# Assignment06

May 9, 2019

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[K-means clustering on the spatial domain]

Apply K-means algorithm to the regular grid of a spatial domain in two dimension with varying number of clusters.

The spatial domain can be represented by two matrices where one matrix represents the horizontal index and the other matrix represents the vertical index.

Define a distance between each spatial point  $(x_i, y_i)$  and a centroid  $(c_x^k, c_y^k)$  for cluster k using L2-norm square and L1-norm.

Visualize the result using color coding scheme that distinguishes different clusters.

Observe the trajectory of centroid during the optimization and the shape of the clusters depending on the distance.

### 1 Initial array

```
In [1]: import matplotlib.pyplot as plt
        import numpy as np
        import random
        import statistics
        height = 70
        width = 70
        im_array = np.zeros((height, width), dtype = float)
        x = np.zeros((height, width), dtype = int)
        y = np.zeros((height, width), dtype = int)
        for i in range(height):
            for j in range(width):
                x[i][j] = i
                y[i][j] = j
        im_label1 = np.zeros((height, width), dtype = int)
        im_label2 = np.zeros((height, width), dtype = int)
        E1 = []
        E2 = []
```

#### 2 Nomalization

## 3 Compute distance based on L2-norm (x, y)

$$||x - y||_2 = \sqrt{x^2 + y^2}$$

## 4 Compute distance based on L1-norm (x, y)

#### 5 Initialize Label

initialize all images with random label

#### 6 Initialize Centroid

```
im_count[im_label[i,j]] += 1

for i in range(cluster_num):
    im_average_x[i] /= im_count[i]
    im_average_y[i] /= im_count[i]

return im_average_x, im_average_y, im_count
```

### 7 Plot the Final Image

```
In [7]: def plot_average(im_label1, im_label2, cluster_num, x1, y1, x2, y2, iter_num):
            plt.subplot(121)
            plt.title("L1-norm Image" + str(iter_num))
            plt.imshow(im_label1)
            for i in range(cluster_num):
                plt.plot(y1[i], x1[i],'r')
                plt.plot([y1[i][(len(y1[i])-1)]], [x1[i][(len(x1[i])-1)]], '*r')
            plt.axis('off')
            plt.subplot(122)
            plt.title("L2-norm square Image" + str(iter_num))
            plt.imshow(im_label2)
            for i in range(cluster_num):
                plt.plot(y2[i], x2[i],'r')
                plt.plot([y2[i][(len(y2[i])-1)]], [x2[i][(len(x2[i])-1)]], '*r')
            plt.axis('off')
            plt.show()
```

# 8 Clustering

```
In [8]: def clustering(im_label1, im_label2, cluster_num, im_average_x1, im_average_y1, im_average_y1, im_average_x1, im_average_y1, im_average_x1, im_avera
                                                                dist1 = []
                                                                 dist2 = []
                                                                 for i in range(height):
                                                                                       for j in range(width):
                                                                                                             dist1.clear()
                                                                                                             dist2.clear()
                                                                                                             for k in range(cluster_num):
                                                                                                                                  nx = x[i][j]
                                                                                                                                  ny = y[i][j]
                                                                                                                                   cx1 = im_average_x1[k]
                                                                                                                                   cy1 = im_average_y1[k]
                                                                                                                                   cx2 = im_average_x2[k]
                                                                                                                                   cy2 = im_average_y2[k]
                                                                                                                                   dist2.append(distance(nx, cx2) + distance(ny, cy2))
                                                                                                                                   dist1.append(L1_distance(nx, cx1) + L1_distance(ny, cy1))
```

```
im_label2[i][j] = np.argmin(dist2)
im_label1[i][j] = np.argmin(dist1)
return im_label1, im_label2
```

# 9 Compute Centroid

```
In [9]: def computeCentroid(cluster_num, im_label1, im_label2):
            label_x = []
            label_y = []
            num1 = np.zeros(cluster_num, dtype = int)
            num2 = np.zeros(cluster_num, dtype = int)
            cx1 = np.zeros(cluster_num, dtype = float)
            cy1 = np.zeros(cluster_num, dtype = float)
            cx2 = np.zeros(cluster num, dtype = float)
            cy2 = np.zeros(cluster_num, dtype = float)
            for k in range(cluster_num):
                label_x.append([])
                label_y.append([])
            for i in range(height):
                for j in range(width):
                    cx2[im_label2[i][j]] += x[i][j]
                    cy2[im_label2[i][j]] += y[i][j]
                    num2[im\_label2[i][j]] += 1
                    label_x[im_label1[i][j]].append(x[i][j])
                    label_y[im_label1[i][j]].append(y[i][j])
                    num1[im_label1[i][j]] += 1
            for k in range(cluster_num):
                if num2[k] != 0:
                    cx2[k] /= num2[k]
                    cy2[k] /= num2[k]
                if num1[k] != 0:
                    cx1[k] = statistics.median(label_x[k])
                    cy1[k] = statistics.median(label_y[k])
            return cx1, cy1, cx2, cy2
```

## 10 Compute Energy

```
E = \frac{1}{n} \sum_{x \in \Omega} ||f(x) - m_c||^2
```

where  $\Omega$  denotes the image domain and the number of pixels  $|\Omega|$  is n, and  $m_c$  denotes the centroid for cluster c that is the cluster label of f(x).

```
In [10]: def computeEnergy(im_label1,im_label2, cluster_num, im_average_x1, im_average_y1, im_itotal_sum1 = 0
    total_sum2 = 0
    num = 0
    for i in range(height):
        for j in range(width):
            total_sum1 += distance(x[i][j], im_average_x1[im_label1[i][j]])
            total_sum1 += distance(y[i][j], im_average_y1[im_label1[i][j]])
            total_sum2 += distance(x[i][j], im_average_x2[im_label2[i][j]])
            total_sum2 += distance(y[i][j], im_average_y2[im_label2[i][j]])
            num += 1

total_sum1 /= num
    total_sum2 /= num
    return total_sum1, total_sum2
```

#### 11 Initialize Cluster

```
In [11]: def initialCluster(cluster_num):
    im_average_x1 = np.zeros(cluster_num, dtype=float)
    im_average_y1 = np.zeros(cluster_num, dtype=float)
    im_count1 = np.zeros(cluster_num, dtype=int)

im_average_x2 = np.zeros(cluster_num, dtype=float)
    im_average_y2 = np.zeros(cluster_num, dtype=float)
    im_count2 = np.zeros(cluster_num, dtype=int)

im_label1, im_label2 = initialLabel(cluster_num)

im_average_x1, im_average_y1, im_count1 = initialCentroid(im_average_x1, im_average_x2, im_average_x2, im_average_y2, im_count2 = initialCentroid(im_average_x2, im_average_x2, im
```

# 12 Clustering until no change

no change means energy is maintained

return im\_average\_x1, im\_average\_y1, im\_count1, im\_average\_x2, im\_average\_y2, im\_co

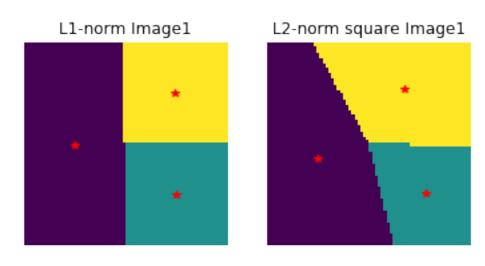
```
im_trajectory_y2 = []
for i in range(cluster_num):
    im_trajectory_x1.append([])
    im_trajectory_y1.append([])
    im_trajectory_x2.append([])
    im_trajectory_y2.append([])
while True:
    im_label1, im_label2 = clustering(im_label1, im_label2, cluster_num, im_avera
    im_average_x1, im_average_y1,im_average_x2, im_average_y2 = computeCentroid(
    for i in range(cluster_num):
        im_trajectory_x1[i].append(im_average_x1[i])
        im_trajectory_y1[i].append(im_average_y1[i])
        im_trajectory_x2[i].append(im_average_x2[i])
        im_trajectory_y2[i].append(im_average_y2[i])
    a,b = computeEnergy(im_label1,im_label2, cluster_num, im_average_x1, im_average
    E1.append(a)
    E2.append(b)
    if iter_num > 0:
        if E2[iter_num -1] - E2[iter_num] < 0.00001:</pre>
            break
    iter_num += 1
    plot_average(im_label1, im_label2, cluster_num, im_trajectory_x1,im_trajectory
return iter_num
```

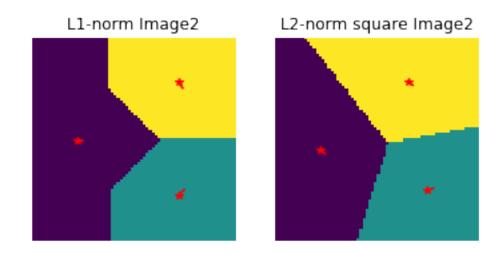
# 13 Energy graph per each iteration

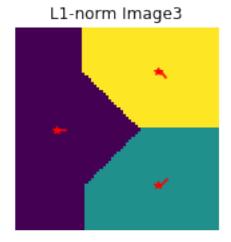
# 14 K = 3

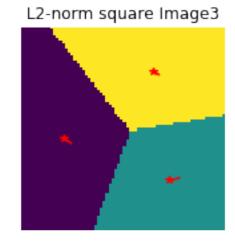
## 14.1 Final average image

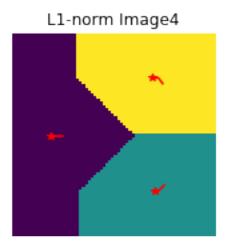
In [14]: im\_average\_x1, im\_average\_y1, im\_count1,im\_average\_x2, im\_average\_y2, im\_count2,im\_la
 iter\_num = iteration(3,im\_average\_x1, im\_average\_y1, im\_count1, im\_average\_x2, im\_average\_x2)

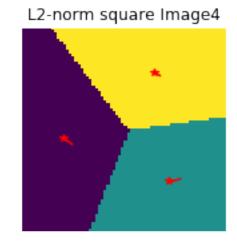


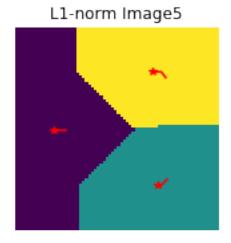


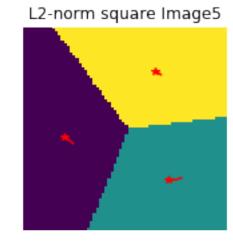


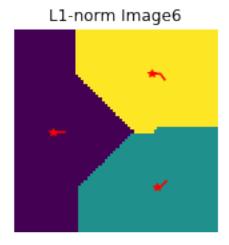


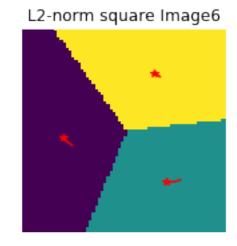


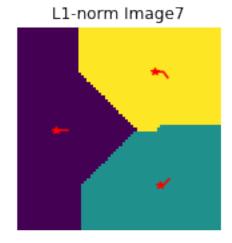


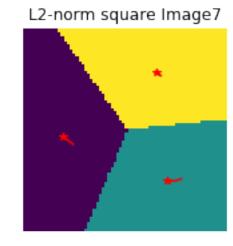


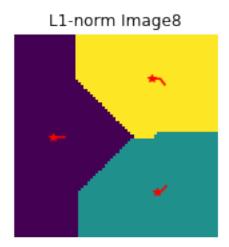


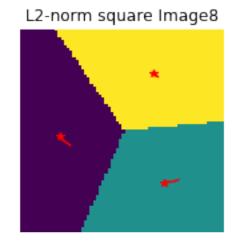


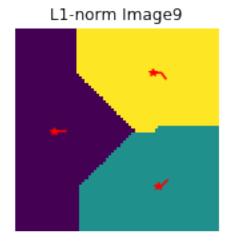


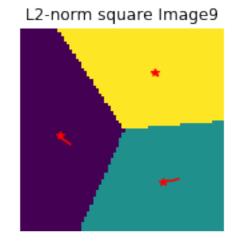


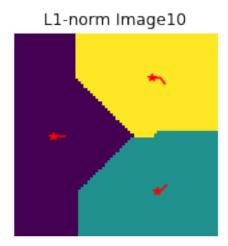


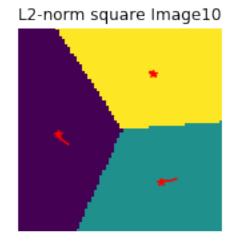


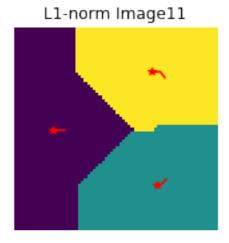


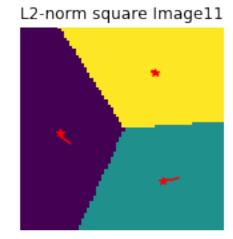


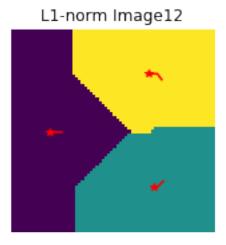


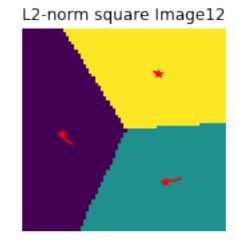


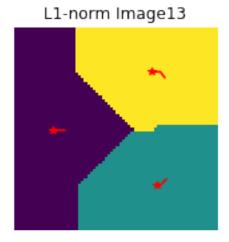


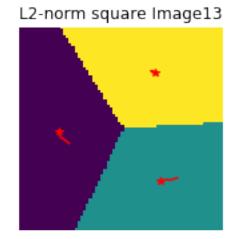


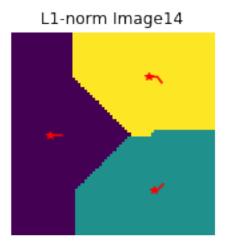


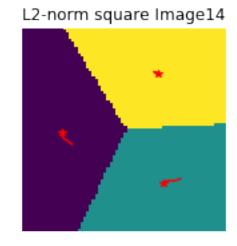




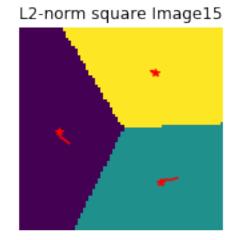


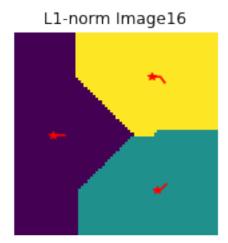


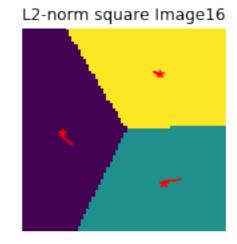




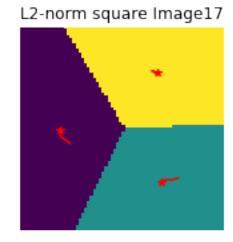
L1-norm Image15

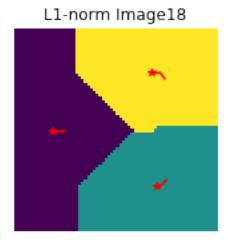


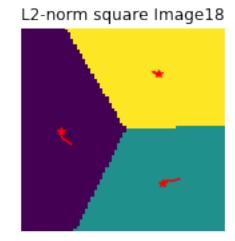




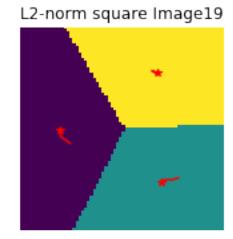
L1-norm Image17

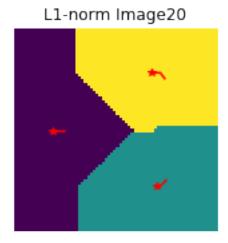


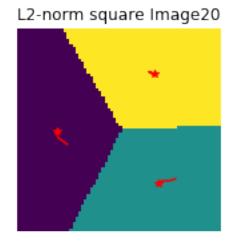


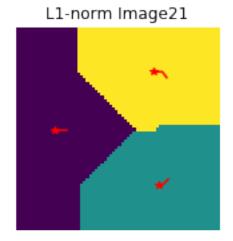


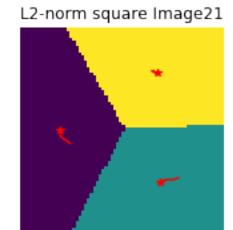
L1-norm Image19

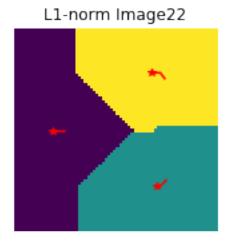


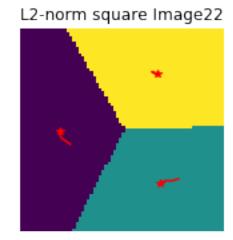


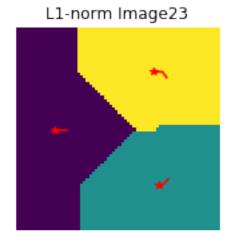


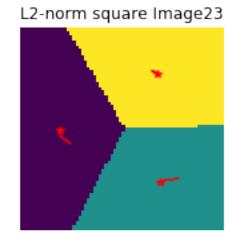


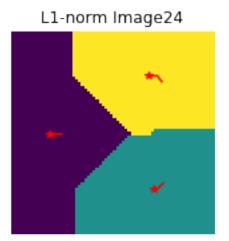


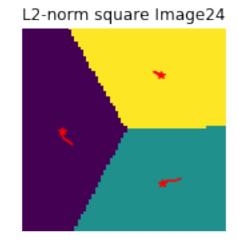


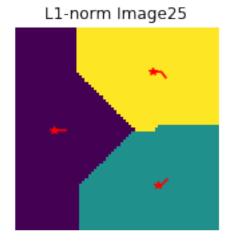


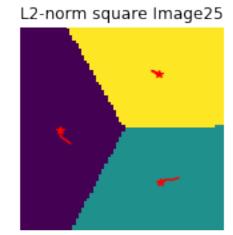


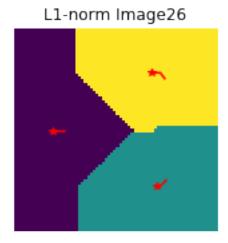


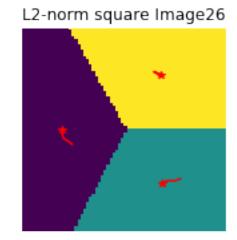




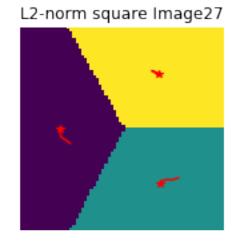


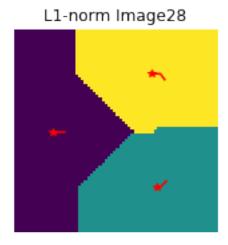


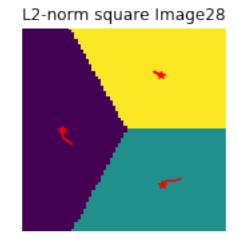


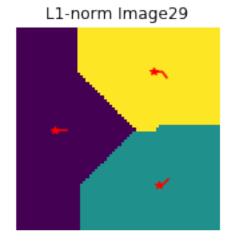


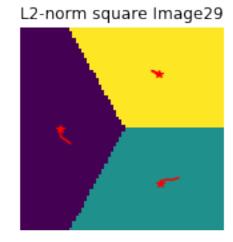
L1-norm Image27

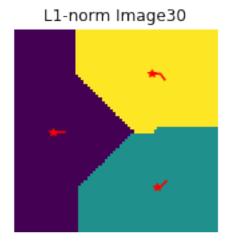


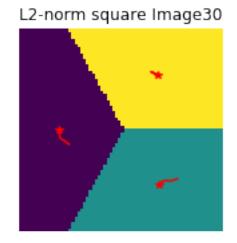








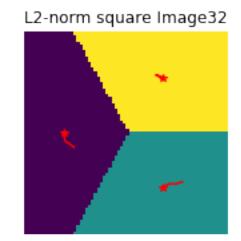




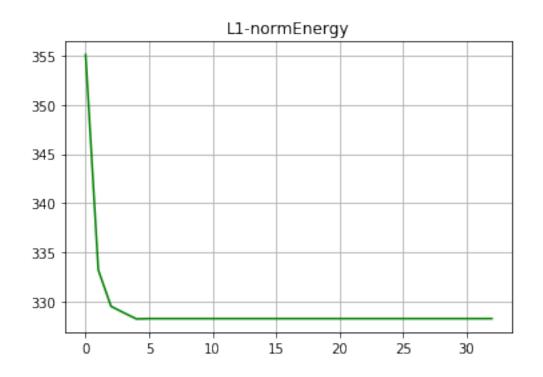
L1-norm Image31

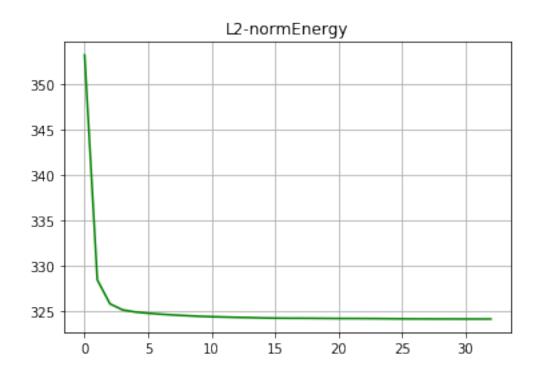
L2-norm square Image31

L1-norm Image32



# 14.2 Energy graphs

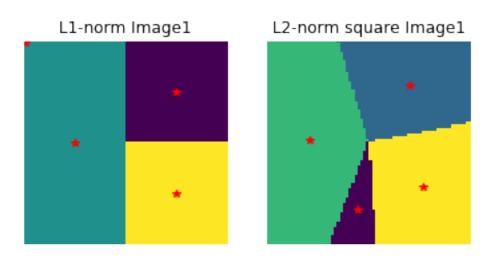


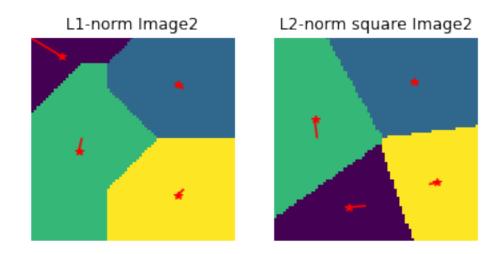


## 15 K = 4

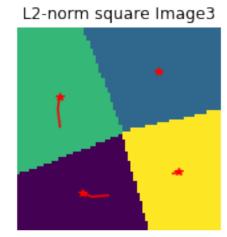
## 15.1 Final average image

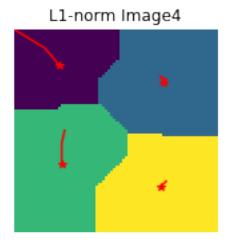
In [16]: im\_average\_x1, im\_average\_y1, im\_count1,im\_average\_x2, im\_average\_y2, im\_count2,im\_la
iter\_num = iteration(4,im\_average\_x1, im\_average\_y1, im\_count1, im\_average\_x2, im\_average\_x2, im\_average\_x2

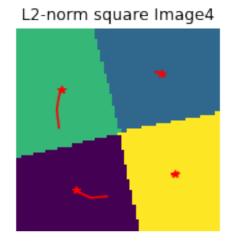




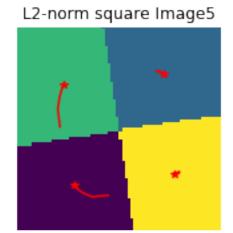
L1-norm Image3

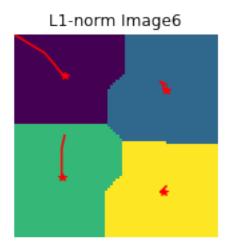


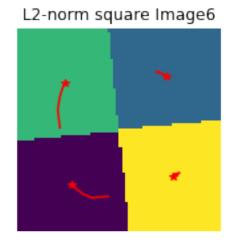




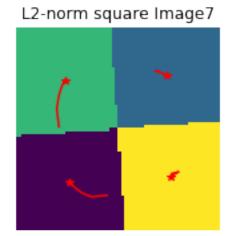
L1-norm Image5

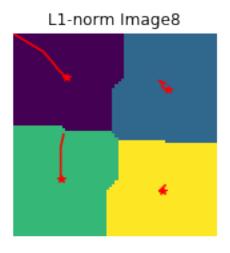


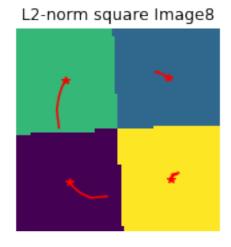




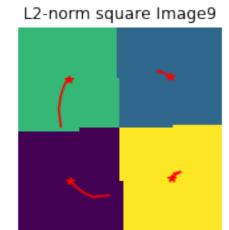
L1-norm Image7

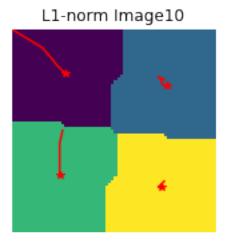


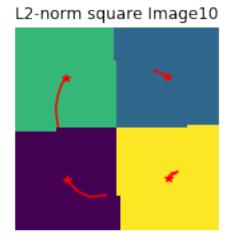




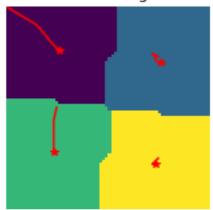
L1-norm Image9



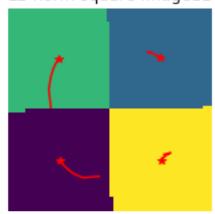




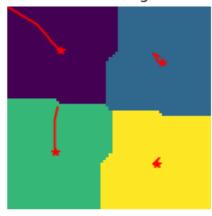
L1-norm Image11



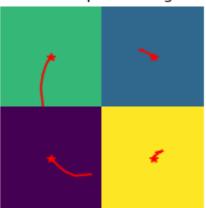
L2-norm square Image11



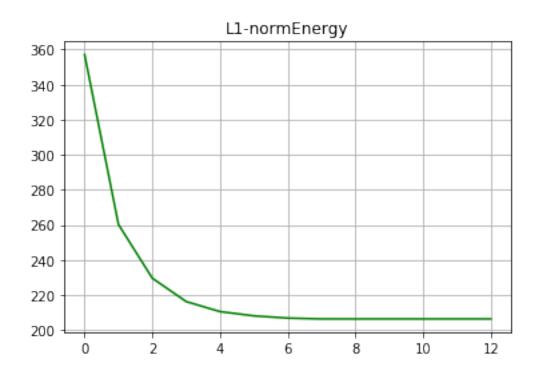
L1-norm Image12

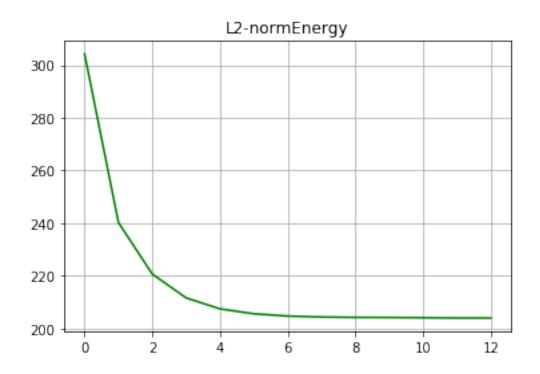


L2-norm square Image12



# 15.2 Energy graphs

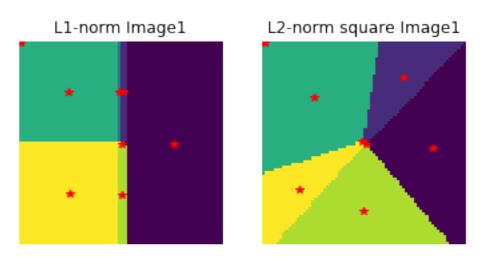


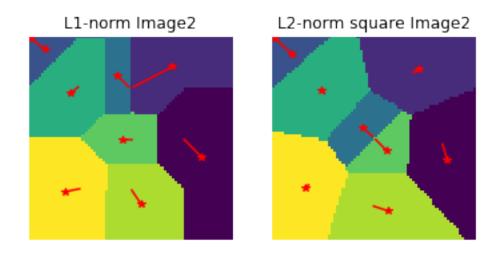


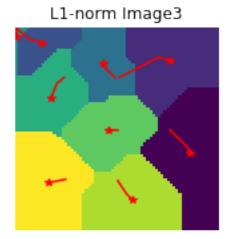
# 16 K = 9

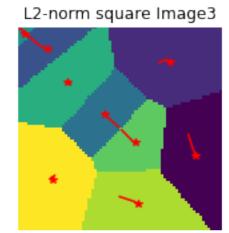
## 16.1 Final average image

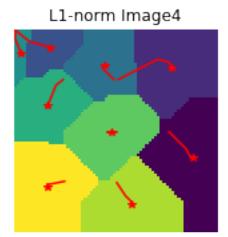
In [18]: im\_average\_x1, im\_average\_y1, im\_count1,im\_average\_x2, im\_average\_y2, im\_count2,im\_la
 iter\_num = iteration(9,im\_average\_x1, im\_average\_y1, im\_count1, im\_average\_x2, im\_average\_x2, im\_average\_x2)

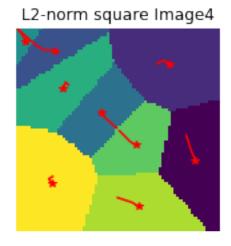




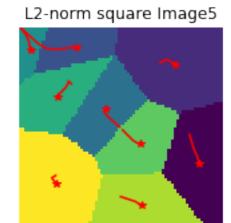


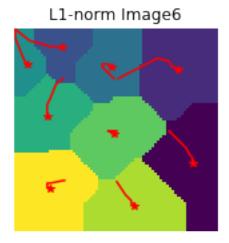


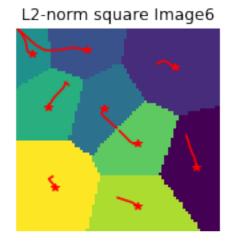




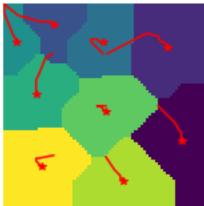
L1-norm Image5



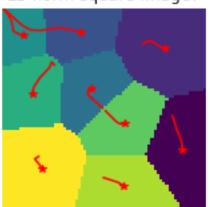




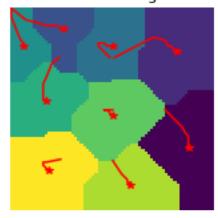
L1-norm Image7



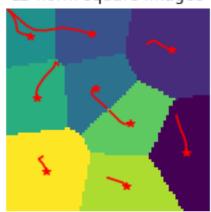
L2-norm square Image7



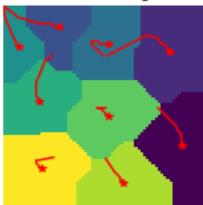
L1-norm Image8



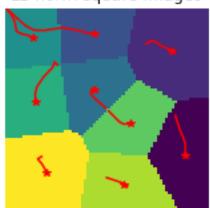
L2-norm square Image8



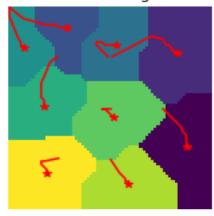
L1-norm Image9



L2-norm square Image9



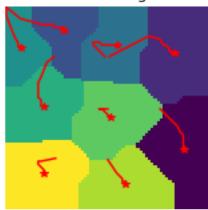
L1-norm Image10



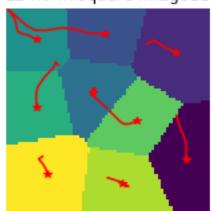
L2-norm square Image10



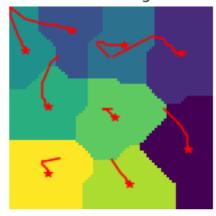
L1-norm Image11



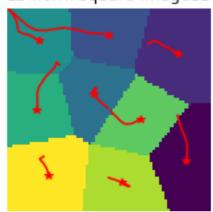
L2-norm square Image11



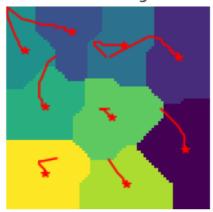
L1-norm Image12



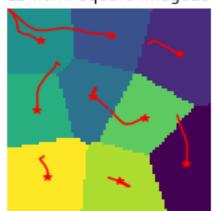
L2-norm square Image12



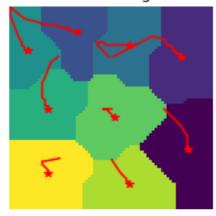
L1-norm Image13



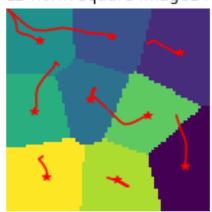
L2-norm square Image13



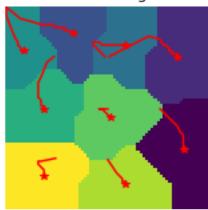
L1-norm Image14



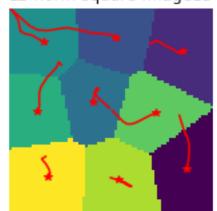
L2-norm square Image14



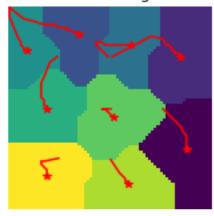
L1-norm Image15



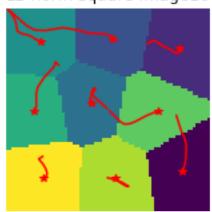
L2-norm square Image15



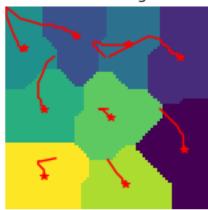
L1-norm Image16



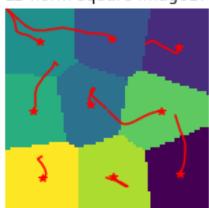
L2-norm square Image16



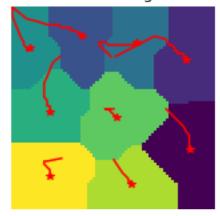
L1-norm Image17



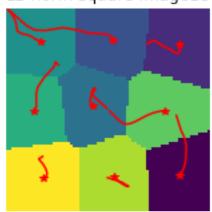
L2-norm square Image17



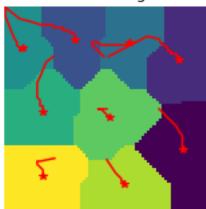
L1-norm Image18



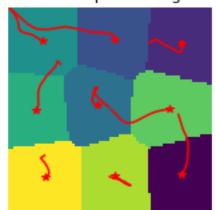
L2-norm square Image18



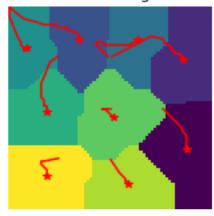
L1-norm Image19



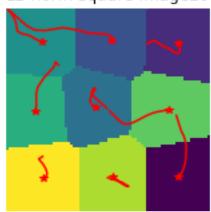
L2-norm square Image19



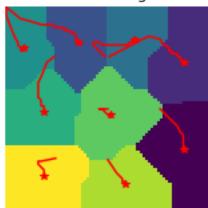
L1-norm Image20



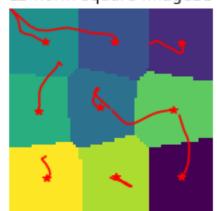
L2-norm square Image20



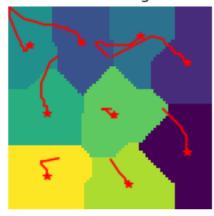
L1-norm Image21



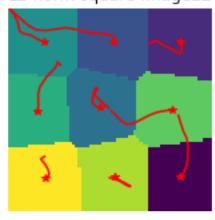
L2-norm square Image21



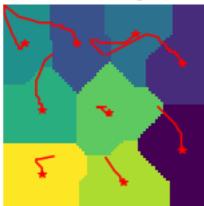
L1-norm Image22



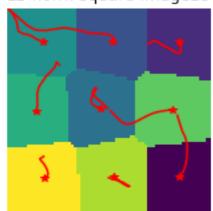
L2-norm square Image22



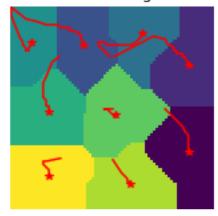
L1-norm Image23



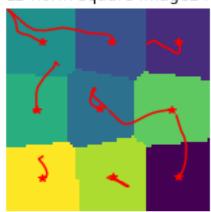
L2-norm square Image23



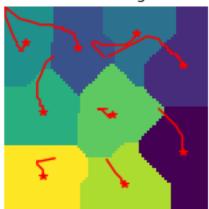
L1-norm Image24



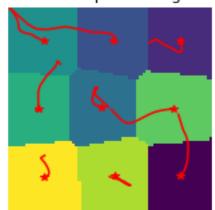
L2-norm square Image24



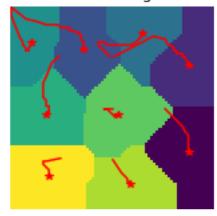
L1-norm Image25



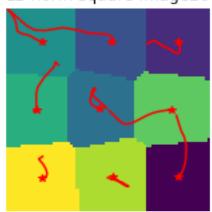
L2-norm square Image25



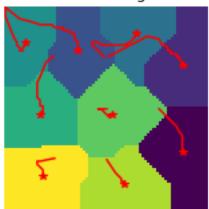
L1-norm Image26



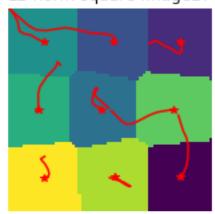
L2-norm square Image26



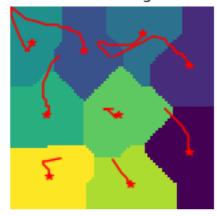
L1-norm Image27



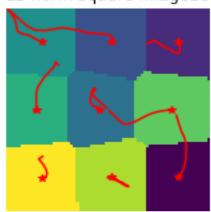
L2-norm square Image27



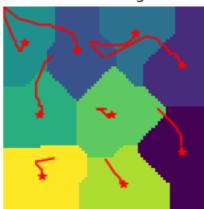
L1-norm Image28



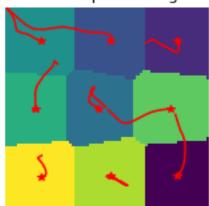
L2-norm square Image28



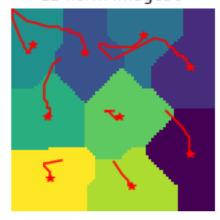
L1-norm Image29



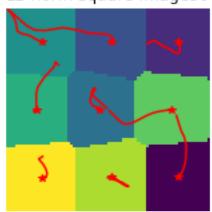
L2-norm square Image29



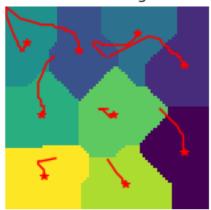
L1-norm Image30



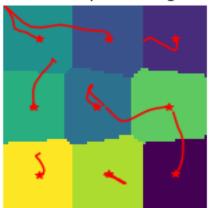
L2-norm square Image30



L1-norm Image31

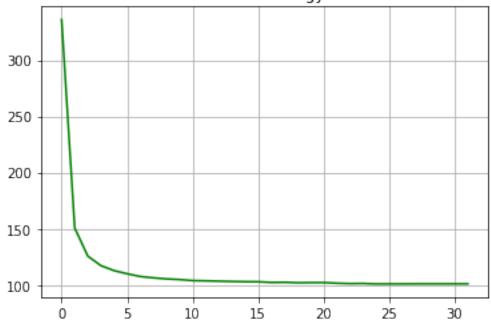


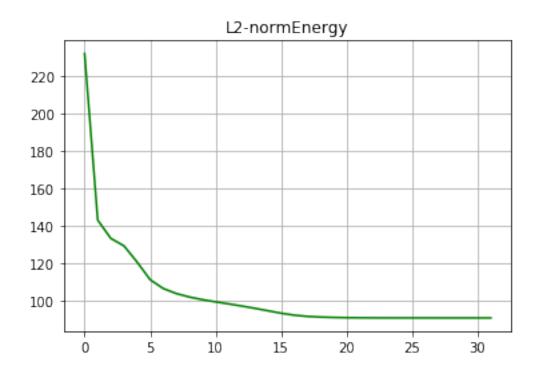
L2-norm square Image31



## 16.2 Energy graphs



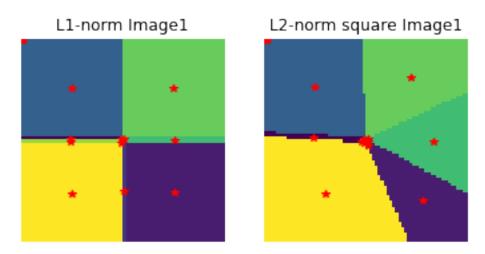




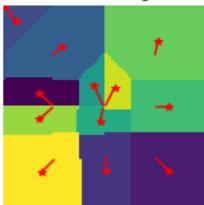
## 17 K = 15

## 17.1 Final average image

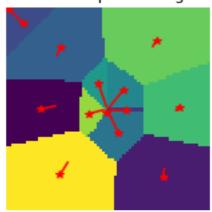
In [20]: im\_average\_x1, im\_average\_y1, im\_count1,im\_average\_x2, im\_average\_y2, im\_count2,im\_la
 iter\_num = iteration(15,im\_average\_x1, im\_average\_y1, im\_count1, im\_average\_x2, im\_average\_x2)



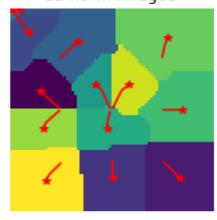
L1-norm Image2



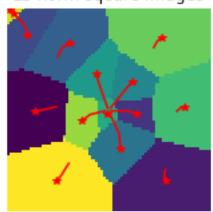
L2-norm square Image2



L1-norm Image3

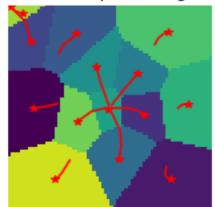


L2-norm square Image3



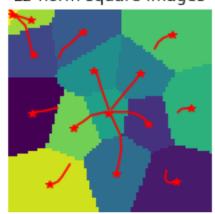
L1-norm Image4

e4 L2-norm square Image4

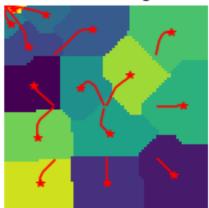


L1-norm Image5

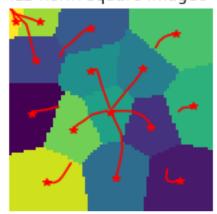
L2-norm square Image5



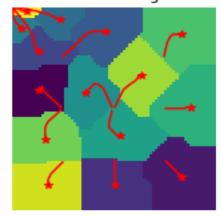
L1-norm Image6



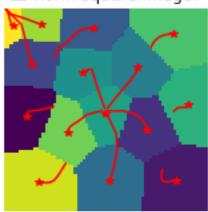
L2-norm square Image6



L1-norm Image7

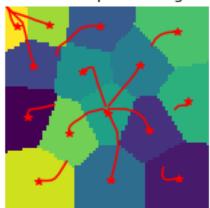


L2-norm square Image7

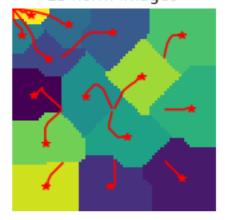


L1-norm Image8

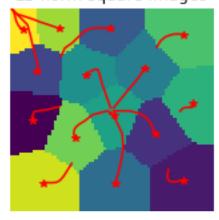
L2-norm square Image8



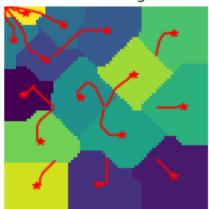
L1-norm Image9



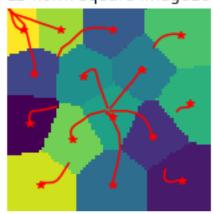
L2-norm square Image9



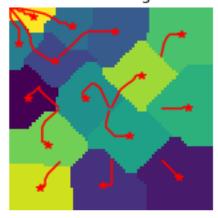
L1-norm Image10



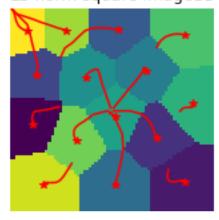
L2-norm square Image10



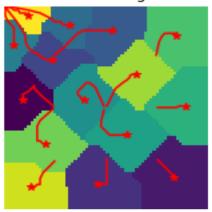
L1-norm Image11



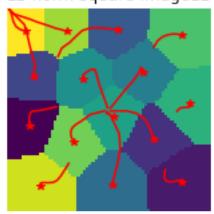
L2-norm square Image11



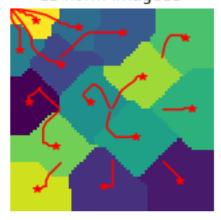
L1-norm Image12



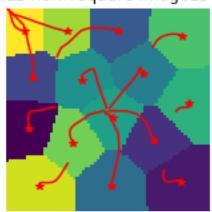
L2-norm square Image12



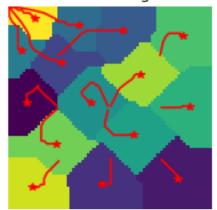
L1-norm Image13



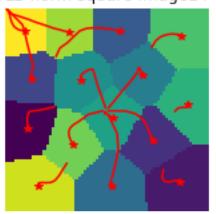
L2-norm square Image13



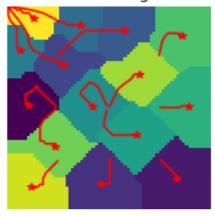
L1-norm Image14



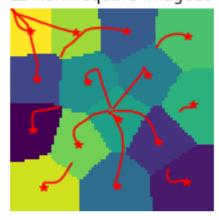
L2-norm square Image14



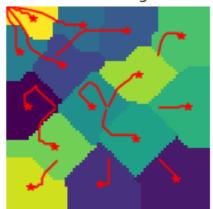
L1-norm Image15



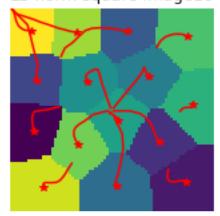
L2-norm square Image15



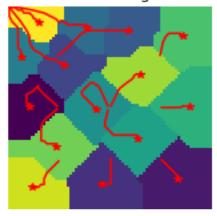
L1-norm Image16



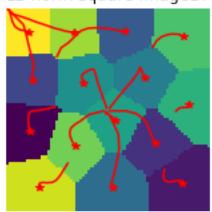
L2-norm square Image16



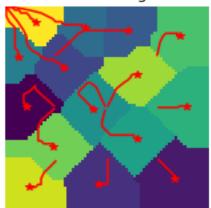
L1-norm Image17



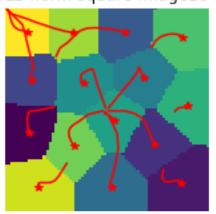
L2-norm square Image17



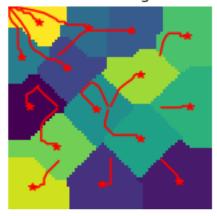
L1-norm Image18



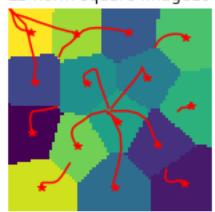
L2-norm square Image18



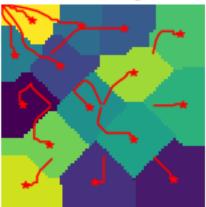
L1-norm Image19



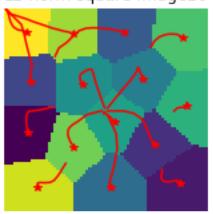
L2-norm square Image19



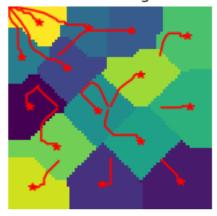
L1-norm Image20



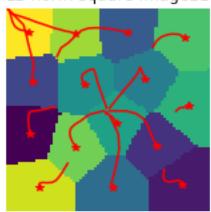
L2-norm square Image20



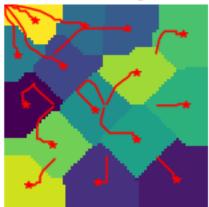
L1-norm Image21



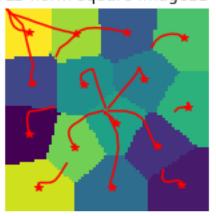
L2-norm square Image21



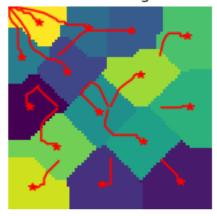
L1-norm Image22



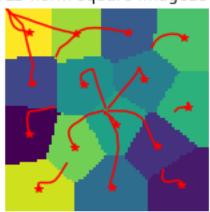
L2-norm square Image22



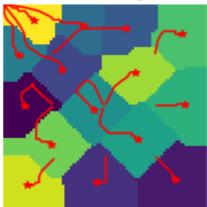
L1-norm Image23



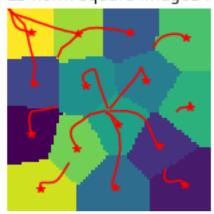
L2-norm square Image23



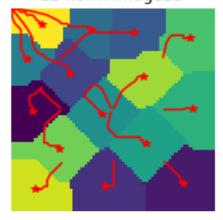
L1-norm Image24



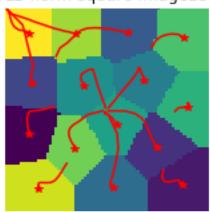
L2-norm square Image24



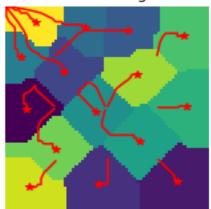
L1-norm Image25



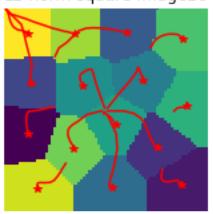
L2-norm square Image25



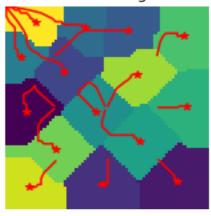
L1-norm Image26



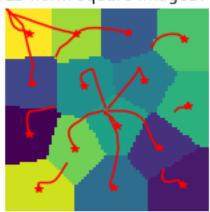
L2-norm square Image26



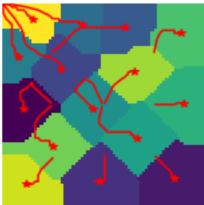
L1-norm Image27



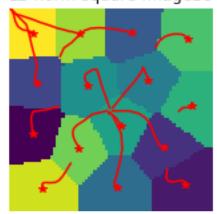
L2-norm square Image27



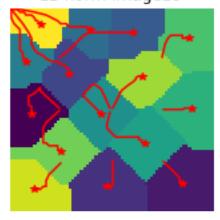
L1-norm Image28



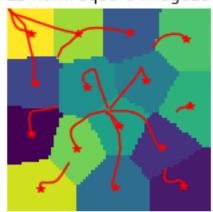
L2-norm square Image28



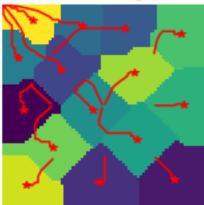
L1-norm Image29



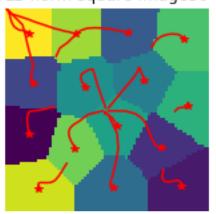
L2-norm square Image29



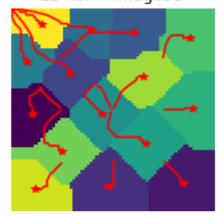
L1-norm Image30



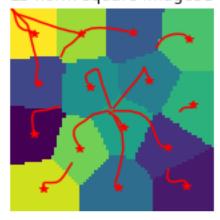
L2-norm square Image30



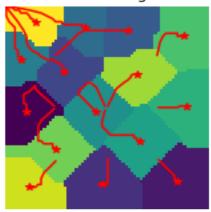
L1-norm Image31



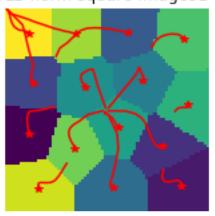
L2-norm square Image31



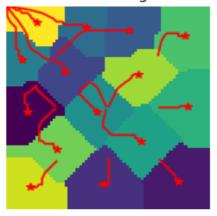
L1-norm Image32



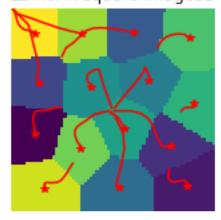
L2-norm square Image32



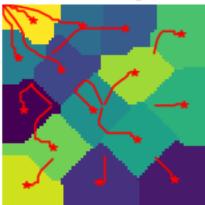
L1-norm Image33



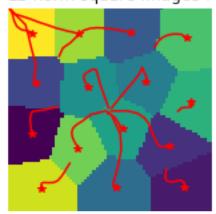
L2-norm square Image33



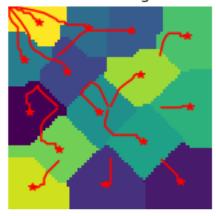
L1-norm Image34



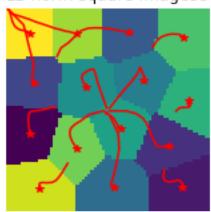
L2-norm square Image34



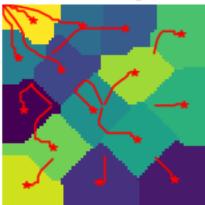
L1-norm Image35



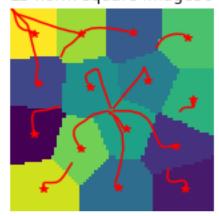
L2-norm square Image35



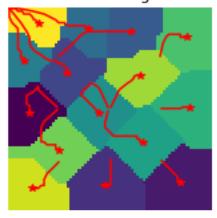
L1-norm Image36



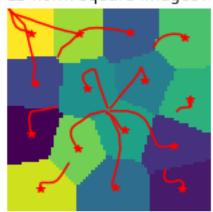
L2-norm square Image36



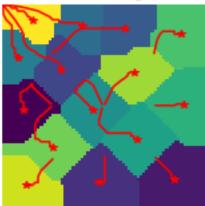
L1-norm Image37



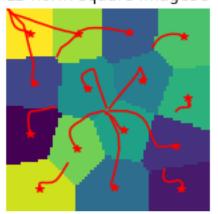
L2-norm square Image37



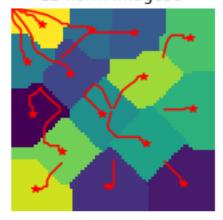
L1-norm Image38



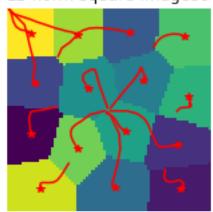
L2-norm square Image38



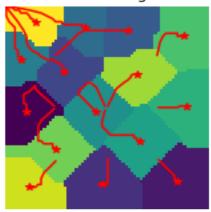
L1-norm Image39



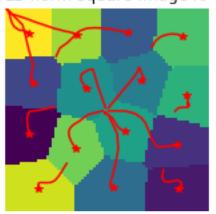
L2-norm square Image39



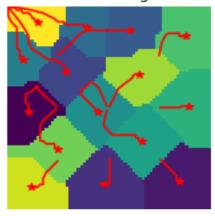
L1-norm Image40



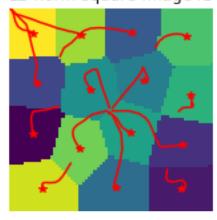
L2-norm square Image40



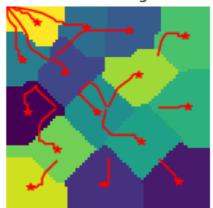
L1-norm Image41



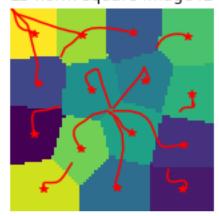
L2-norm square Image41



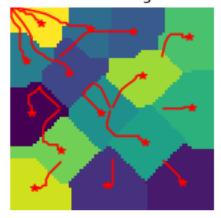
L1-norm Image42



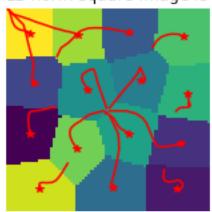
L2-norm square Image42



L1-norm Image43



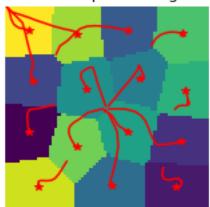
L2-norm square Image43



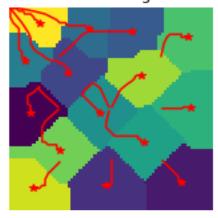
L1-norm Image44



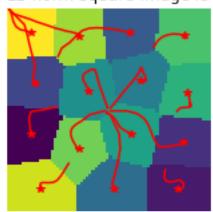
L2-norm square Image44



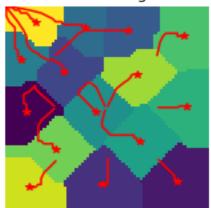
L1-norm Image45



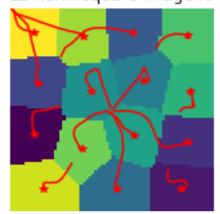
L2-norm square Image45



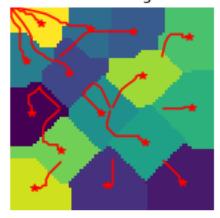
L1-norm Image46



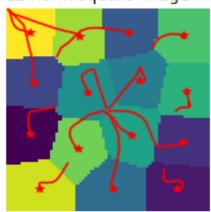
L2-norm square Image46



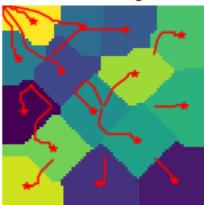
L1-norm Image47



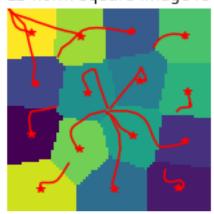
L2-norm square Image47



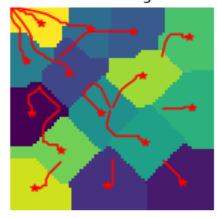
L1-norm Image48



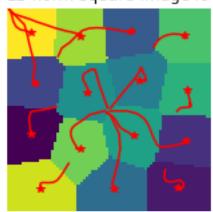
L2-norm square Image48



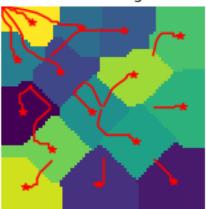
L1-norm Image49



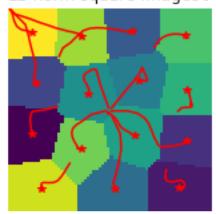
L2-norm square Image49



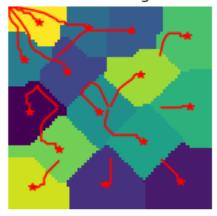
L1-norm Image50



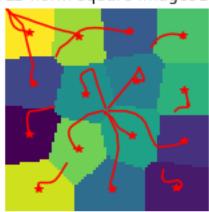
L2-norm square Image50



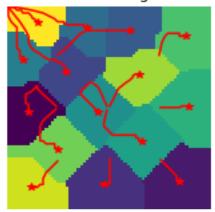
L1-norm Image51



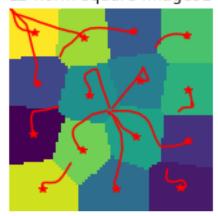
L2-norm square Image51



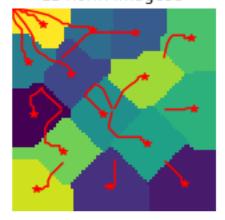
L1-norm Image52



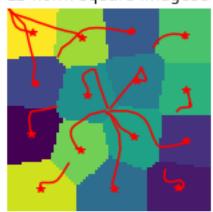
L2-norm square Image52



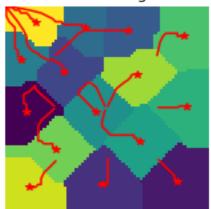
L1-norm Image53



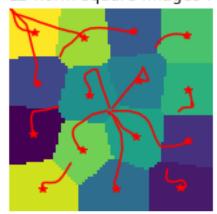
L2-norm square Image53



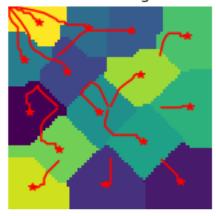
L1-norm Image54



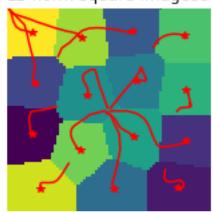
L2-norm square Image54



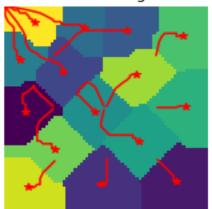
L1-norm Image55



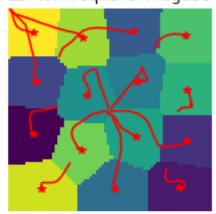
L2-norm square Image55



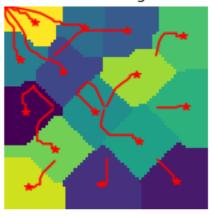
L1-norm Image56



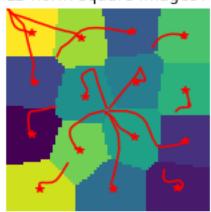
L2-norm square Image56



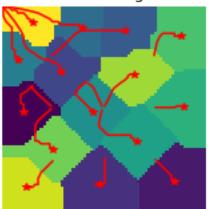
L1-norm Image57



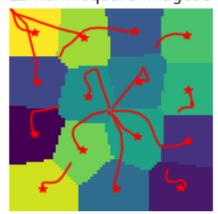
L2-norm square Image57



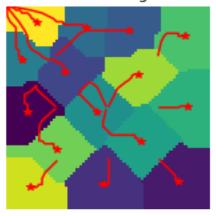
L1-norm Image58



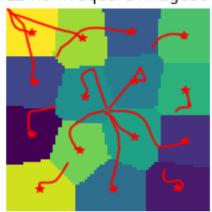
L2-norm square Image58



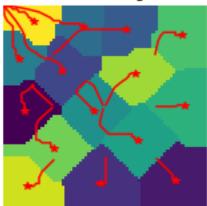
L1-norm Image59



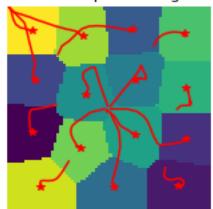
L2-norm square Image59



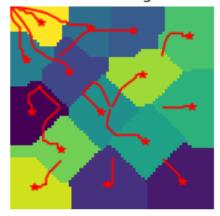
L1-norm Image60



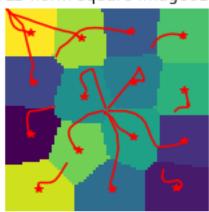
L2-norm square Image60



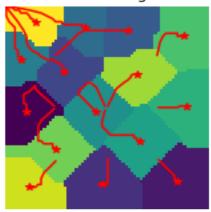
L1-norm Image61



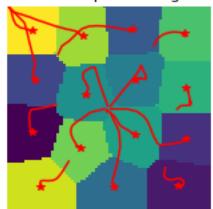
L2-norm square Image61



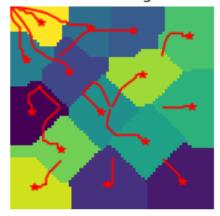
L1-norm Image62



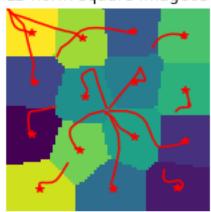
L2-norm square Image62



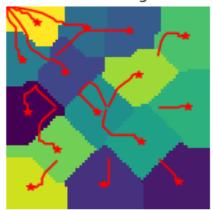
L1-norm Image63



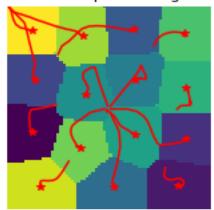
L2-norm square Image63



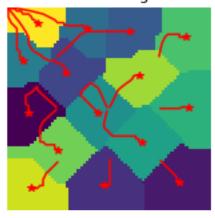
L1-norm Image64



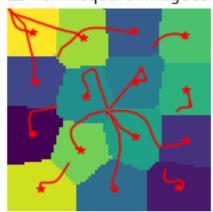
L2-norm square Image64



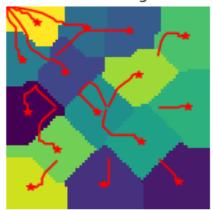
L1-norm Image65



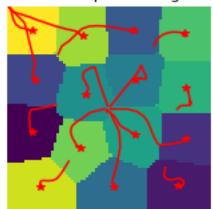
L2-norm square Image65



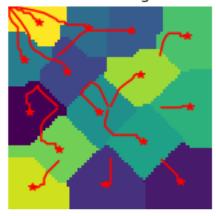
L1-norm Image66



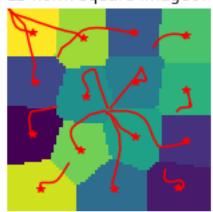
L2-norm square Image66



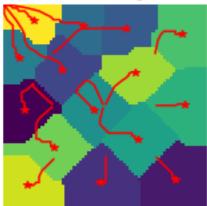
L1-norm Image67



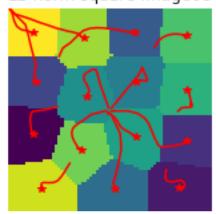
L2-norm square Image67



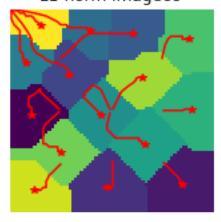
L1-norm Image68



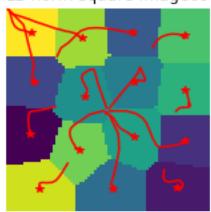
L2-norm square Image68



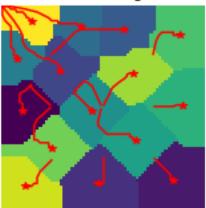
L1-norm Image69



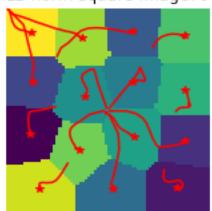
L2-norm square Image69



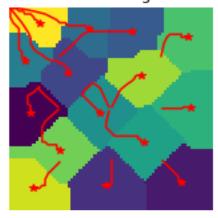
L1-norm Image70



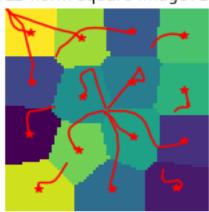
L2-norm square Image70



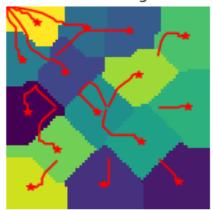
L1-norm Image71



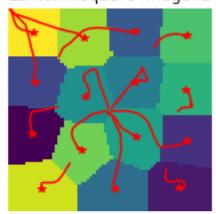
L2-norm square Image71



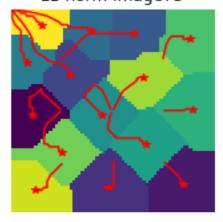
L1-norm Image72



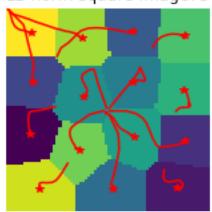
L2-norm square Image72



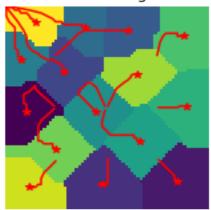
L1-norm Image73



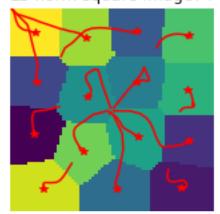
L2-norm square Image73



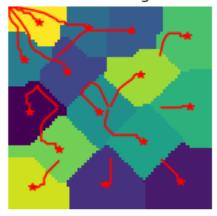
L1-norm Image74



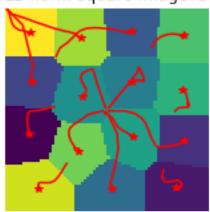
L2-norm square Image74



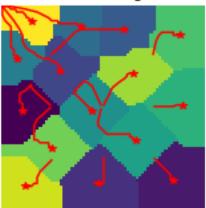
L1-norm Image75



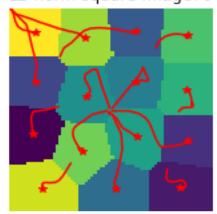
L2-norm square Image75



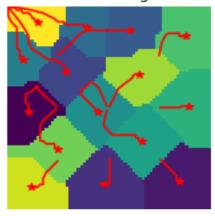
L1-norm Image76



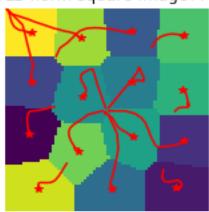
L2-norm square Image76



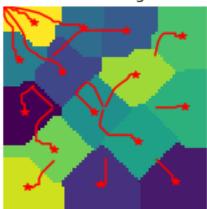
L1-norm Image77



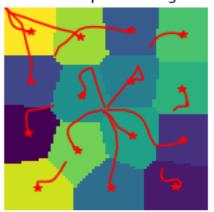
L2-norm square Image77



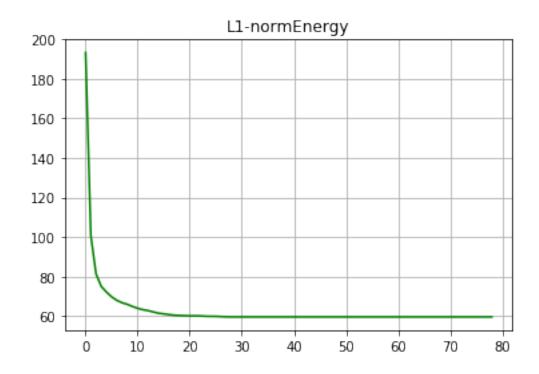
L1-norm Image78

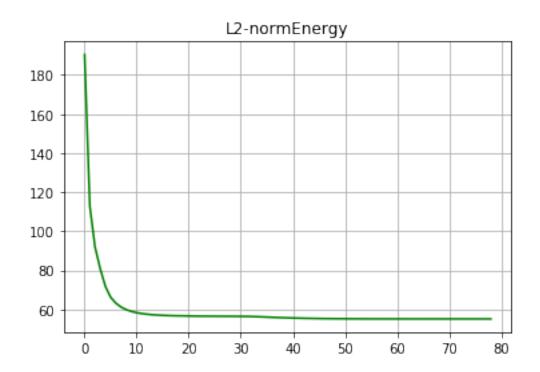


L2-norm square Image78



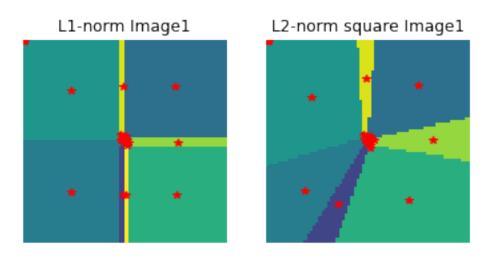
## 17.2 Energy graphs



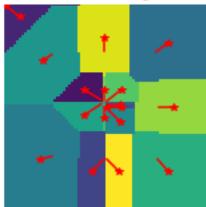


## K = 20

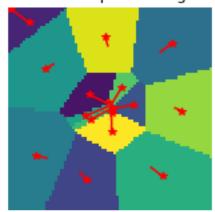
In [22]: im\_average\_x1, im\_average\_y1, im\_count1,im\_average\_x2, im\_average\_y2, im\_count2,im\_la
 iter\_num = iteration(20,im\_average\_x1, im\_average\_y1, im\_count1, im\_average\_x2, im\_average\_x2)



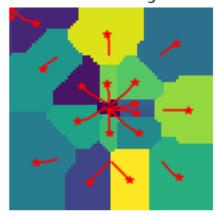
L1-norm Image2



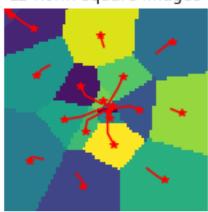
L2-norm square Image2



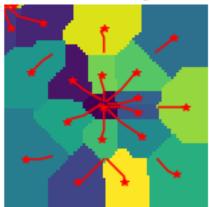
L1-norm Image3



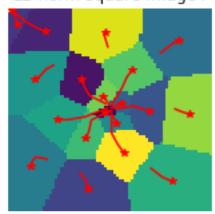
L2-norm square Image3



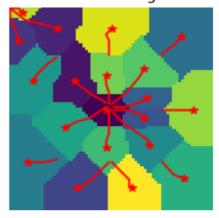
L1-norm Image4



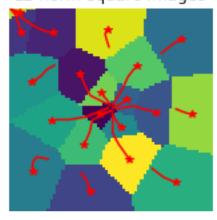
L2-norm square Image4



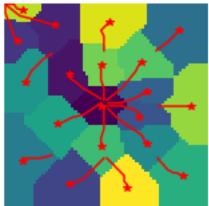
L1-norm Image5



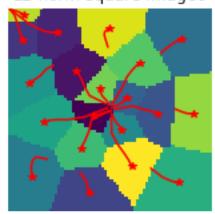
L2-norm square Image5



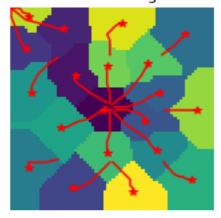
L1-norm Image6



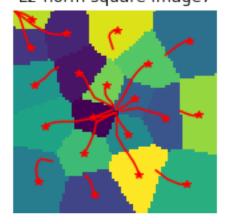
L2-norm square Image6



L1-norm Image7

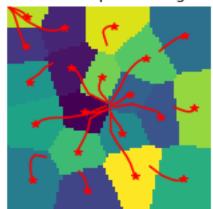


L2-norm square Image7



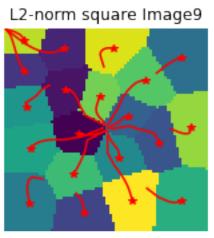
L1-norm Image8

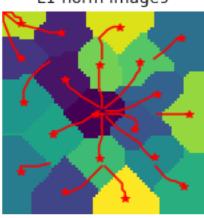
L2-norm square Image8



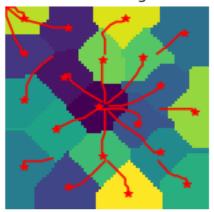
L1-norm Image9

ge9 L2-no

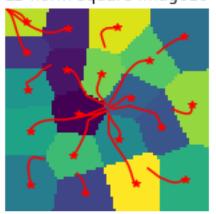




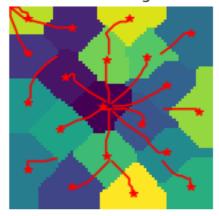
L1-norm Image10



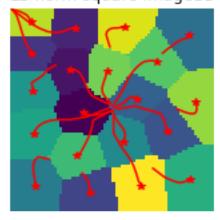
L2-norm square Image10



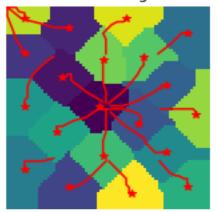
L1-norm Image11



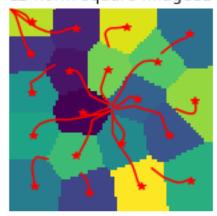
L2-norm square Image11



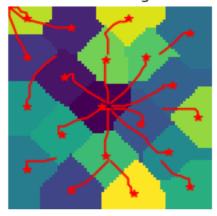
L1-norm Image12



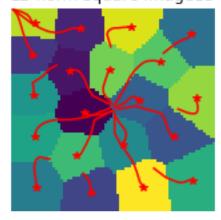
L2-norm square Image12



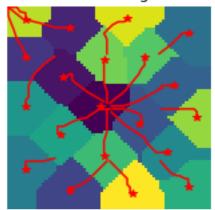
L1-norm Image13



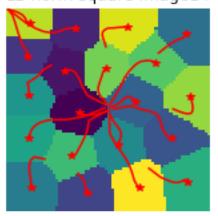
L2-norm square Image13



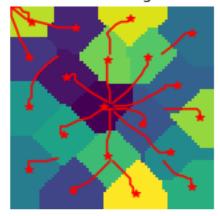
L1-norm Image14



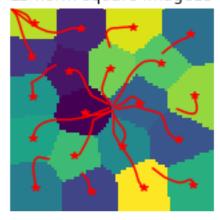
L2-norm square Image14



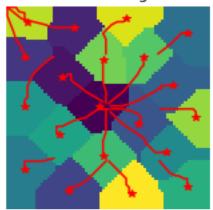
L1-norm Image15



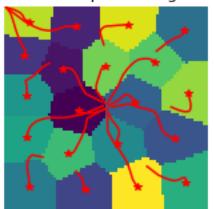
L2-norm square Image15



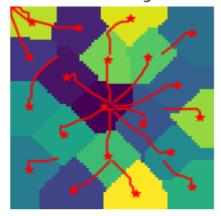
L1-norm Image16



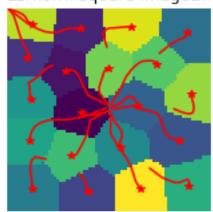
L2-norm square Image16



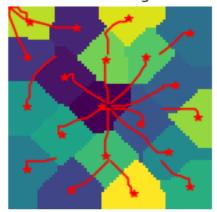
L1-norm Image17



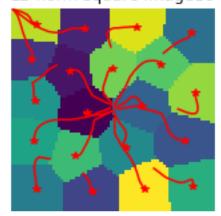
L2-norm square Image17



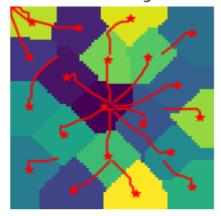
L1-norm Image18



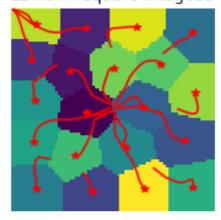
L2-norm square Image18



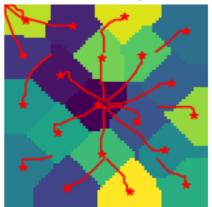
L1-norm Image19



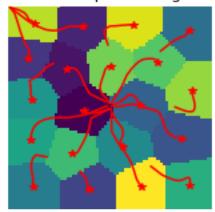
L2-norm square Image19



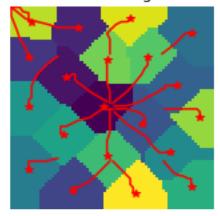
L1-norm Image20



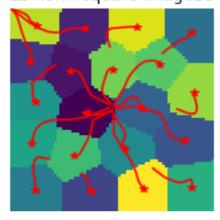
L2-norm square Image20



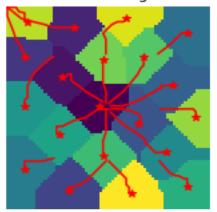
L1-norm Image21



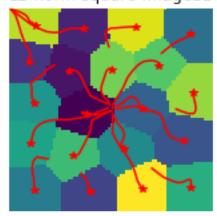
L2-norm square Image21



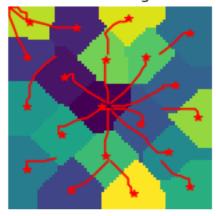
L1-norm Image22



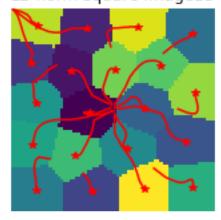
L2-norm square Image22



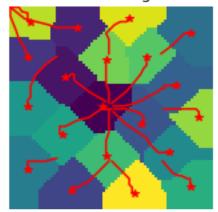
L1-norm Image23



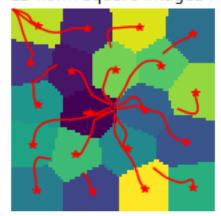
L2-norm square Image23



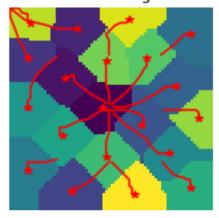
L1-norm Image24



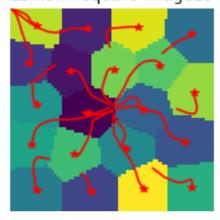
L2-norm square Image24



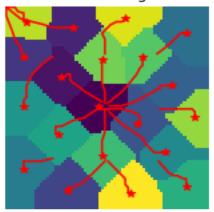
L1-norm Image25



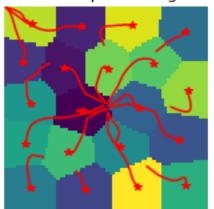
L2-norm square Image25



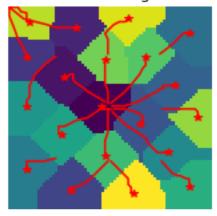
L1-norm Image26



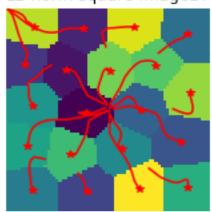
L2-norm square Image26



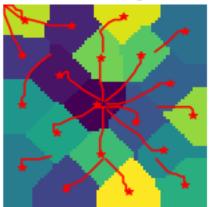
L1-norm Image27



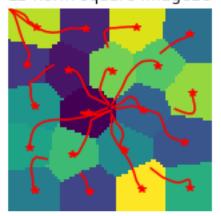
L2-norm square Image27



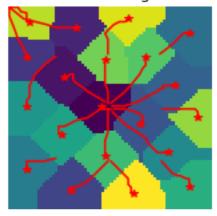
L1-norm Image28



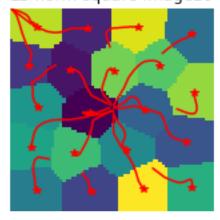
L2-norm square Image28



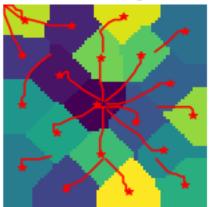
L1-norm Image29



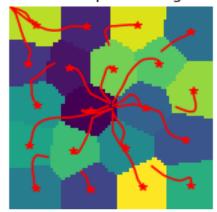
L2-norm square Image29



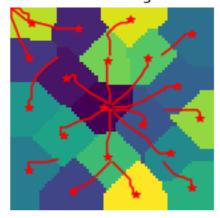
L1-norm Image30



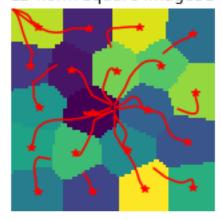
L2-norm square Image30



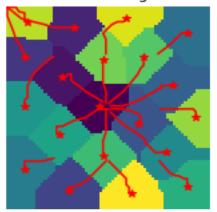
L1-norm Image31



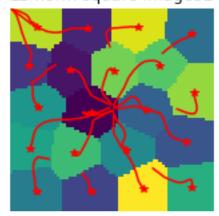
L2-norm square Image31



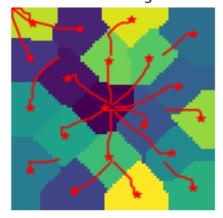
L1-norm Image32



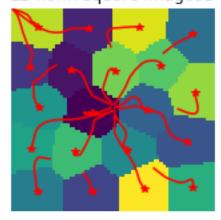
L2-norm square Image32



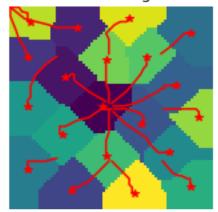
L1-norm Image33



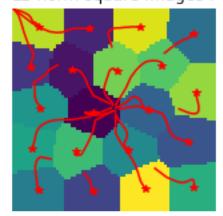
L2-norm square Image33



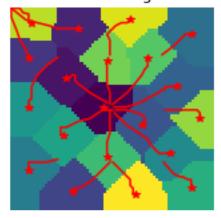
L1-norm Image34



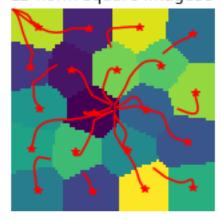
L2-norm square Image34



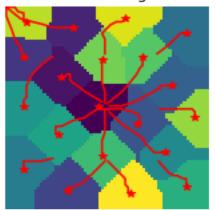
L1-norm Image35



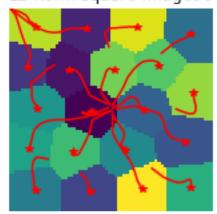
L2-norm square Image35



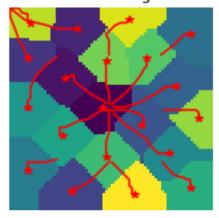
L1-norm Image36



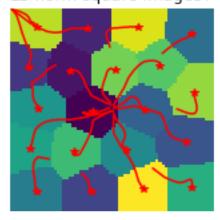
L2-norm square Image36



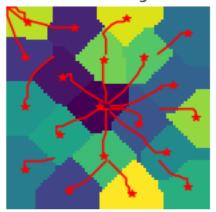
L1-norm Image37



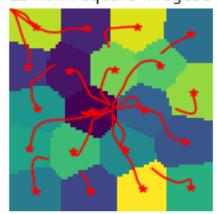
L2-norm square Image37



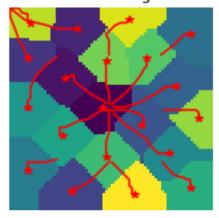
L1-norm Image38



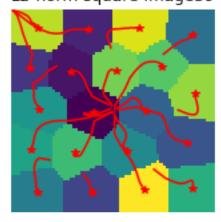
L2-norm square Image38



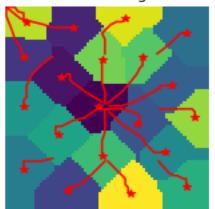
L1-norm Image39



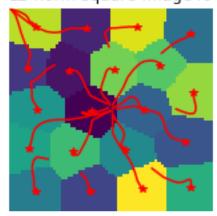
L2-norm square Image39



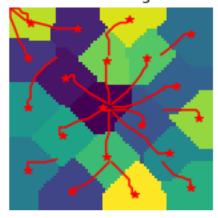
L1-norm Image40



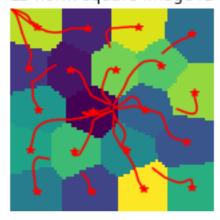
L2-norm square Image40



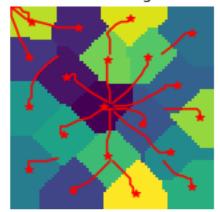
L1-norm Image41



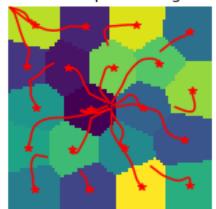
L2-norm square Image41



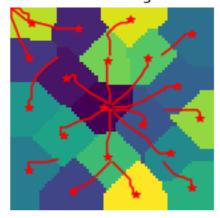
L1-norm Image42



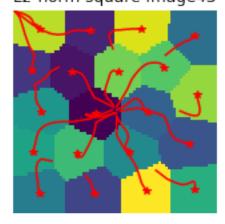
L2-norm square Image42



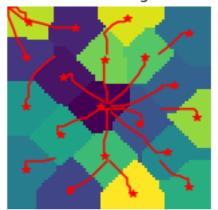
L1-norm Image43



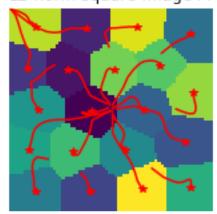
L2-norm square Image43



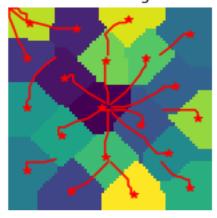
L1-norm Image44



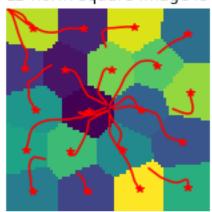
L2-norm square Image44



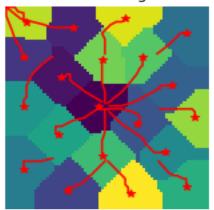
L1-norm Image45



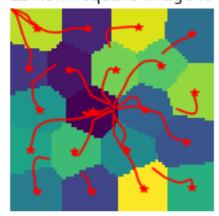
L2-norm square Image45



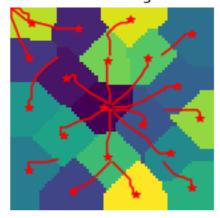
L1-norm Image46



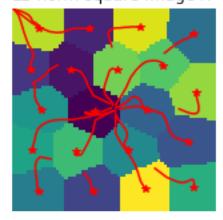
L2-norm square Image46



L1-norm Image47



L2-norm square Image47



L1-norm Image48

L2-norm square Image48

## 19 Energy graph

