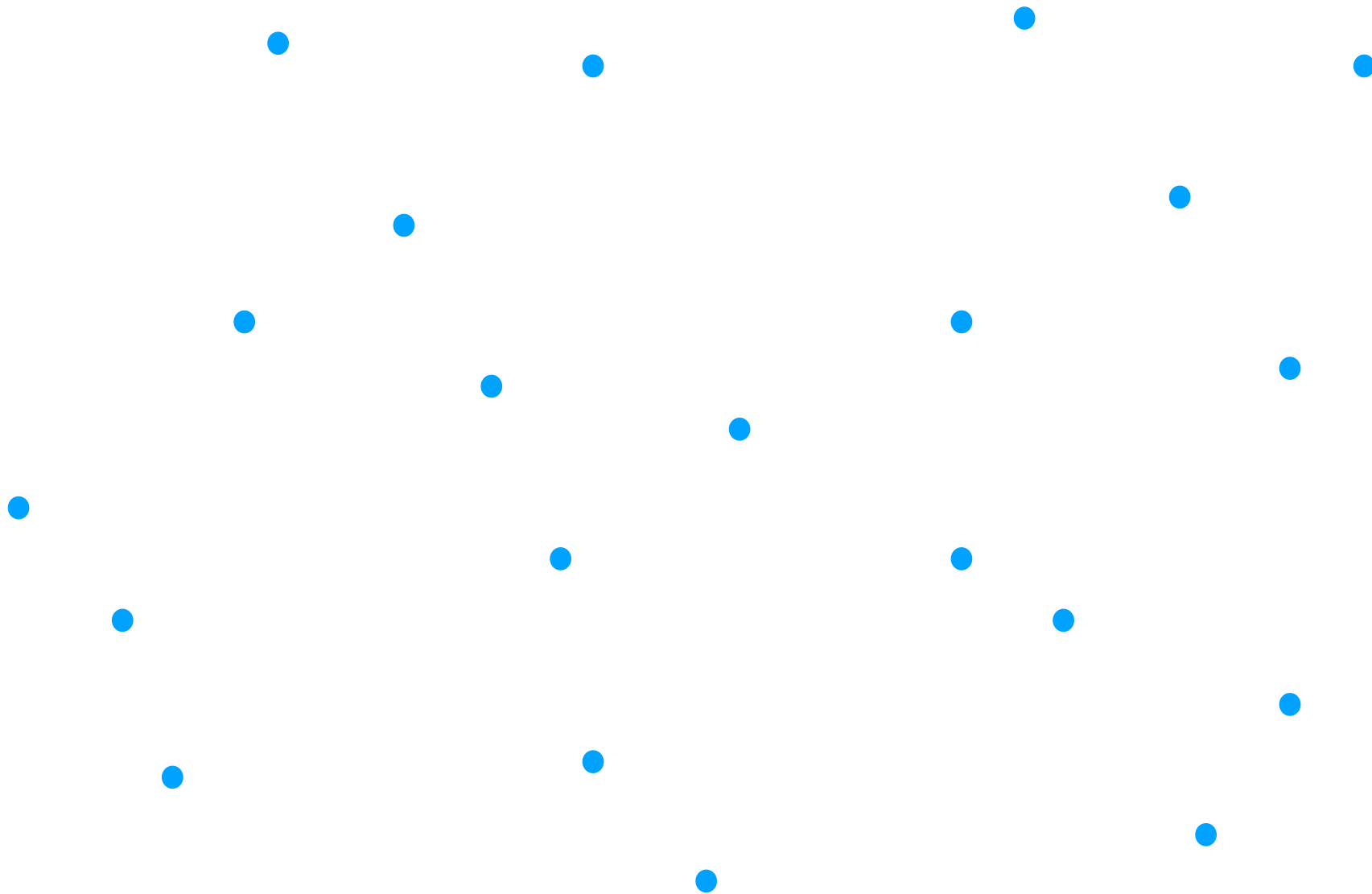
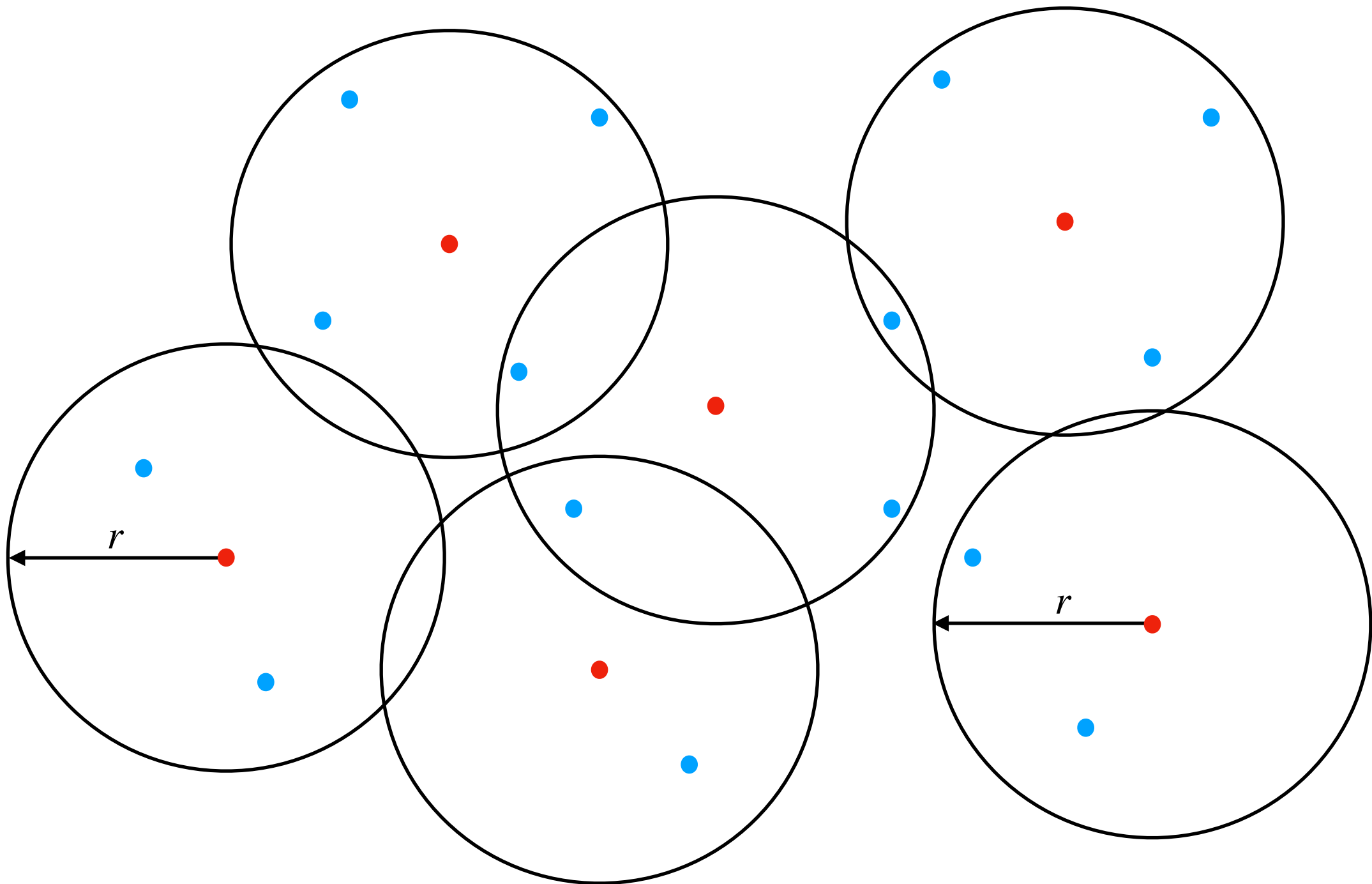


# High Dimensional Clustering and Applications

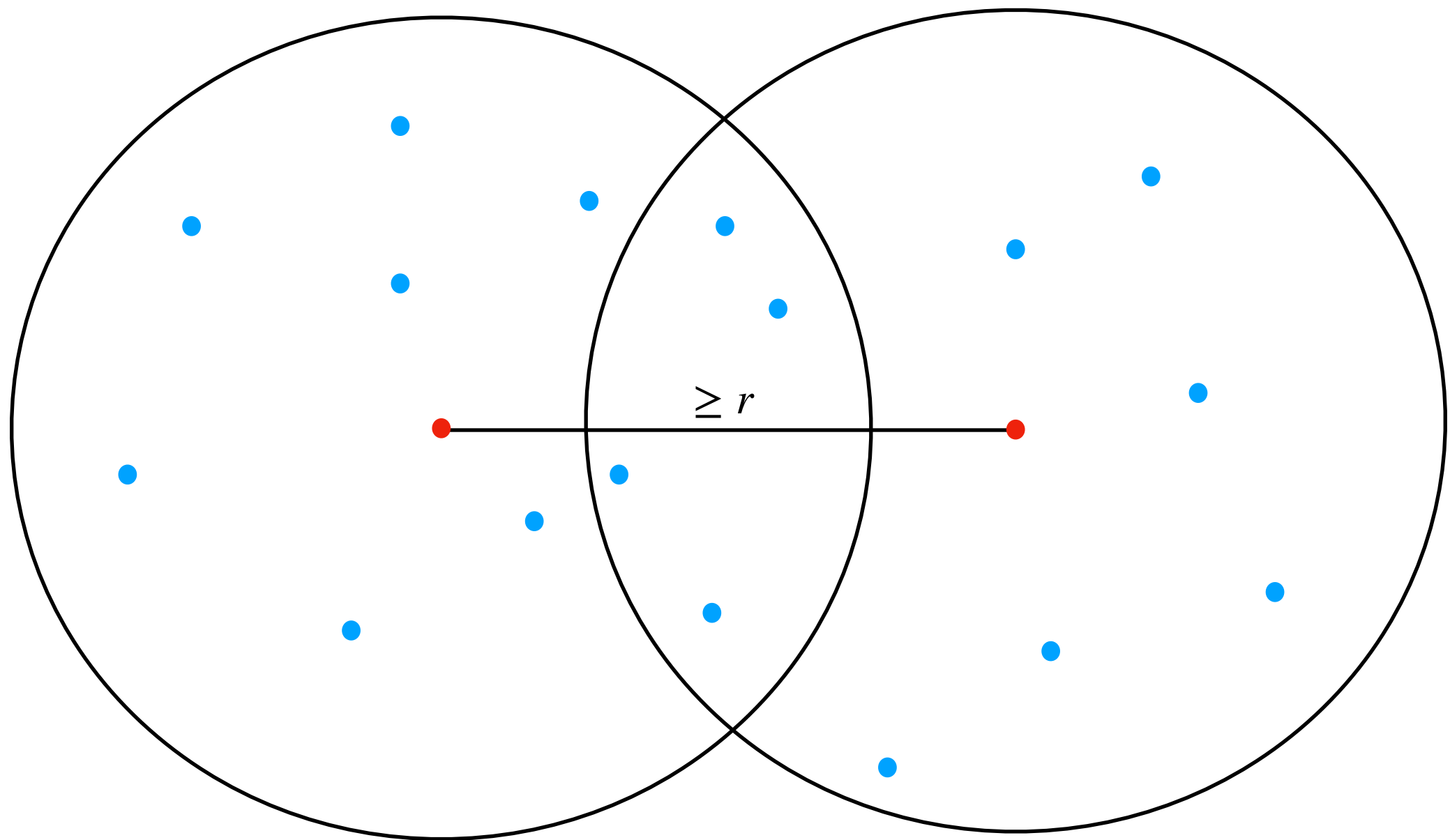
# R-nets



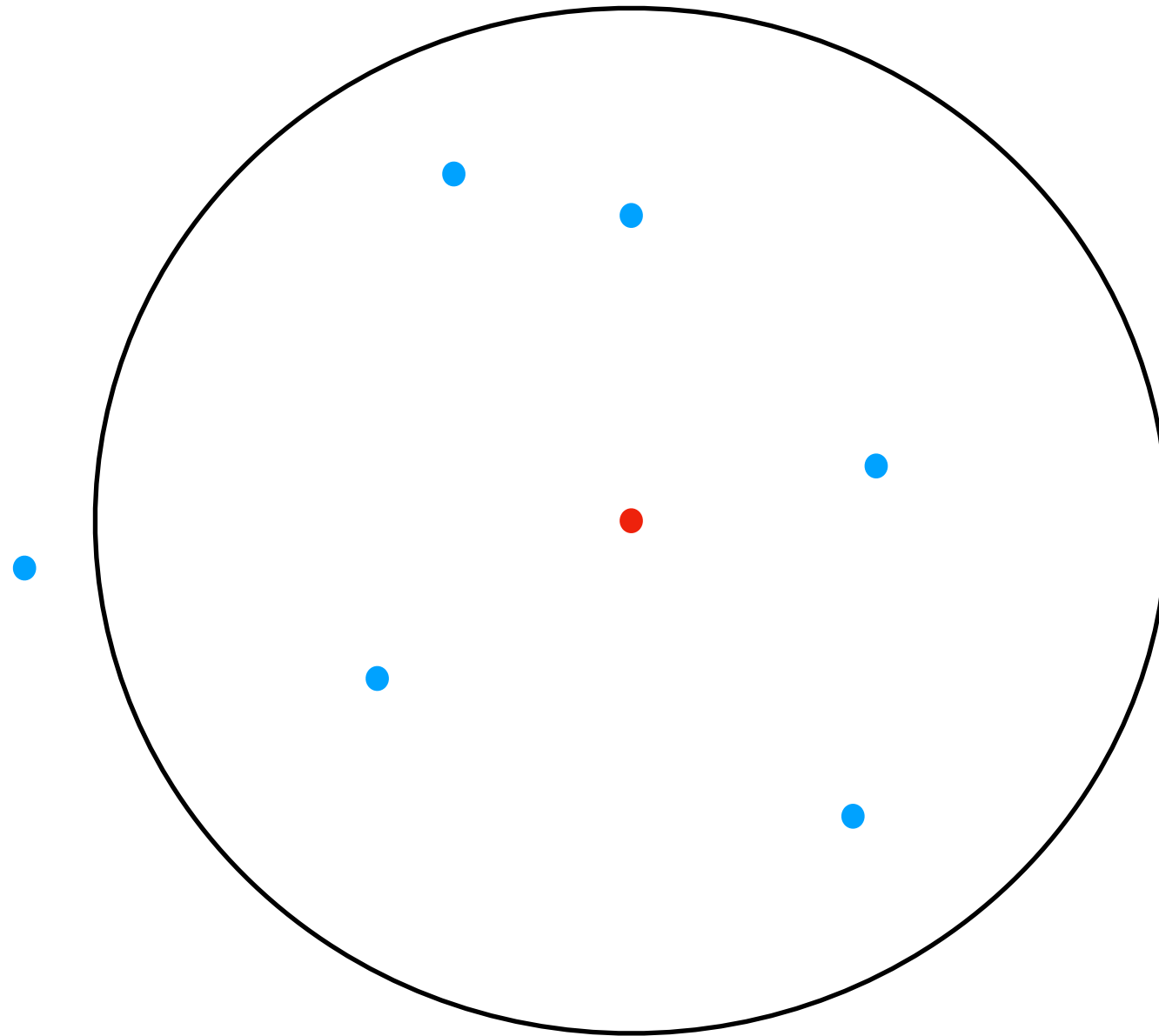
# Covering



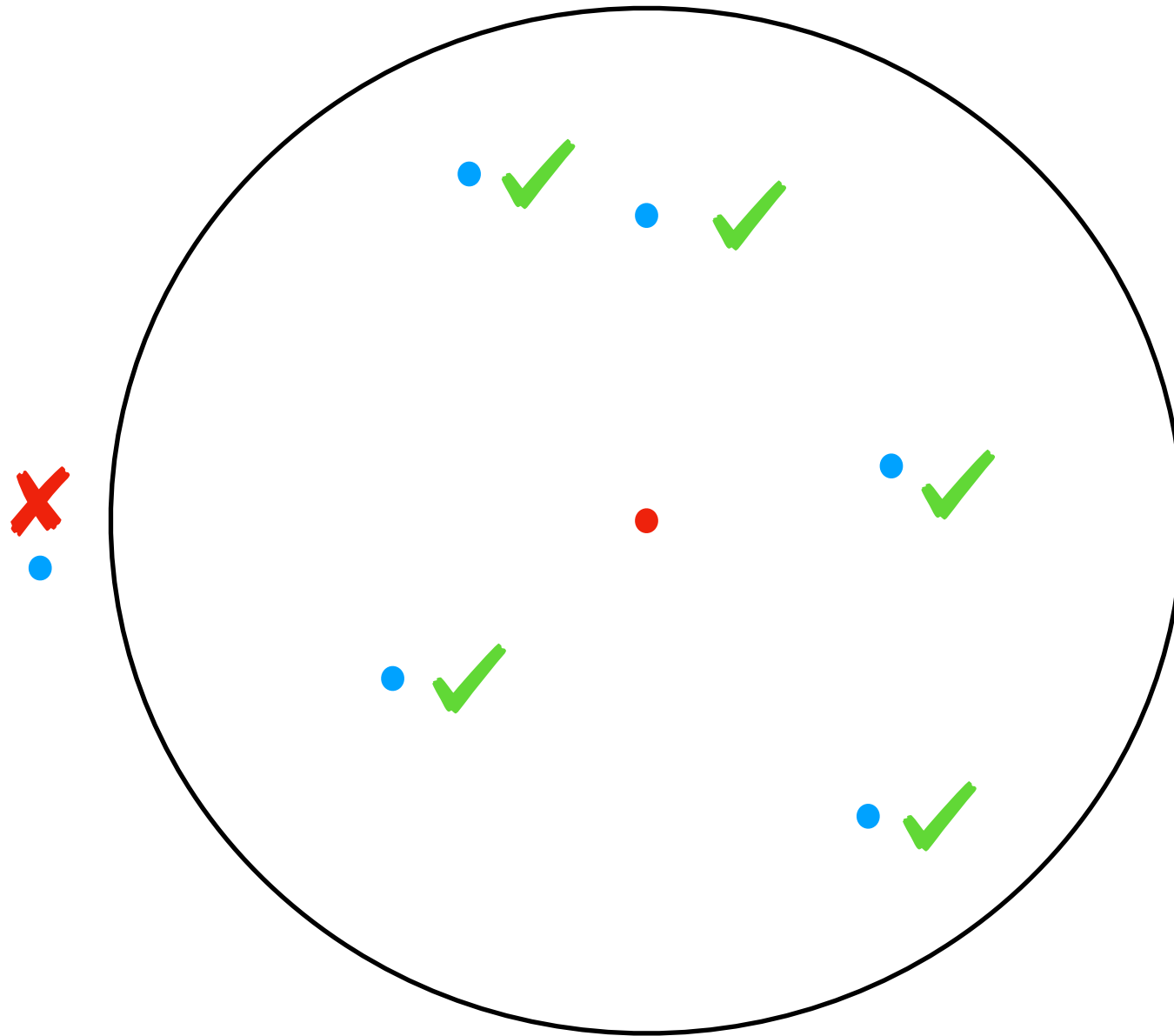
# Packing



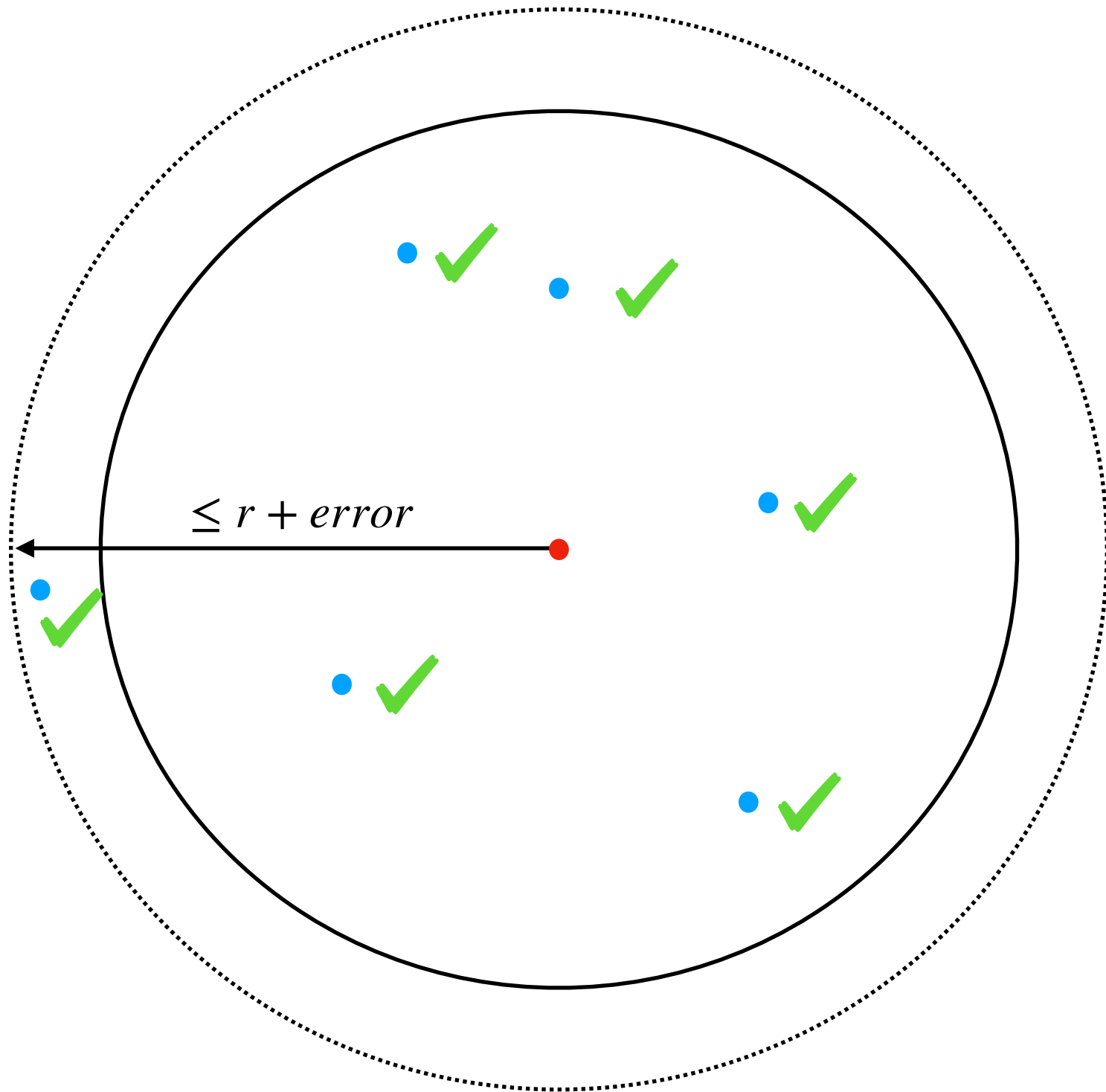
# Approximate r-net



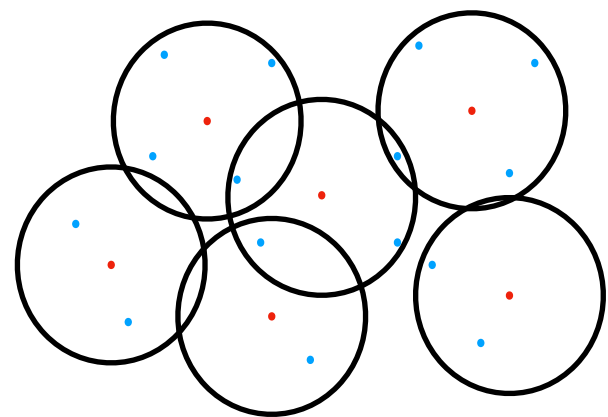
# Approximate r-net



# Lift covering



R-nets

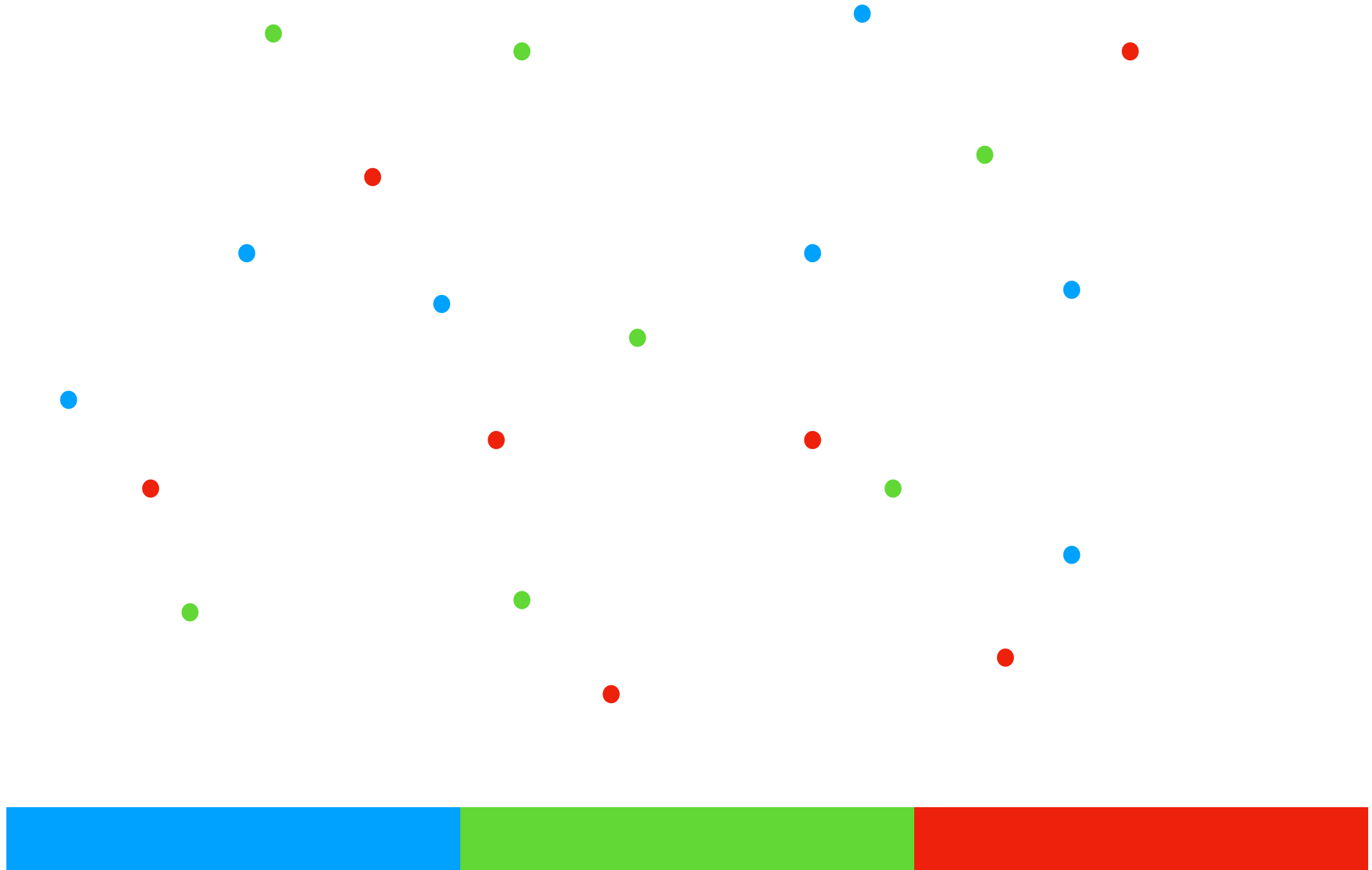


Distance matrix

	$p_1$	$p_2$		$p_n$
	yes	no	...	yes
	no	no	...	yes
	no	yes	...	no



# Grouping points

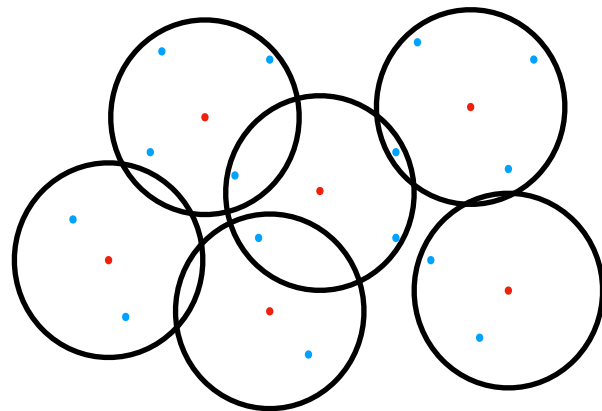


# Distance matrix

	$p_1$	$p_2$	...	$p_n$
	yes	no	...	yes
	no	no	...	yes
	no	yes	...	no

How do we efficiently build  
such a matrix?

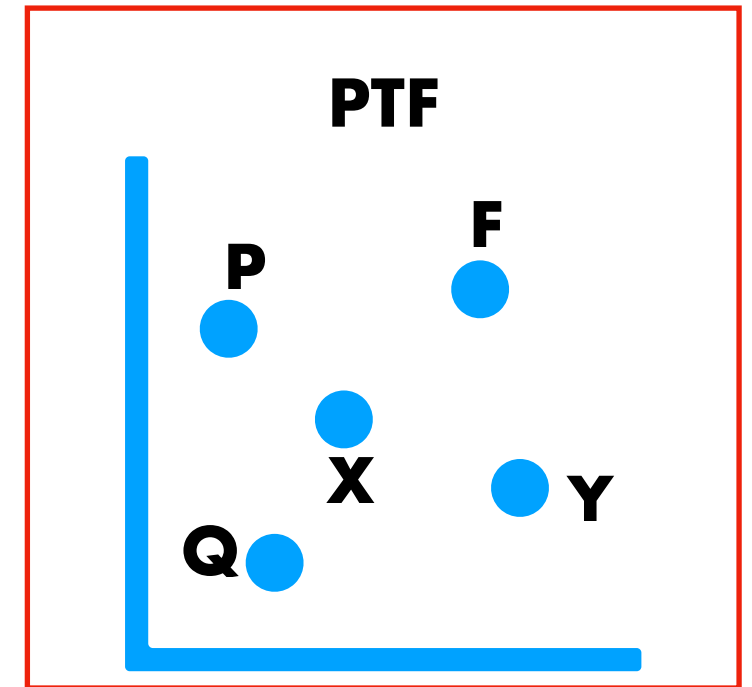
## R-nets



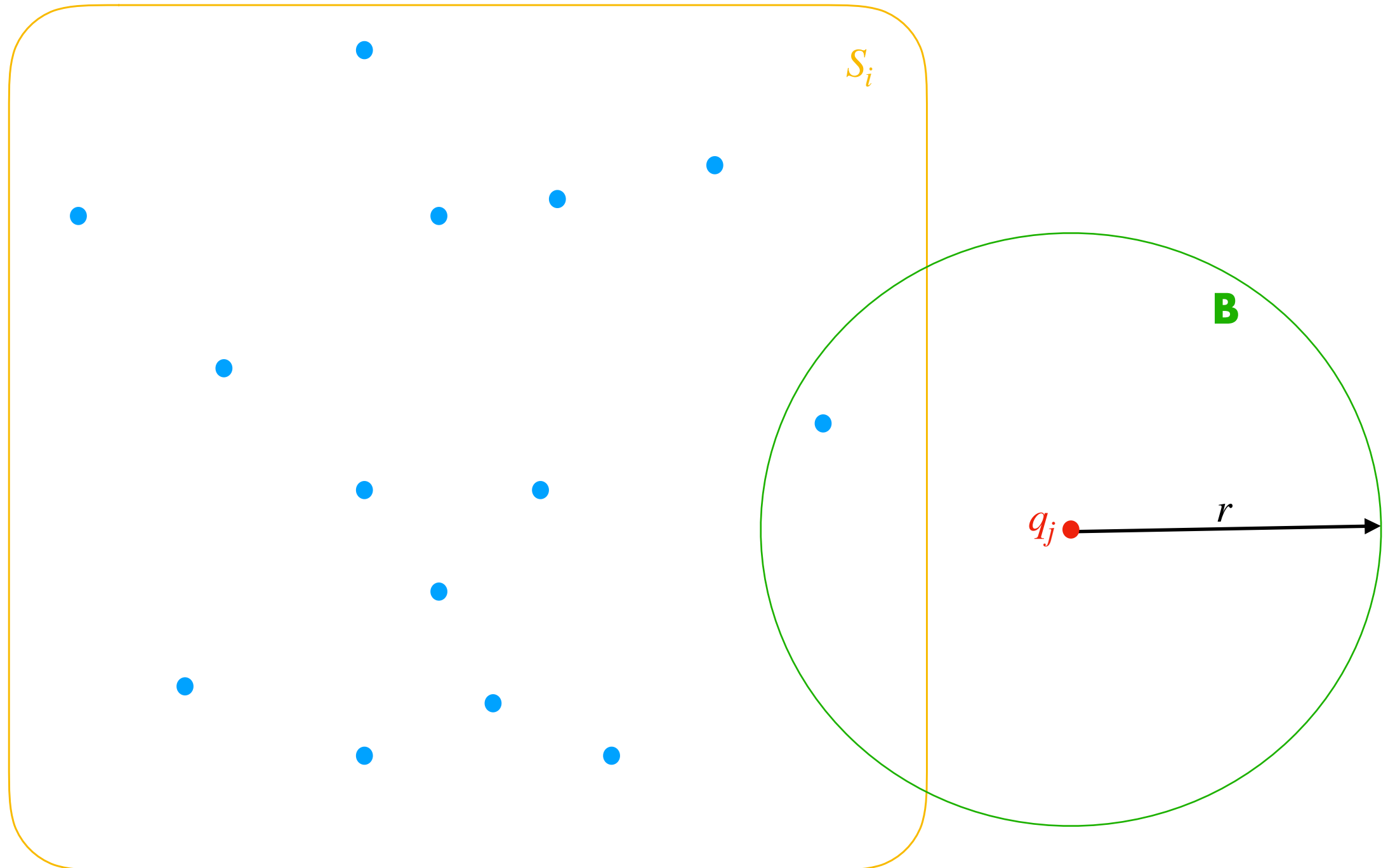
## Distance matrix

	$p_1$	$p_2$		$p_n$
	yes	no	...	yes
	no	no	...	yes
	no	yes	...	no

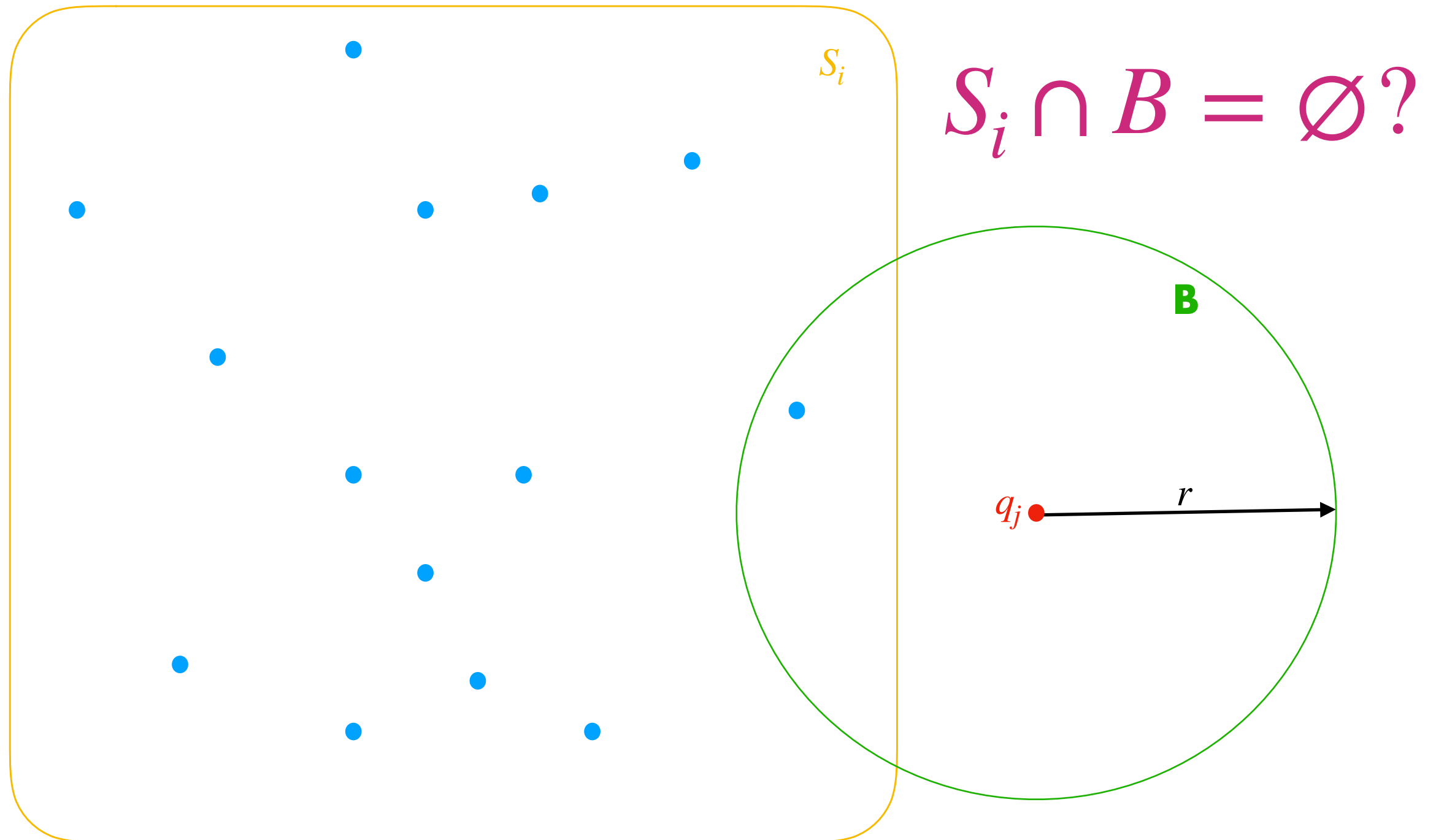
## PTF



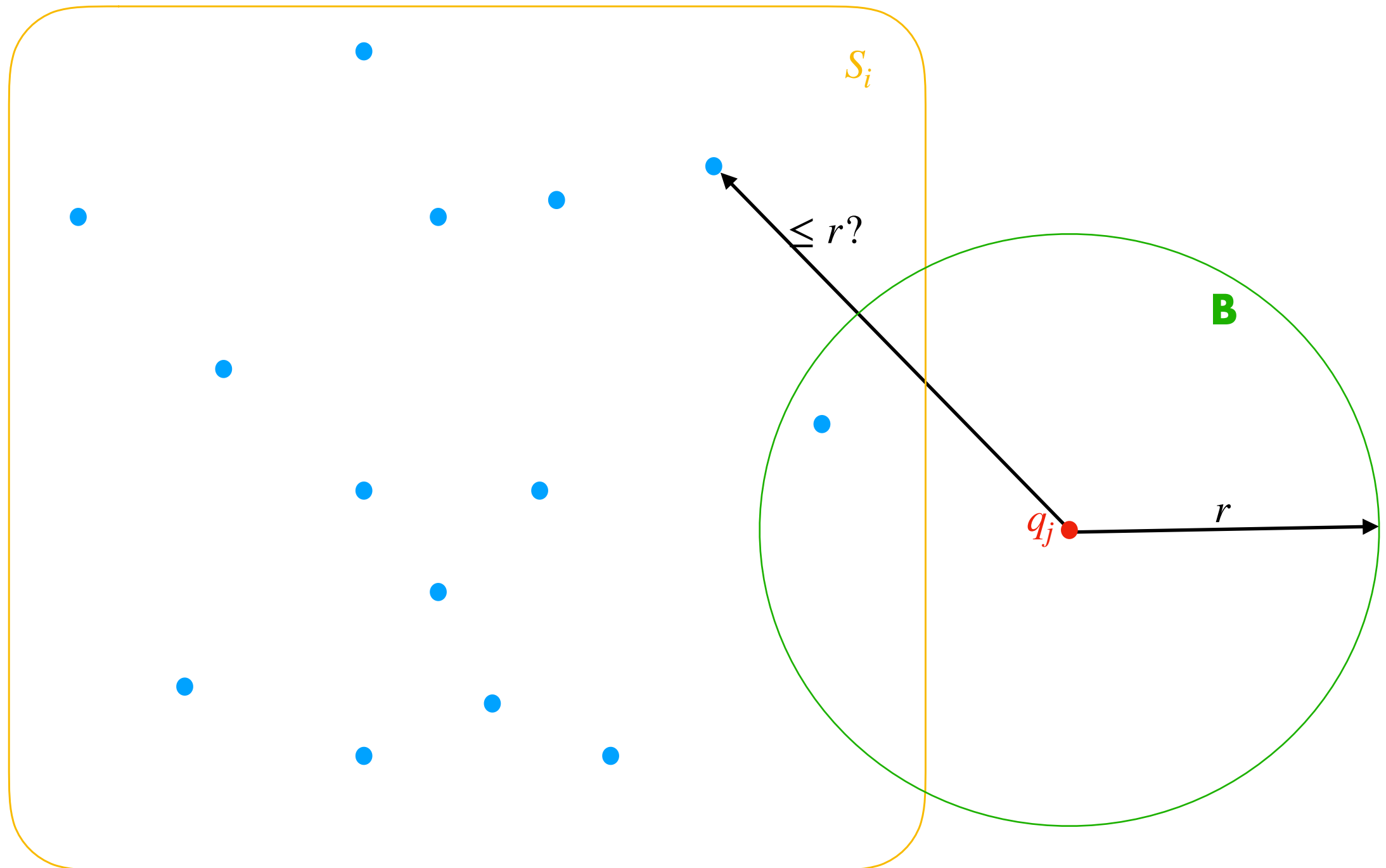
# Imagine one entry



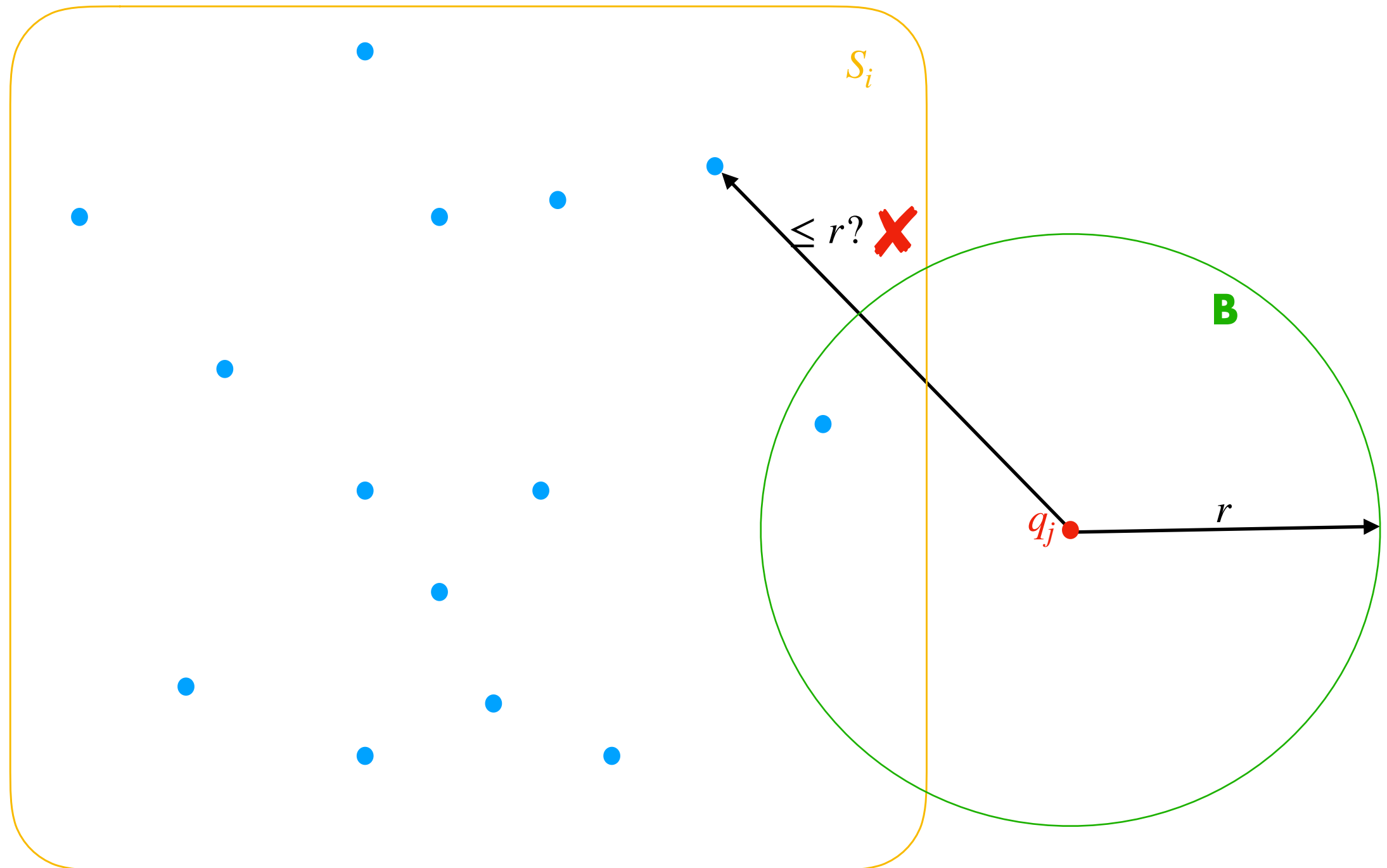
# Naive search algorithm



# Naive search algorithm

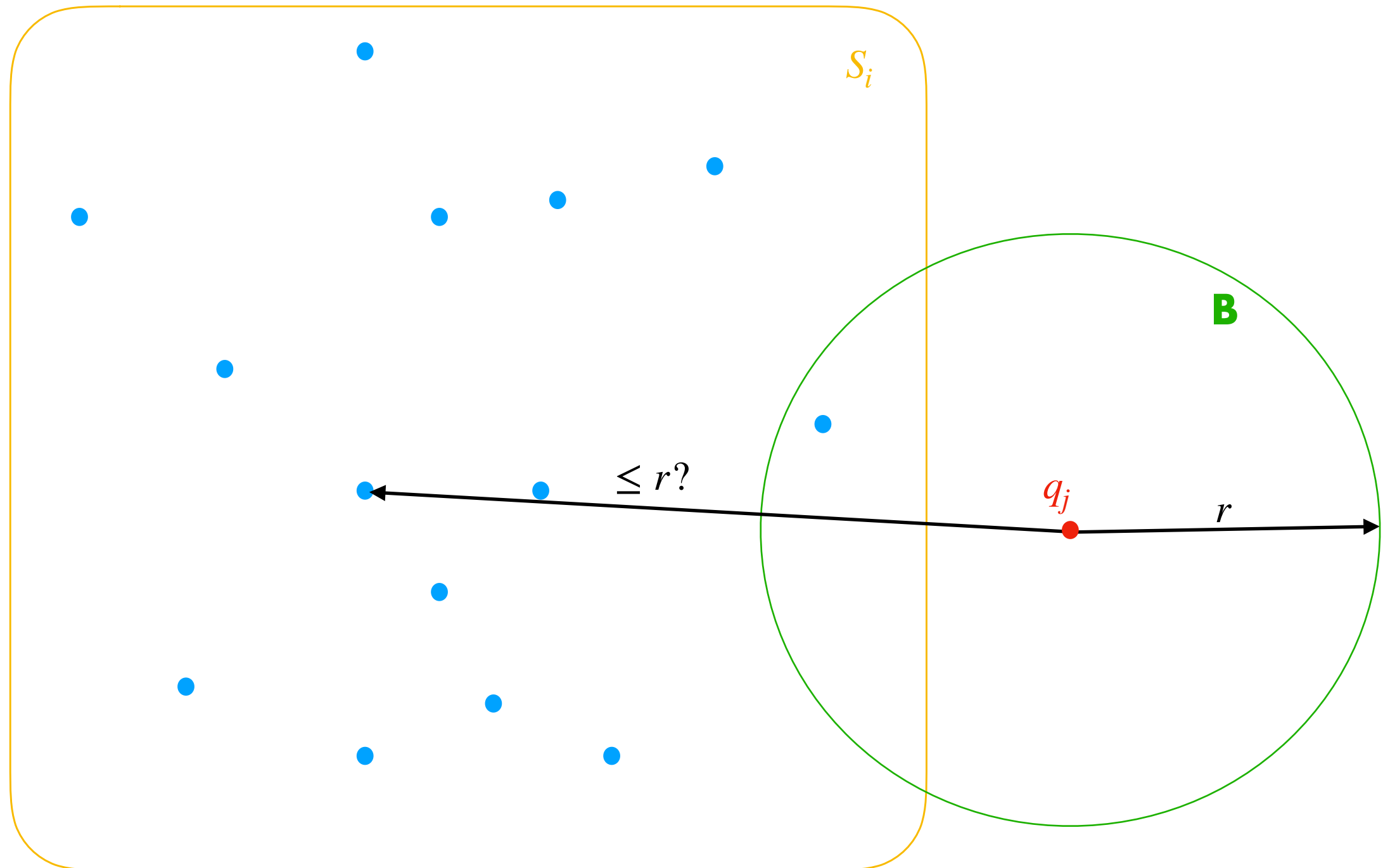


# Naive search algorithm

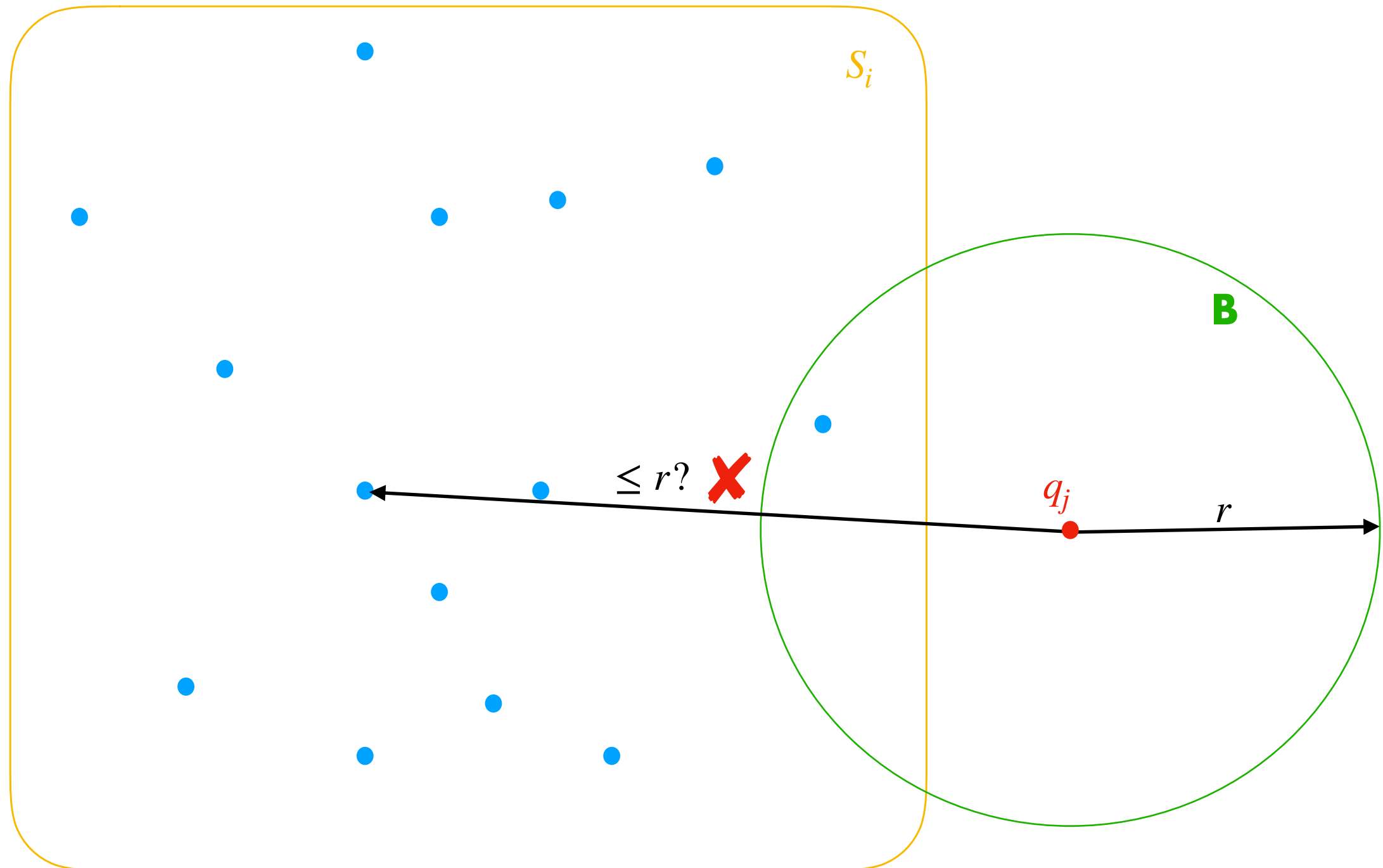




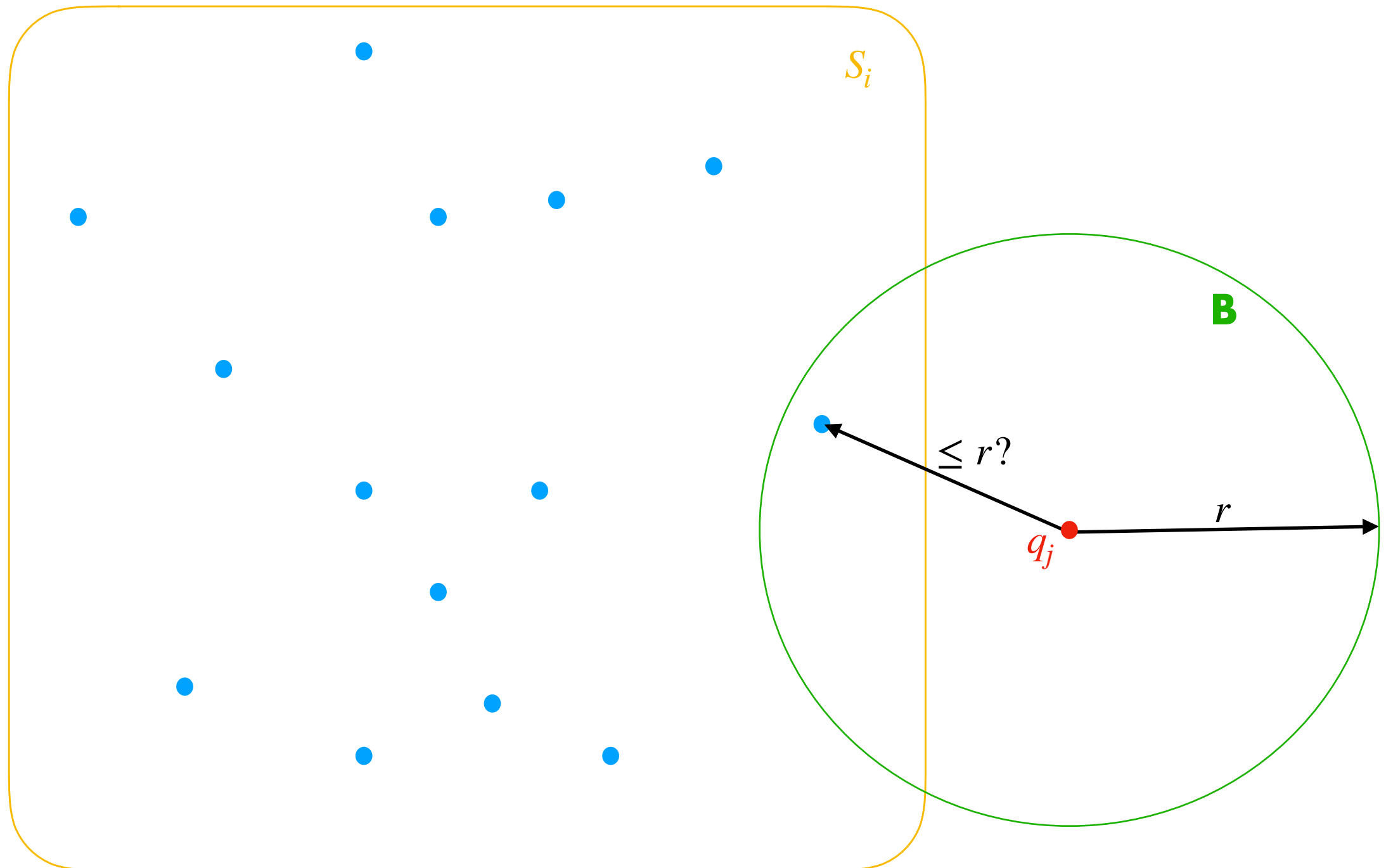
# Naive search algorithm



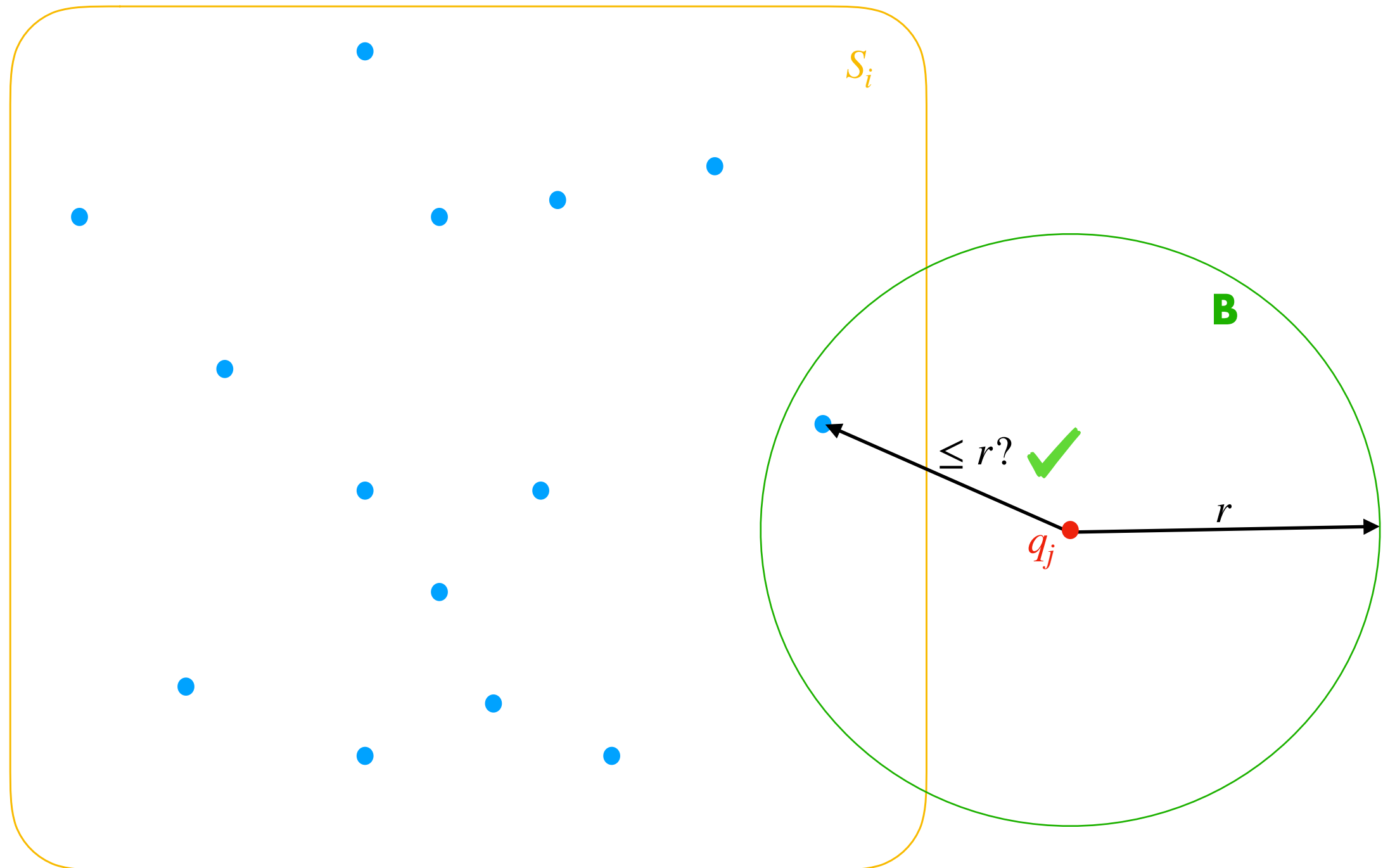
# Naive search algorithm



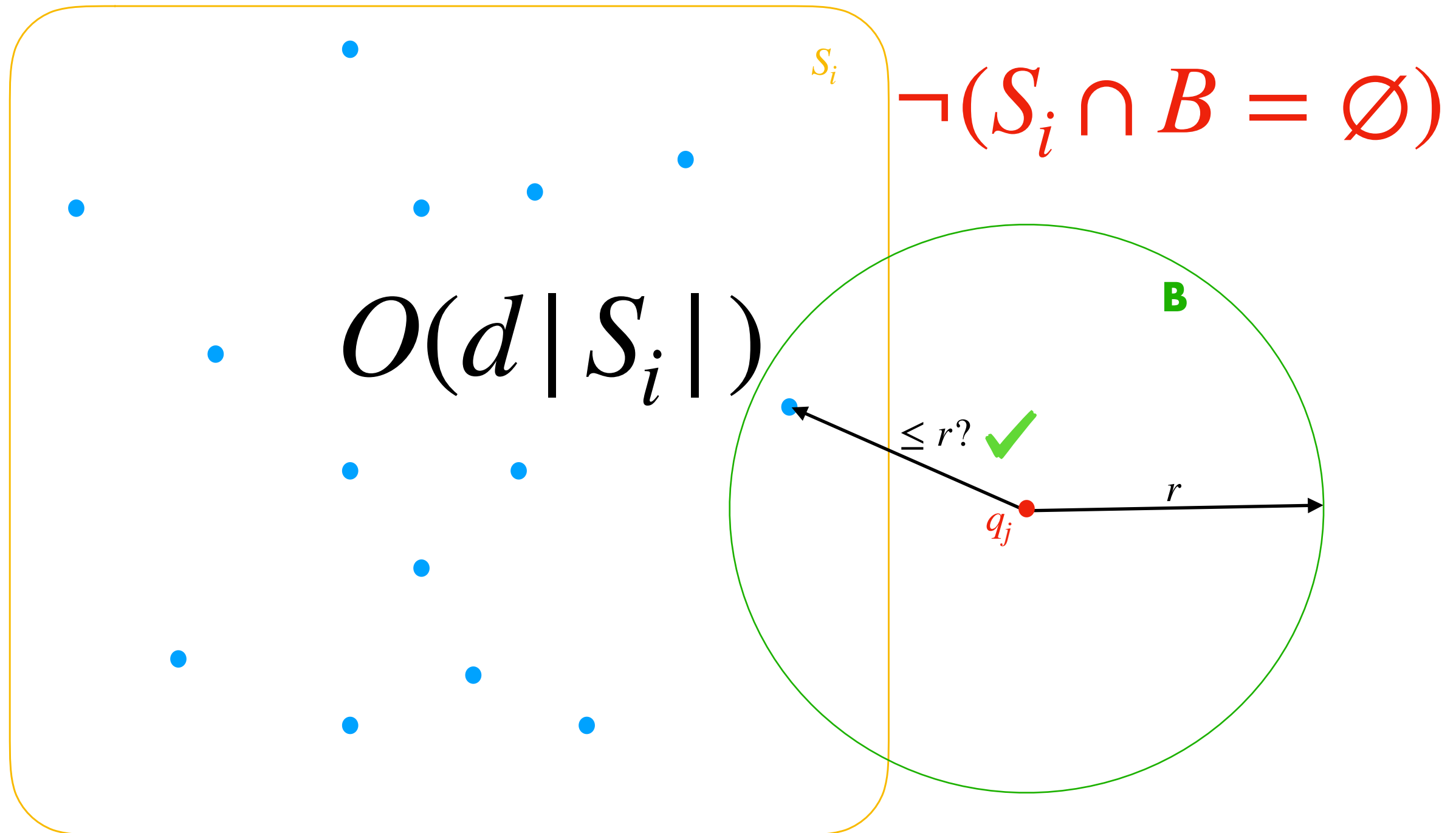
# Naive search algorithm



# Naive search algorithm

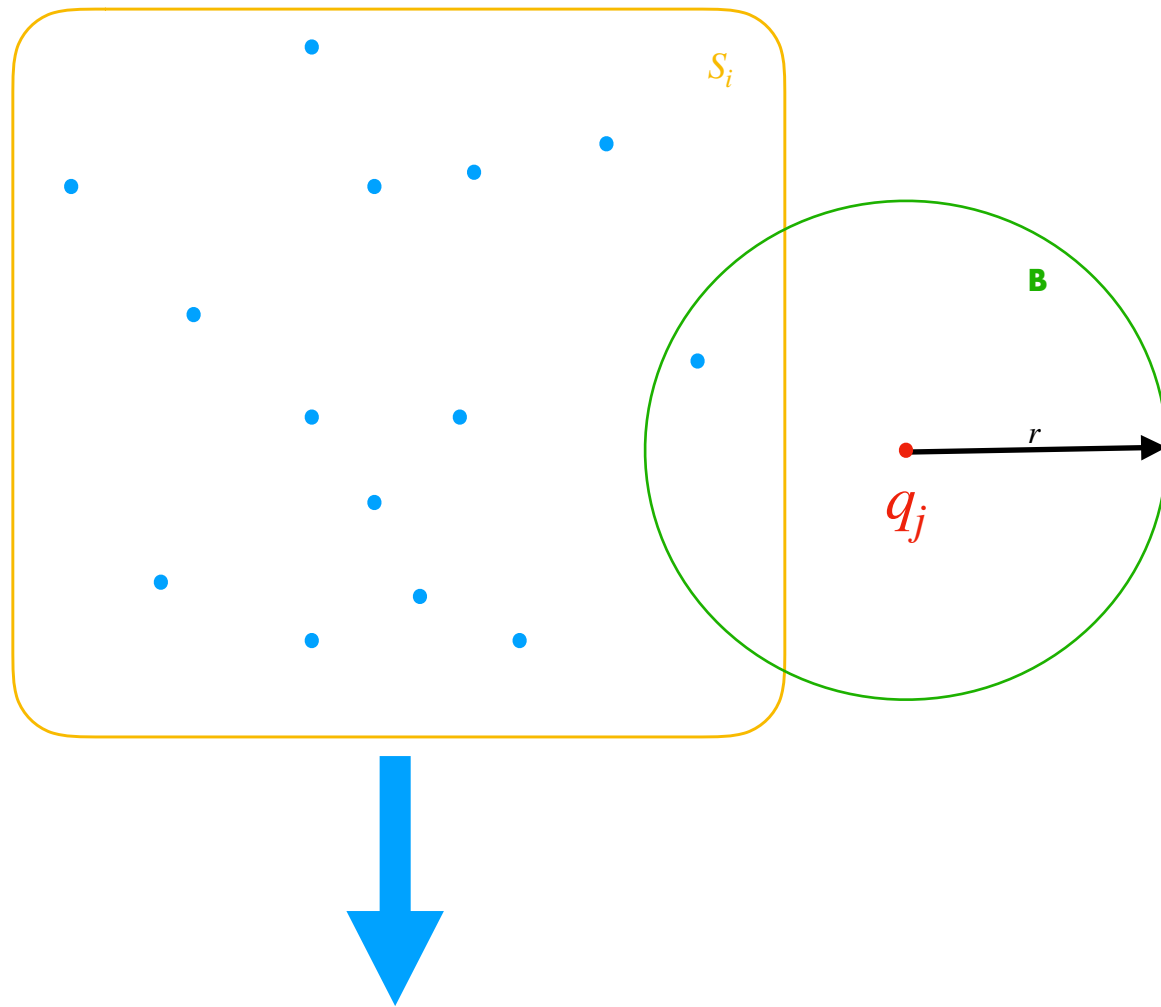


# Naive search algorithm



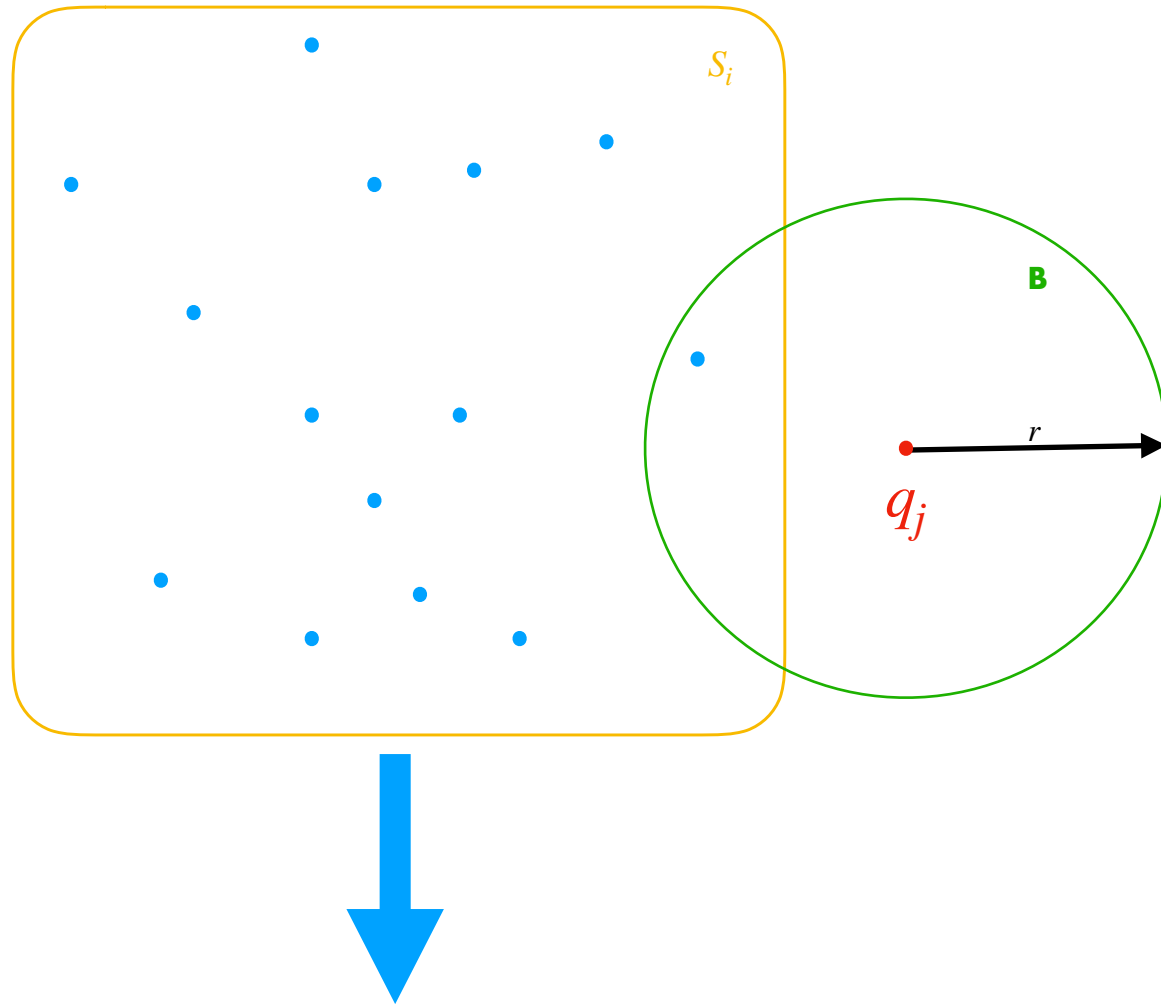
Building the matrix costs as much as a naive algorithm for building the r-net!

# Probabilistic polynomial thresholding functions (PTF)



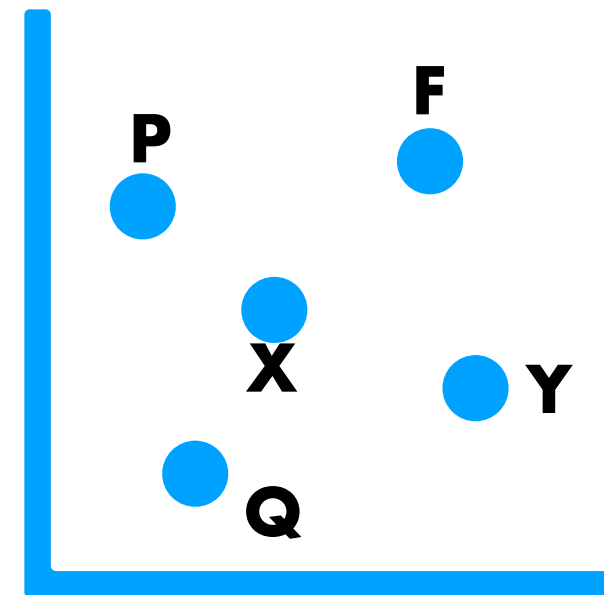
$$Z := \min_{p \in S_i} (\text{dist}(p, q_j)) \leq r$$

# Probabilistic polynomial thresholding functions (PTF)



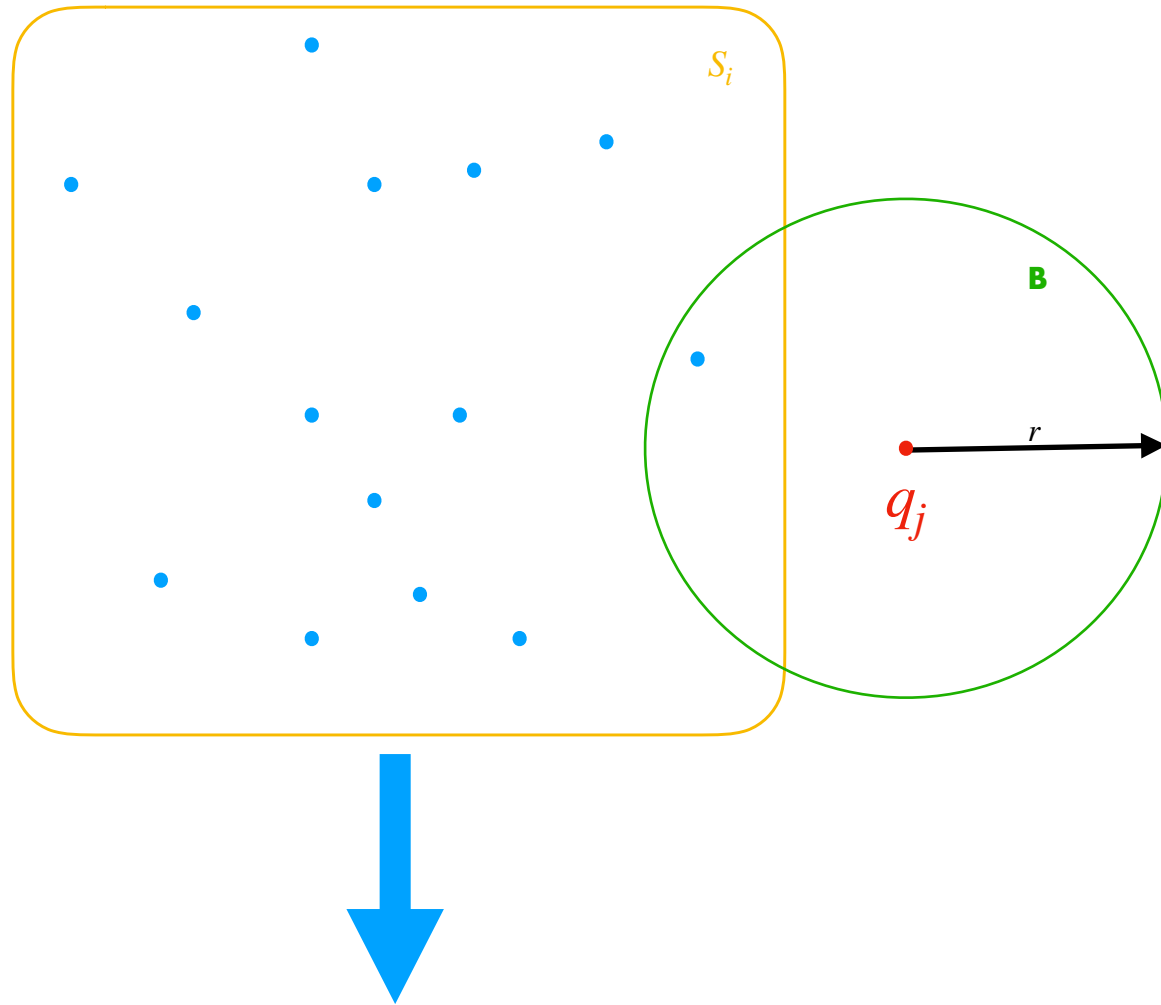
$$Z := \min_{p \in S_i} (\text{dist}(p, q_j)) \leq r$$

Distribution of Polynomials

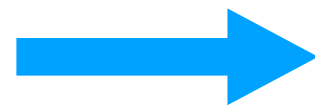




# Probabilistic polynomial thresholding functions (PTF)

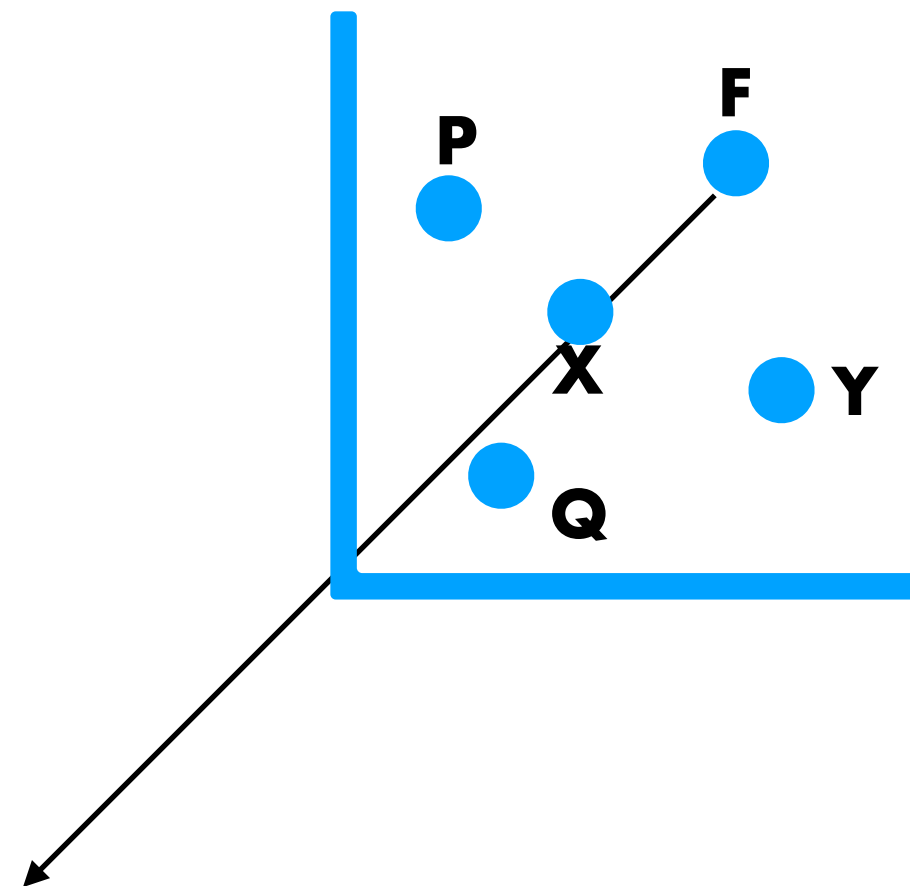


$$Z := \min_{p \in S_i} (\text{dist}(p, q_j)) \leq r$$



if  $Z \leq r$  then  $F(p_1, p_2, \dots, p_n, q_j) \geq 2|S_i|$  with some probability  
 if  $Z > r + \epsilon d$  then  $F(p_1, p_2, \dots, p_n, q_j) \leq |S_i|$  with some probability

Distribution of Polynomials

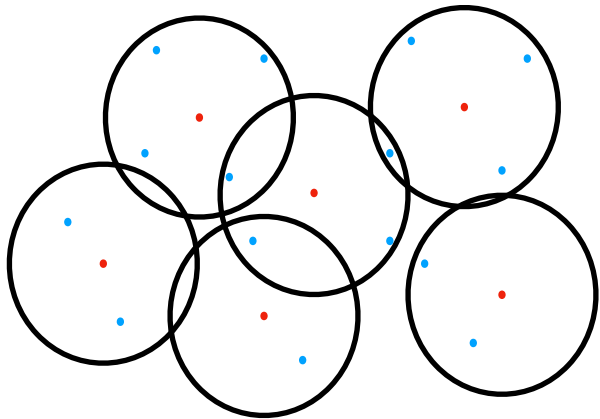


Leads to improvement of runtime  
for building the distance matrix

	$p_1$	$p_2$	...	$p_n$
	yes	yes	...	yes
	yes	yes	...	yes
	yes	yes	...	yes

**What now?**

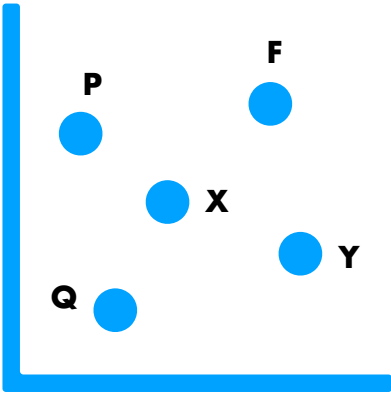
**R-nets**



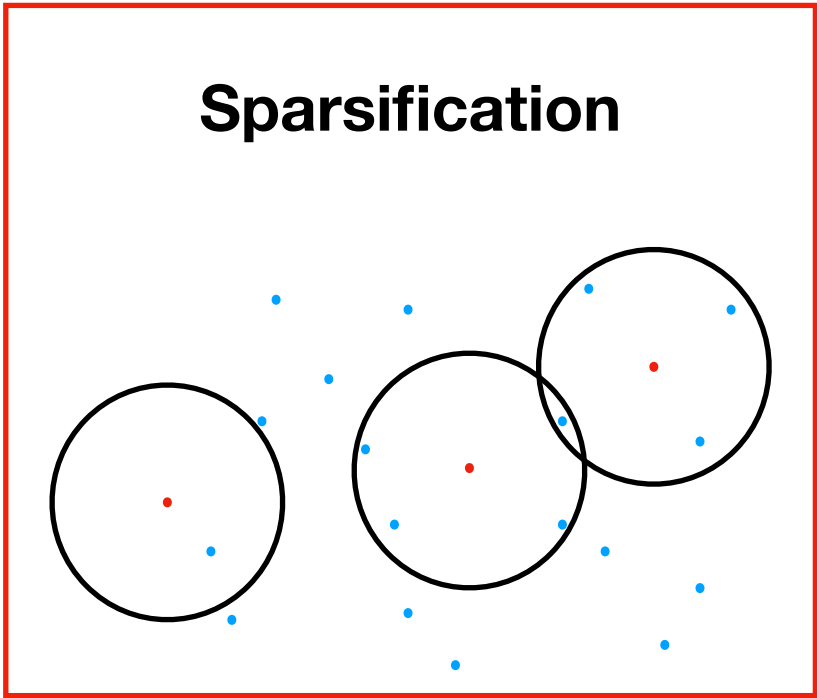
**Distance matrix**

	$p_1$	$p_2$		$p_n$
blue	yes	no	...	yes
green	no	no	...	yes
red	no	yes	...	no

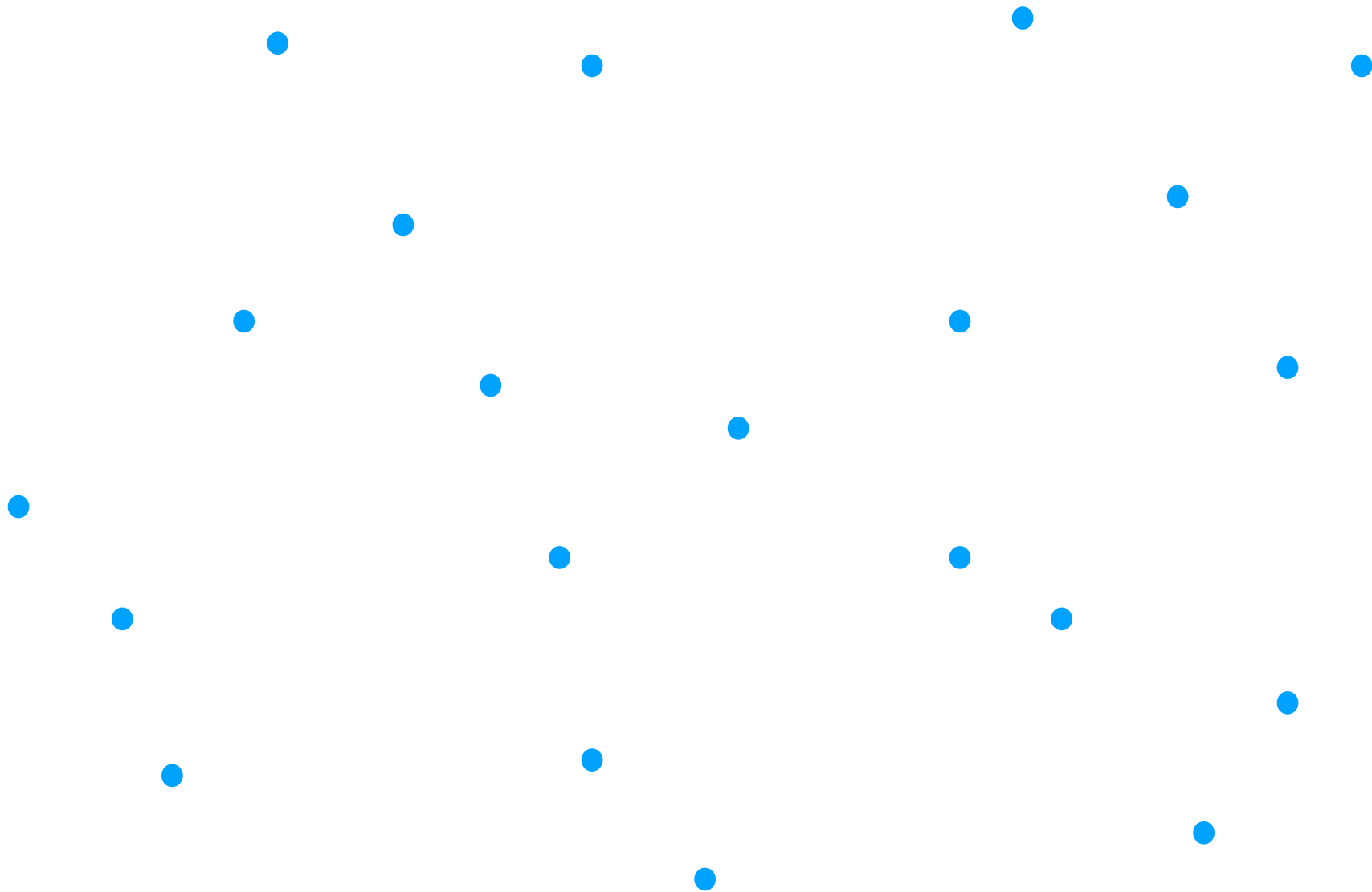
**PTF**



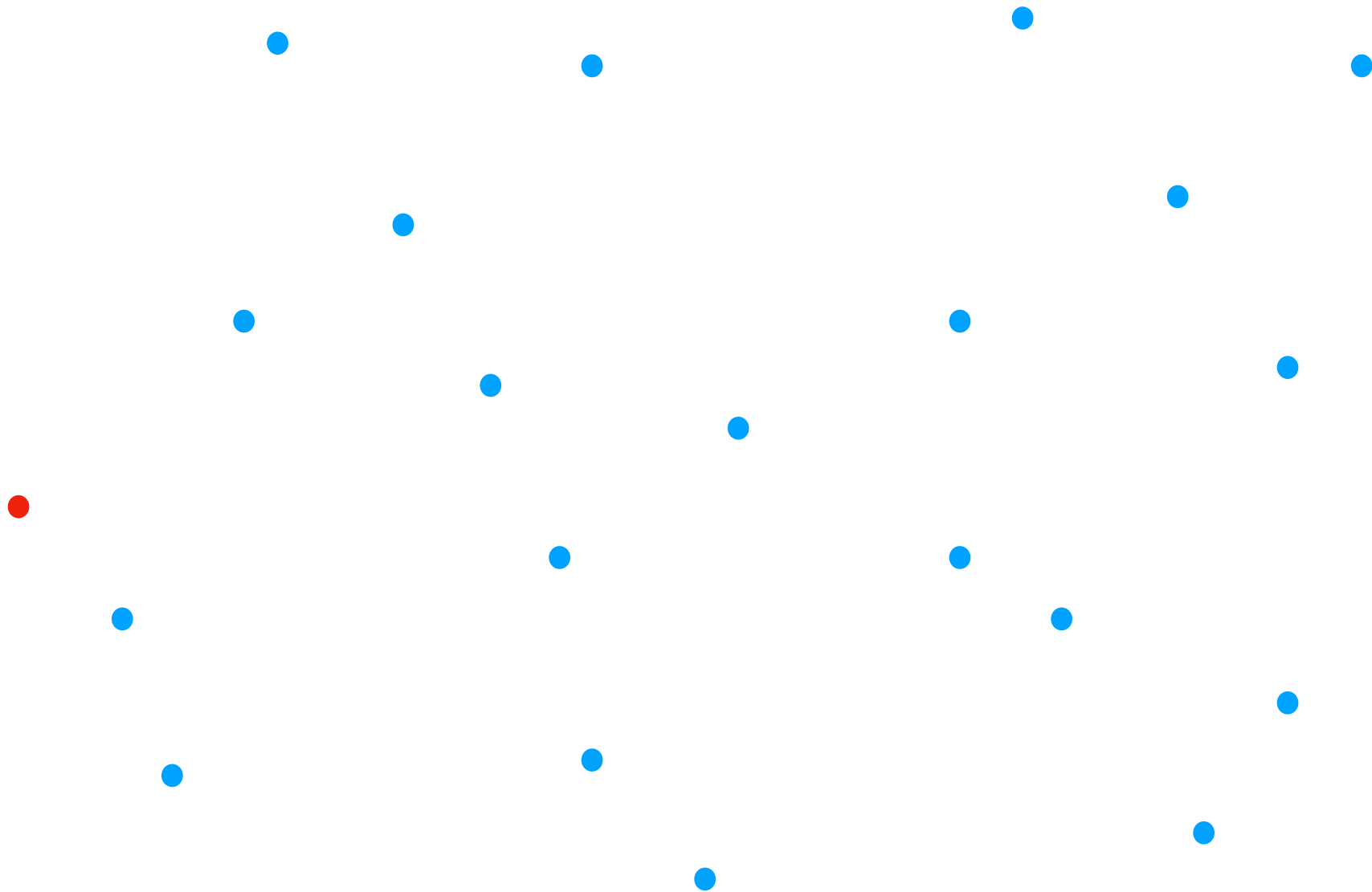
**Sparsification**



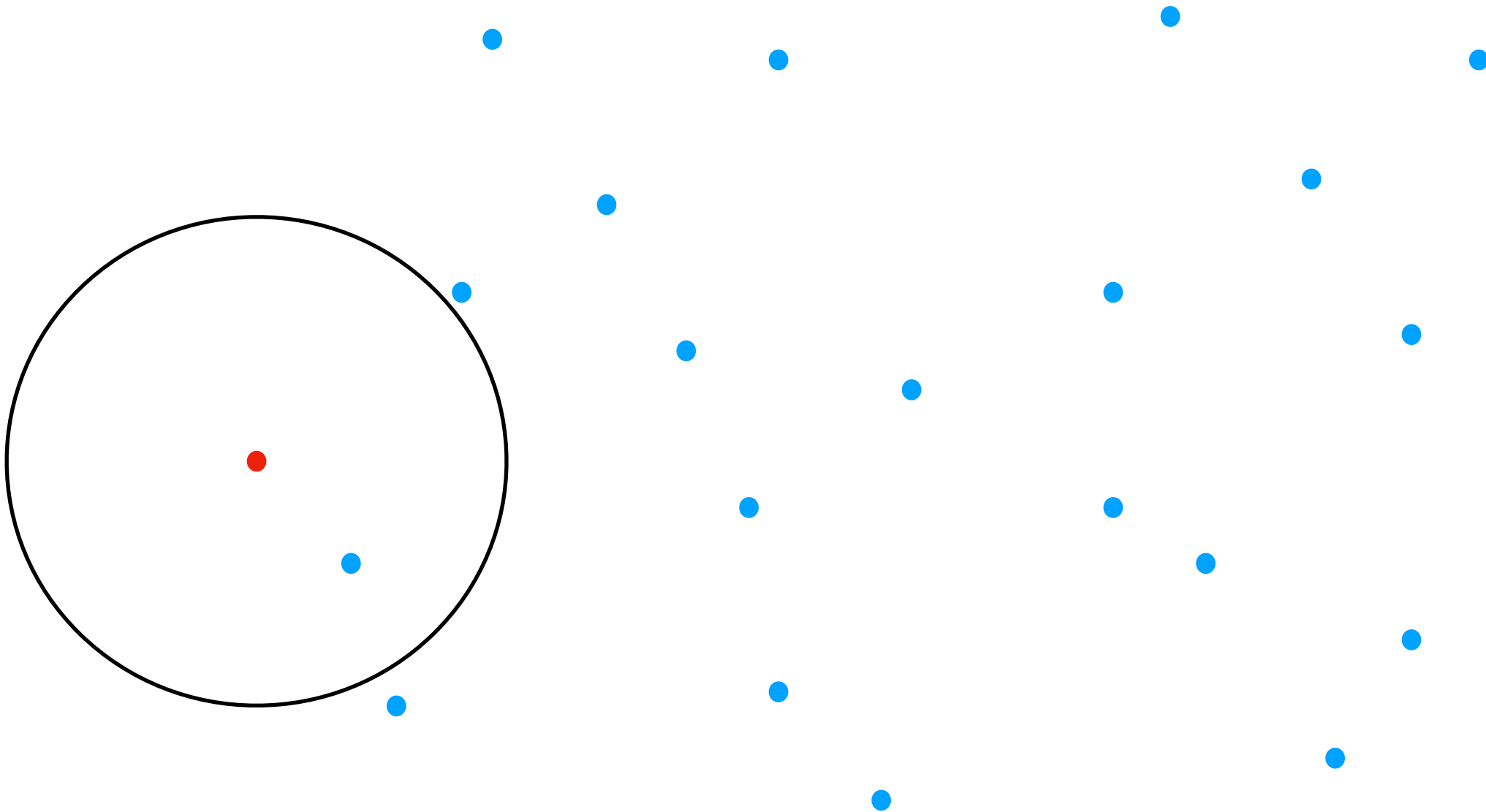
# Sparsify



# Sparsify

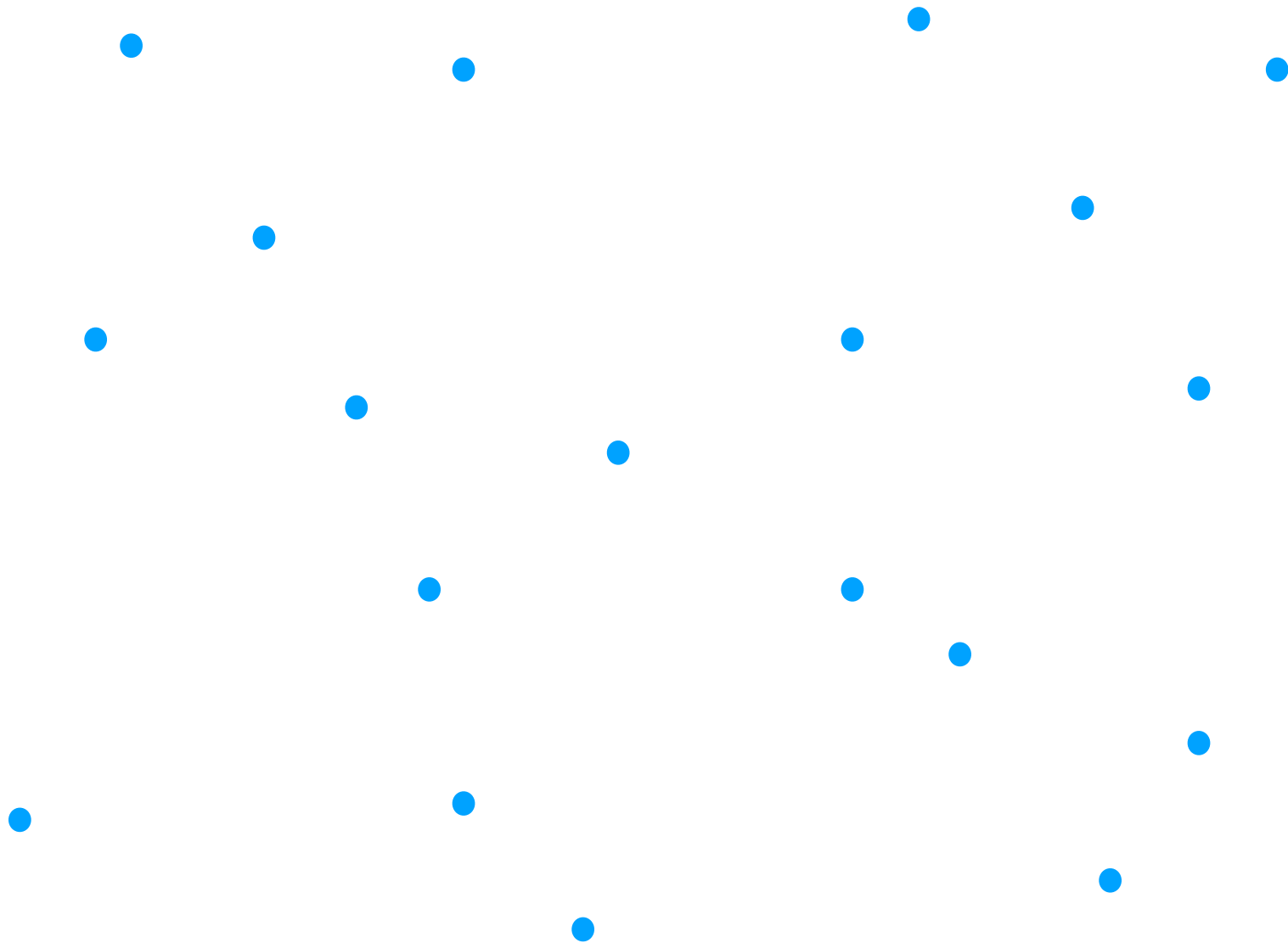


# Sparsify

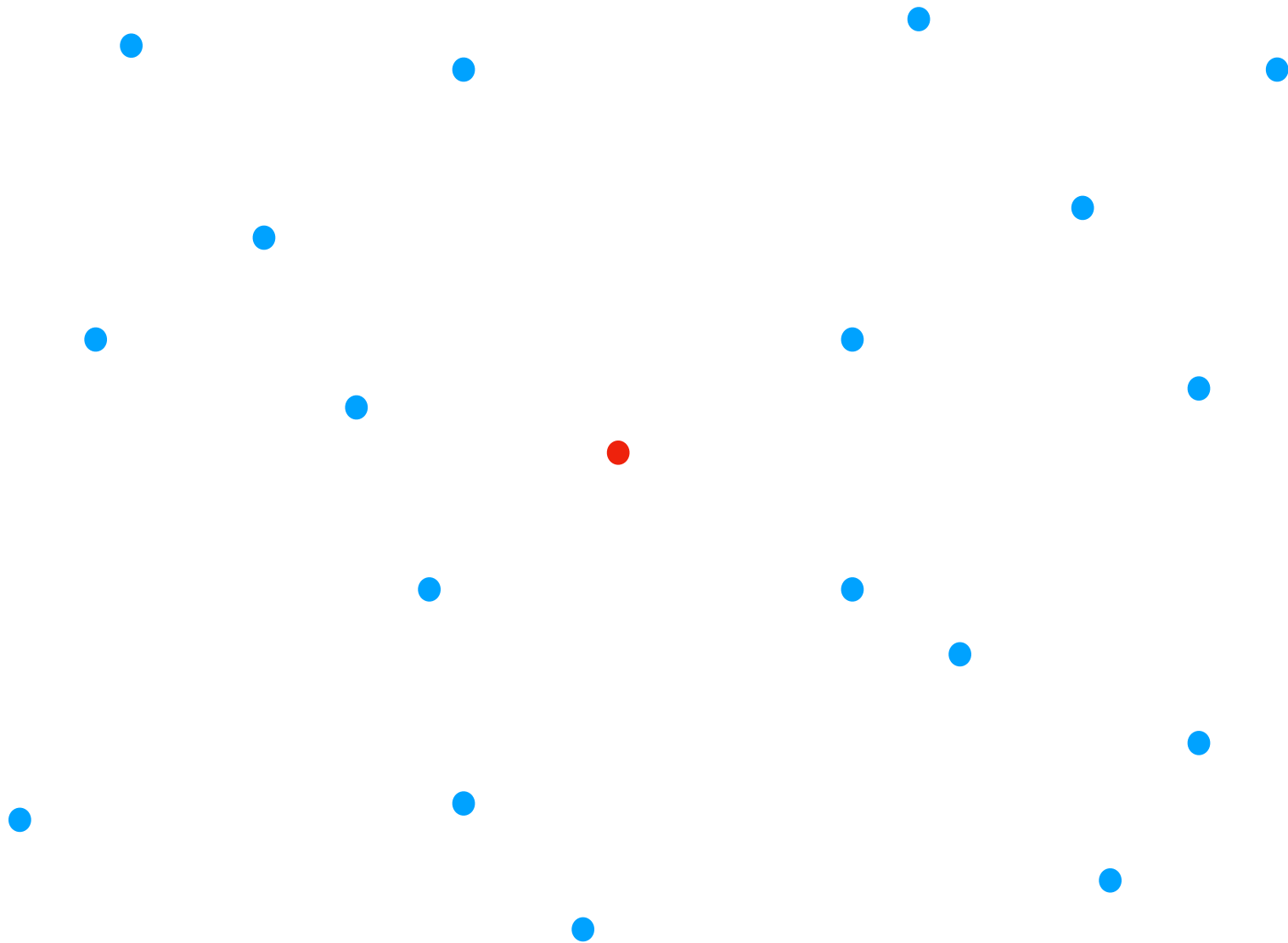




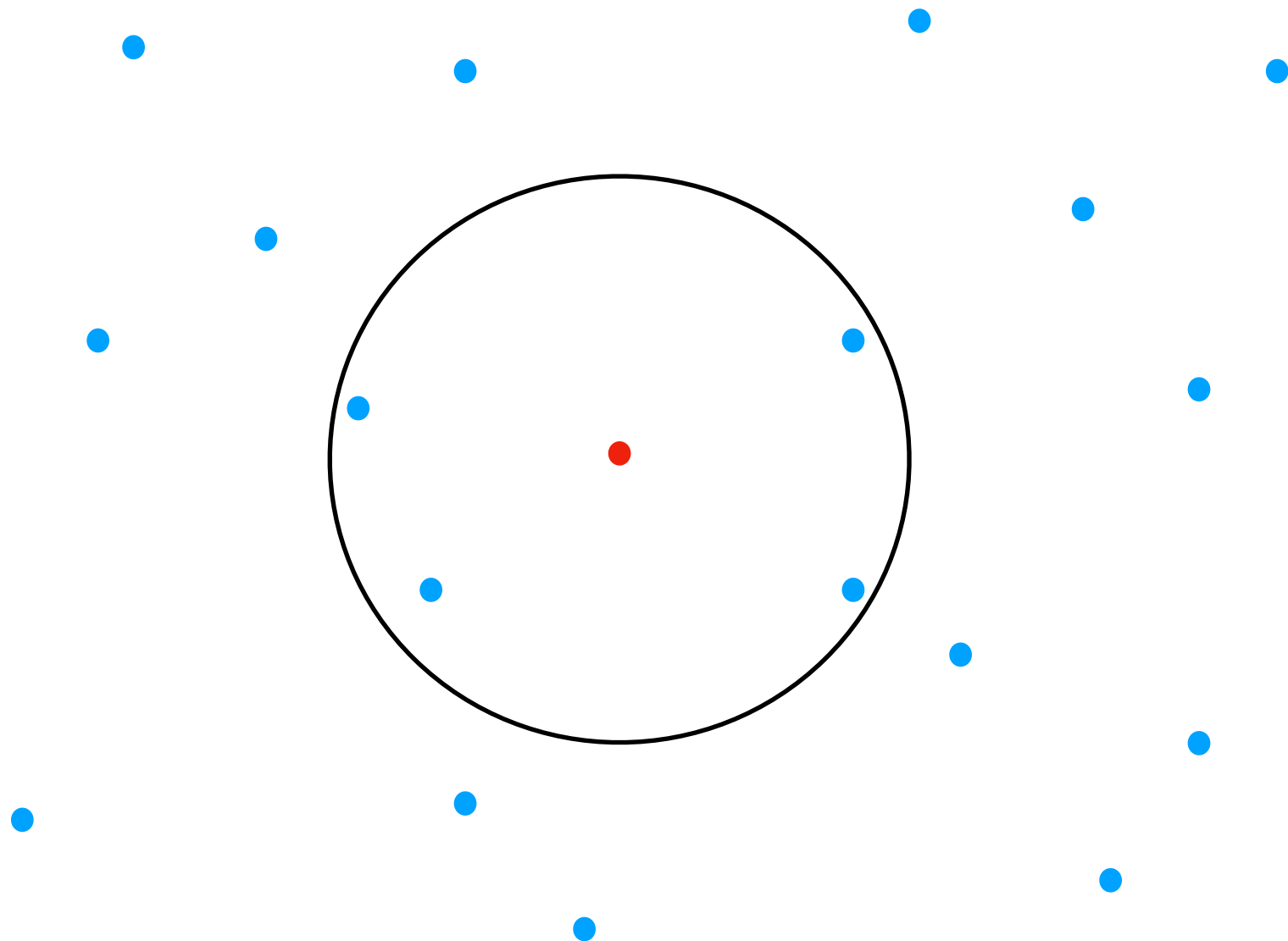
# Sparsify



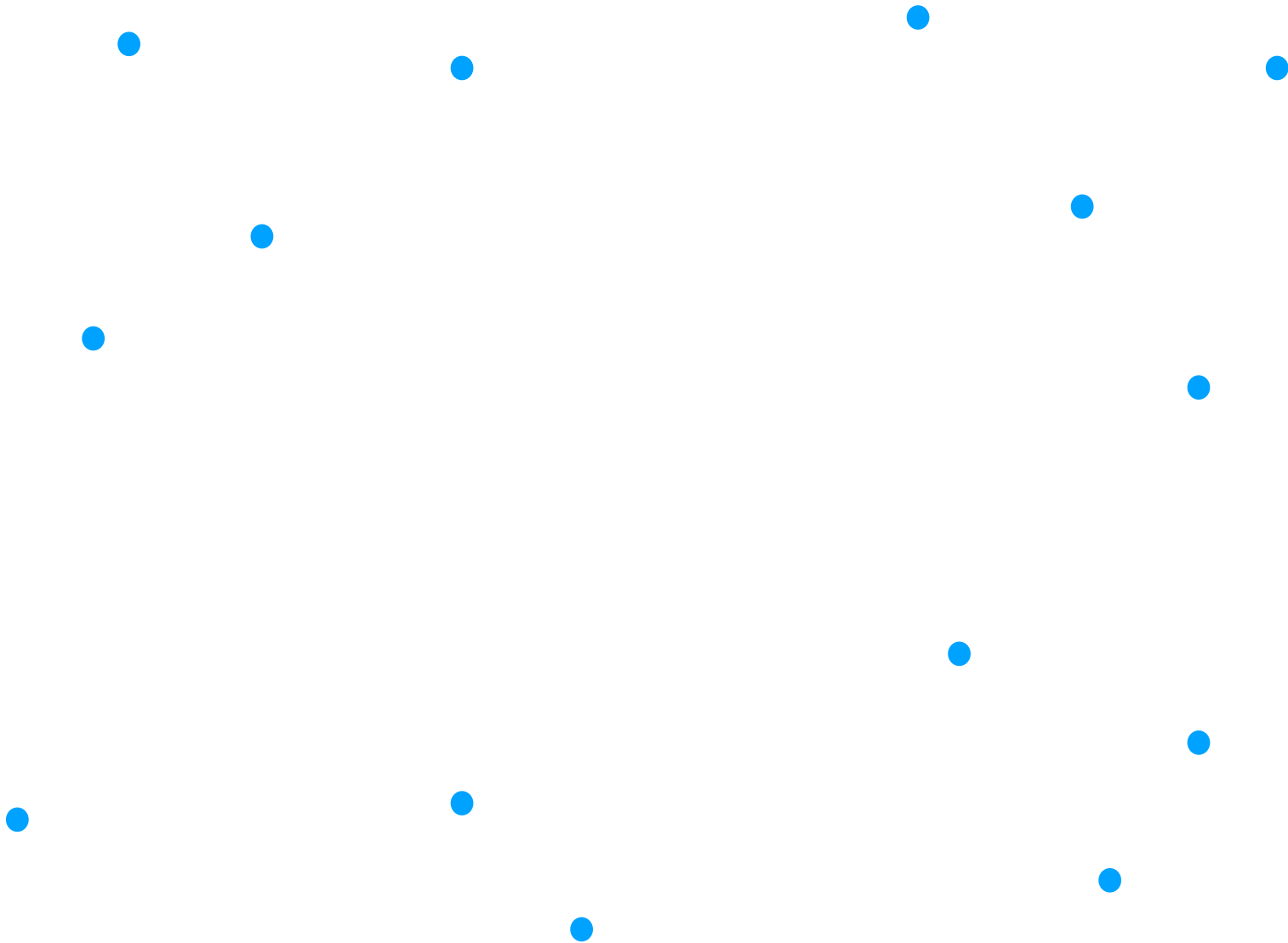
# Sparsify



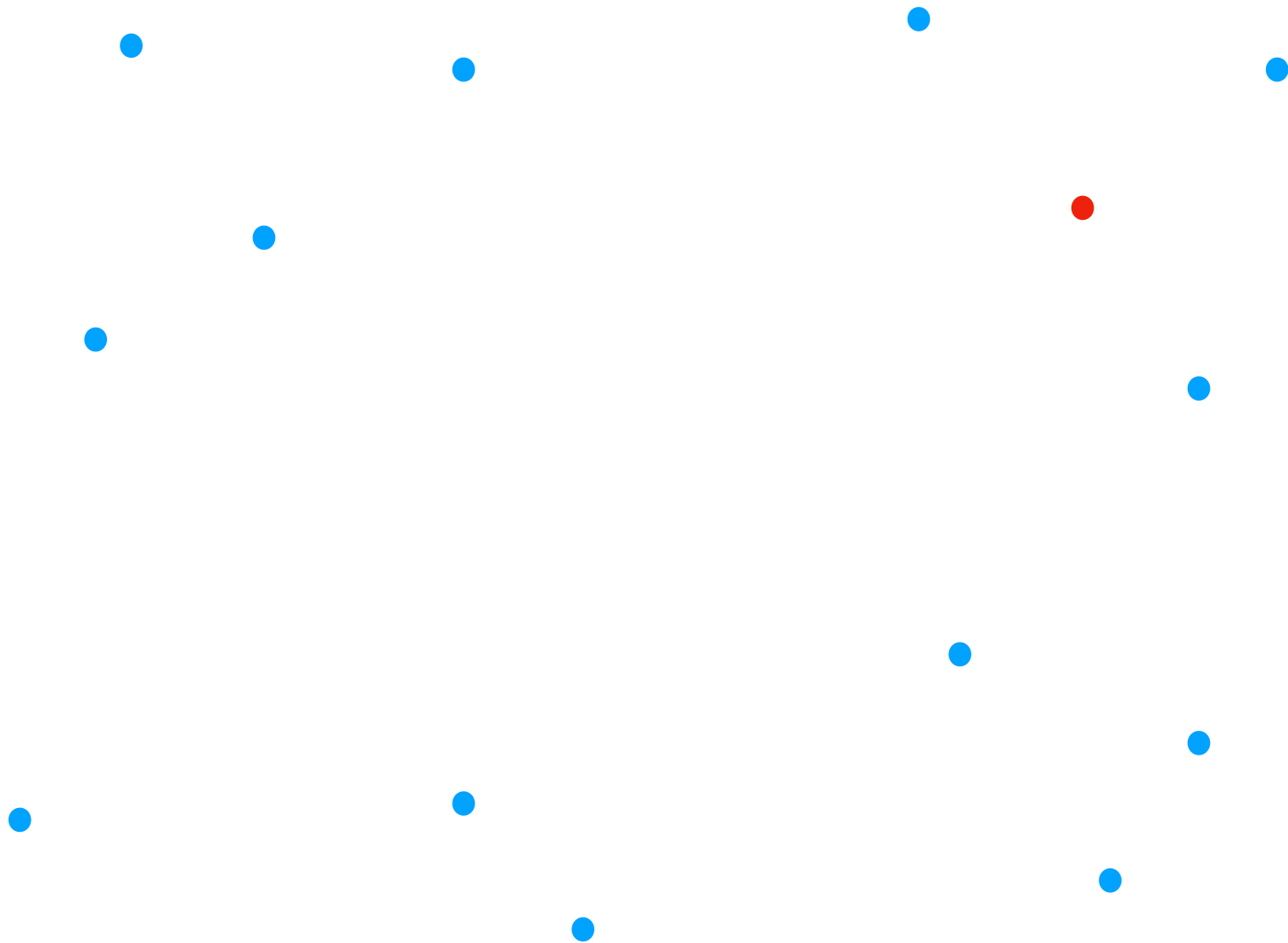
# Sparsify



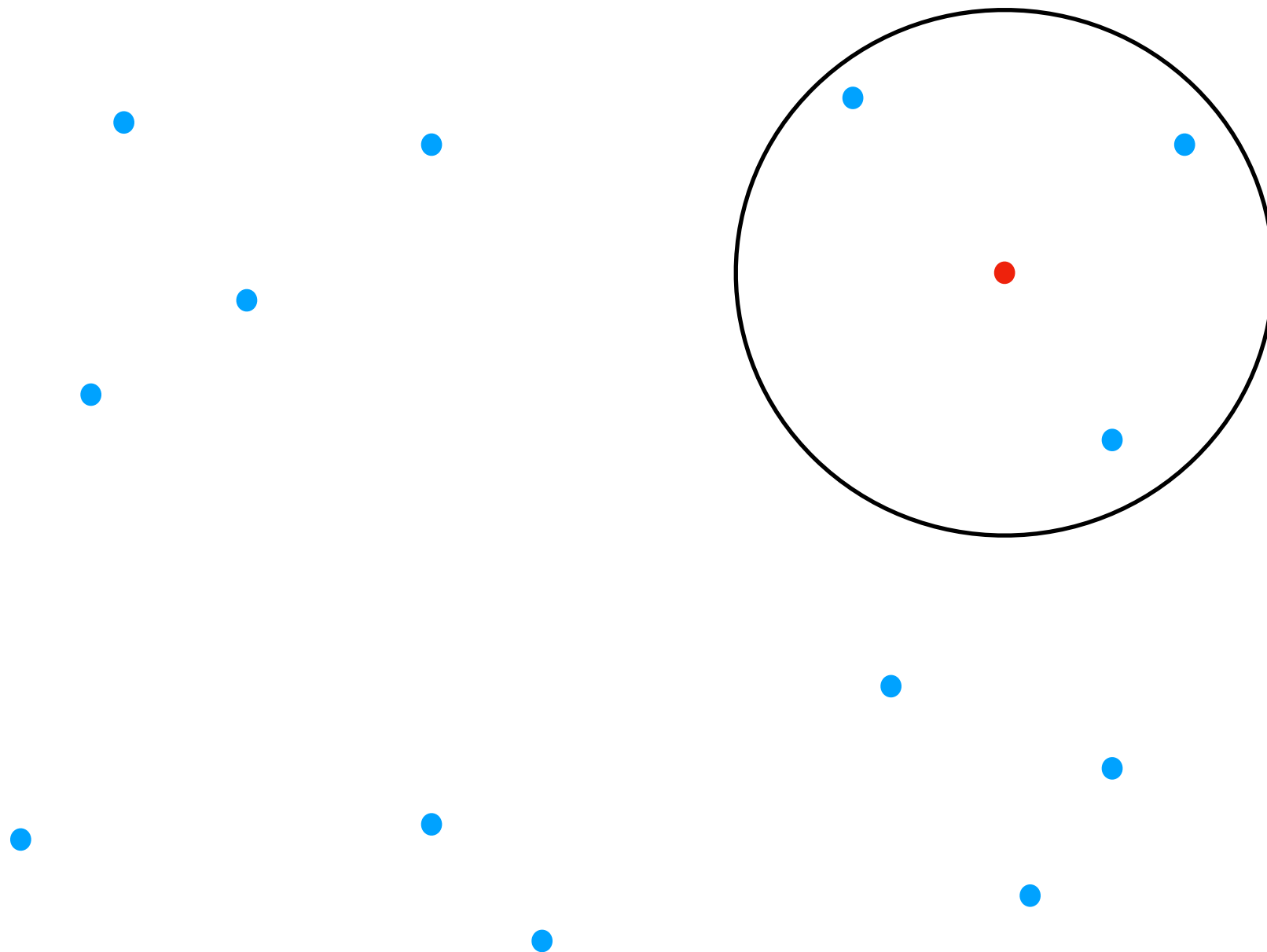
# Sparsify



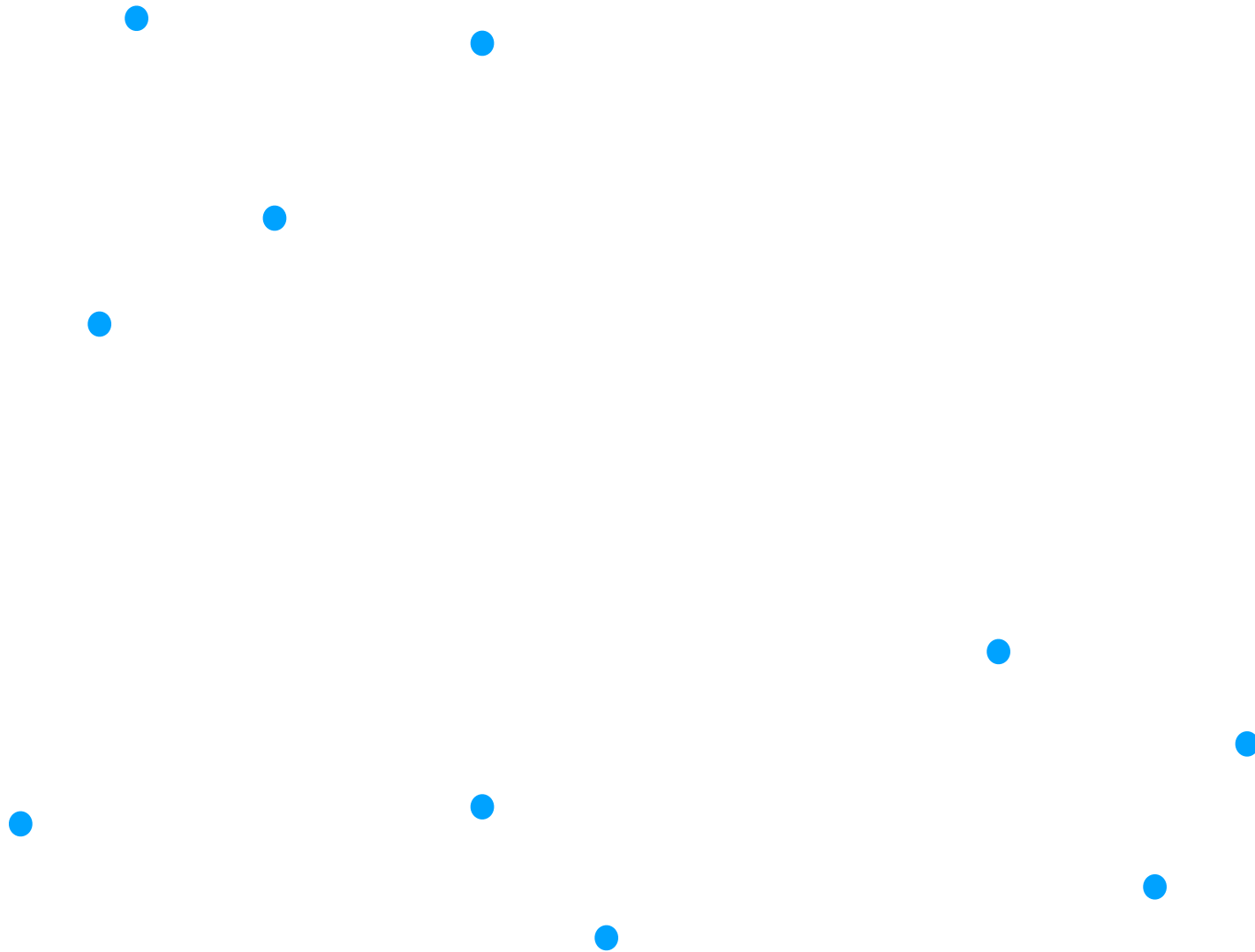
# Sparsify



# Sparsify



# Sparsify

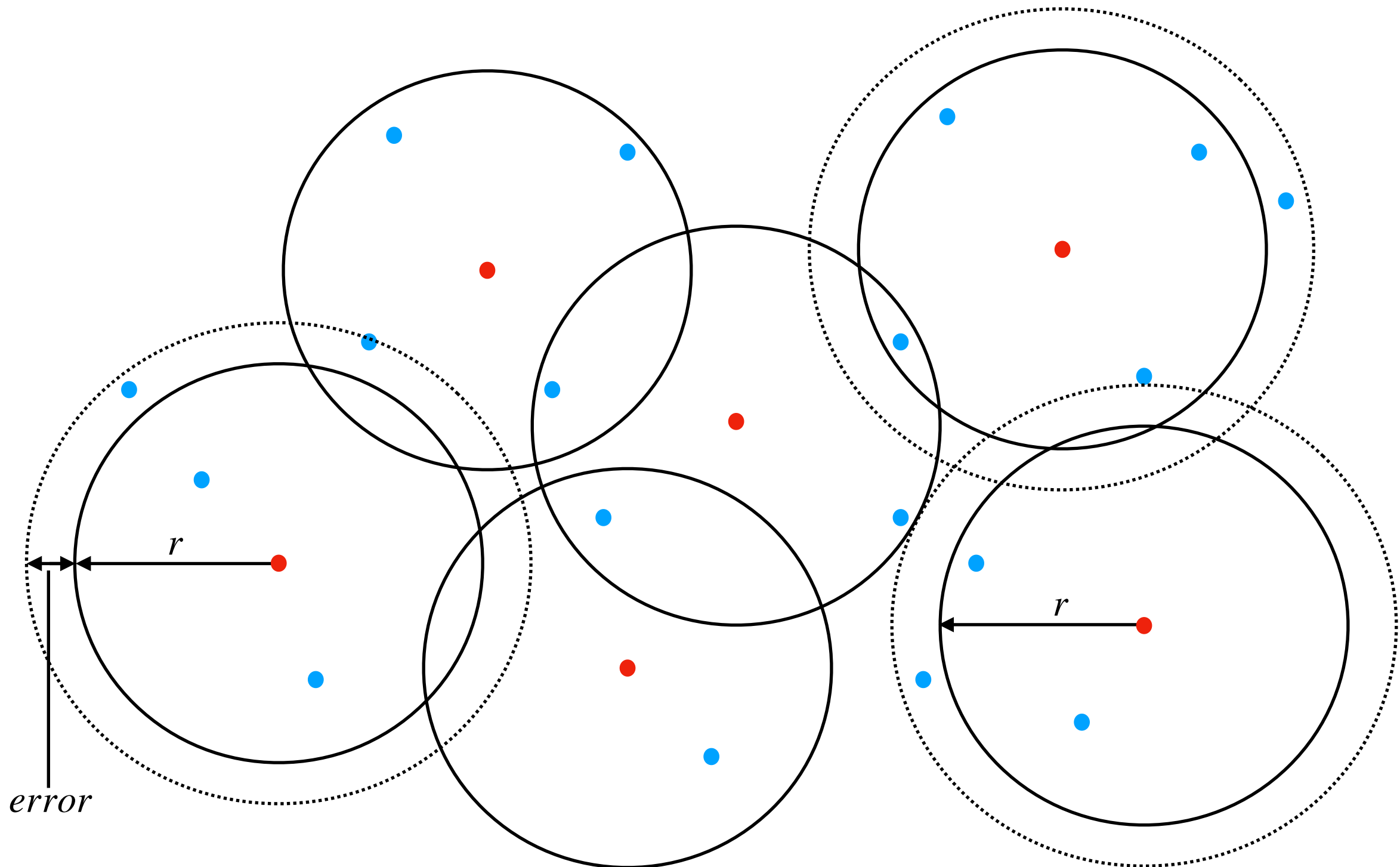


# Sparsity with high probability

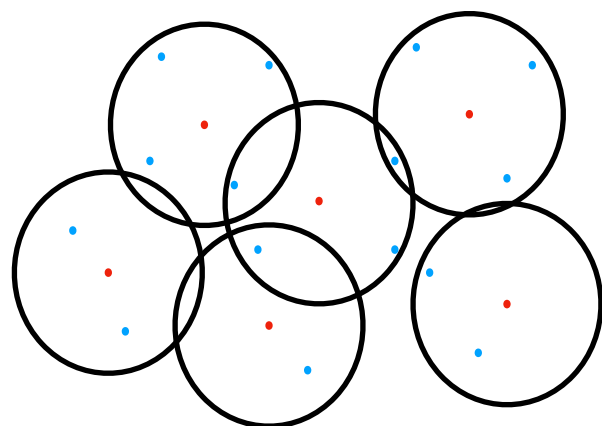
	$p_1$	$p_2$	...	$p_n$
	no	no	...	yes
	no	no	...	no
	no	yes	...	no



# Approximate r-net



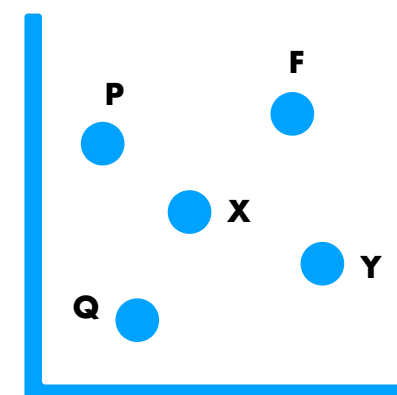
## R-nets



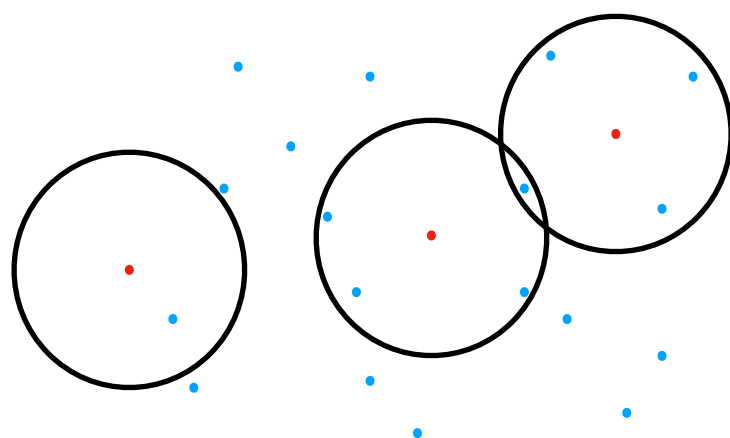
## Distance matrix

	$p_1$	$p_2$		$p_n$
Blue	yes	no	...	yes
Green	no	no	...	yes
Red	no	yes	...	no

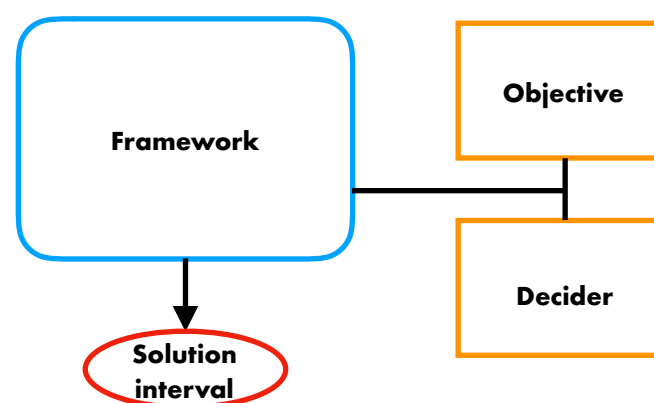
## PTF



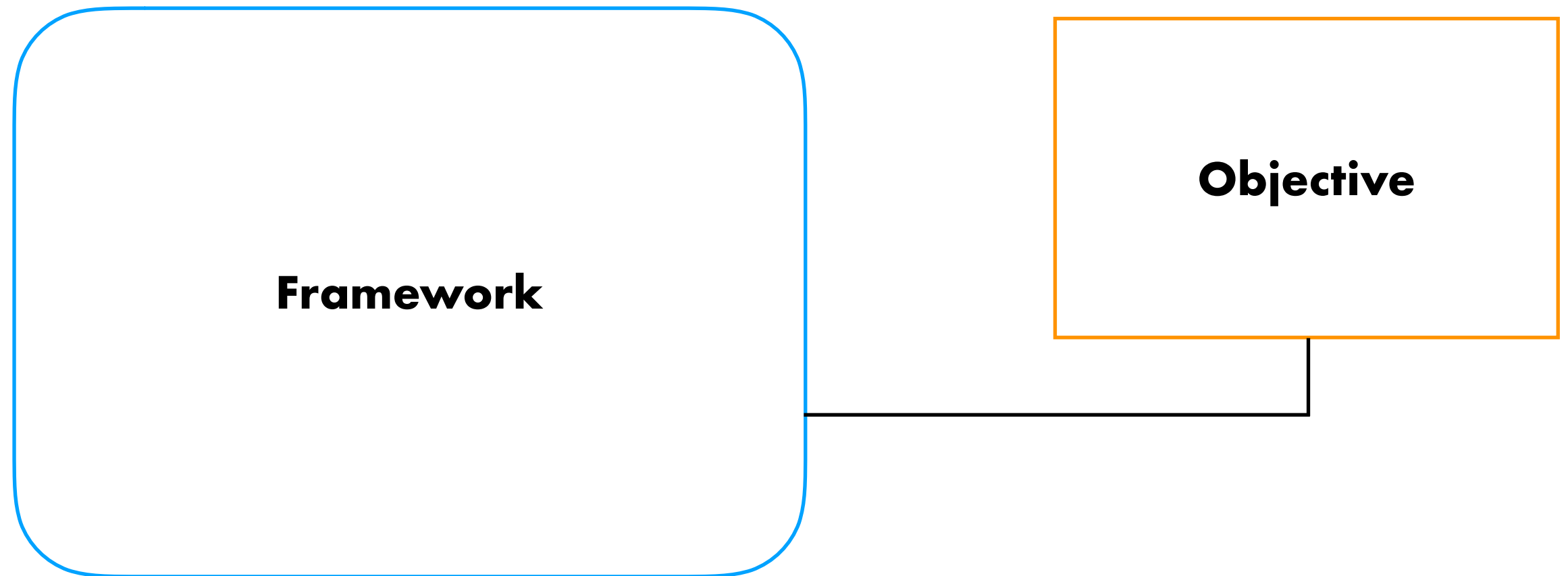
## Sparsification



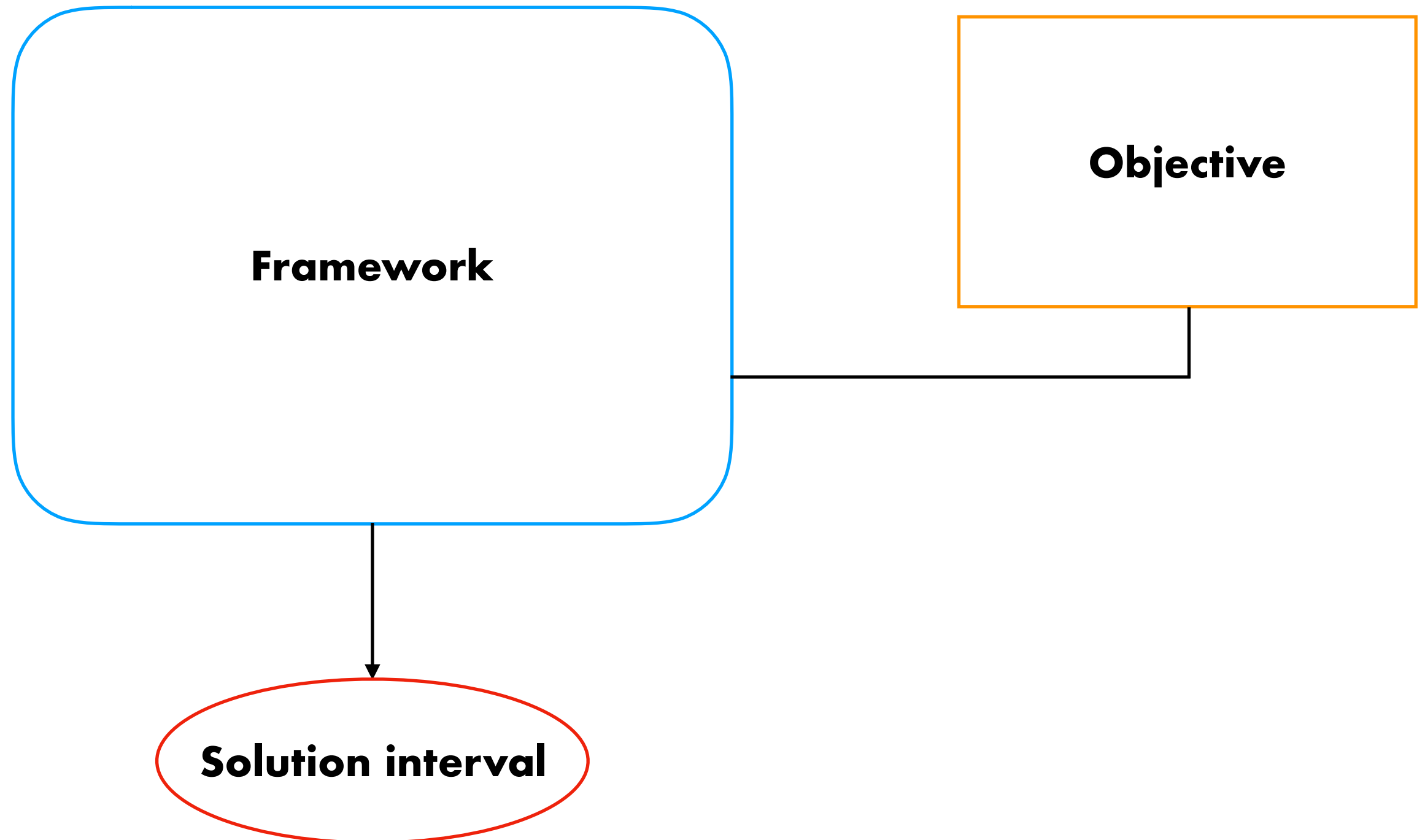
## Net & Prune



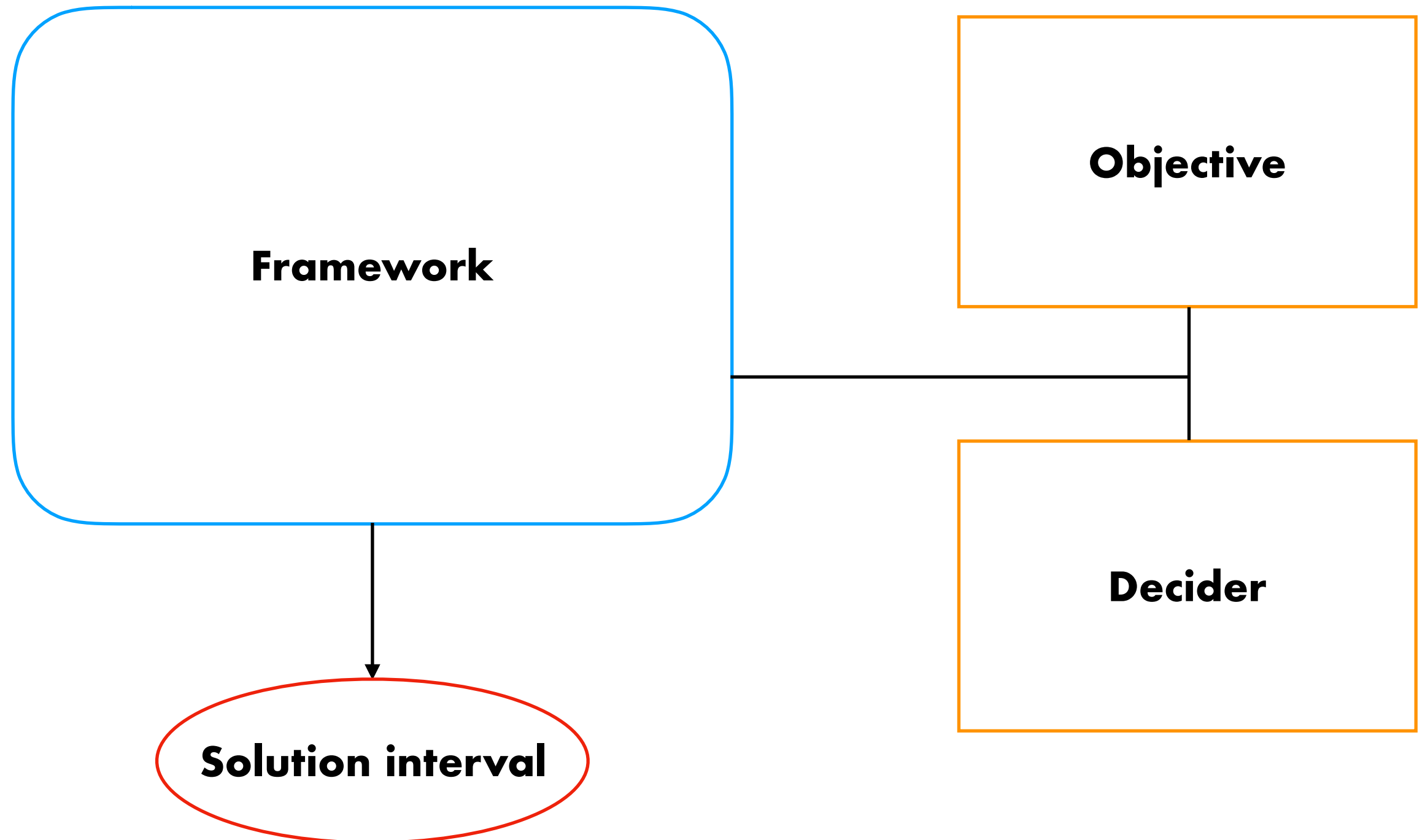
# Net & Prune framework



# Net & Prune framework



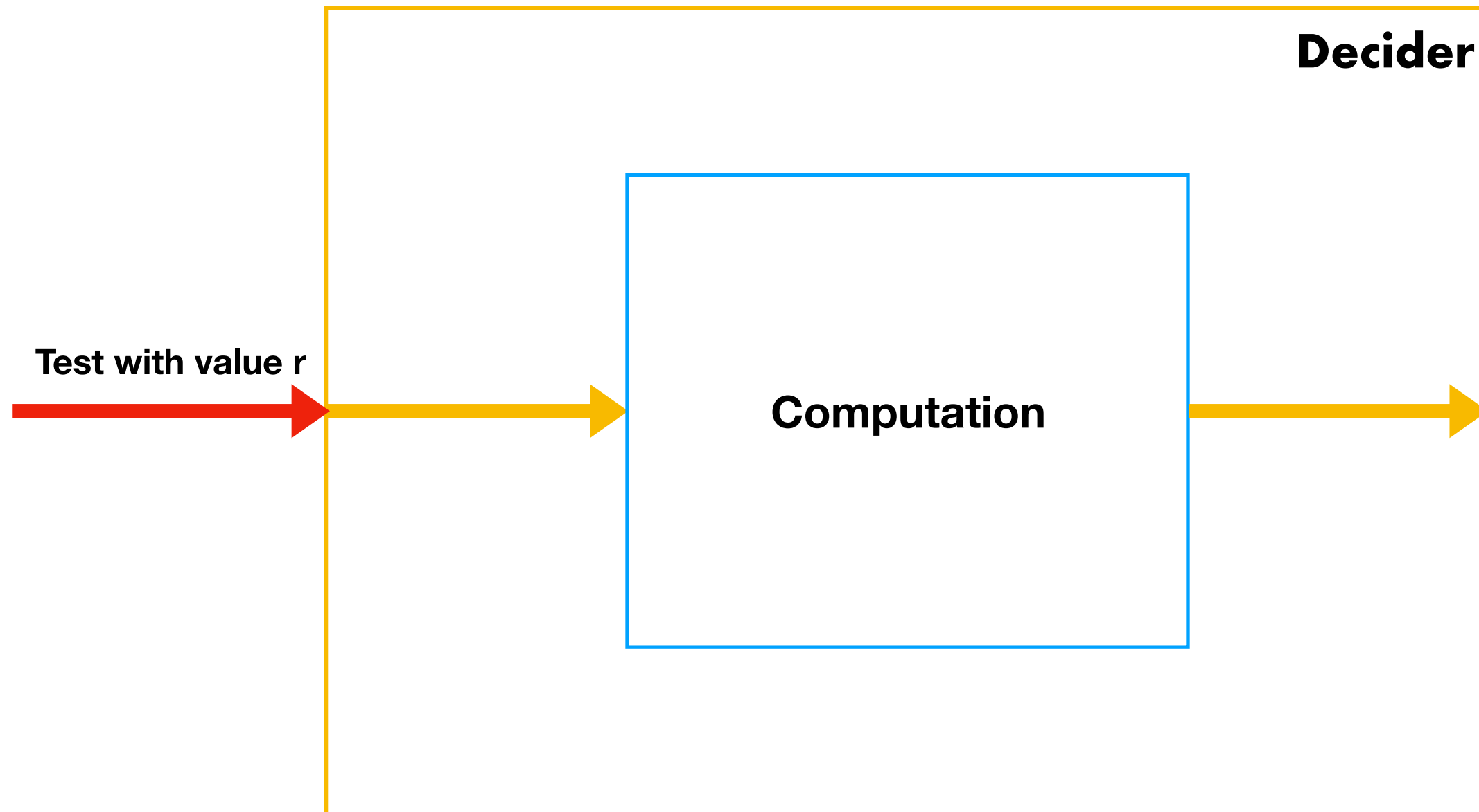
# Net & Prune framework



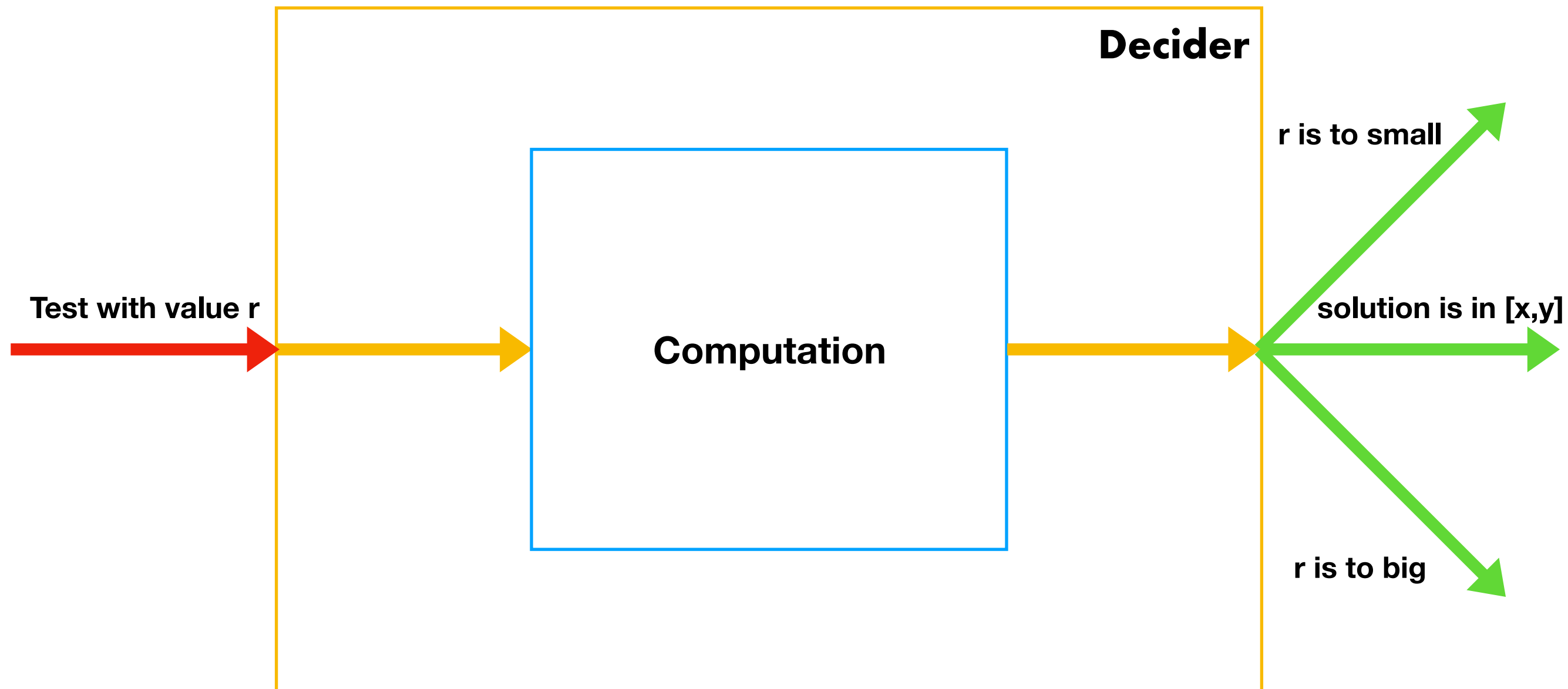
# Decider



# Decider

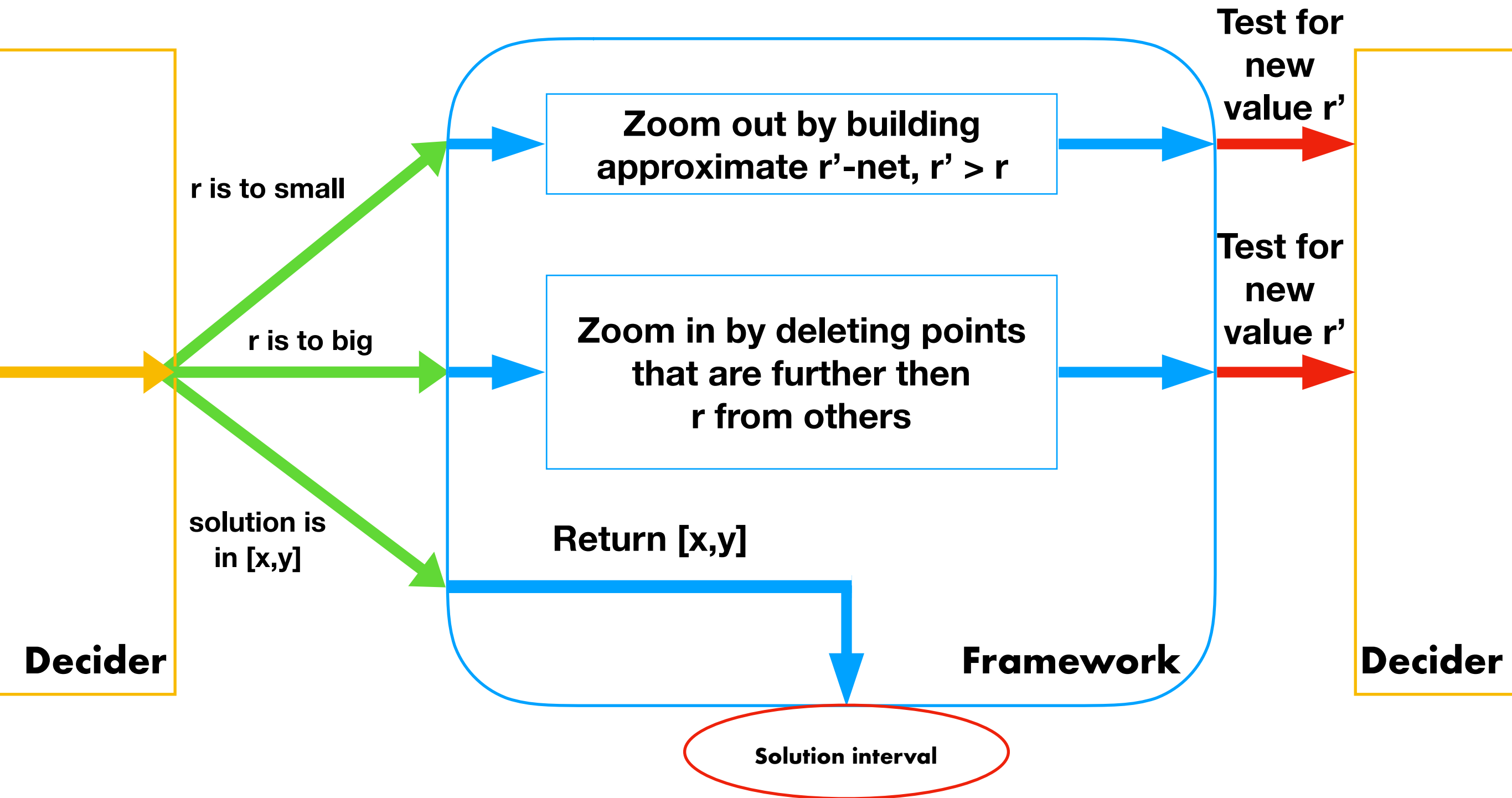


# Decider

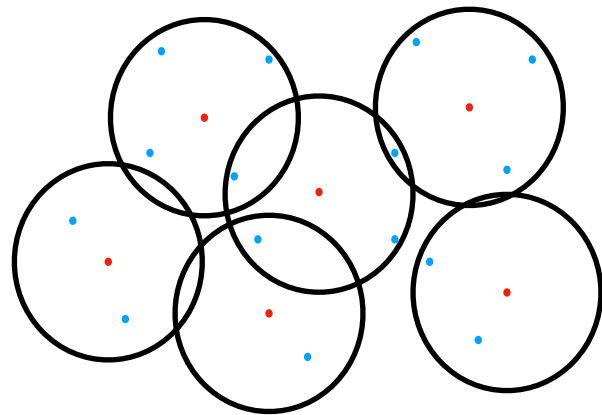




# Decider



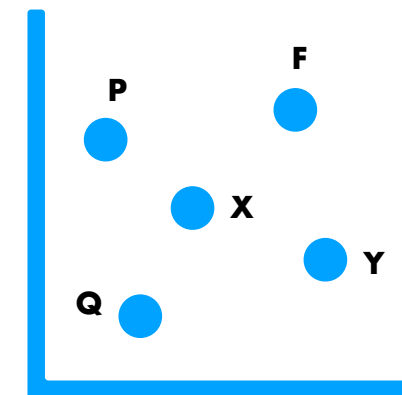
## R-nets



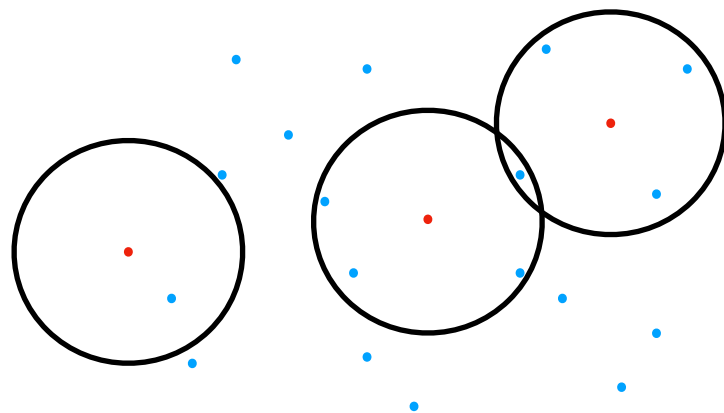
## Distance matrix

	$p_1$	$p_2$	...	$p_n$
Blue	yes	no	...	yes
Green	no	no	...	yes
Red	no	yes	...	no

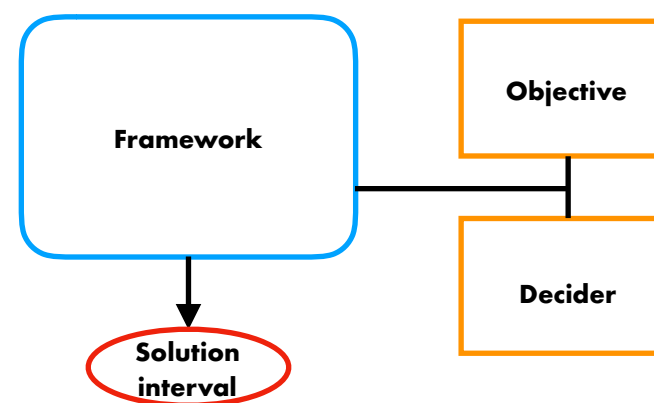
## PTF



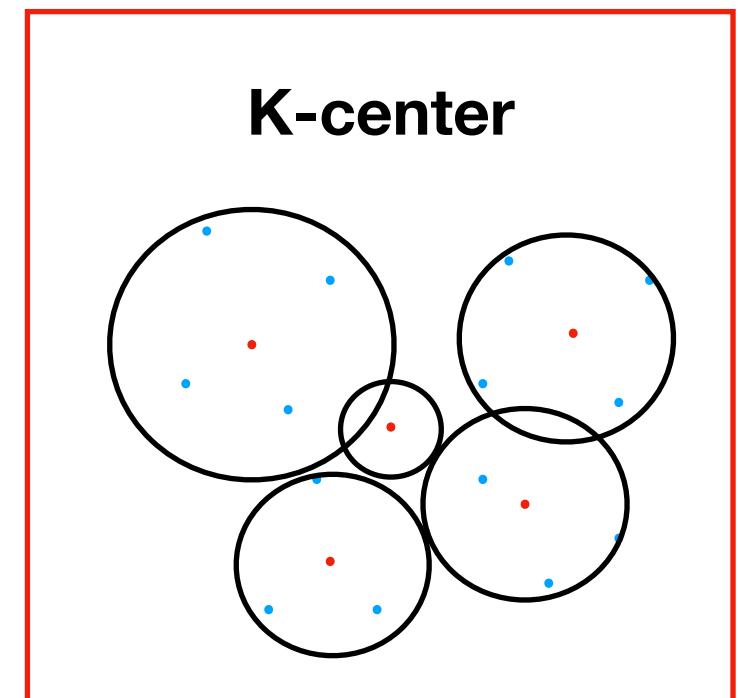
## Sparsification



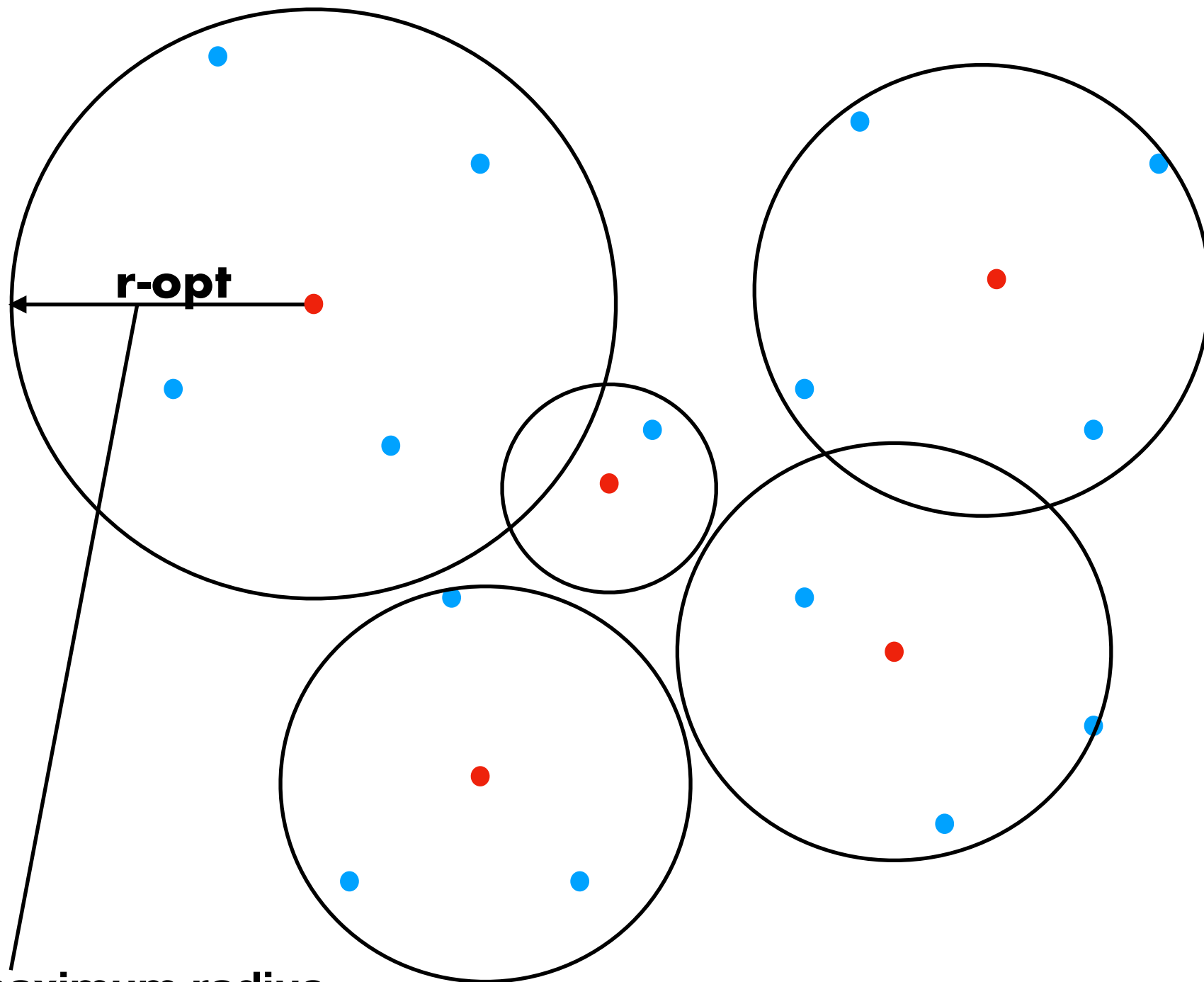
## Net & Prune



## K-center

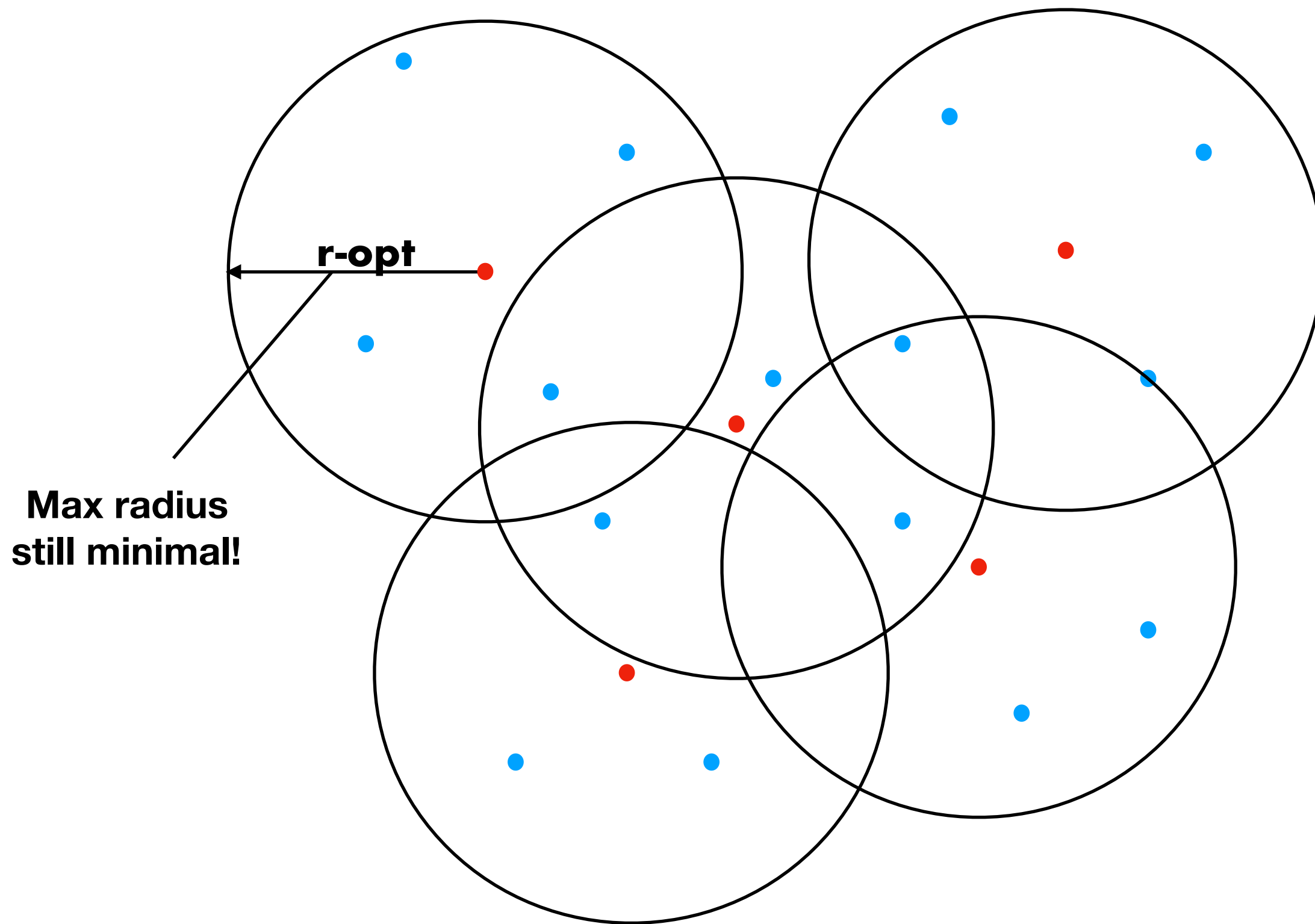


# K-Center with $k=5$



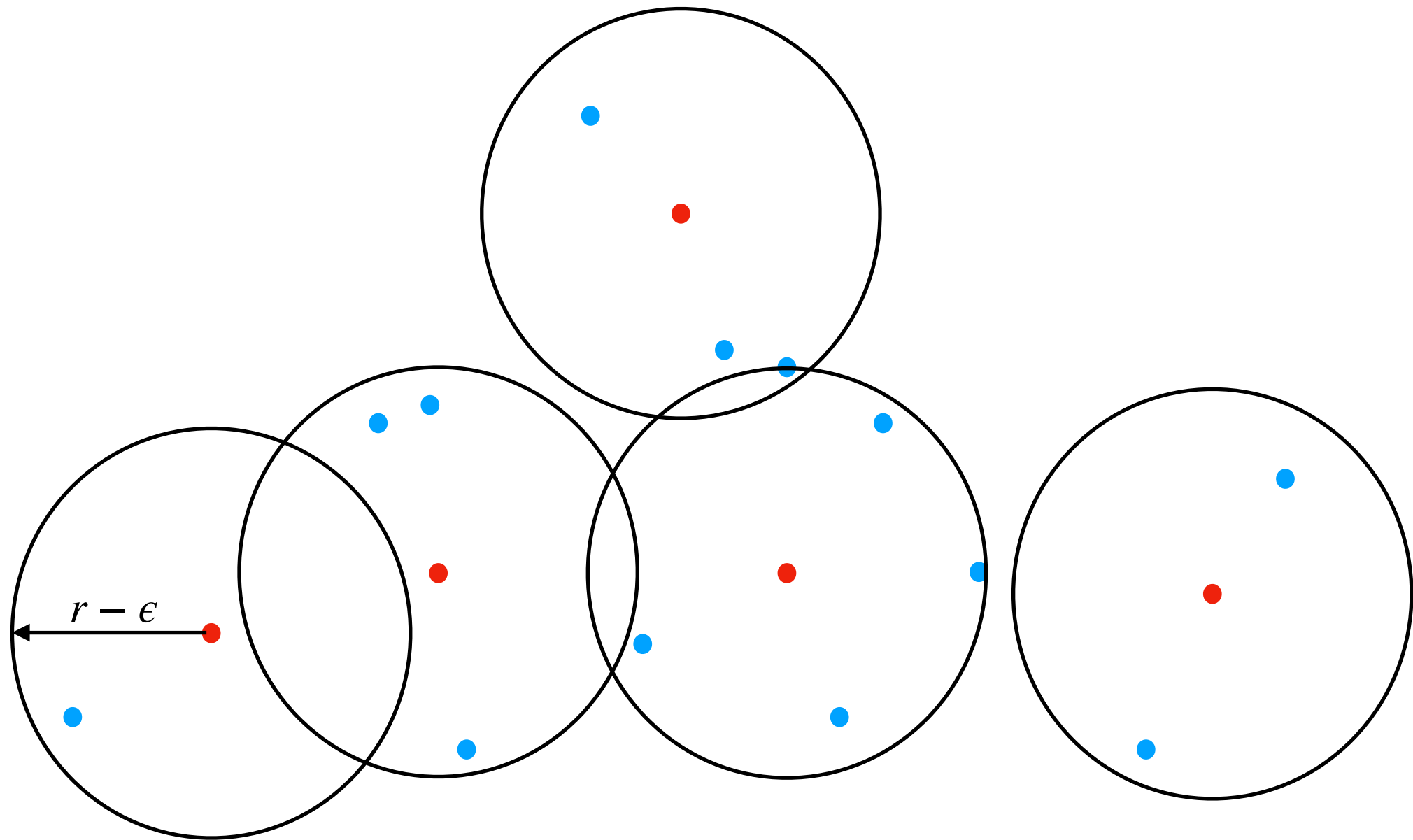
**Minimize maximum radius**

# r-opt net = 5-center cluster



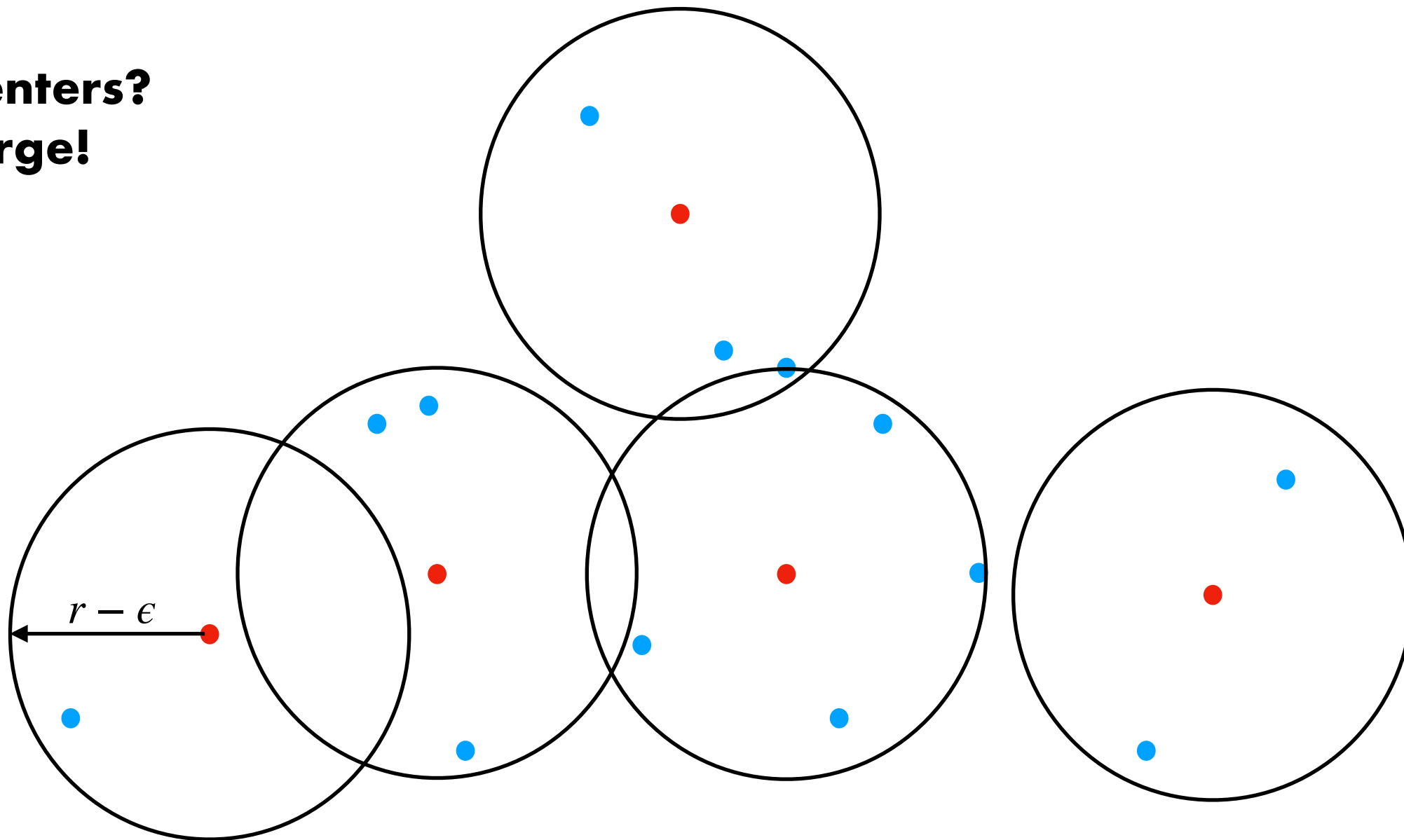
Solve with the framework by  
building a decider!

# Test for value $r$ ?

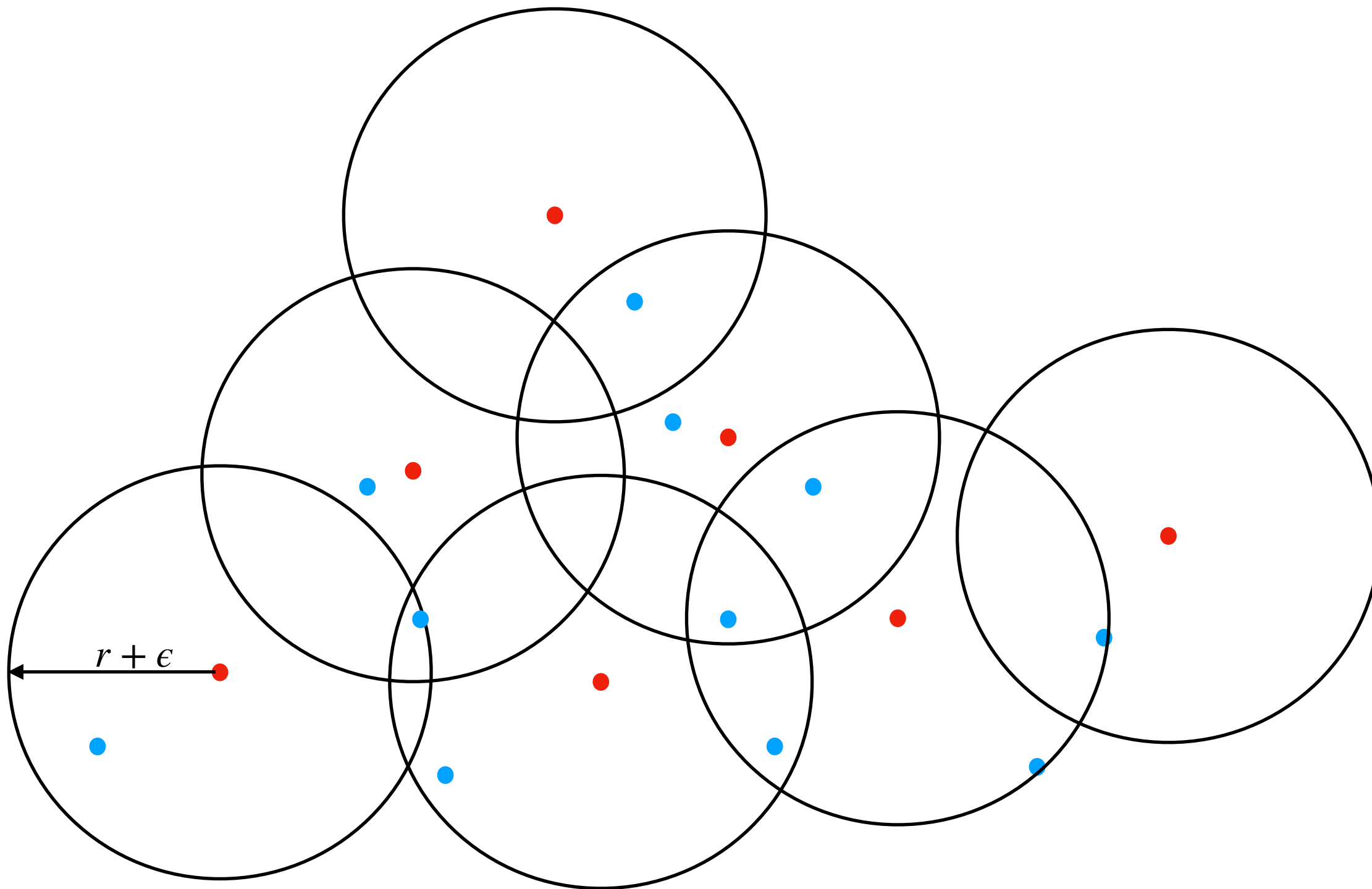


# Test for value $r$ ?

**k or less centers?**  
 **$r$  is too large!**



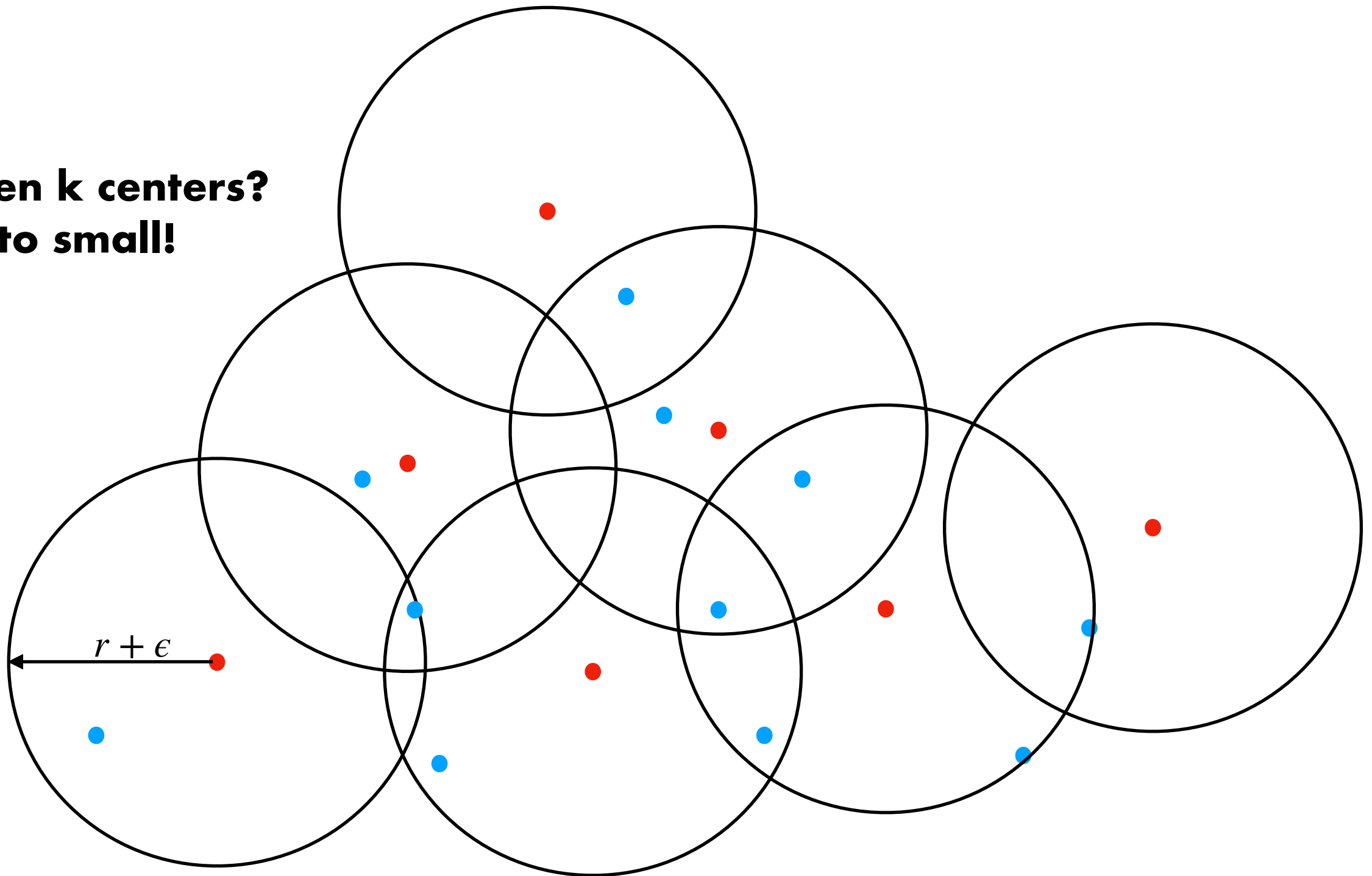
# Test for value $r$ ?

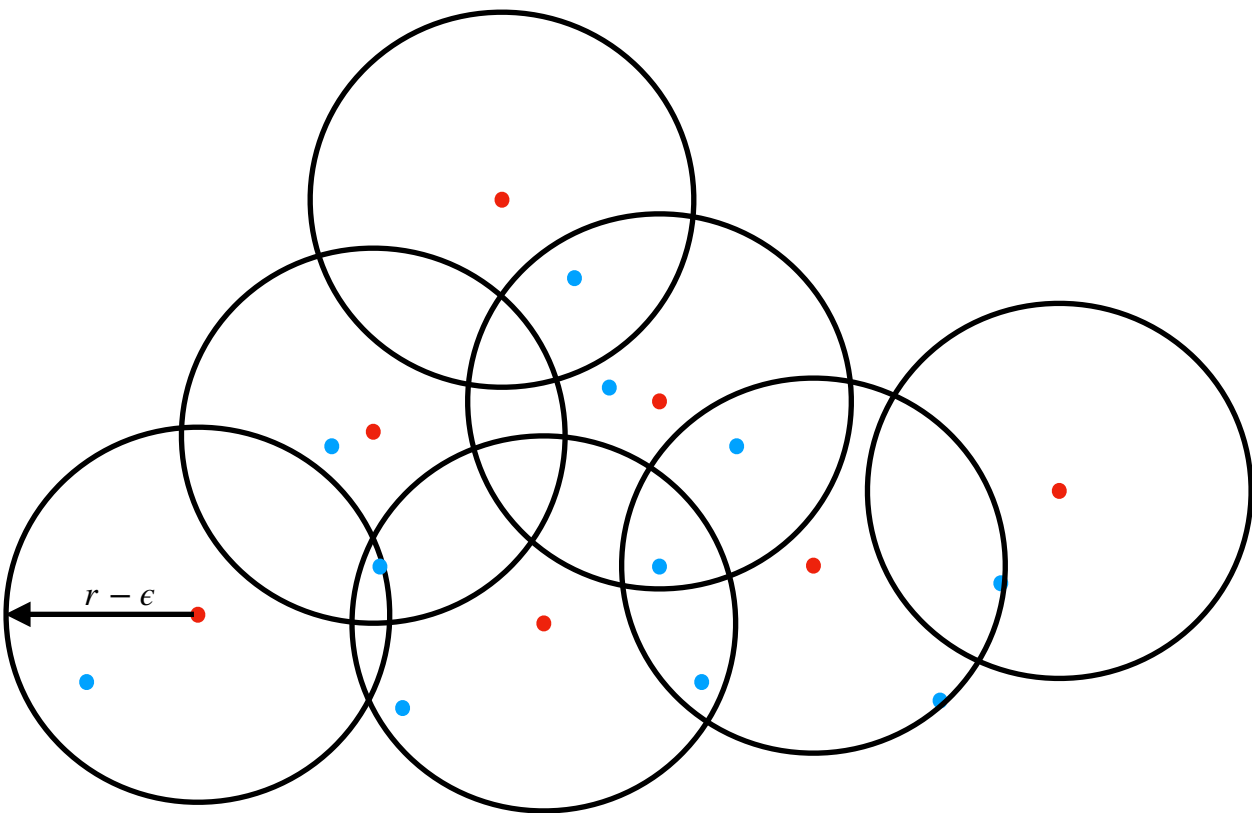




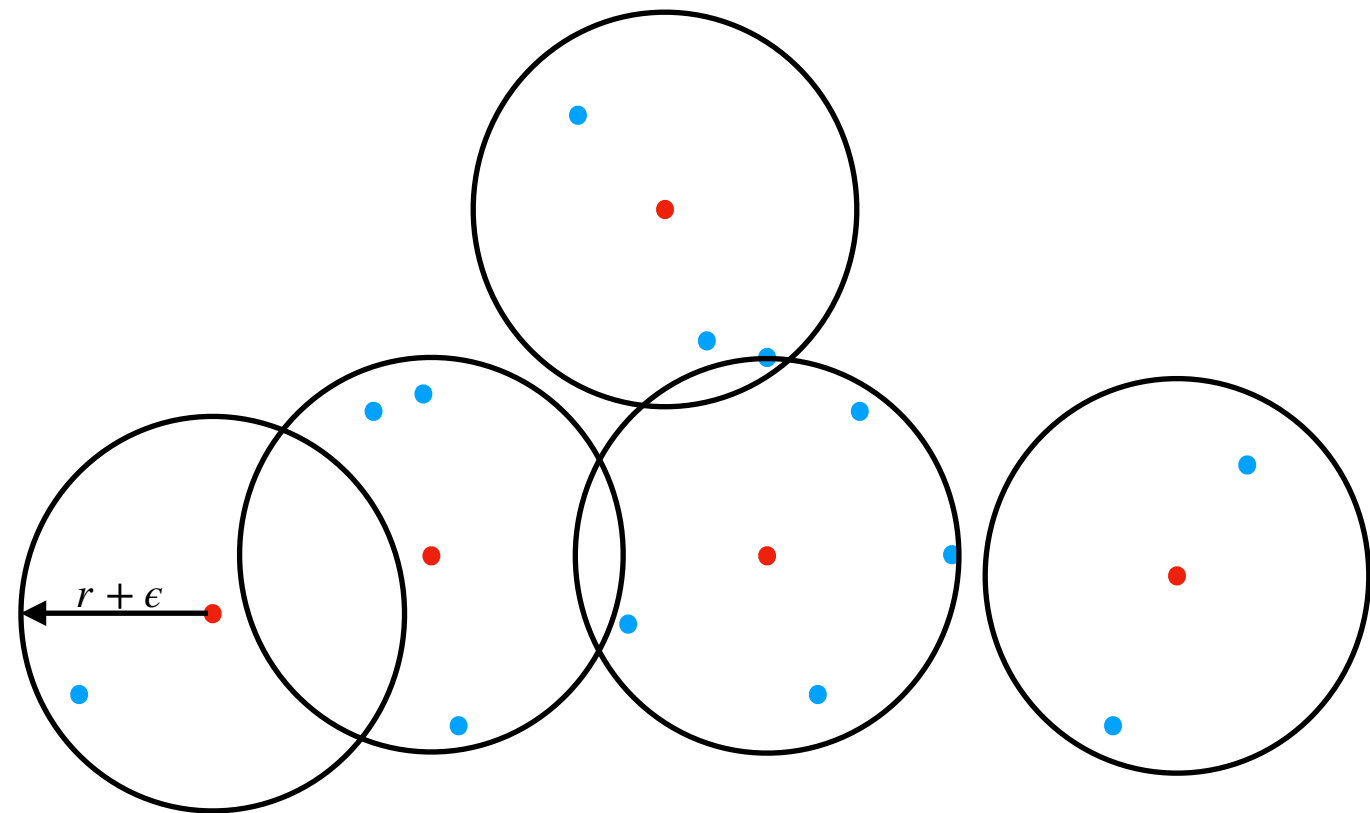
# Test for value $r$ ?

**more than  $k$  centers?  
 $r$  is too small!**





**More than  $k$  centers?**



**$k$  or less centers?**

**Optimal solution  
in  $[x, y]$**

**Binary search with  
the decider**

**Return  $r$  where  
 $r_{opt} \in [r, r(4 + \epsilon)]$**

**Optimal solution  
in  $[x,y]$**



**Slice the interval into pieces**



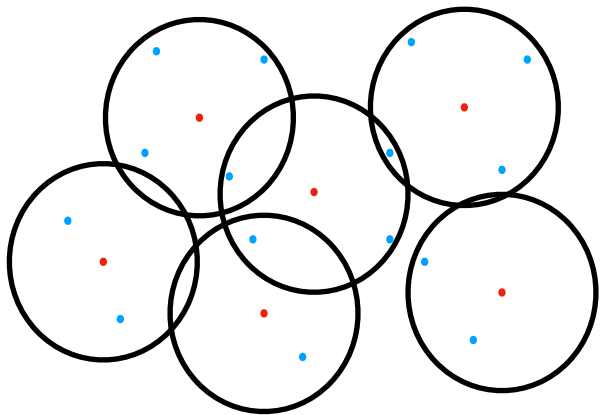
**Binary search with  
the decider**



**Get a smaller interval  $[x',y']$ ,  
that contains the solution**



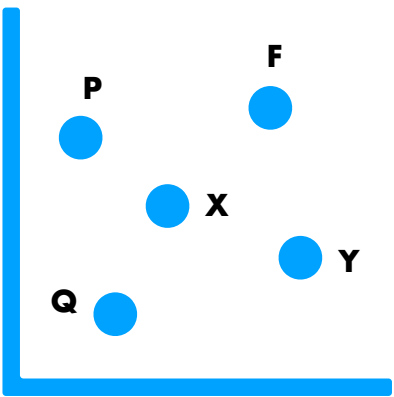
**R-nets**



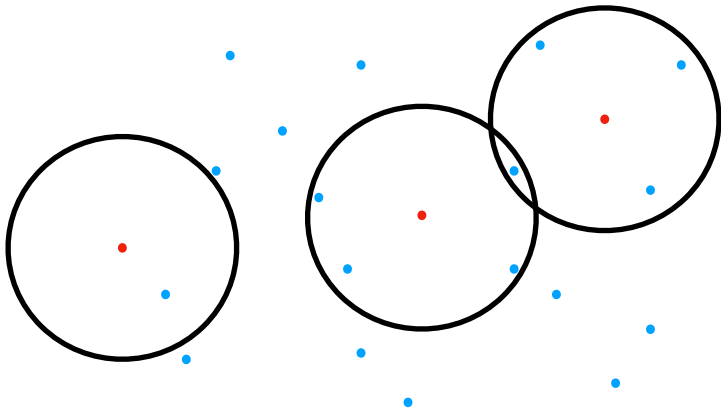
**Distance matrix**

	$p_1$	$p_2$		$p_n$
Blue	yes	no	...	yes
Green	no	no	...	yes
Red	no	yes	...	no

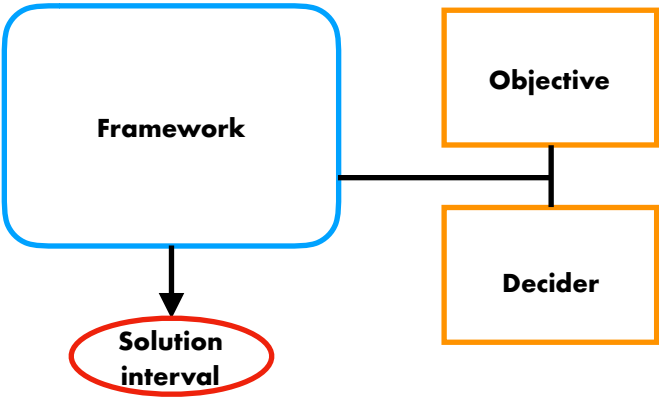
**PTF**



**Sparsification**



**Net & Prune**



**K-center**

