

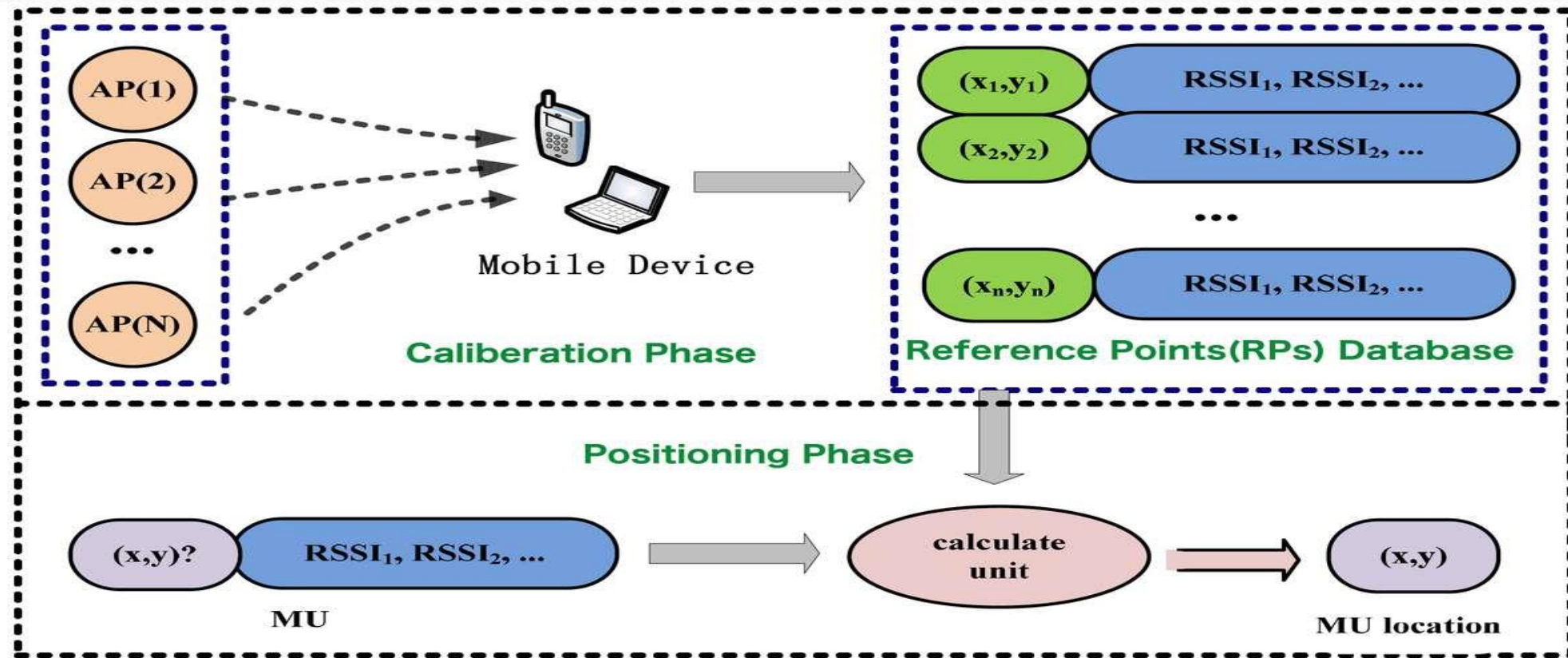
# Wifi positioning based on machine learning approach

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Rssi(reciving singnal strength index)&Ap(access point)&Rps(reference points)





# ***k*-nearest neighbors algorithm (*k*-NN)**

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This method was first proposed by Evelyn Fix and Joseph Hodges in 1951 and later developed by Thomas Cove

A classification (classification) is non-numerical, and here  $k$  is the closest fixed value to the test, which is a small and correct number chosen based on the fact that it is close to the majority of test samples.

For example, suppose we have the following mechanisms:

$$(X_1, Y_1), (X_2, Y_2), \dots, (X_n, Y_n)$$

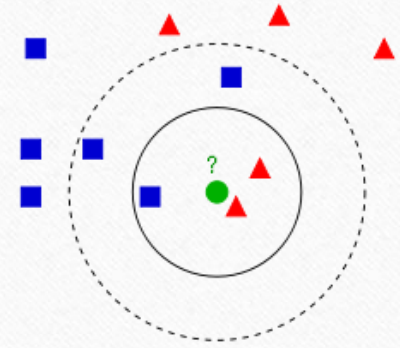
which takes those high values which are our test values to the that  $Y$  is the label infinite dimensional environment  $R^d \times \{1,2\}$  of class  $X$ . That is:

$$r=1,2 \text{ و } X|Y = r \sim P_r$$

$P_r$  The distribution is probability, so by using the which software suitable for the space, we get  $X$  and  $Y$  information as follows:

$$|X(1) - x| < |X(2) - x| < |X(3) - x| < \dots < |X(n) - x|$$

For example, in the figure above, if  $k=3$ , the straight line becomes our decision area, and we vote for the triangle, and if  $k=5$ , the dotted line becomes our decision area, and we vote for the square, and we vote for the circle. We are green



#### Distance functions

Euclidean

$$\sqrt{\sum_{i=1}^k (x_i - y_i)^2}$$

Manhattan

$$\sum_{i=1}^k |x_i - y_i|$$

Minkowski

$$\left( \sum_{i=1}^k (|x_i - y_i|)^q \right)^{1/q}$$

To calculate these distances, if the numbers are continuous, we can use face-to-face relationships:

In fact, we are counting their softness

We have used the first relation



X	Y	Distance
Male	Male	0
Male	Female	1

#### Hamming Distance

$$D_H = \sum_{i=1}^k |x_i - y_i|$$

$$x = y \Rightarrow D = 0$$

$$x \neq y \Rightarrow D = 1$$

For categories such as male and female, we use face-to-face relationship:

## weighted k-nearest neighbors algorithm (k-NN)

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In this method, we weight  $K$  nearest neighbors with  $1/K$  weight and the rest with zero weight so that  $\sum \omega_{ni} = 1$ . It has the following relations

$$\mathcal{R}_{\mathcal{R}}(C_n^{runn}) - \mathcal{R}_{\mathcal{R}}(C^{Bayes}) = (B_1 s_n^2 + B_2 t_n^2) \{1 + o(1)\},$$

for constants  $B_1$  and  $B_2$  where  $s_n^2 = \sum_{i=1}^n w_{ni}^2$  and  $t_n = n^{-2/d} \sum_{i=1}^n w_{ni} \{i^{1+2/d} - (i-1)^{1+2/d}\}$ .

The optimal weighting scheme  $\{w_{ni}^*\}_{i=1}^n$ , that balances the two terms in the display above, is given as follows: set  $k^* = \lfloor Bn^{\frac{4}{d+4}} \rfloor$ ,

$$w_{ni}^* = \frac{1}{k^*} \left[ 1 + \frac{d}{2} - \frac{d}{2k^{*2/d}} \{i^{1+2/d} - (i-1)^{1+2/d}\} \right] \text{ for } i = 1, 2, \dots, k^* \text{ and}$$

$$w_{ni}^* = 0 \text{ for } i = k^* + 1, \dots, n.$$

