```
In [1]: import numpy as np
        import math
        import matplotlib.pyplot as plt
        from scipy.spatial.distance import cdist as cd
In [2]: pos = np.arange(-5.0, 5, 0.1)
```

Kernel Density Estimation Using Gaussian Kernel

```
In [3]: def kernel(x, xi, h):
             return 1/(2*math.pi*h**2)*math.exp((-(x-xi)**2)/(2*h**2))
In [4]: def kde(samples, h):
            size = np.size(samples)
             density = np.zeros(size)
             for i in range(0, size):
                for j in range(0, size):
                     x += kernel(samples[i], samples[j], h)
                density[i] = x/size
             return density
```

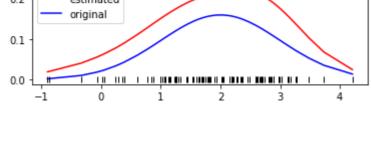
Visualization

```
In [5]: print(kde(pos, 0.2))
        [0.11962994 0.15474337 0.17887646 0.19179397 0.19717879 0.19892699
         0.199369 0.19945604 0.19946939 0.19947098 0.19947113 0.19947114
         0.19947114 0.19947114 0.19947114 0.19947114 0.19947114 0.19947114
         0.19947114 0.19947114 0.19947114 0.19947114 0.19947114 0.19947114
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         0.19947114 0.19947114 0.19947114 0.19947114 0.19947114 0.19947113
         0.19947098 0.19946939 0.19945604 0.199369 0.19892699 0.19717879
         0.19179397 0.17887646 0.15474337 0.11962994]
```

```
In [6]: plt.plot(pos, kde(pos, 0.3))
          plt.show()
          0.13
          0.12
          0.11
          0.10
           0.09
          0.08
```

Testing on data obtained using Normal Distribution

```
In [7]: rand 1 = np.random.normal(2,1,100)
        rand 1 = np.sort(rand 1)
        probability_density = np.zeros(np.size(rand_1))
        for i in range(0,np.size(rand_1)):
            probability_density[i] = kernel(2,rand_1[i],1)
        plt.subplot(2,1,1)
        a = plt.plot(rand 1, kde(rand 1, 0.6), 'r', label='estimated')
        b = plt.plot(rand_1,probability_density,'b',label='original')
        plt.plot(rand_1, np.zeros(np.size(rand_1)), '|k',markeredgewidth=1)
        plt.legend()
        plt.show()
                 estimated
```



In [8]: def distance(a,b):

KNN

```
return abs (a-b)
In [9]: def knn(samples, k):
            samples = np.sort(samples)
            size = np.size(samples)
            density = np.zeros(size)
            for i in range(0, size):
                distance to k = 0
                distance_vector= np.zeros(size)
                for j in range(0, size):
                    distance_vector[j] = distance(samples[i], samples[j])
                    arg distance = distance vector.argsort()
                    temp2 = arg_distance[k-1]
                    distance_to_k = distance(samples[i], samples[temp2])
                    density[i] = k
                    density[i]/=size
                    density[i]/=((4/3)*math.pi*distance to k**3)
            return density
```

Visualization

```
In [10]: print(knn(pos, 4))
         [0.35367765 1.19366207 1.19366207 1.19366207 1.19366207 1.19366207
          1.19366207 1.19366207 1.19366207 1.19366207 1.19366207 1.19366207
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          1.19366207 1.19366207 1.19366207 1.19366207 1.19366207
          1.19366207 1.19366207 1.19366207 1.19366207 1.19366207 1.19366207
          1.19366207 1.19366207 1.19366207 0.35367765]
         /Library/Frameworks/Python.framework/Versions/3.4/lib/python3.4/site-packages/ipykernel_launche
         r.py:15: RuntimeWarning: divide by zero encountered in double scalars
           from ipykernel import kernelapp as app
In [11]: plt.plot(pos, knn(pos, 10))
         plt.show()
         /Library/Frameworks/Python.framework/Versions/3.4/lib/python3.4/site-packages/ipykernel launche
         r.py:15: RuntimeWarning: divide by zero encountered in double_scalars
           from ipykernel import kernelapp as app
          0.18
          0.16
          0.14
          0.12
          0.10
          0.08
          0.06
          0.04
                  <u>-</u>4
```

In [12]: rand 1 = np.random.normal(2, 1, 500)rand 1 = np.sort(rand 1)probability_density = np.zeros(np.size(rand_1))

for i in range(0,np.size(rand 1)):

k = 50 # knn neighbors

0.4

0.2

probability density[i] = kernel(2, rand 1[i], 1)

Testing on data obtained using Normal Distribution

```
plt.subplot(2,1,1)
         plt.plot(rand 1, probability density, 'b', label='original')
         plt.plot(rand 1, knn(rand 1, 350), 'r', label='estimated')
         plt.plot(rand_1, np.zeros(np.size(rand_1)), '|k',markeredgewidth=1)
         plt.legend()
         plt.show()
         /Library/Frameworks/Python.framework/Versions/3.4/lib/python3.4/site-packages/ipykernel launche
         r.py:15: RuntimeWarning: divide by zero encountered in double scalars
           from ipykernel import kernelapp as app
          0.15
                                                 original
                                                 estimated
          0.10
          0.05
                                                 Apply method
In [13]: def parameters():
```

```
h = 0.6 # kde windowsize / radius
   return h, k
def gauss1D(m, v, N, w):
   pos = np.arange(-w, w - w / N, 2 * w / N)
   insE = -0.5 * ((pos - m) / v) ** 2
   norm = 1 / (v * np.sqrt(2 * np.pi))
   res = norm * np.exp(insE)
   realDensity = np.stack((pos, res), axis=1)
   return realDensity
h, k = parameters()
print('Question: Kernel/K-Nearest Neighborhood Density Estimators')
# Produce the random samples
samples = np.random.normal(0, 1, 100)
samples = np.sort(samples)
# Compute the original normal distribution
realDensity = gauss1D(0, 1, 100, 5)
# Estimate the probability density using the KDE
estDensity = kde(samples, h)
# plot results
plt.subplot(2, 1, 1)
plt.plot(samples, estDensity, 'r', linewidth=1.5, label='KDE Estimated Distribution')
plt.plot(realDensity[:, 0], realDensity[:, 1], 'b', linewidth=1.5, label='Real Distribution')
plt.legend()
# Estimate the probability density using KNN
estDensity = knn(samples, k)
# Plot the distributions
plt.subplot(2, 1, 2)
plt.plot(samples, estDensity, 'r', linewidth=1.5, label='KNN Estimated Distribution')
plt.plot(realDensity[:, 0], realDensity[:, 1], 'b', linewidth=1.5, label='Real Distribution')
plt.legend()
plt.show()
Question: Kernel/K-Nearest Neighborhood Density Estimators
/Library/Frameworks/Python.framework/Versions/3.4/lib/python3.4/site-packages/ipykernel launche
r.py:15: RuntimeWarning: divide by zero encountered in double scalars
 from ipykernel import kernelapp as app
0.4

    KDE Estimated Distribution

       Real Distribution
0.2
```

KNN Estimated Distribution

Real Distribution

<u>-2</u>