)
        distr_dict[sample] += 1

    print(distr_dict)
    if normalize:
        total = sum(distr_dict.values())
        for key in distr_dict:
            distr_dict[key] /= total
```

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    print(distr_dict)

    return distr_dict

```

```

In [ ]: x = np.arange(-5,15,0.01)
        y = [p(idx) for idx in x]
        plt.plot(x, y)

        rng = np.random.default_rng(0)

        nx = np.arange(-5,15,0.01)
        ny = [normal_dist(idx, *NORMAL_PARAMS) for idx in nx]

        # plt.plot(nx,ny)

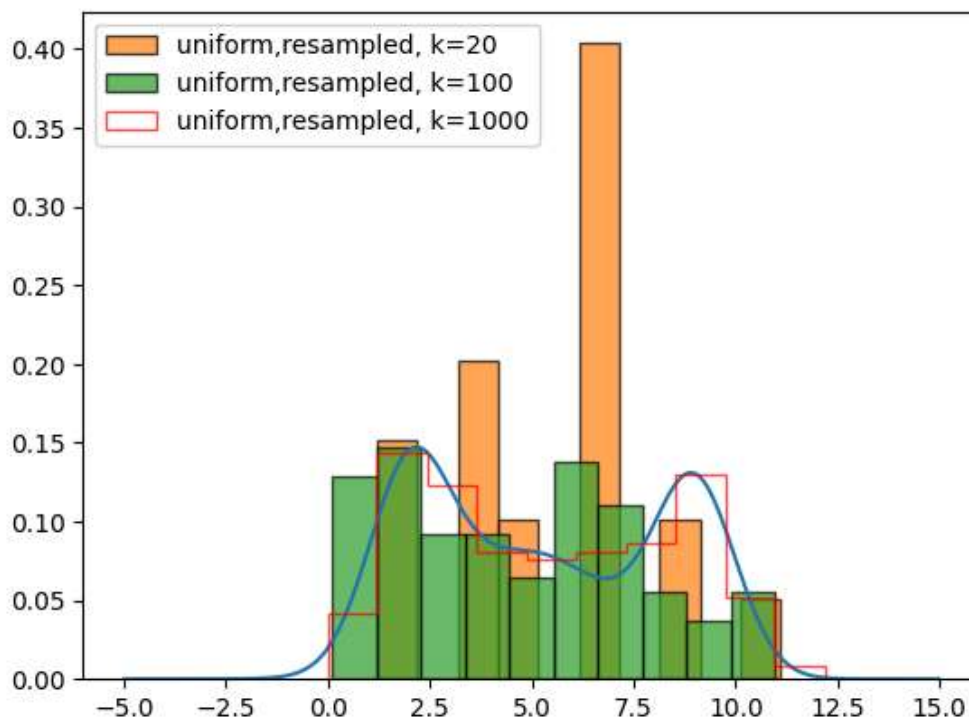
        def uni_q(x):
            return 1/15 if x >= 0 and x <= 15 else 0

        for edgecolor, alpha, histtype, k in zip(("black","black","red"),(1,0.7,1, 1), ("bar","bar"
        samples = gen_uniform_samples(k=k, seed=int(datetime.now().timestamp()))
        resampled = resample(samples, k=k, p_func=p,q_func=uni_q)

        plt.hist(resampled, density=True, label=f"uniform,resampled, k={k}", histtype=histtype,

        plt.legend()
        plt.show()

```



```

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        y = [p(idx) for idx in x]
        plt.plot(x, y)

        rng = np.random.default_rng(0)

        nx = np.arange(-5,15,0.01)
        ny = [normal_dist(idx, *NORMAL_PARAMS) for idx in nx]

        # plt.plot(nx,ny)

        def uni_q(x):

```

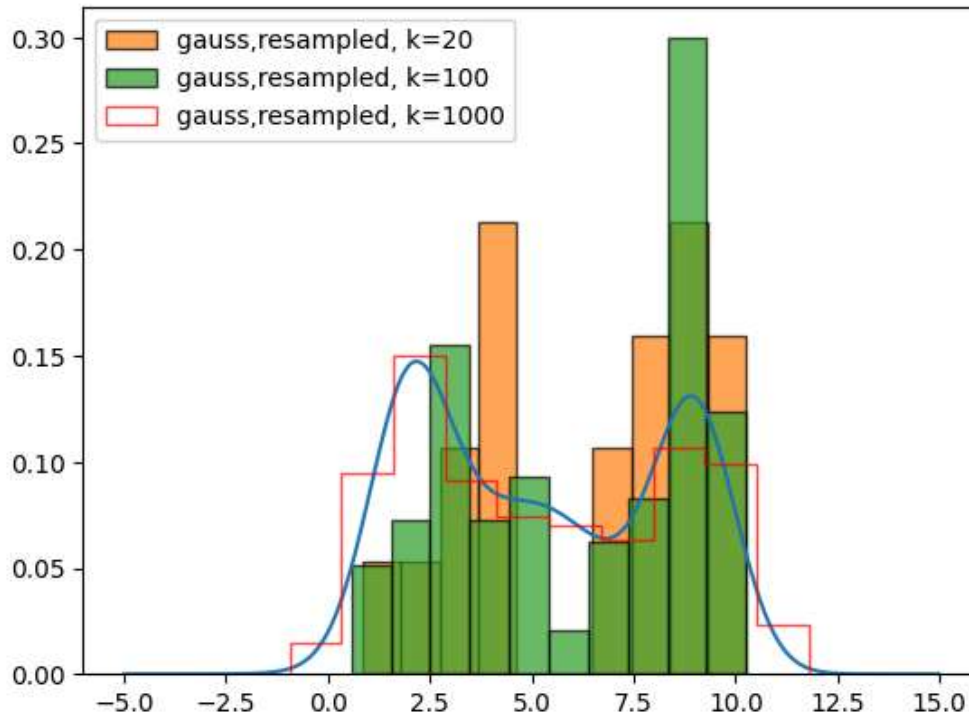
```

return 1/15 if x >= 0 and x <= 15 else 0

for edgecolor, alpha, histtype, k in zip(("black","black","red"),(1,0.7,1, 1), ("bar","bar",
samples = gen_normal_samples(k=k, seed=int(datetime.now().timestamp()))
resampled = resample(samples, k=k)

plt.hist(resampled, density=True, label=f"gauss,resampled, k={k}", histtype=histtype, e
plt.legend()
plt.show()

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The red line for $k = 1000$, fits the best. Keeping the height of the peaks for p and the resampled one.

The green for $k = 100$ shows that the resampled density is located somewhat at the peaks for p .

The yellow for $k = 20$, shows somewhat the same as $k = 1000$, but understandably since q has its mean to be 5, it makes sense that the resampled samples for small k will tend to go a bit to the right with a tendency to be around 5.