```
In [ ]: import math
        import matplotlib.pyplot as plt
        import numpy as np
        from datetime import datetime
In [ ]: SAMPLE_MIN = 0
        SAMPLE_MAX = 15
        SAMPLE_RANGE = (SAMPLE_MIN, SAMPLE_MAX)
        NORMAL_PARAMS = (5,4)
        num samples = 30
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)
In [ ]:
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
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
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
```

```
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]

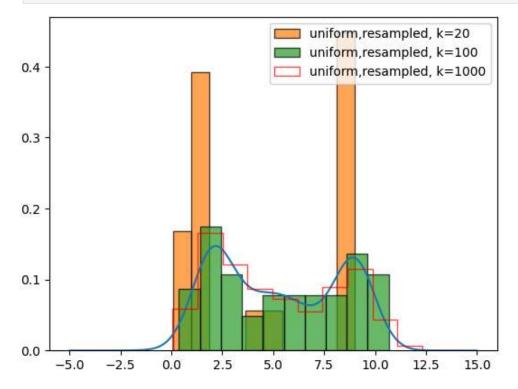
# plt.plot(nx,ny)

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()</pre>
```



The red line for k = 1000, fits the best, but we see that the green for k = 100 also fits similar to p in the peaks, but does not quite follow the middle of p around 4-6.

We see that for k = 20 with the orange has the peaks around where p has its peaks, but since there are so few samples, they are more grouped together around the peaks and very few if any are between the peaks.

```
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)
```

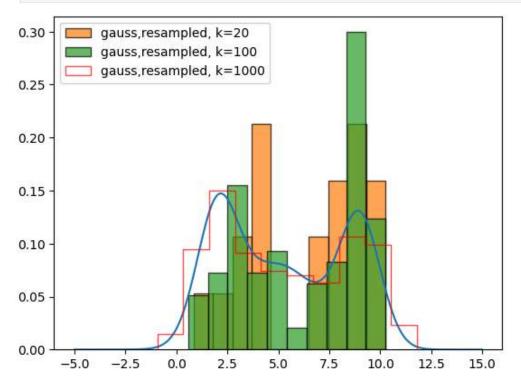
```
nx = np.arange(-5,15,0.01)
ny = [normal_dist(idx, *NORMAL_PARAMS) for idx in nx]

# 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_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, ell.legend()
plt.show()</pre>
```



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.