

# K-MEANS INITIALIZATION PROBLEM

Initial by center or partition

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# CONTENT

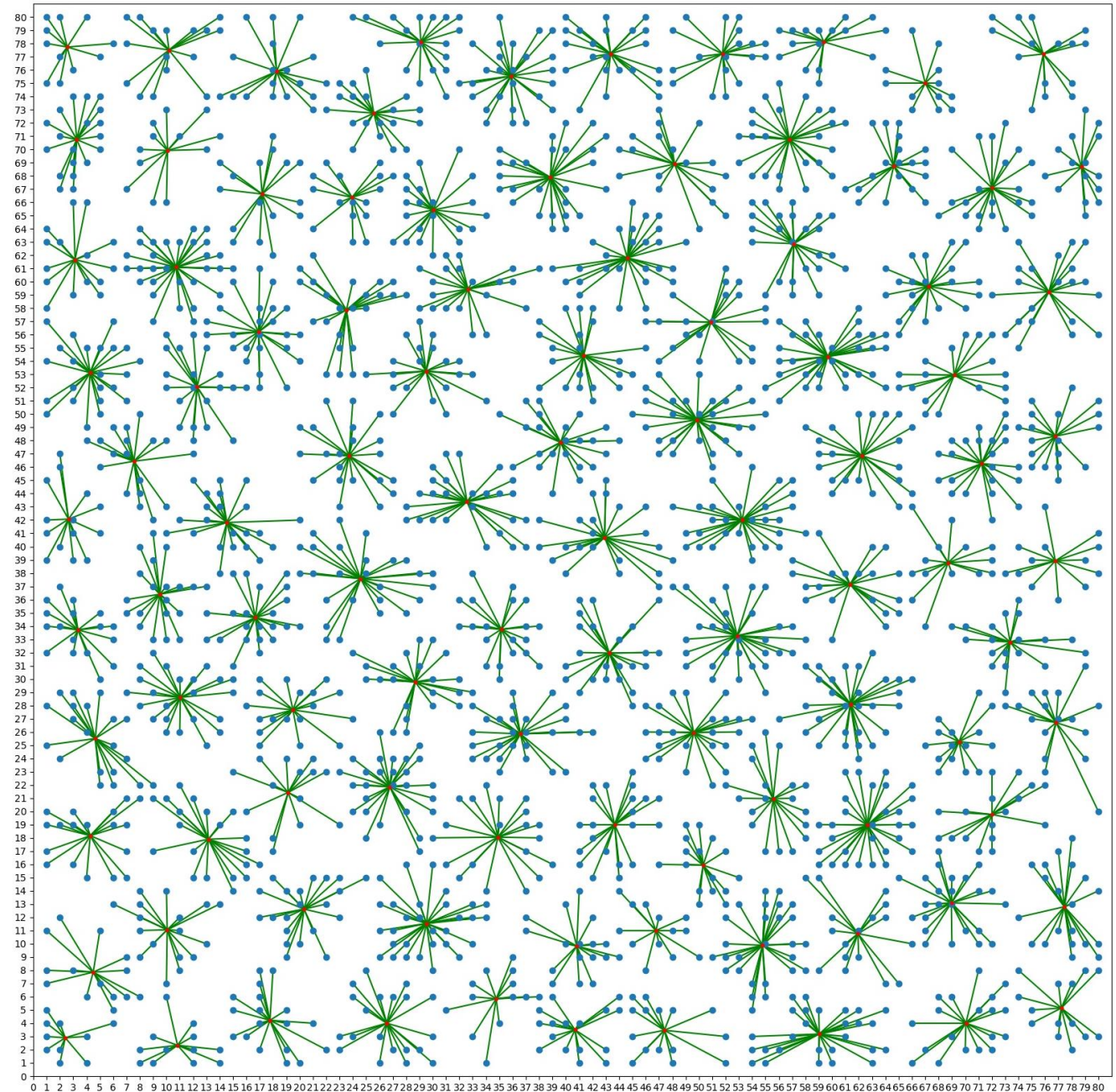
1. K-Means Algorithm
2. Initial K-Means by centers or partition
3. Initialization algorithm for K-Means

# K-MEANS ALGORITHM

An algorithm to solve the clustering problem. It will iterate the following two steps from a random partition of  $S$  into  $k$  subsets:  $S_1, S_2, \dots, S_k$ .

$$(i) c_j = \frac{1}{|S_j|} \sum_{s \in S_j} s$$

$$(ii) S_j = \{s \mid \text{dist}(s, c_j) = \min_{l=1,2,\dots,k} \text{dist}(s, c_l)\}, j = 1, 2, \dots, k$$



# INITIAL K-MEANS

- Initial by partition :

- Generate K partition
- Calculate the average coordinate as the center of partition.

- Initial by centers

- Choice k site as the initial centers
- Each site choice the nearest center as the center and join it's partition.

# INITIALIZATION ALGORITHM FOR K-MEANS

1. Initial by center:
  - I. Use the center select algorithm to choice the initial centers.
  - II. For each site  $S_i$ , select the first  $\frac{1}{3}$  closest sites and let the maximum value of the distance be  $d_i$ , I will choice the site  $S_i$  which can minimize the  $d_i$  as the start center.
  - III. Then, choice the farthest site from the centers as the next center iteratively until we get enough centers.

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## Algorithm 5 Greedy center selection algorithm

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**Require:**  $S$  : all sites,  $k$  : The number of the center

**Ensure:**

```
1: function greedy_center_selection_algorithm( $S, k$ )
2:    $C \leftarrow []$ 
3:    $C.add(S.index(random\_int()))$ 
4:   while  $C.size < k$  do
5:      $C.append(\operatorname{argmax}_s Distance(s, C))$ 
6:   end while
7:   return  $C$ 
8: end function
```

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## Algorithm 6 Initial center selection algorithm

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**Require:**  $S$  : all sites,  $k$  : The number of the center

**Ensure:**

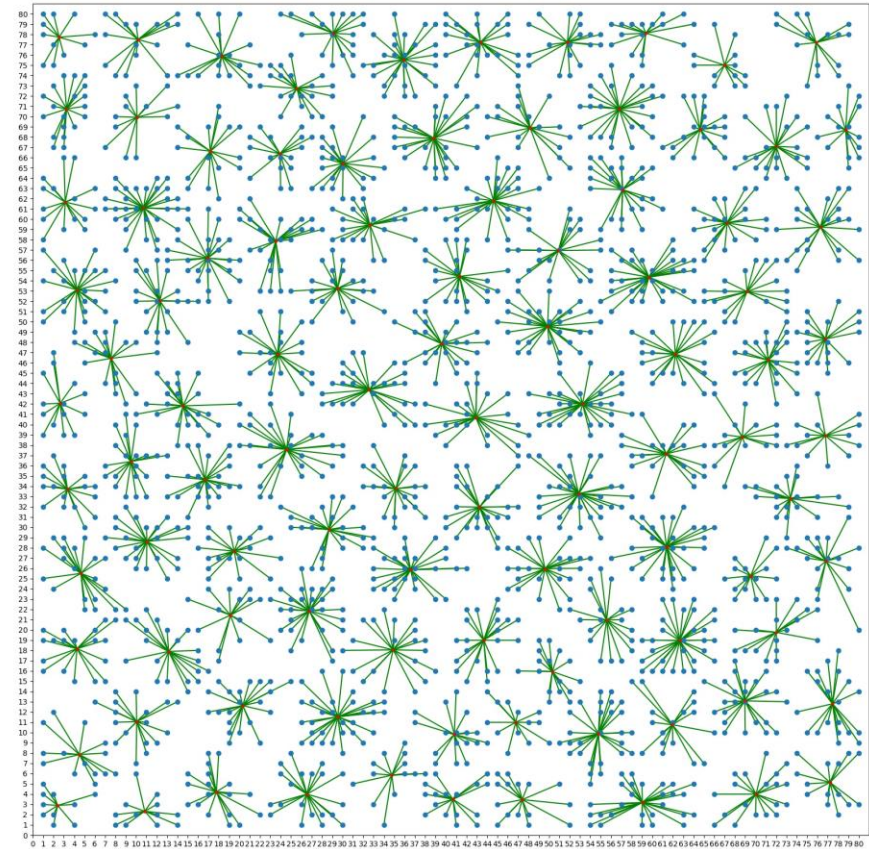
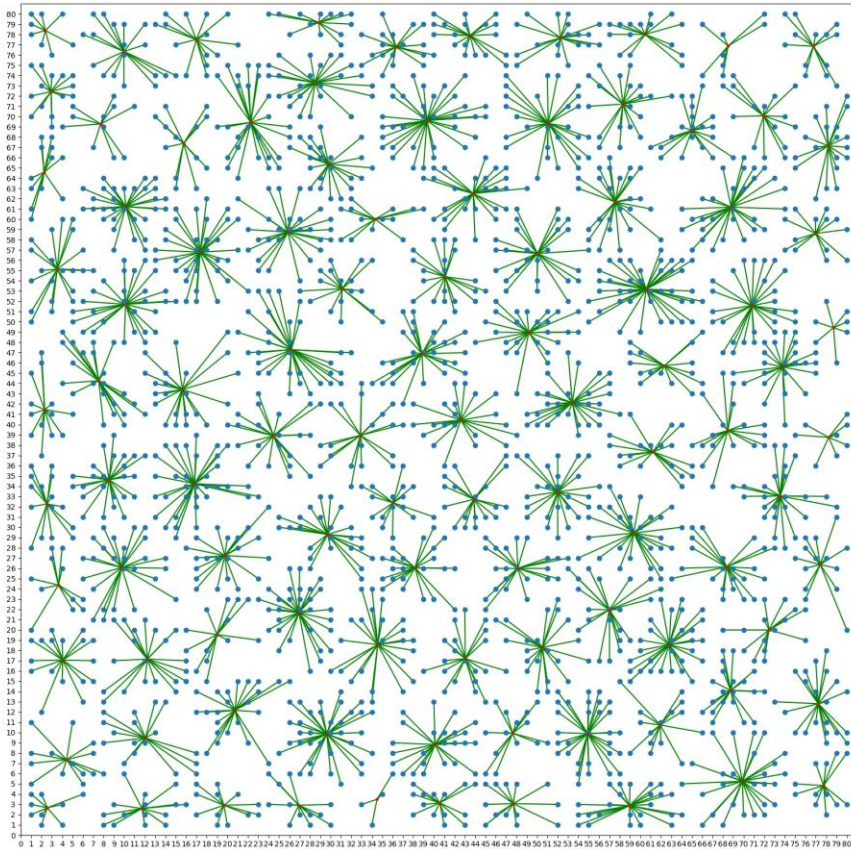
```
1: function initial_center_selection_algorithm( $S, k$ )
2:    $min\_s, min\_value = -1, infinite$ 
3:   for  $s$  in  $S$  do
4:      $s\_closed = [sites \text{ first } 1/3 \text{ closest to } s]$ 
5:      $current\_max = \max([dist(s, s\_c) \text{ for } s\_c \text{ in } s\_closed])$ 
6:     if  $min\_value > current\_max$  then  $min\_s = s$ 
7:     end if
8:   end for
9:   return  $min\_s$ 
10: end function
```

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# INITIALIZATION ALGORITHM FOR K-MEANS



# INITIALIZATION ALGORITHM FOR K-MEANS

1. Initial by cluster:
  - I. Use the way last page to get the  $k$  site, and let them be  $k$  partitions, each cluster has one site.
  - II. Poll each partition, choice the nearest free site and add it into the partition, until all the sites have been added into the partitions.

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**Algorithm 7** Initial partition selection algorithm

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**Require:**  $S$  : all sites,  $k$  : The number of the partitions

**Ensure:**

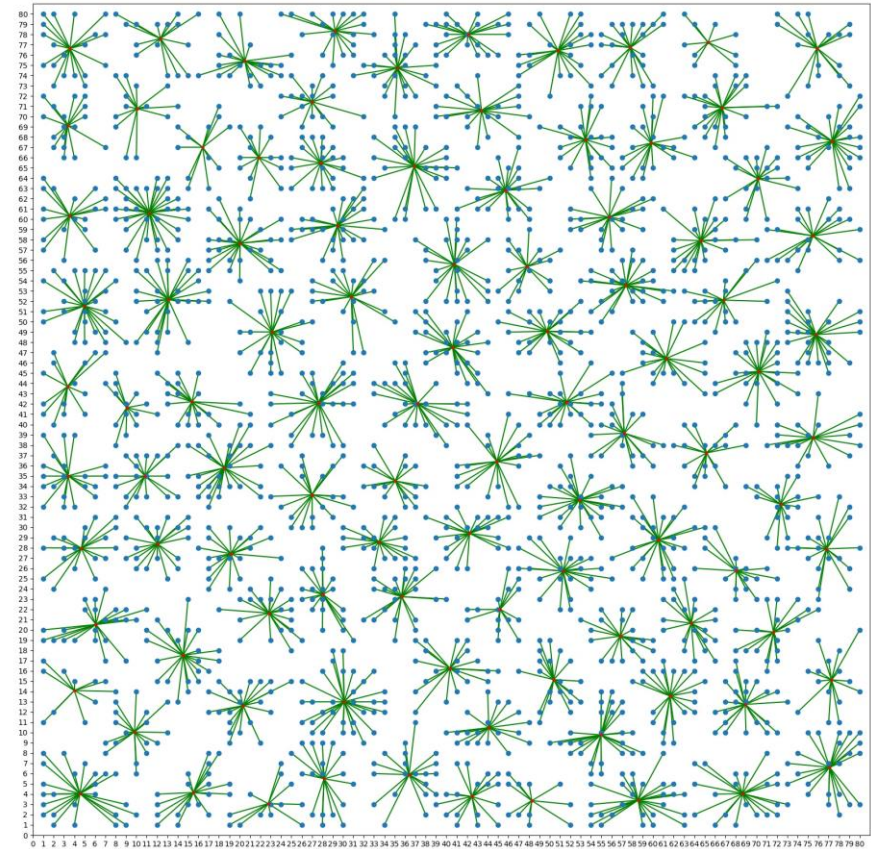
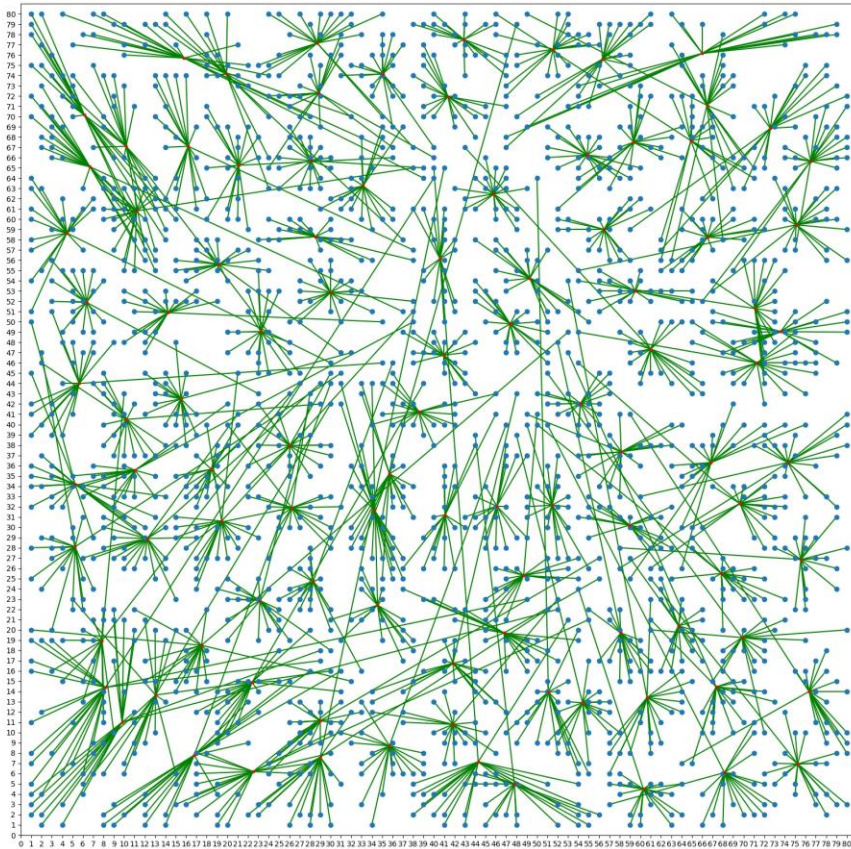
```
1: function initial_partition_selection_algorithm( $S, k$ )
2:    $initial\_partition = initial\_center\_selection\_algorithm(S, k)$ 
3:   while There are sites not in partitions do
4:     for partition in  $initial\_partitions$  do
5:       partition.append(nearest free site)
6:       if There is no free site then
7:         Break
8:       end if
9:     end for
10:  end while
11:  return initial_partition
12: end function
```

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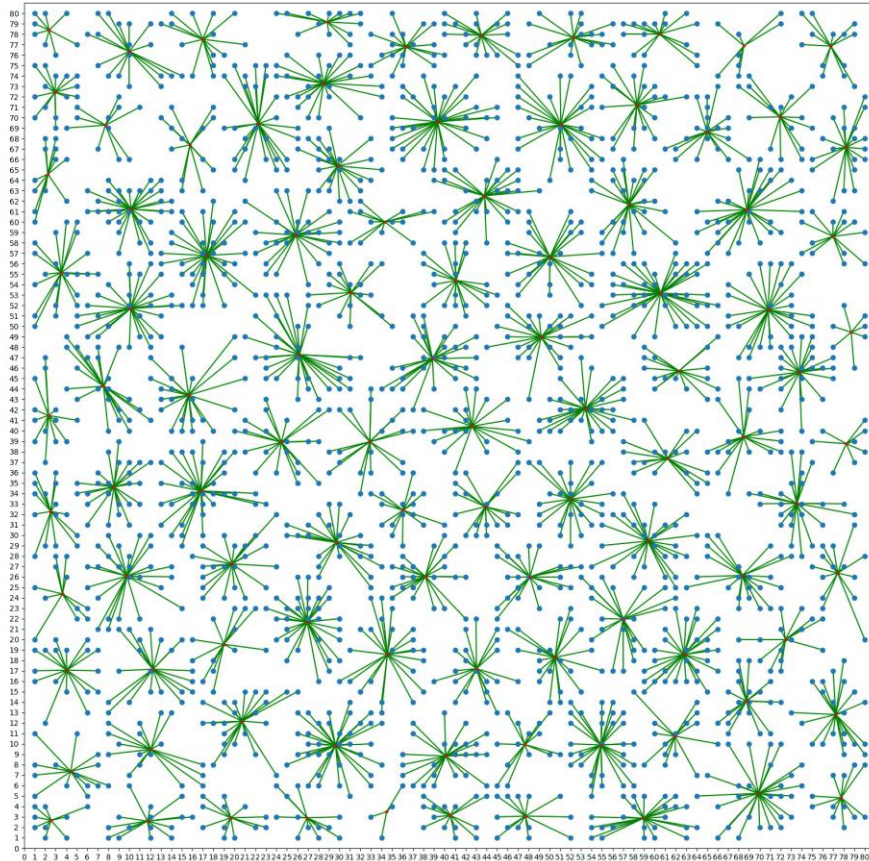


# INITIALIZATION ALGORITHM FOR K-MEANS

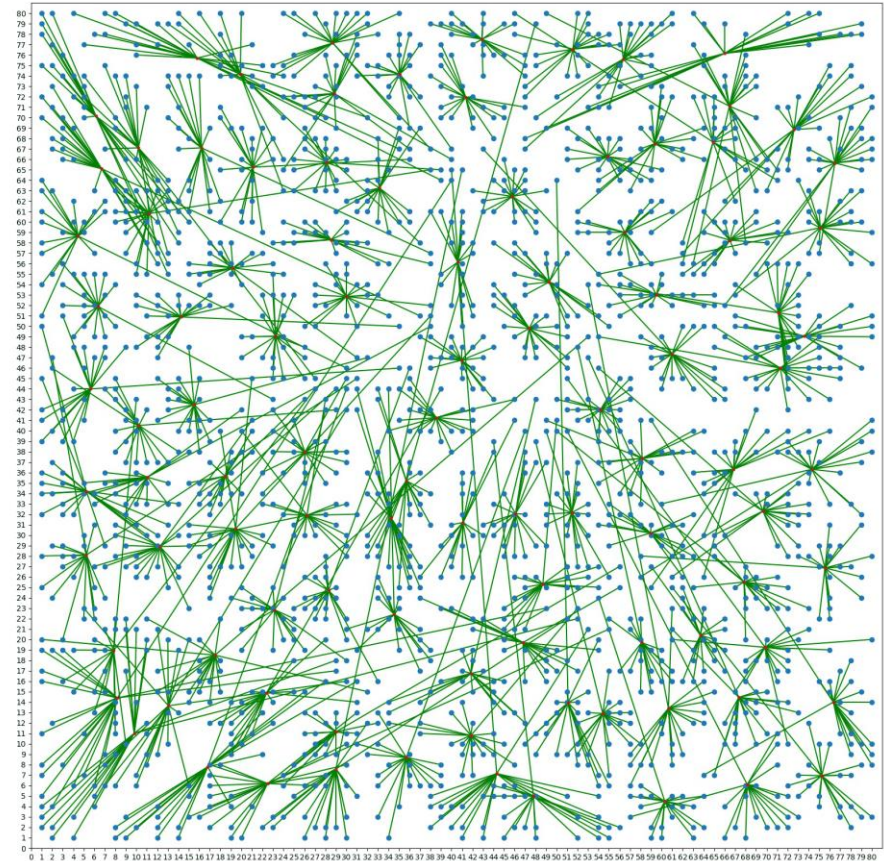




# INITIALIZATION ALGORITHM FOR K-MEANS



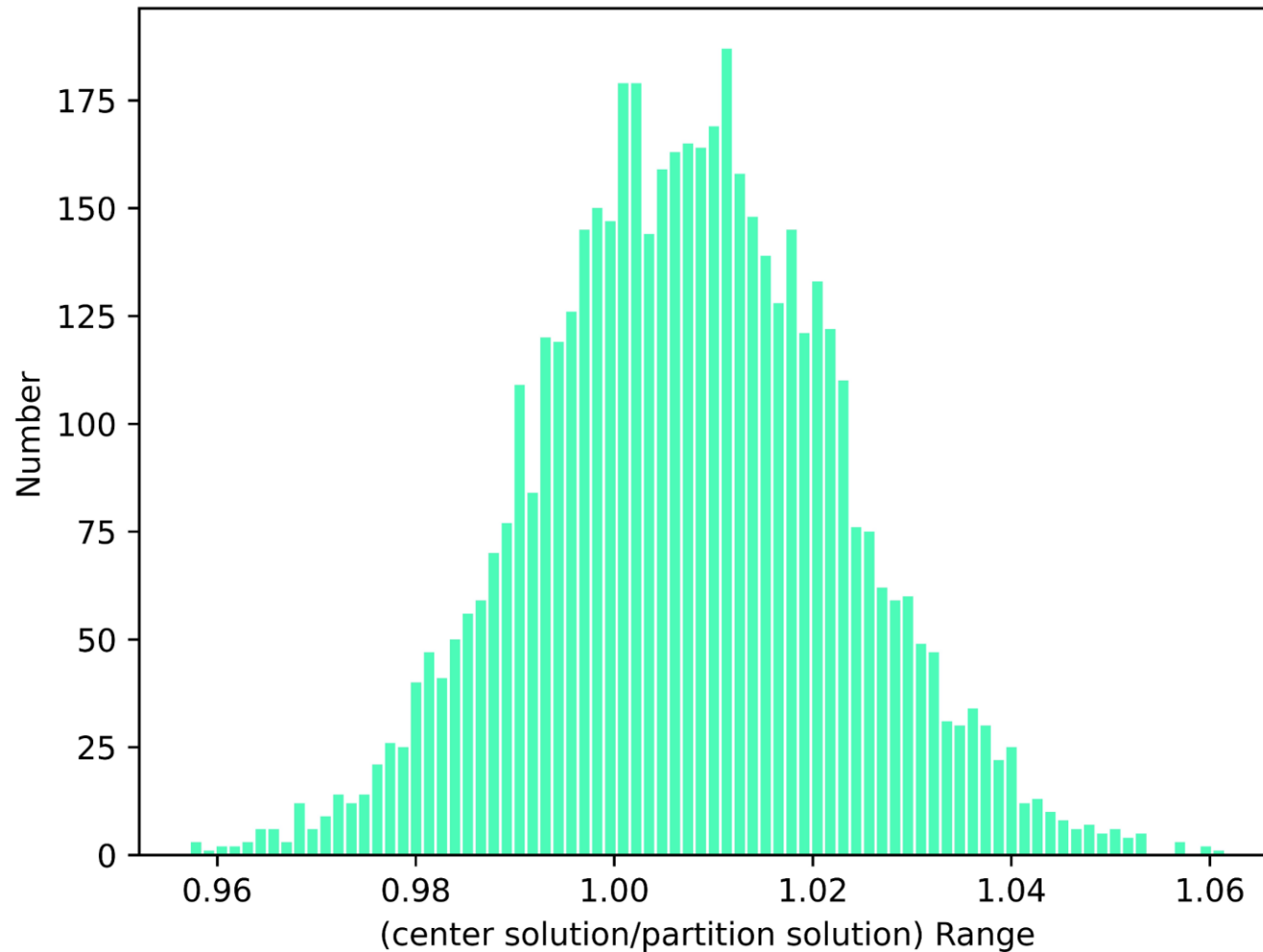
Initial by center

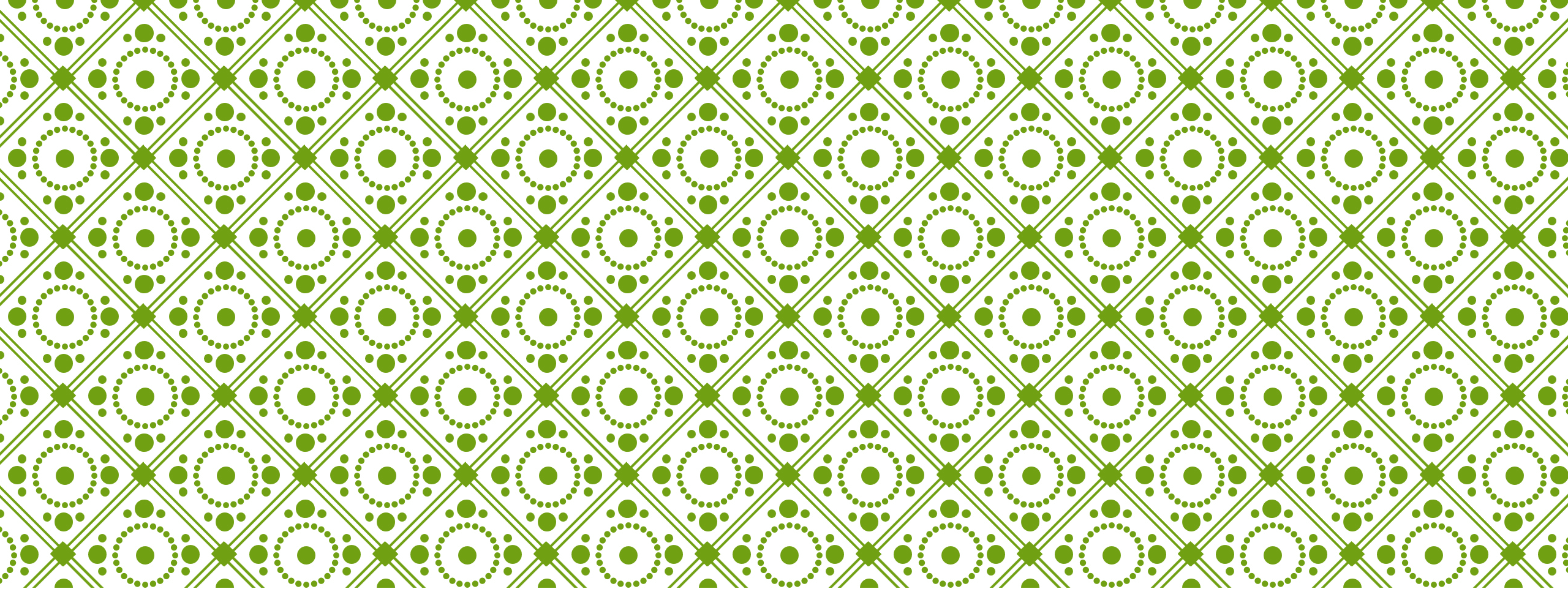


Initial by partition

# THE ALGORITHM TO INITIALIZATION ALGORITHM FOR K-MEANS

I generate 5000 cases. Each case have 2000 sites and will be divided into 100 partitions. I do initialization for them by center and partition respectively, then I do statistic for the solution of cases.





# THANK YOU !

Q&A