

# ZERO-RATE THRESHOLDS & NEW CAPACITY BOUNDS FOR LIST-DECODING AND LIST-RECOVERY

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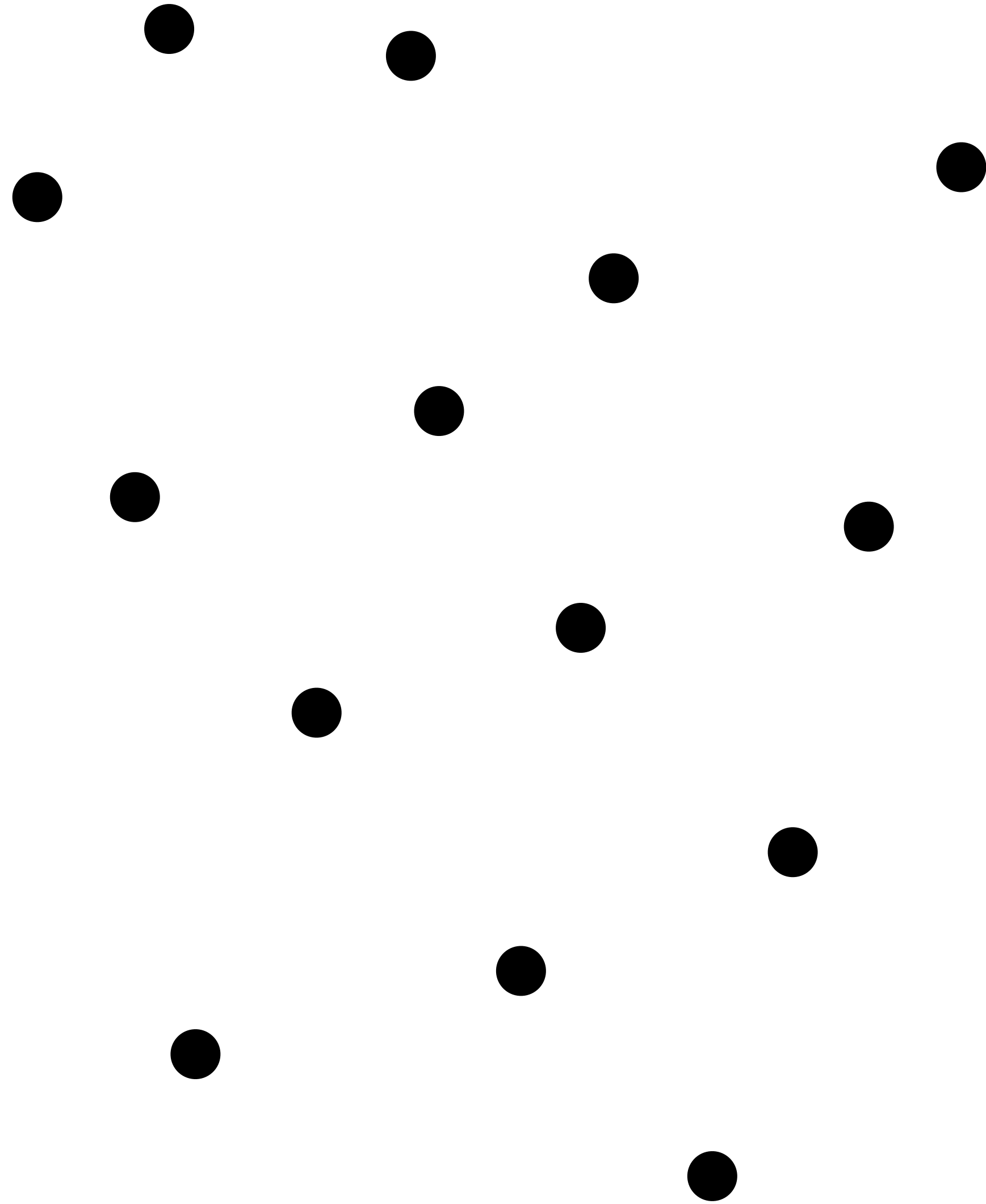
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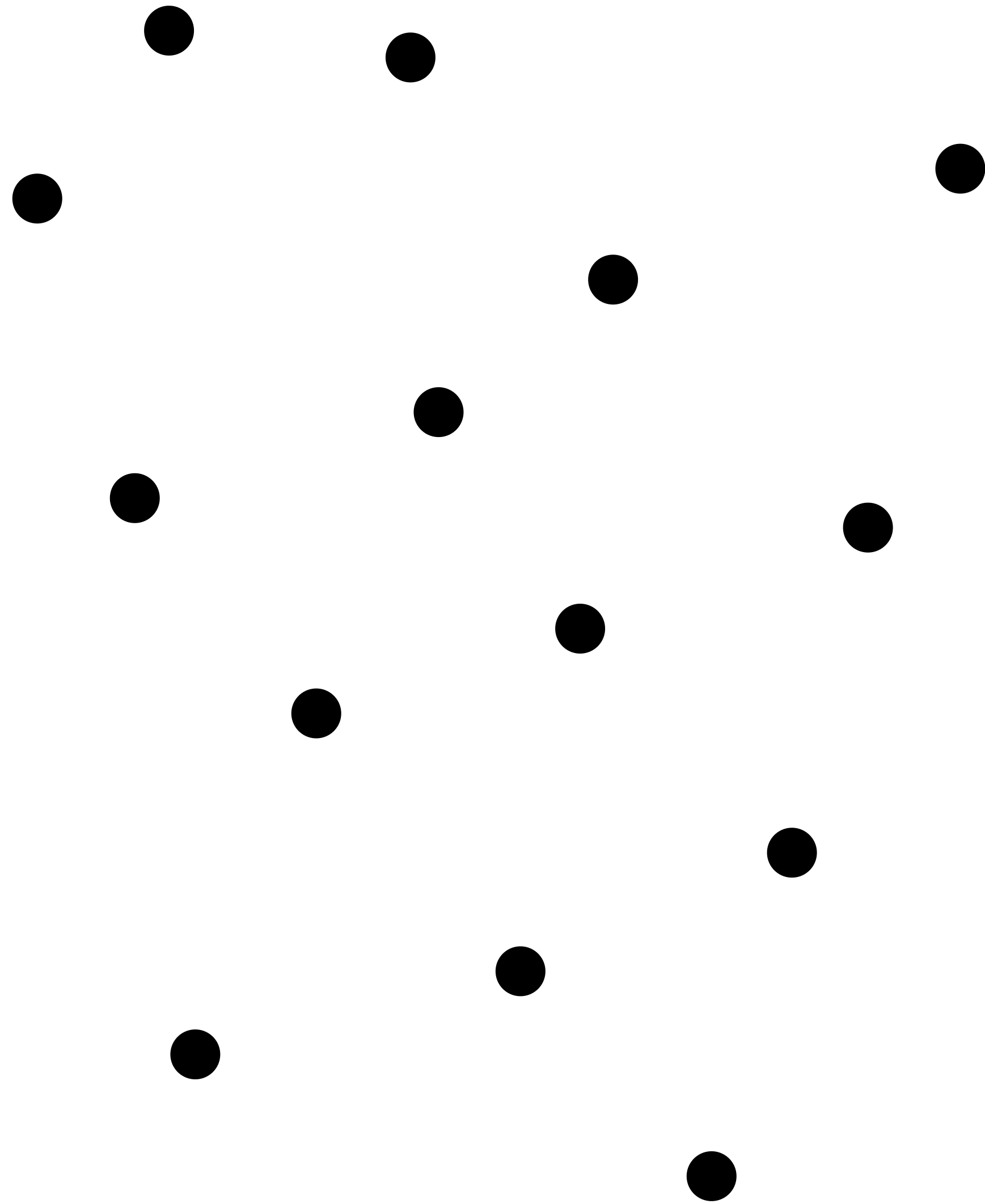
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- However, these desiderata are at odds

# RATE-DISTANCE TRADEOFF

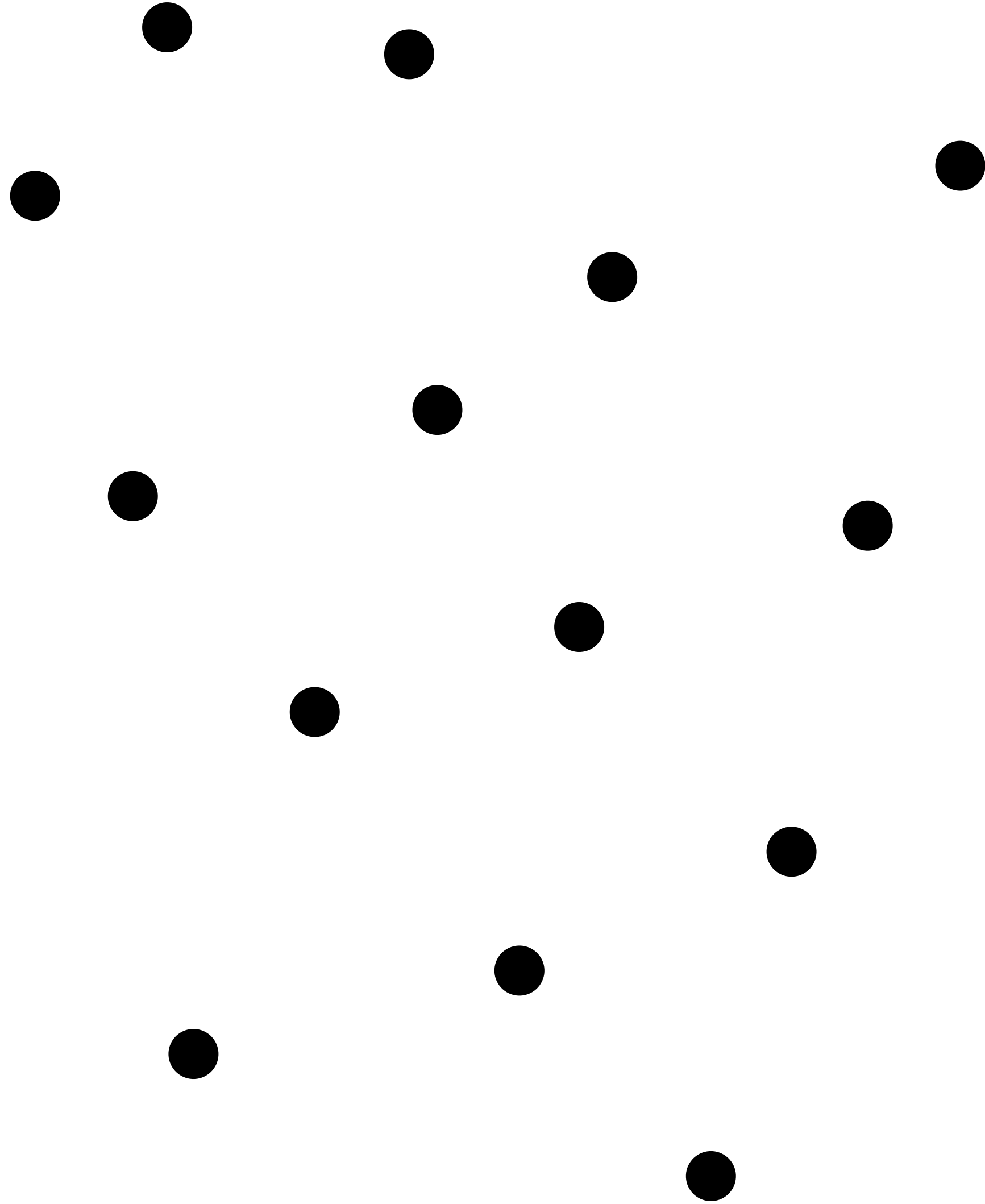


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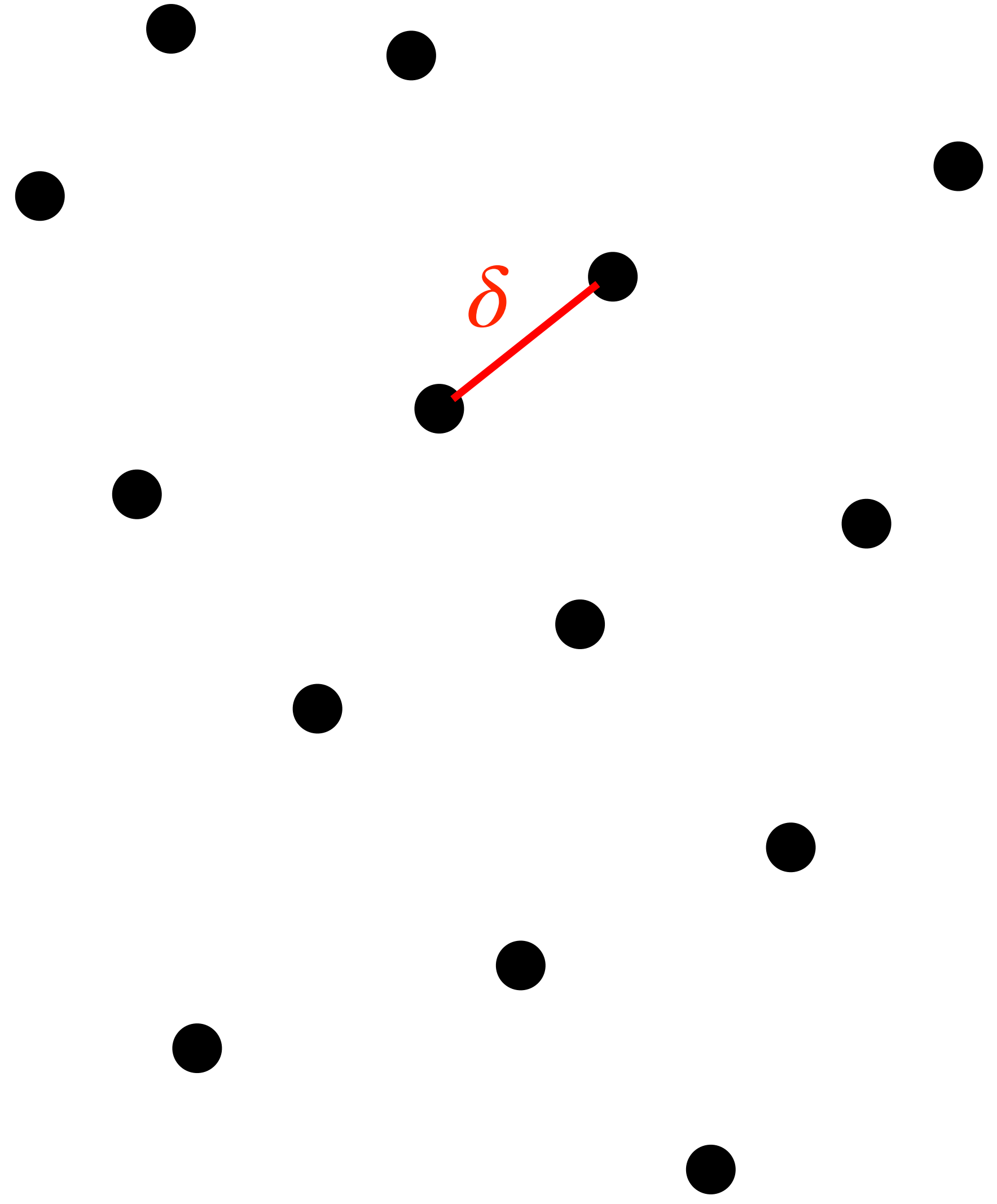
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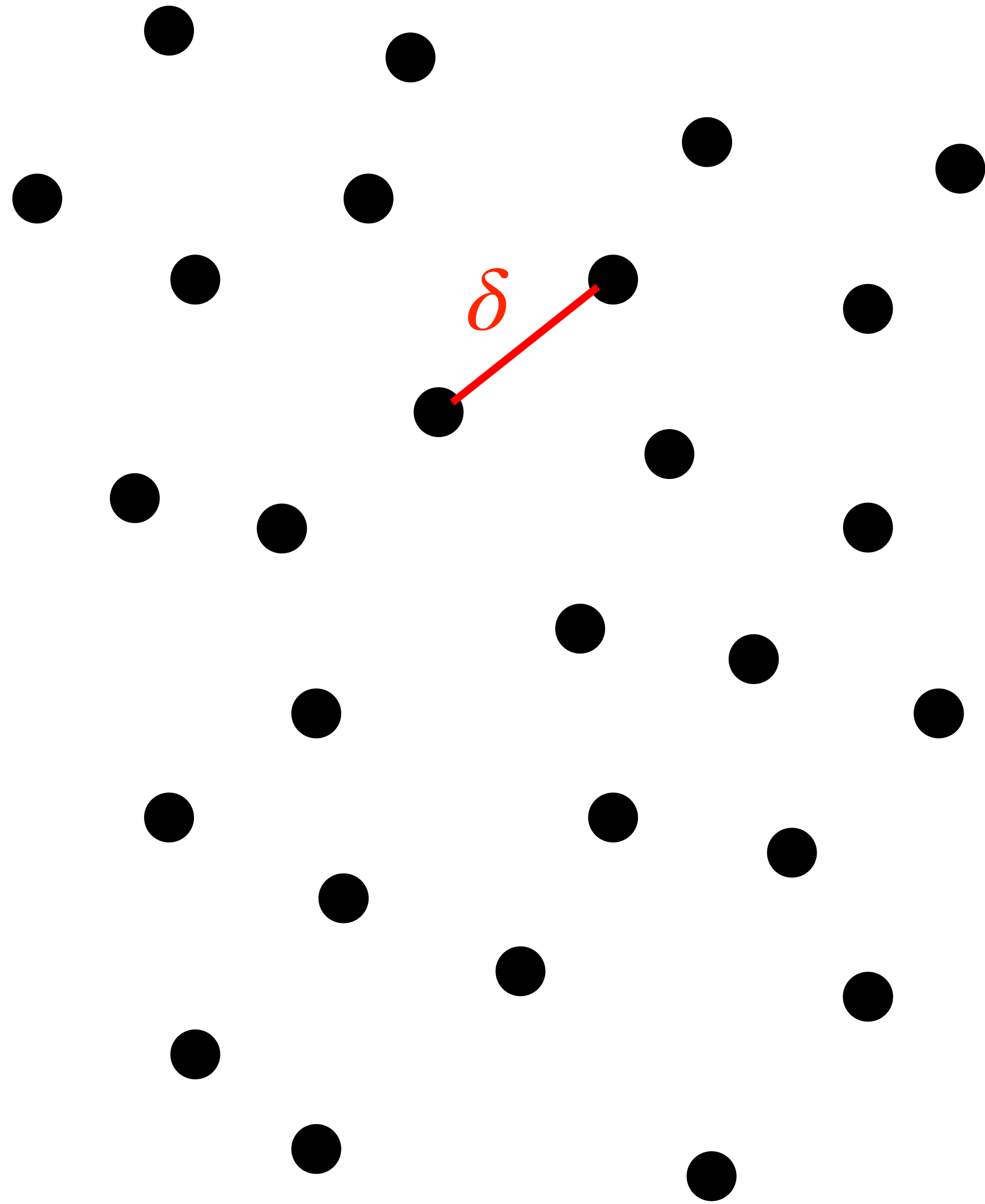
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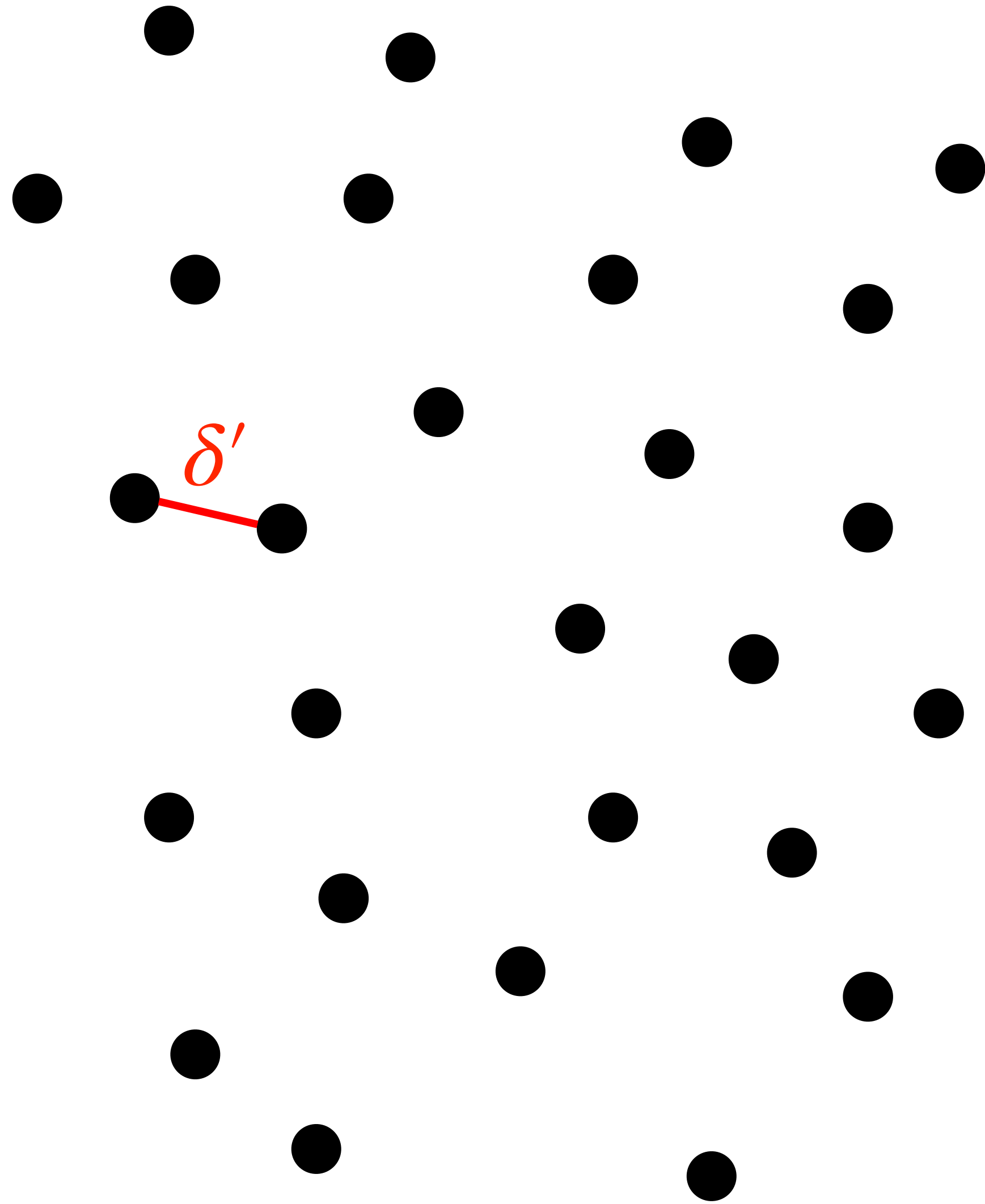
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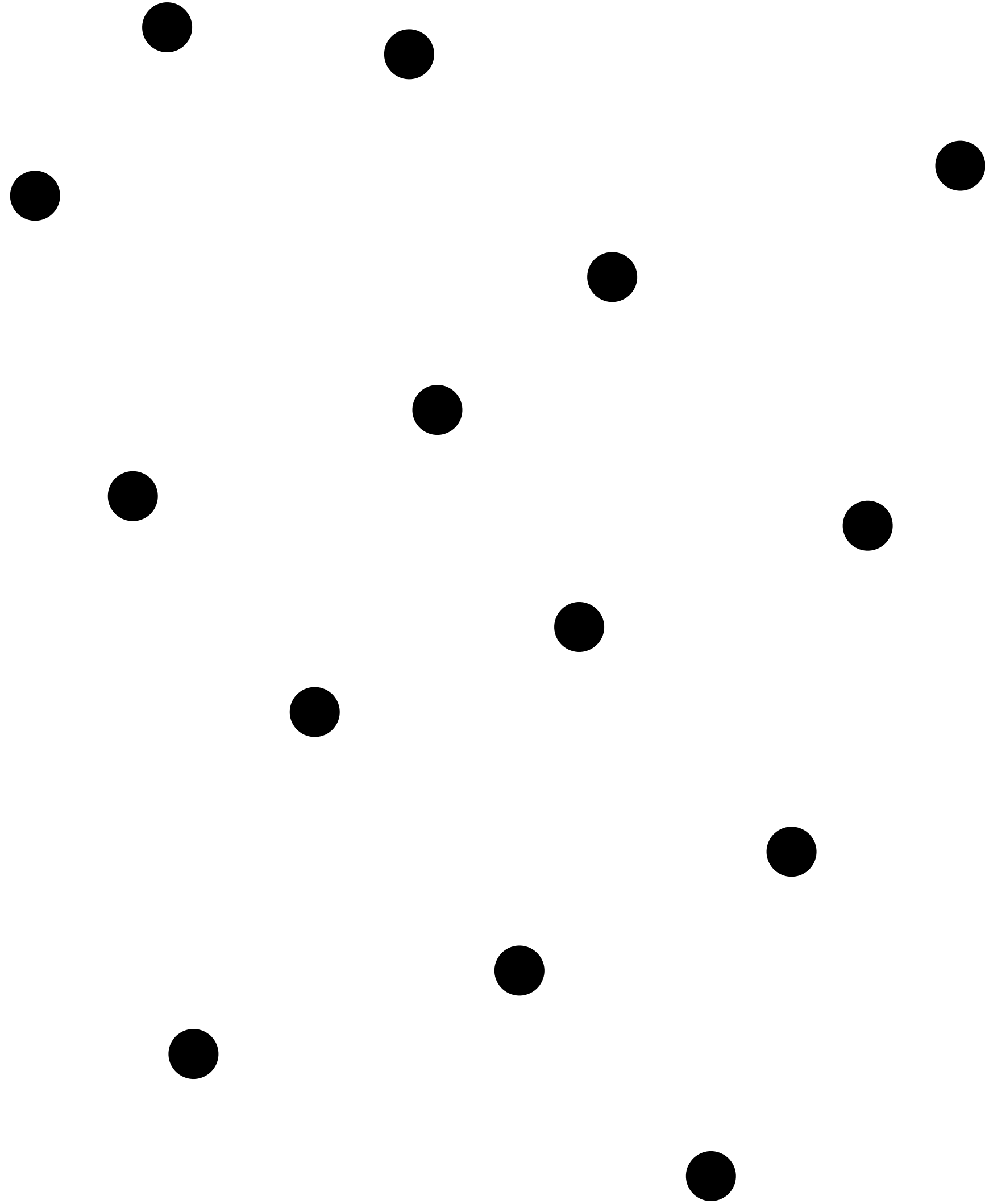


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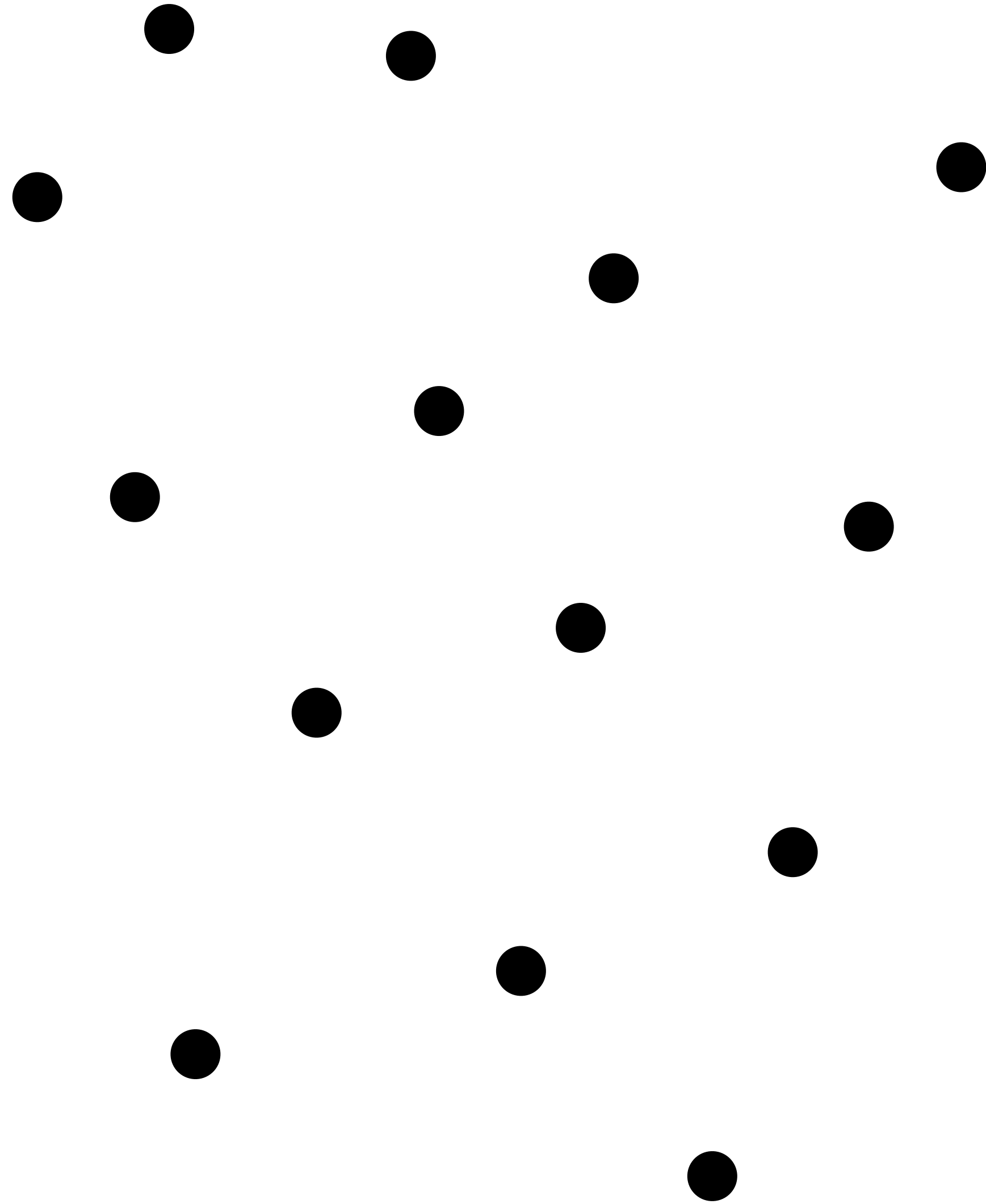
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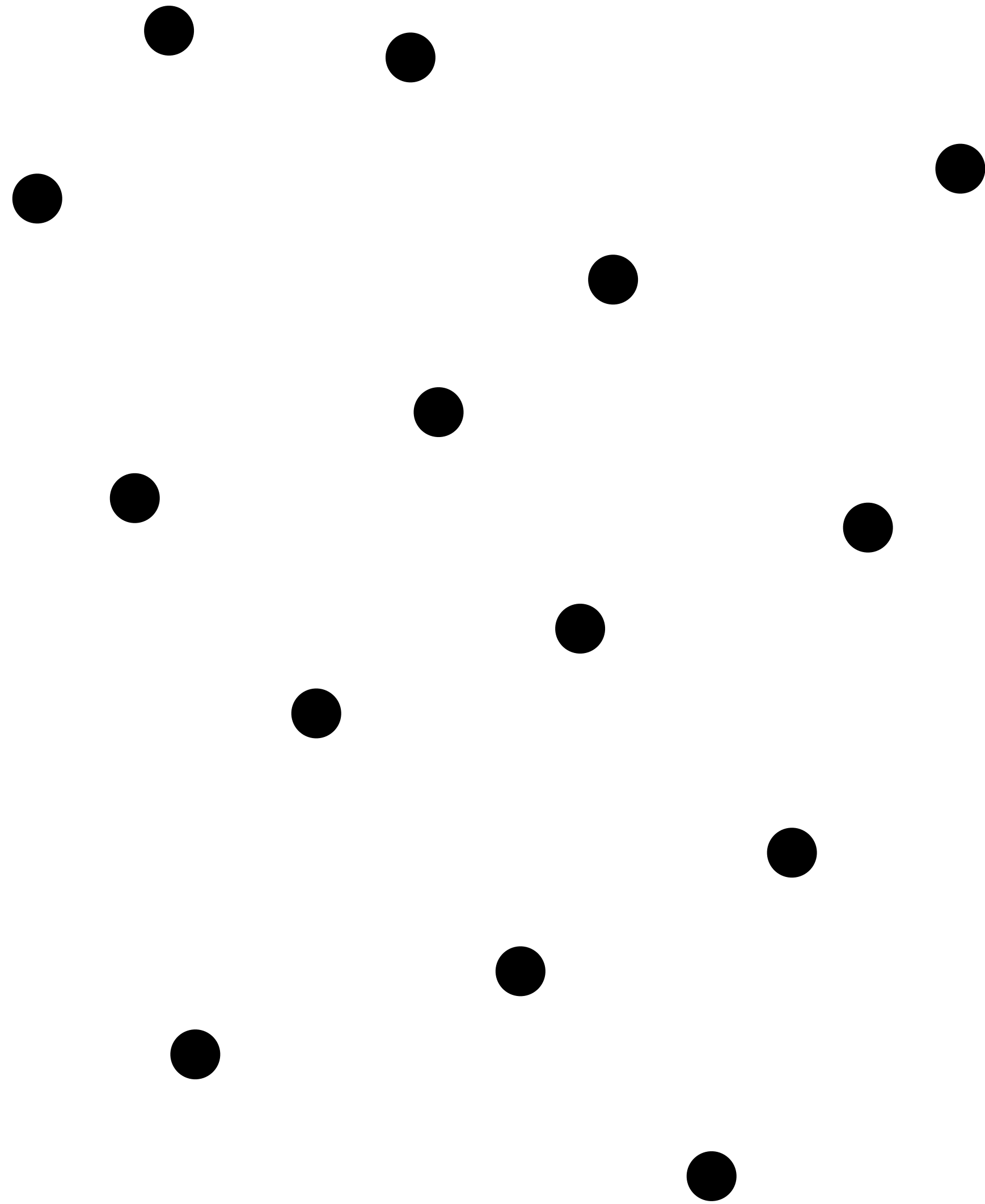
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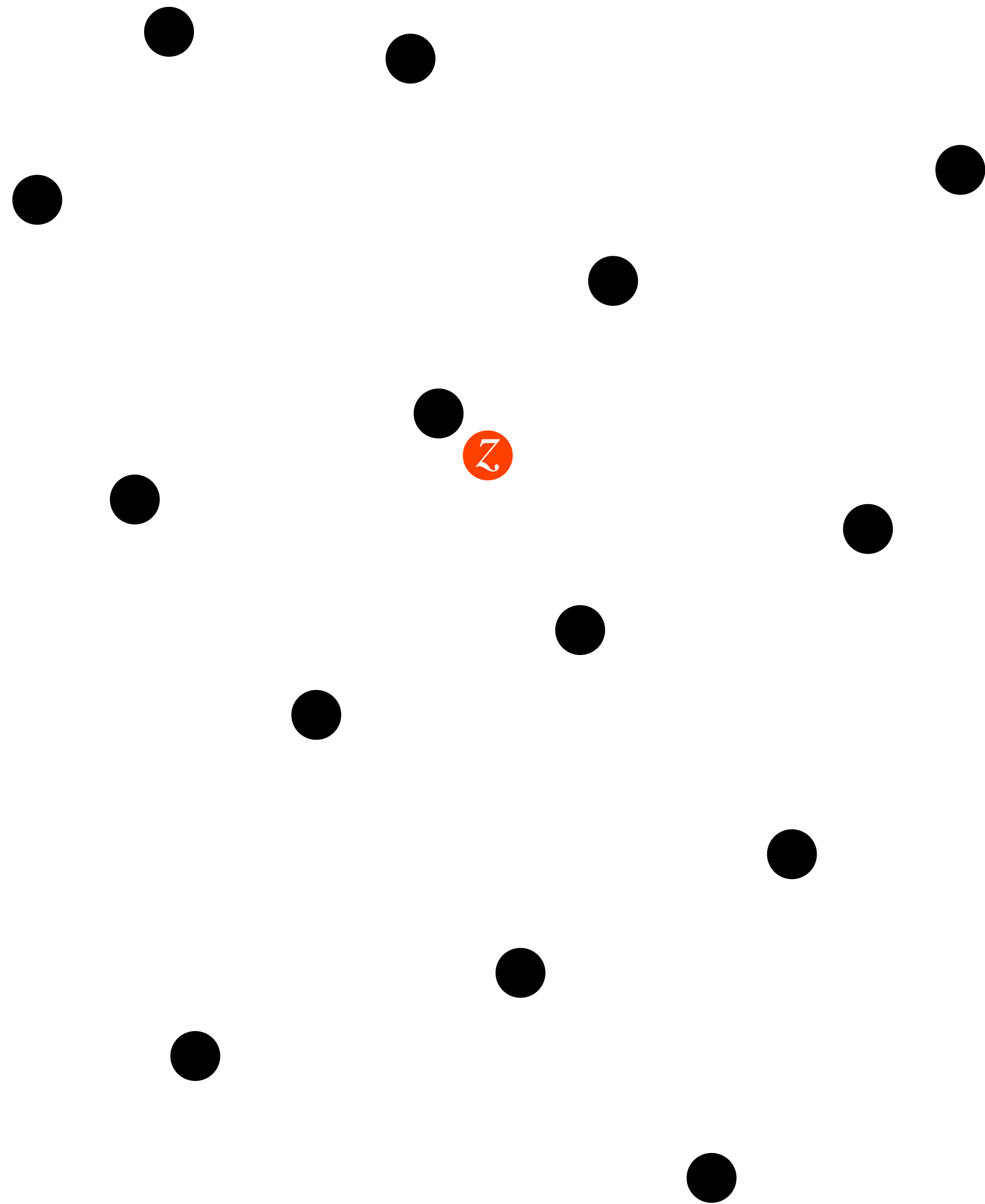
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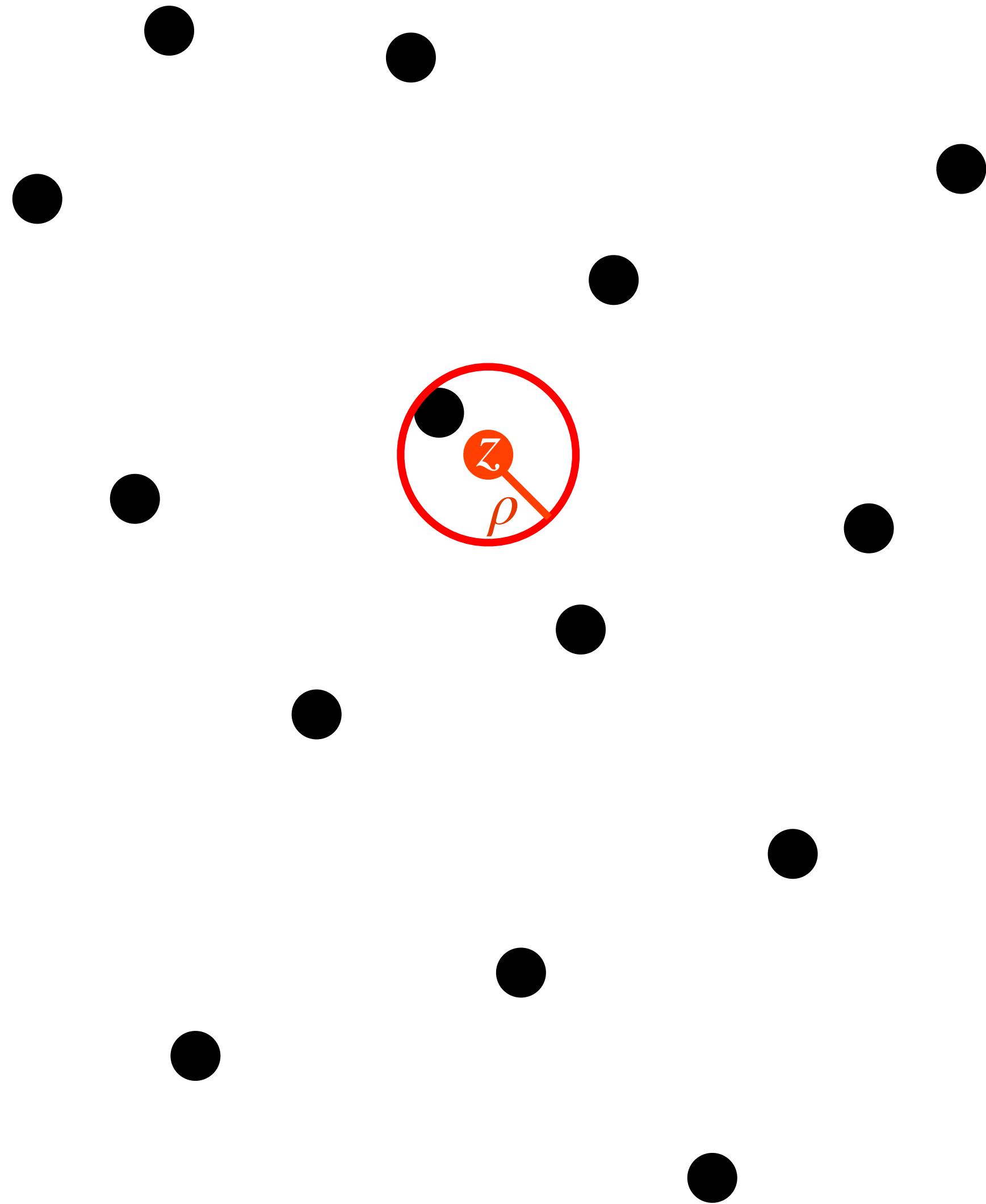
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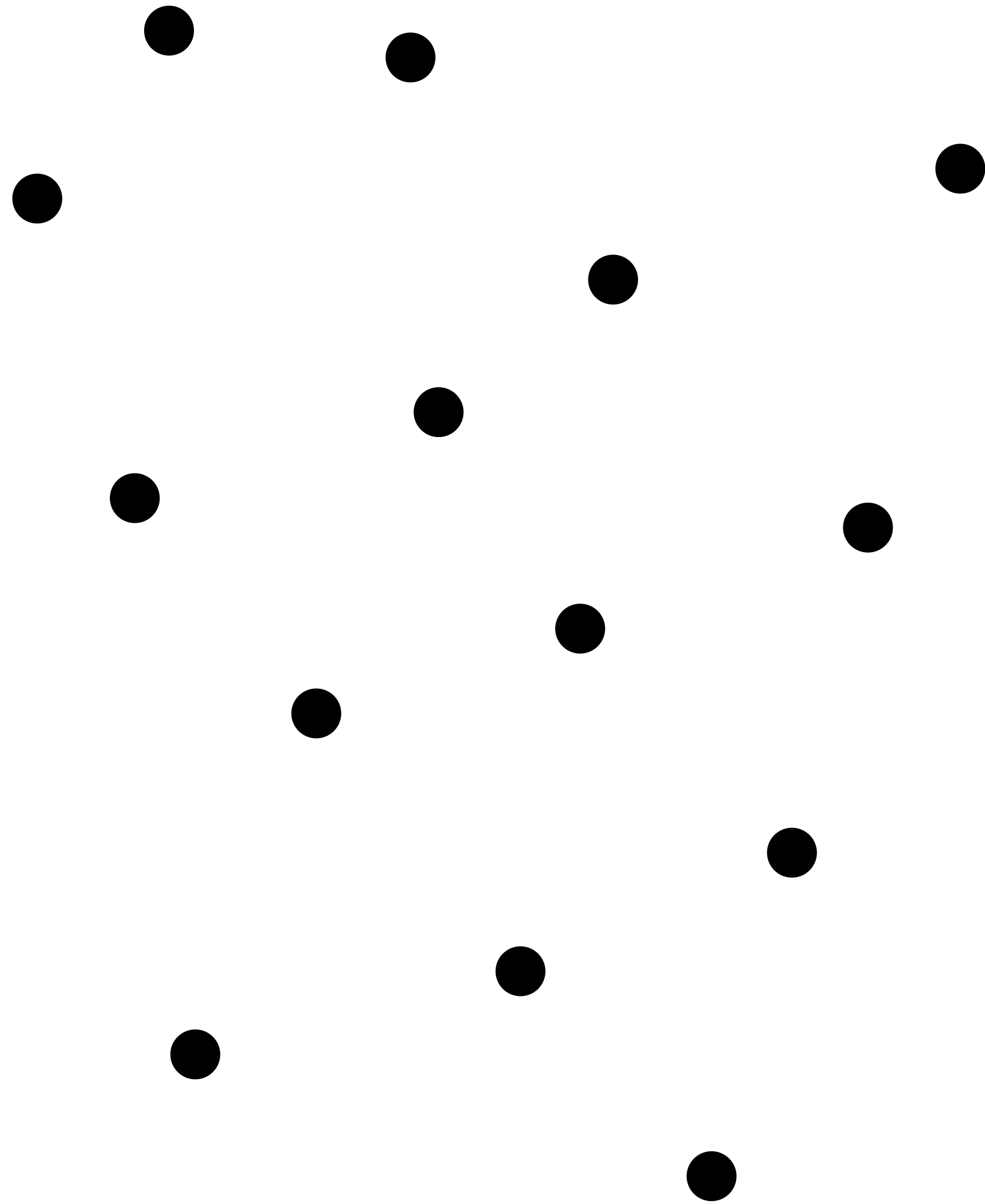
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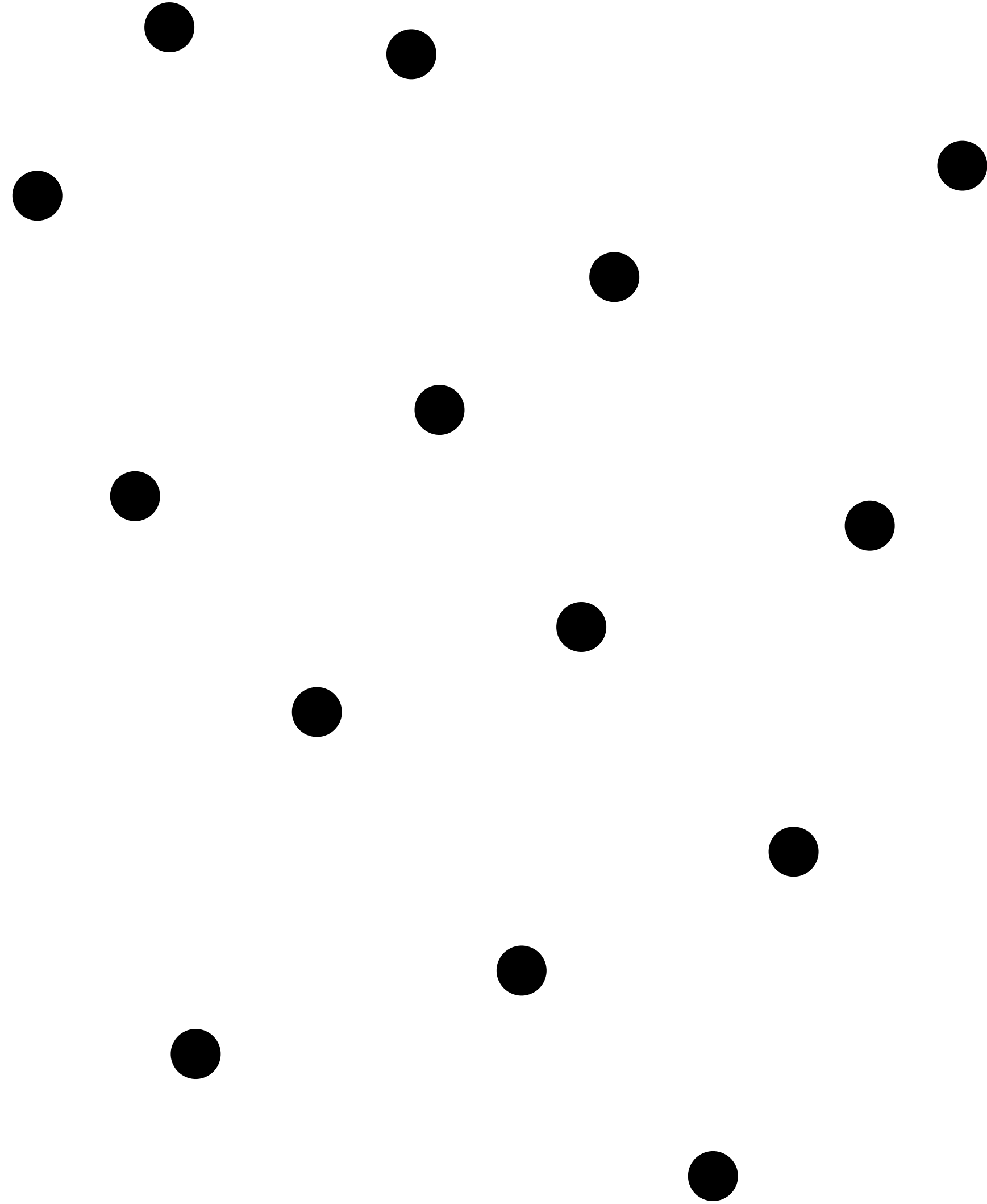
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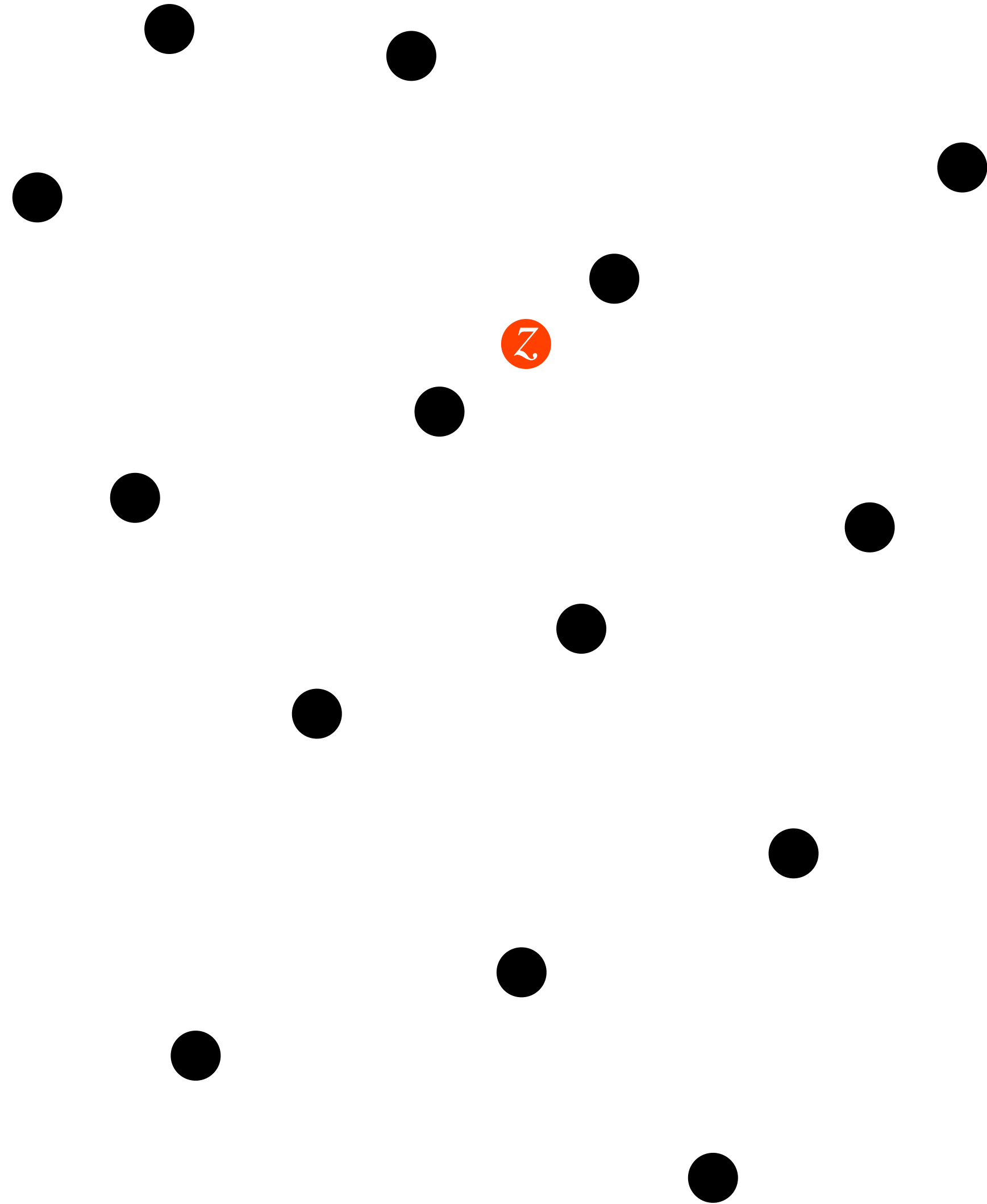
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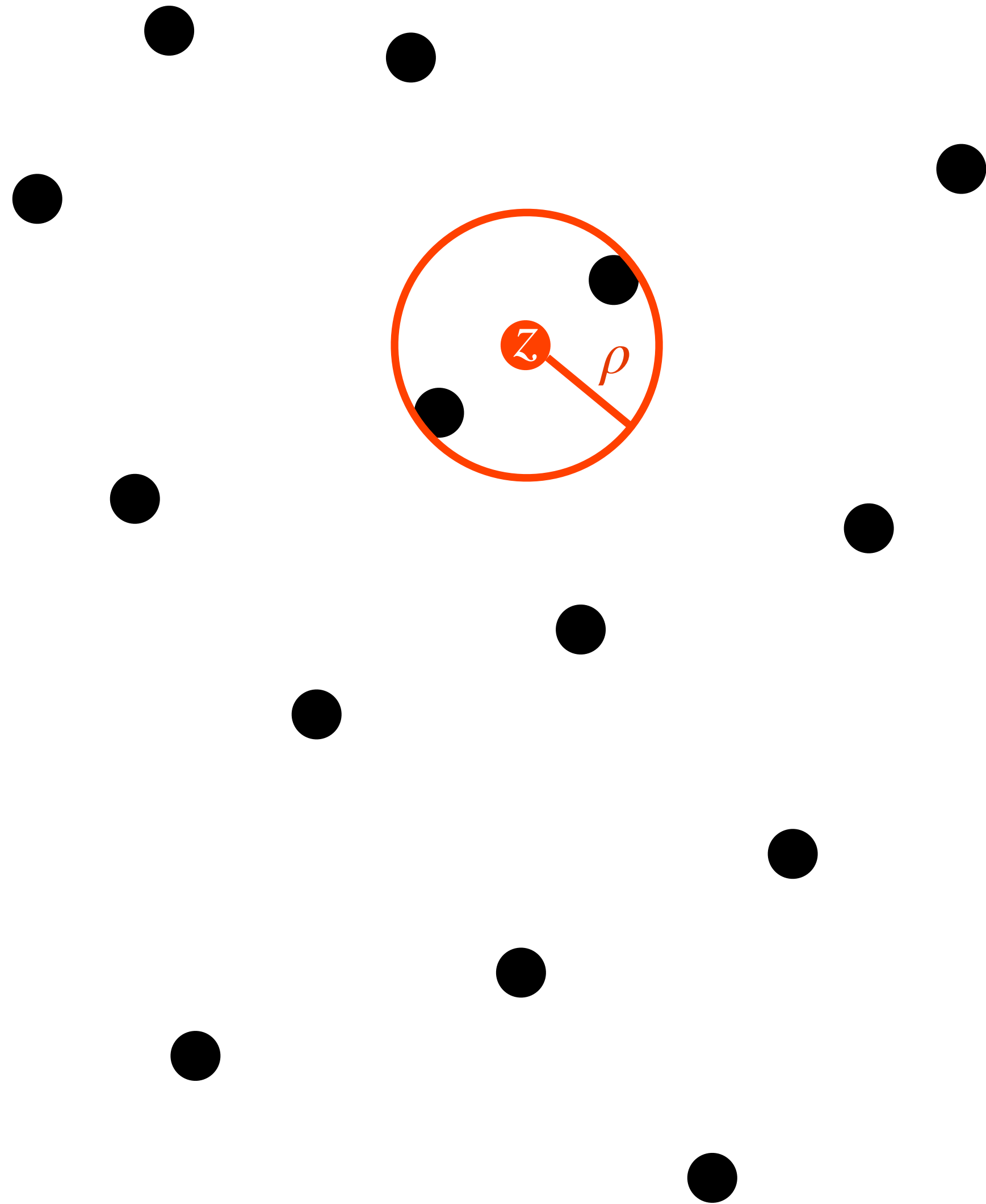


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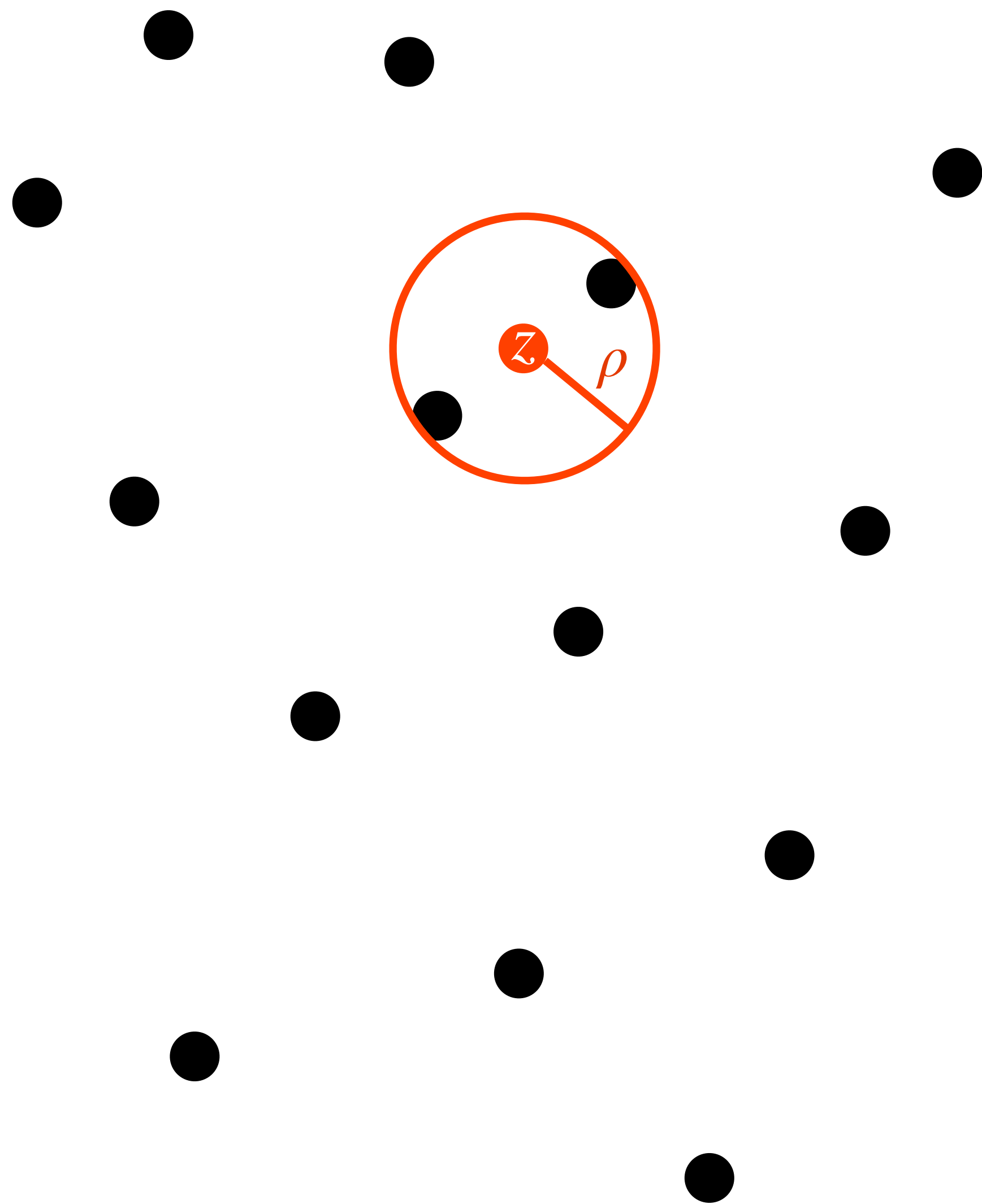
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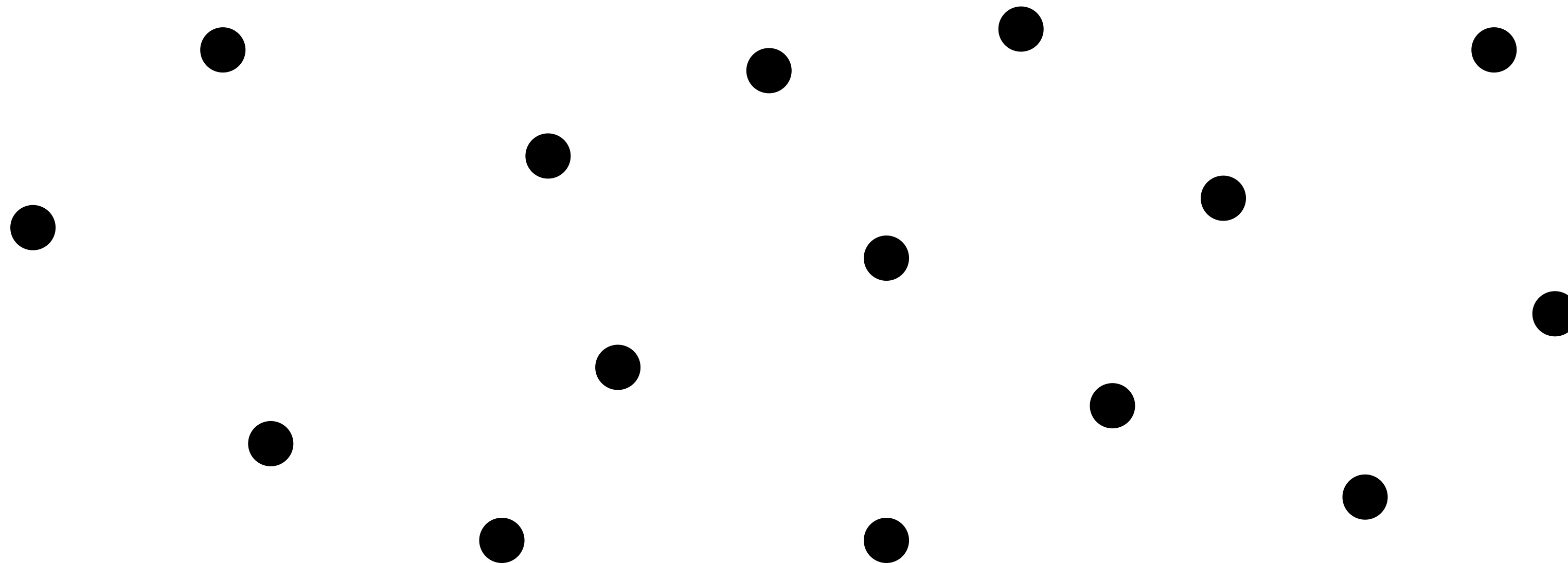
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- “Half-the-distance limit”

# LIST-DECODABILITY

[Elias'57, Wozencraft'58]

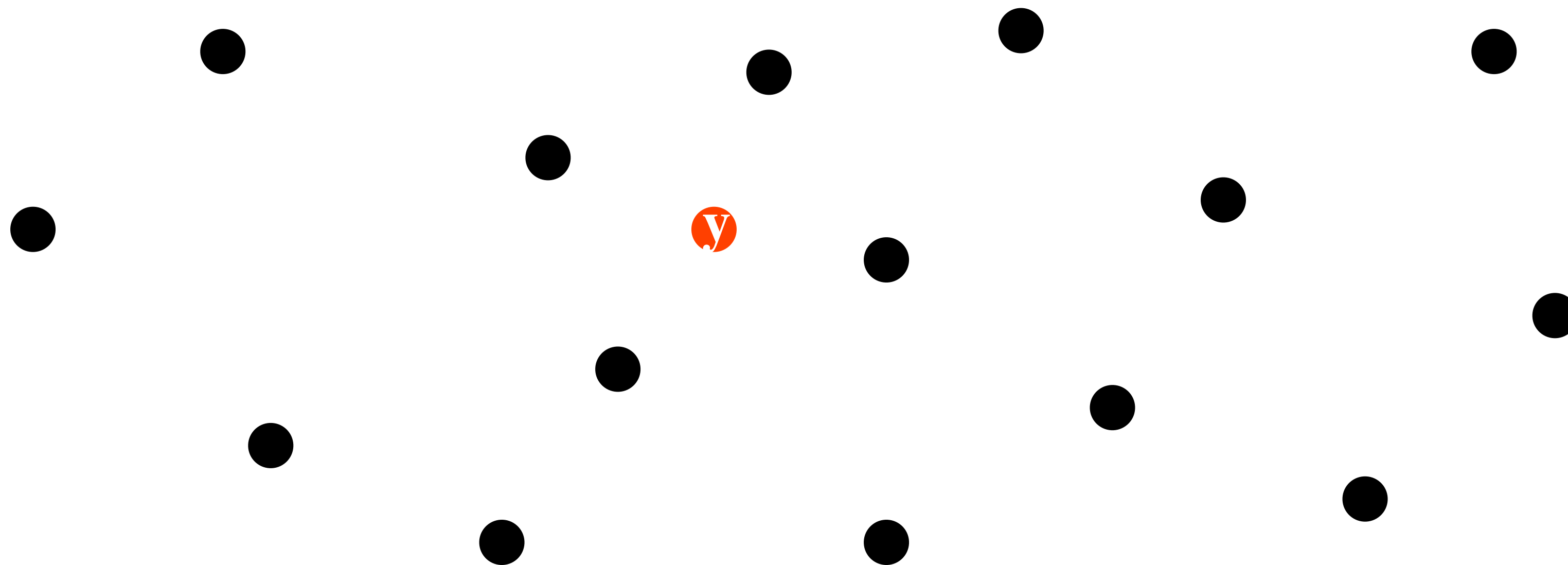
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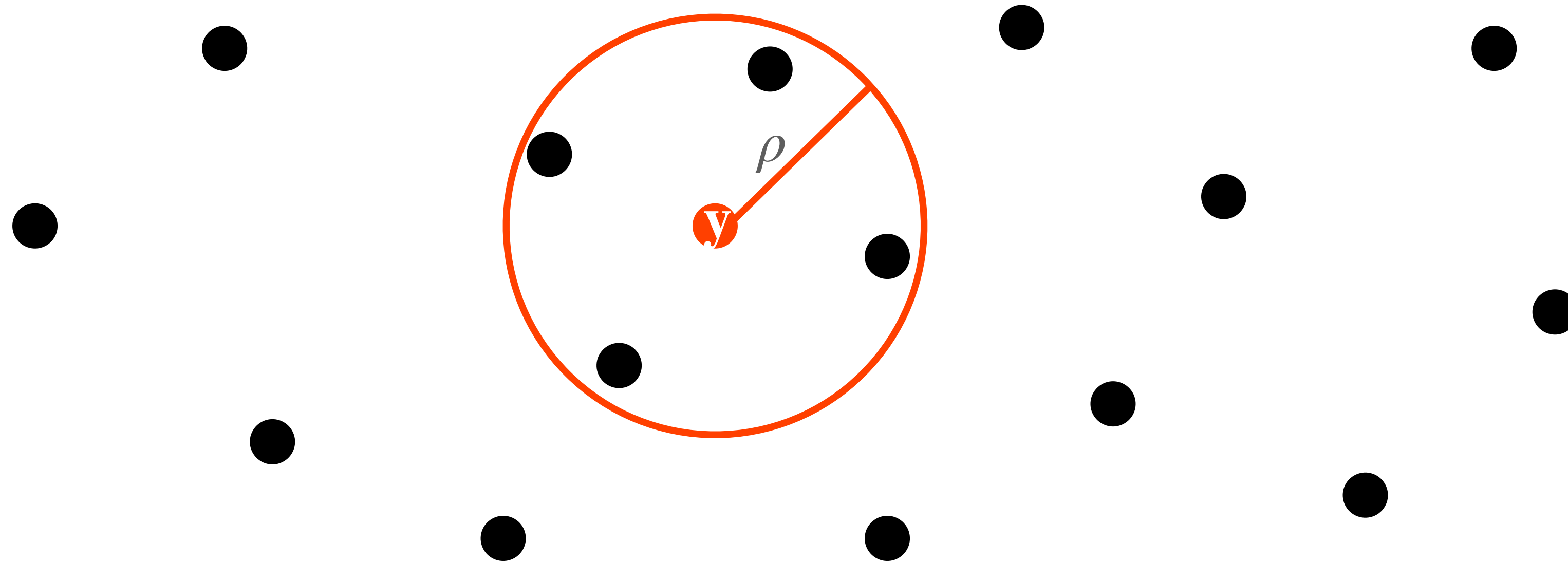
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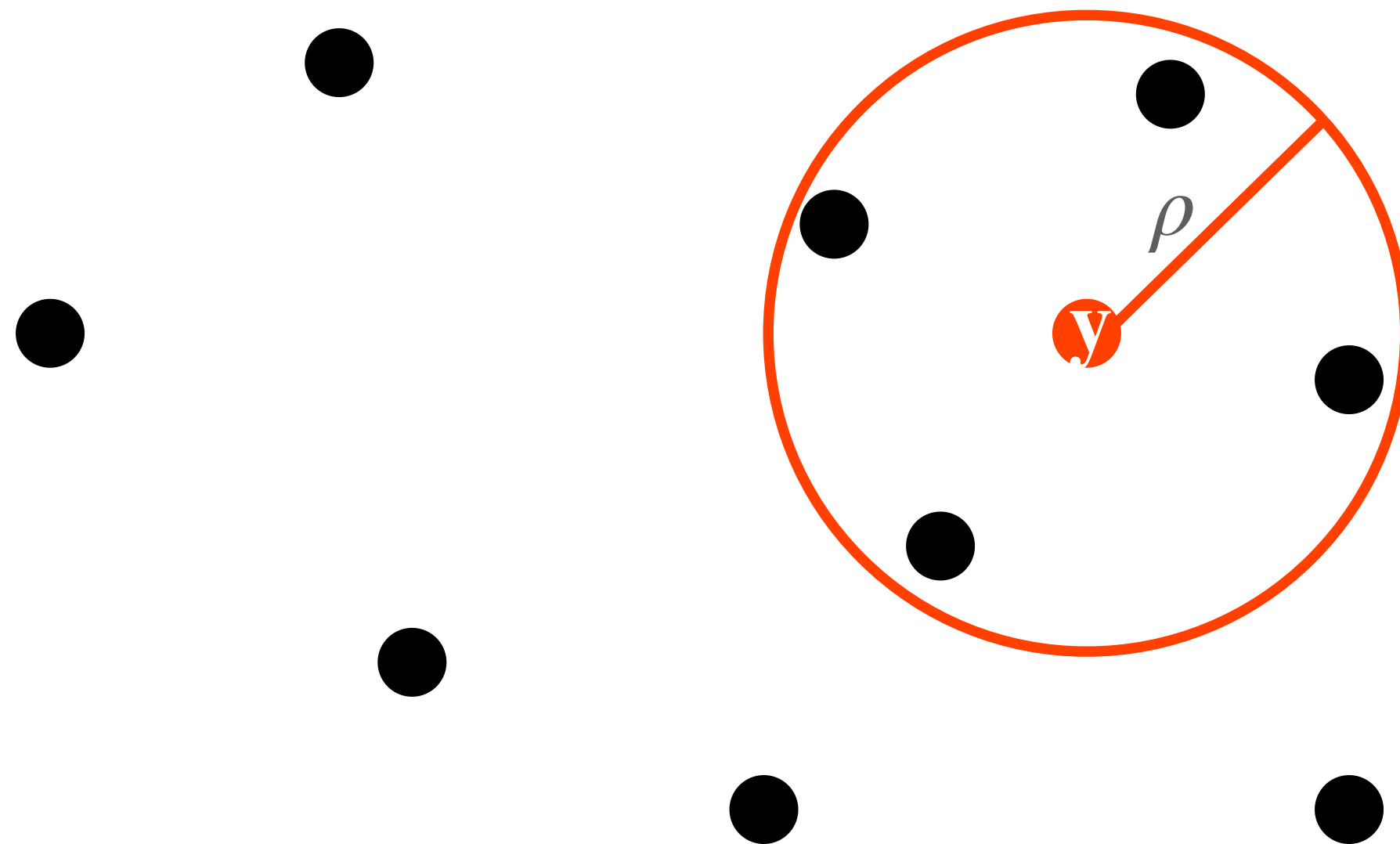
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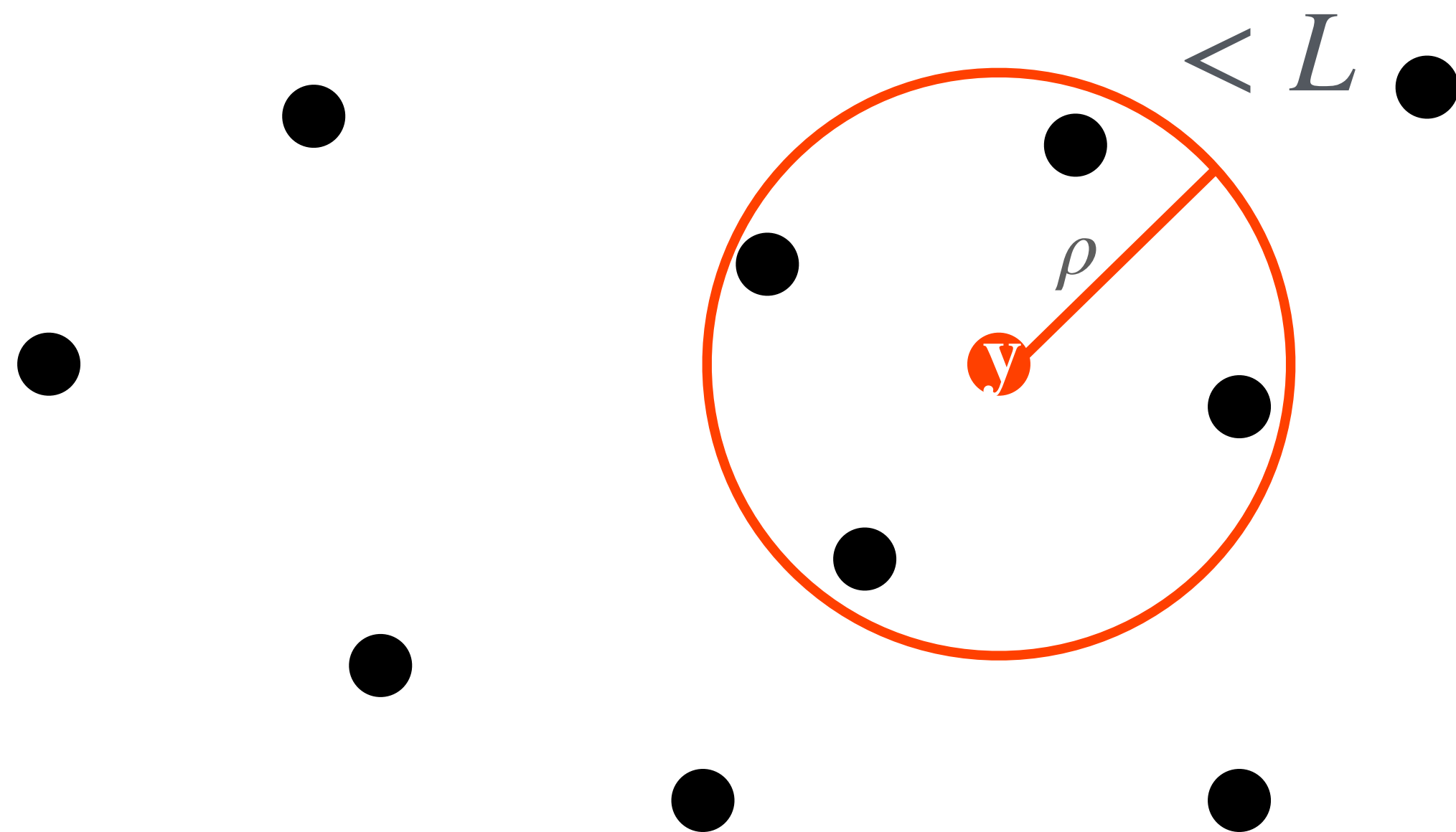
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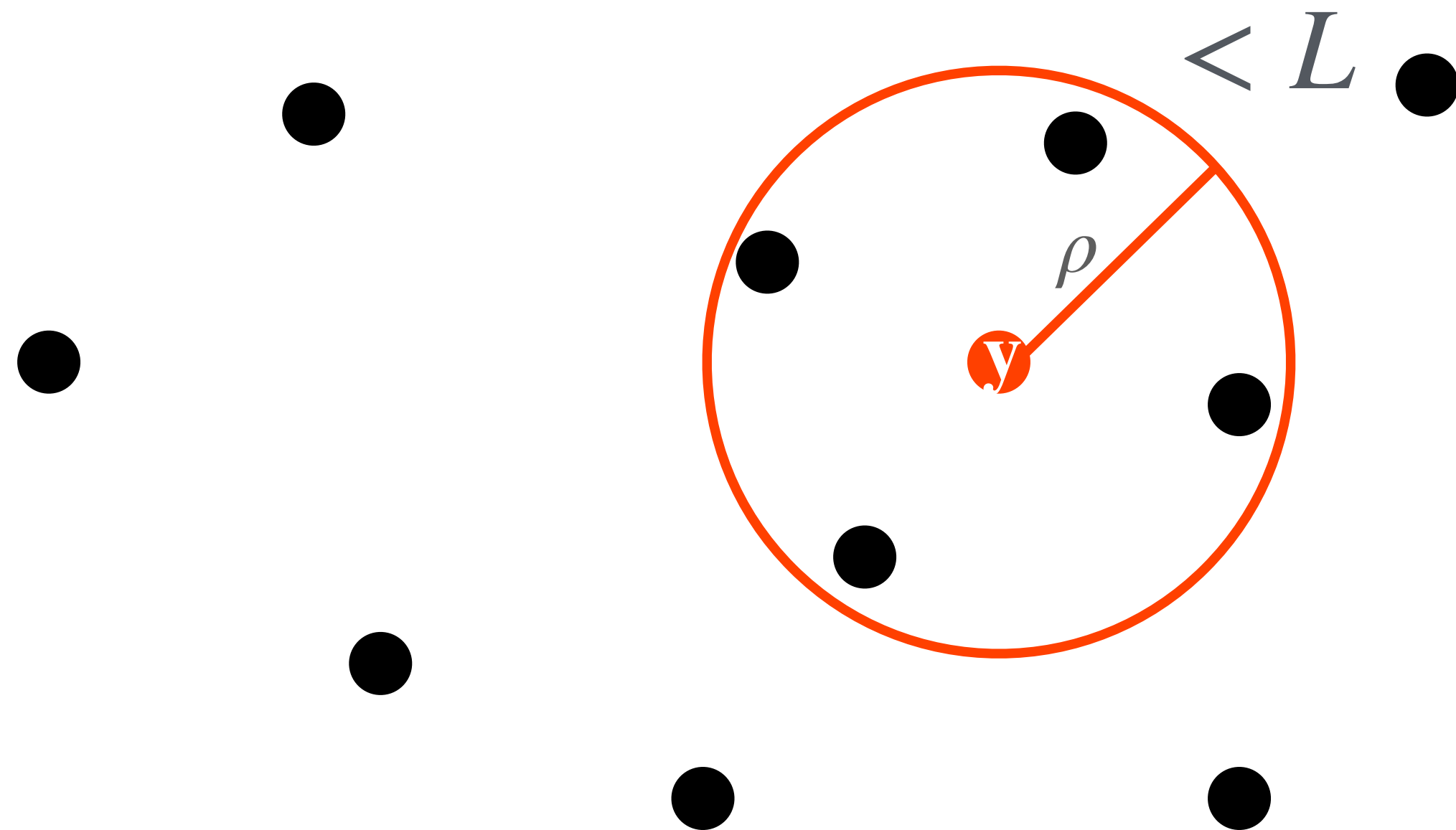
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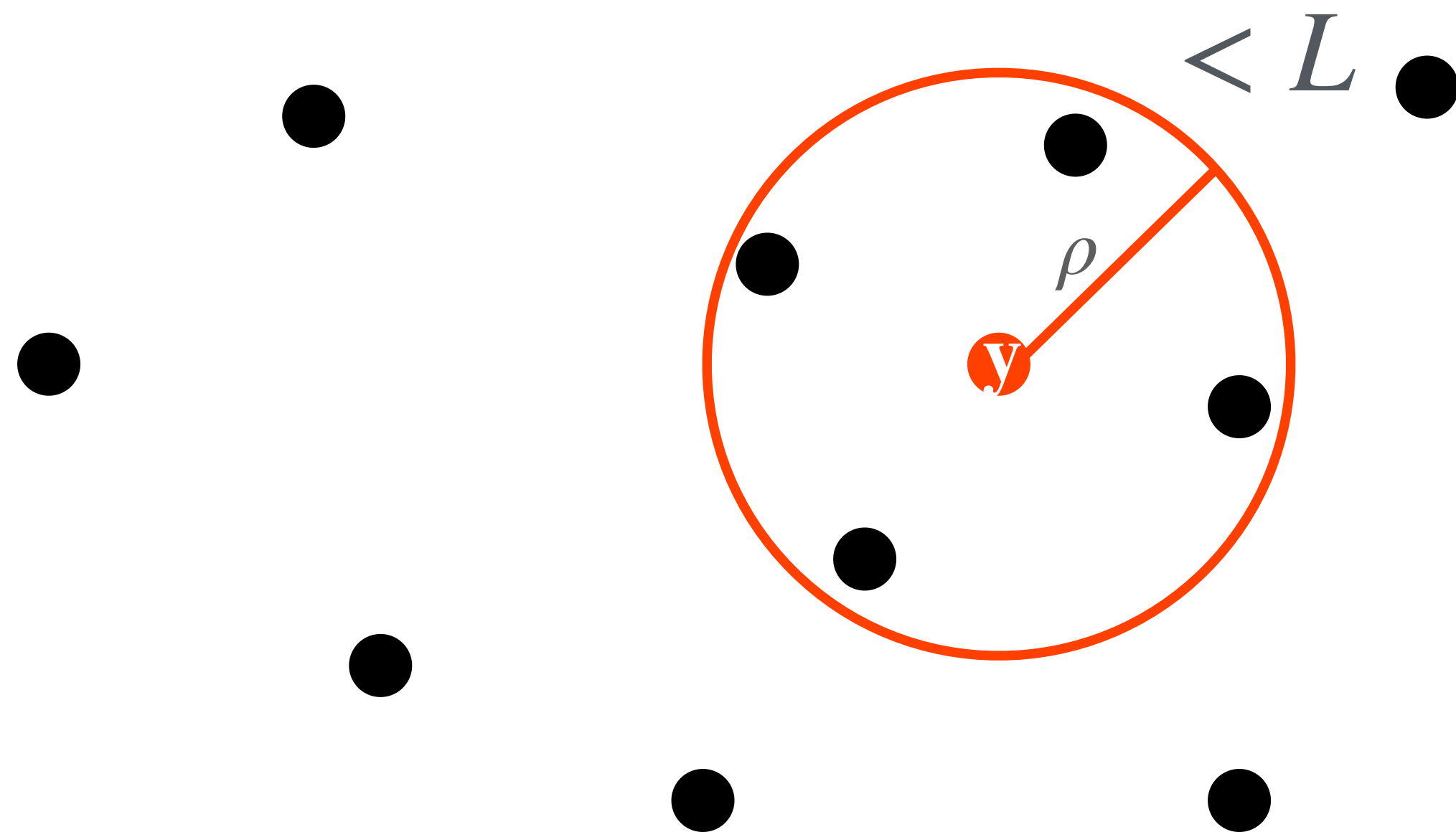
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Tradeoffs between  $R, \rho, L$ ?

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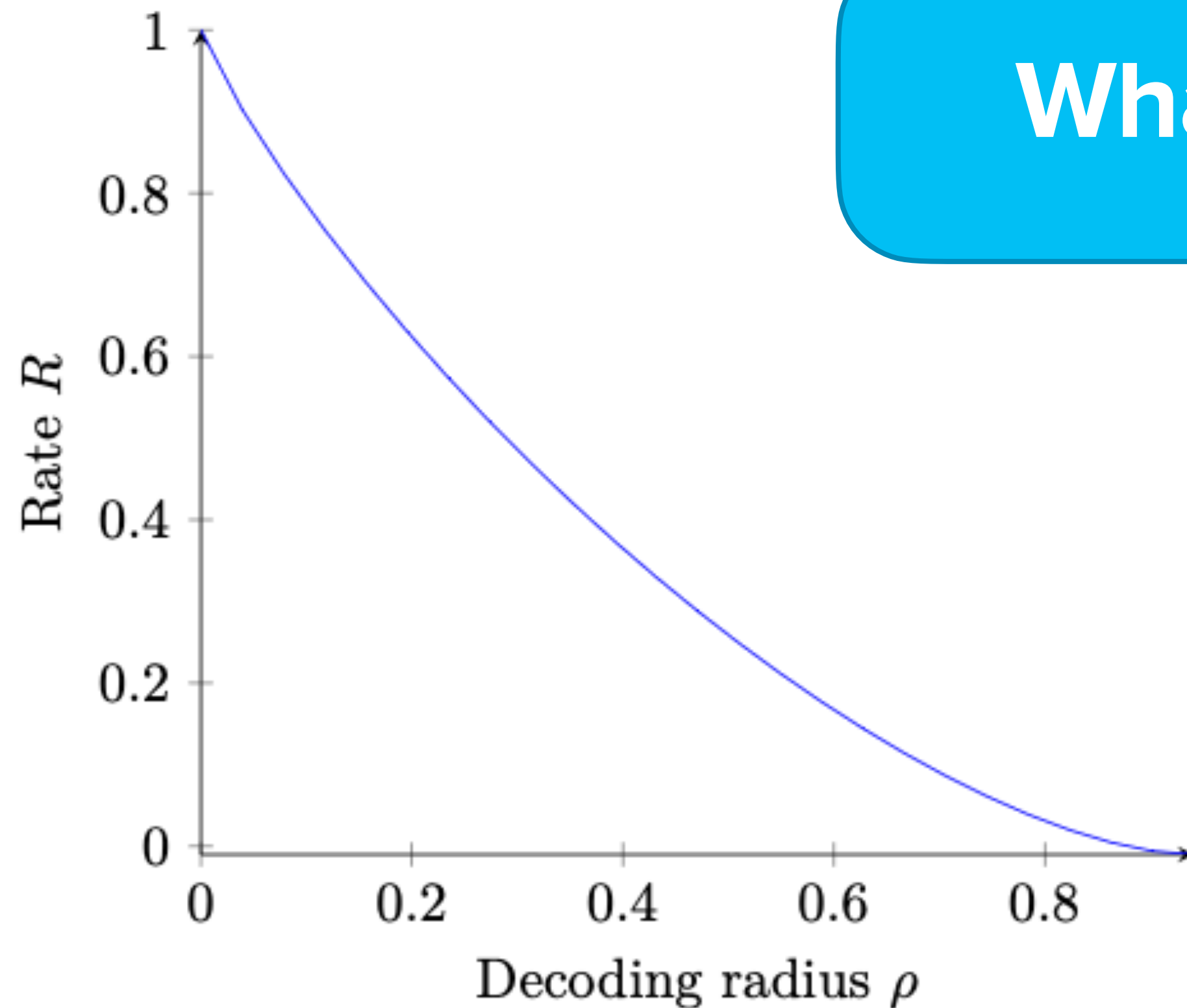
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Describe this curve?



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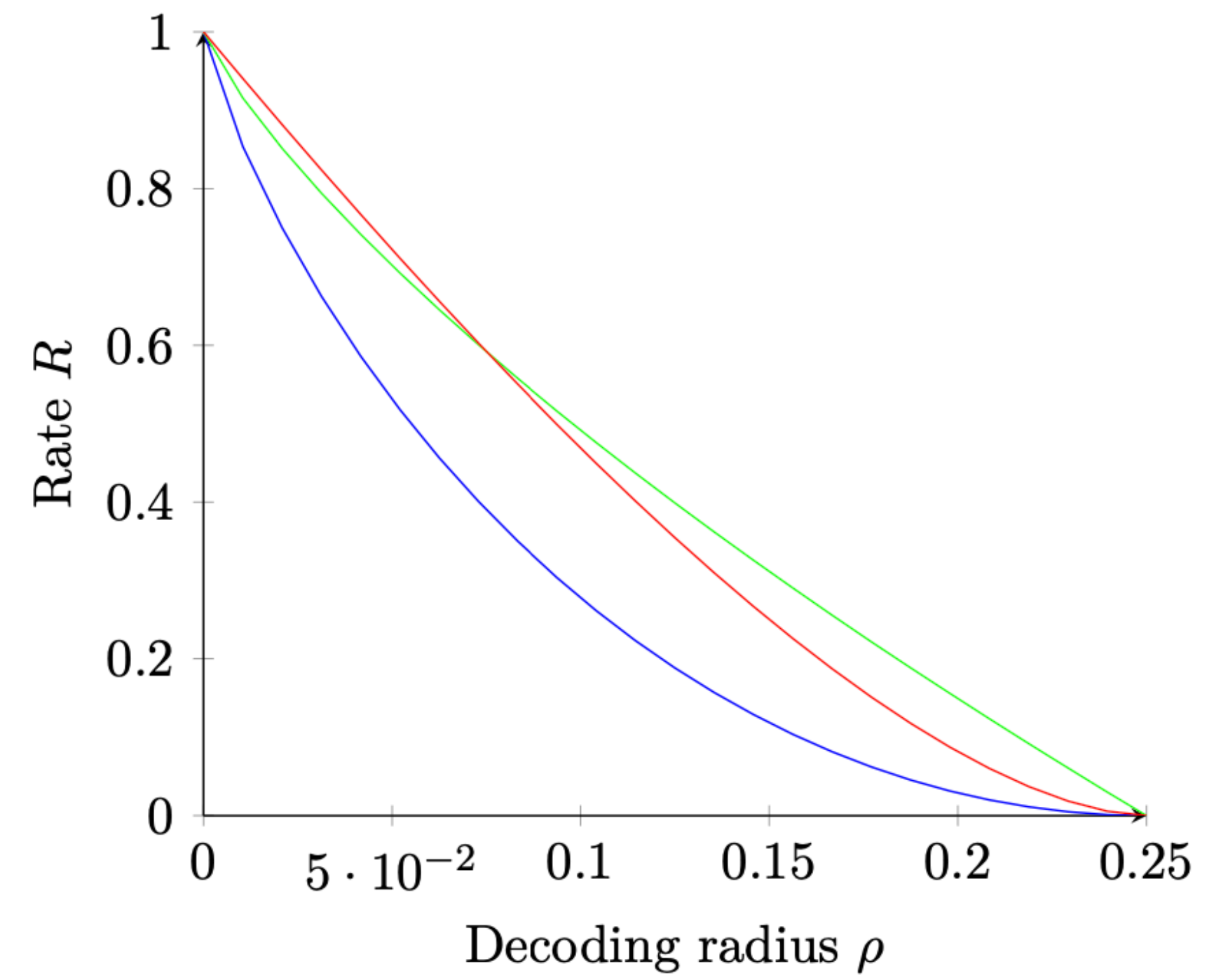
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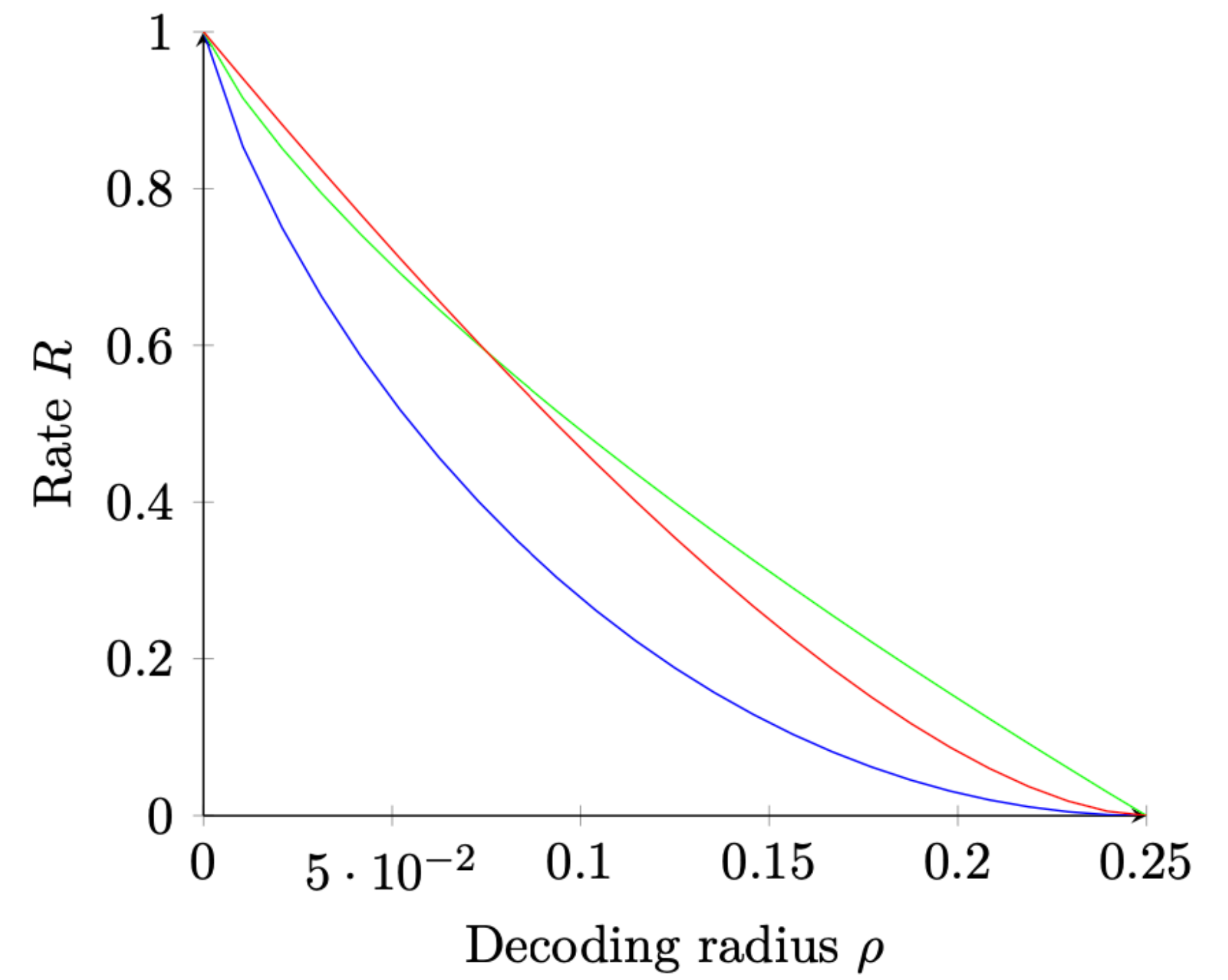


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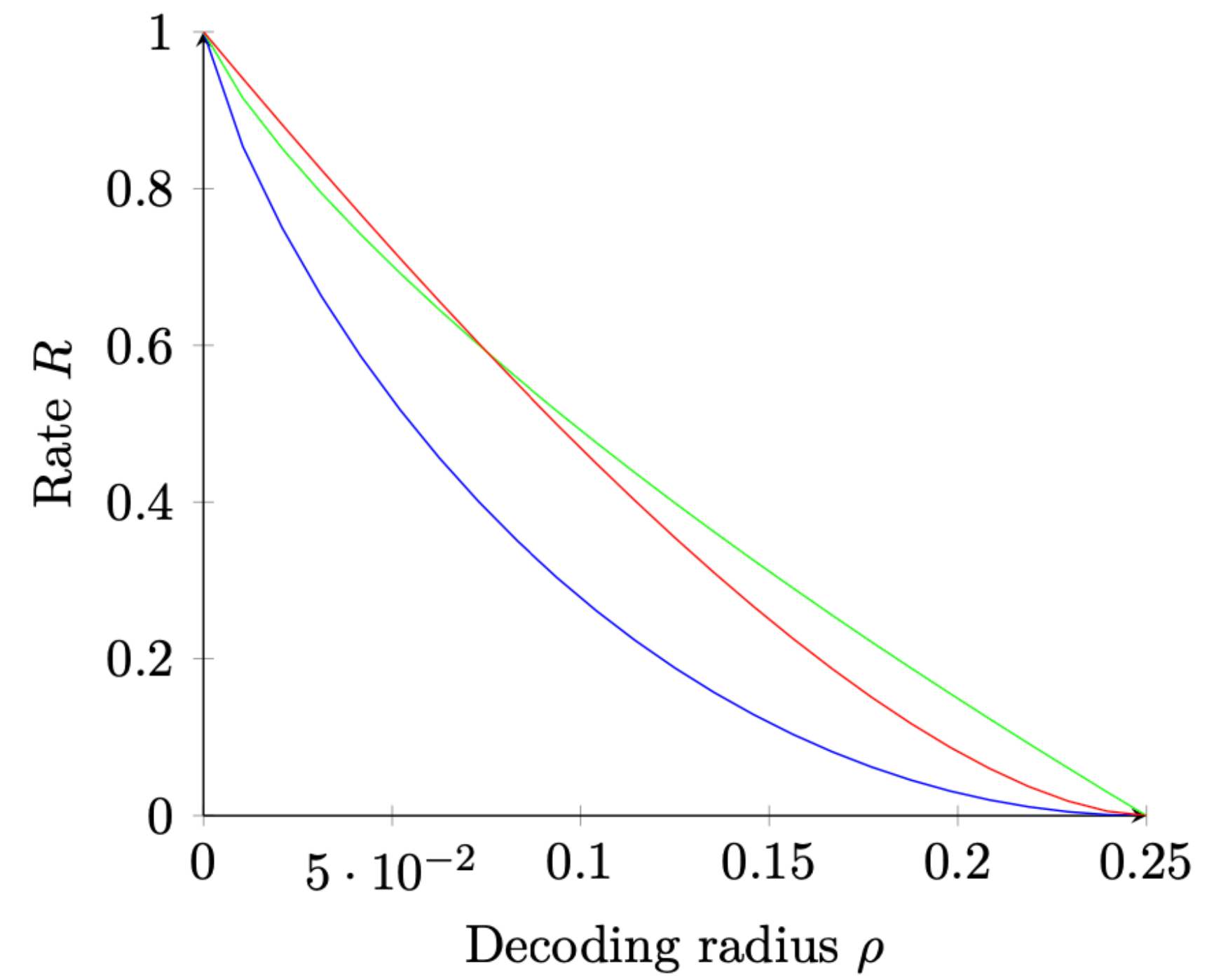
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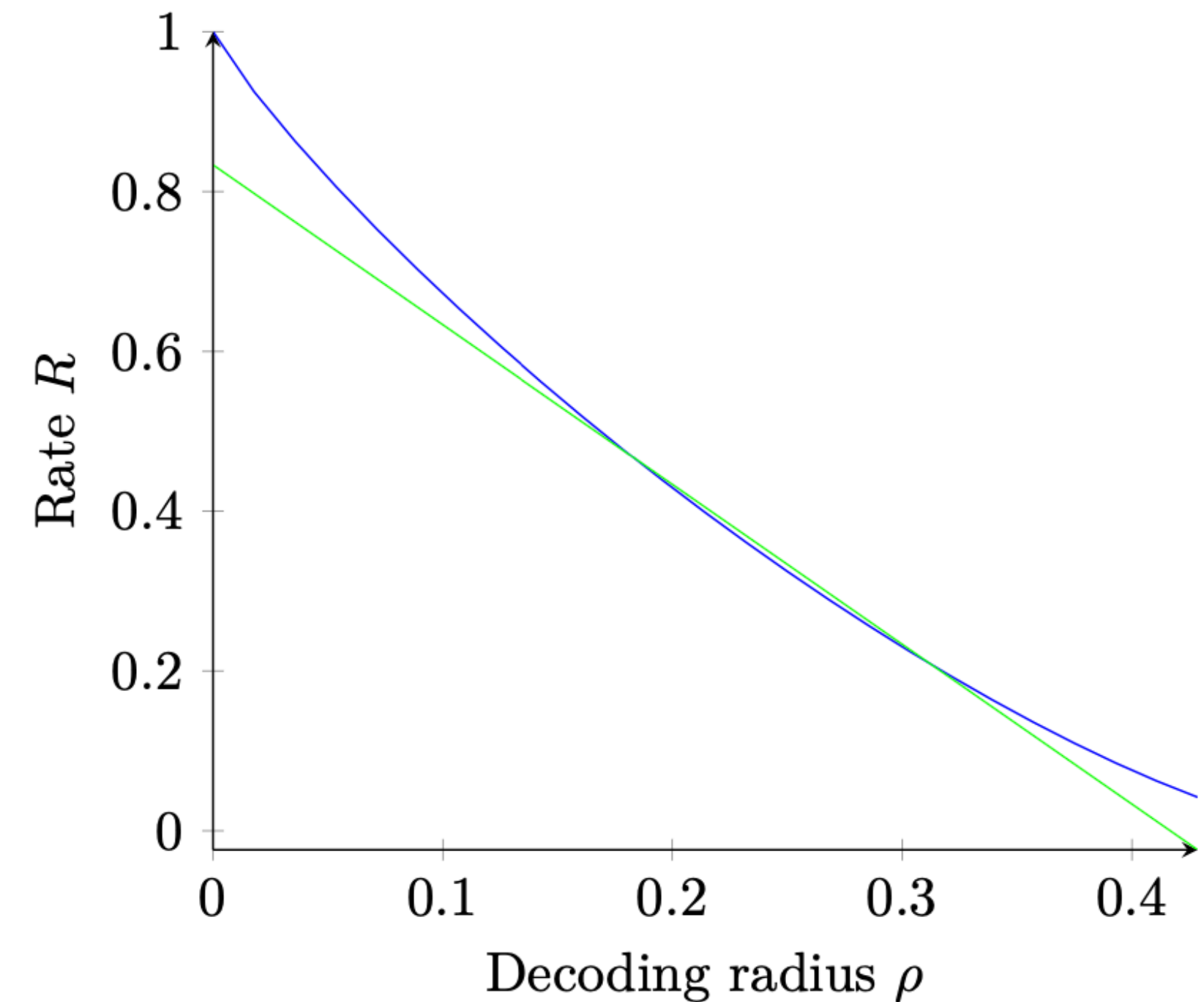
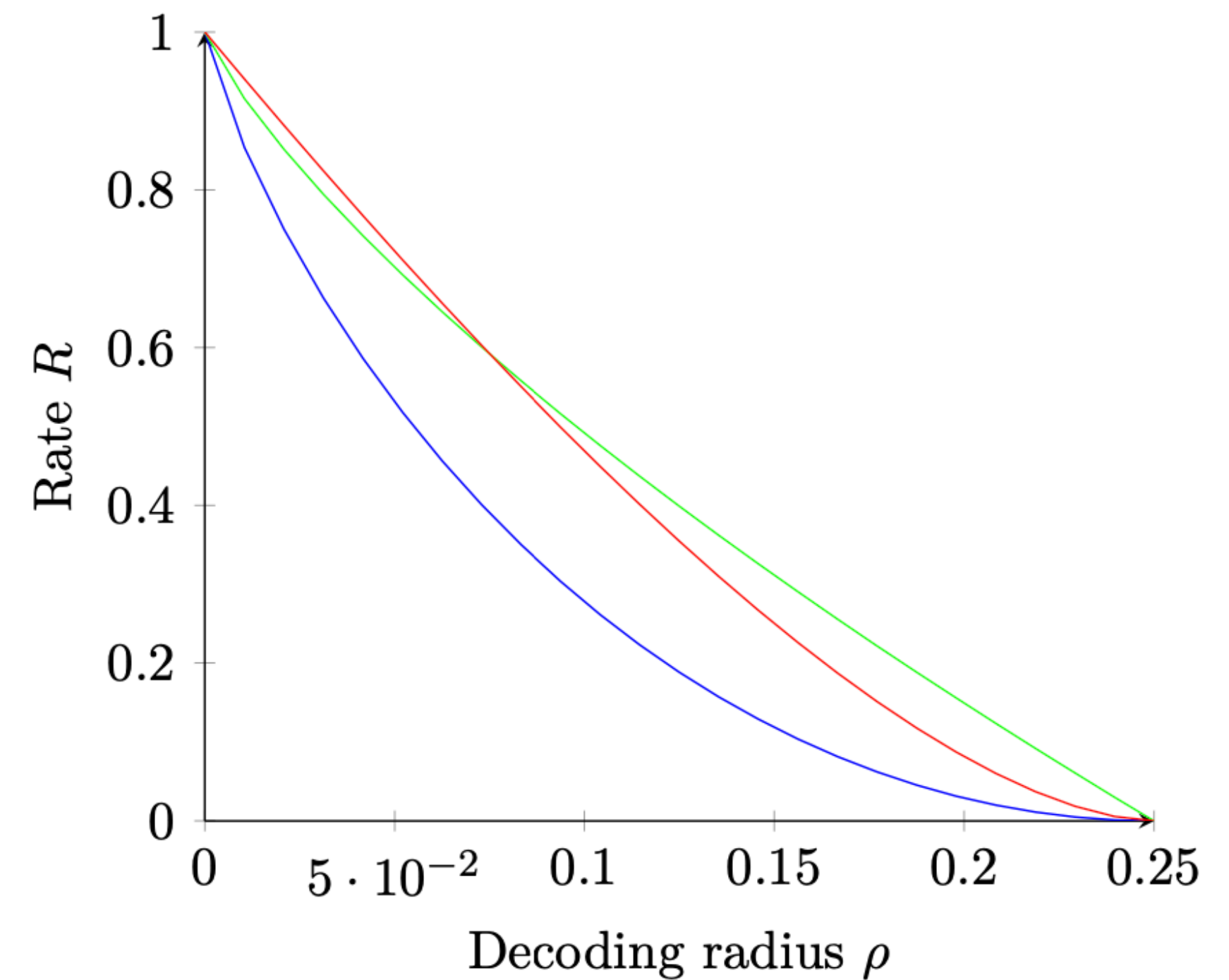
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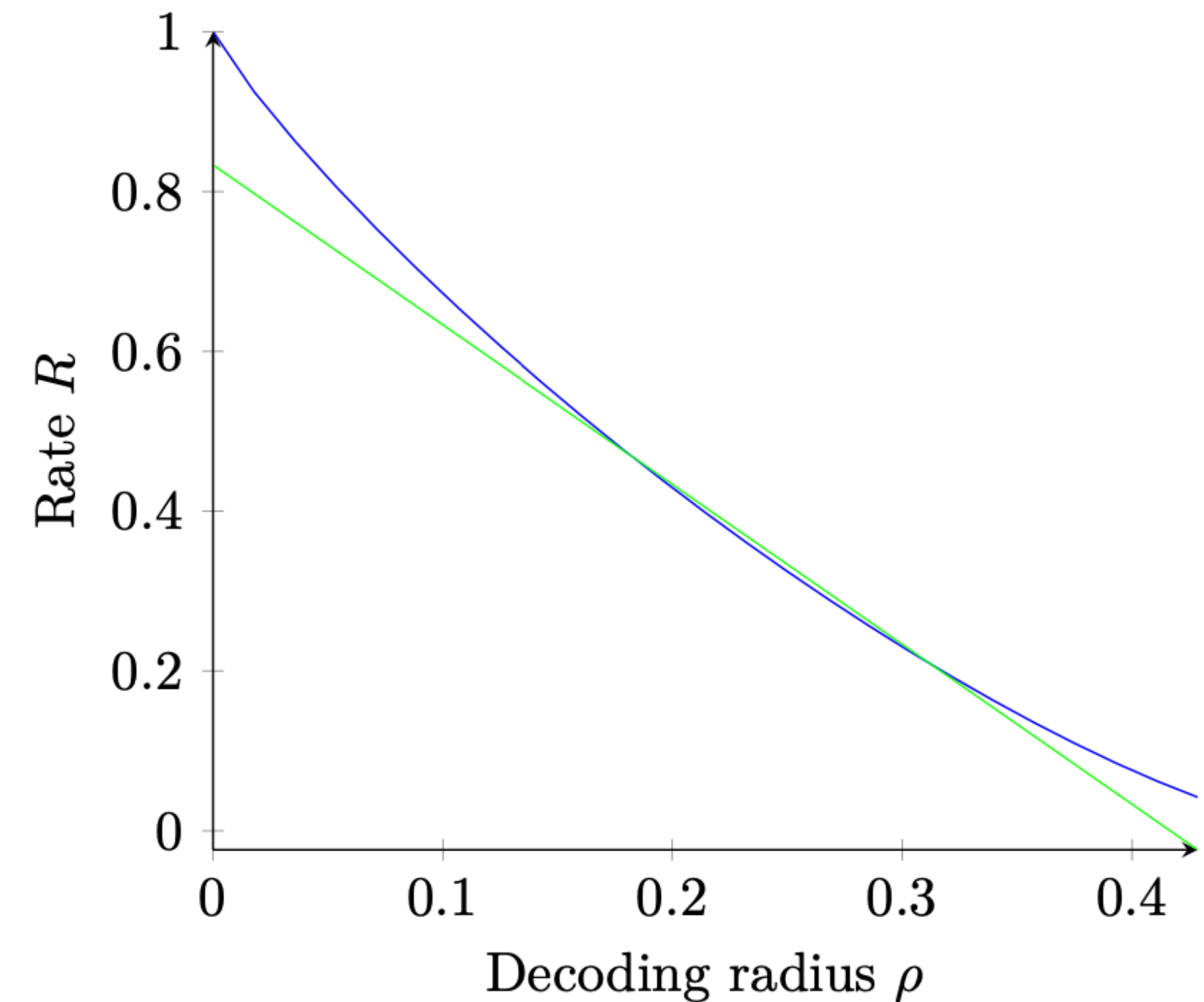
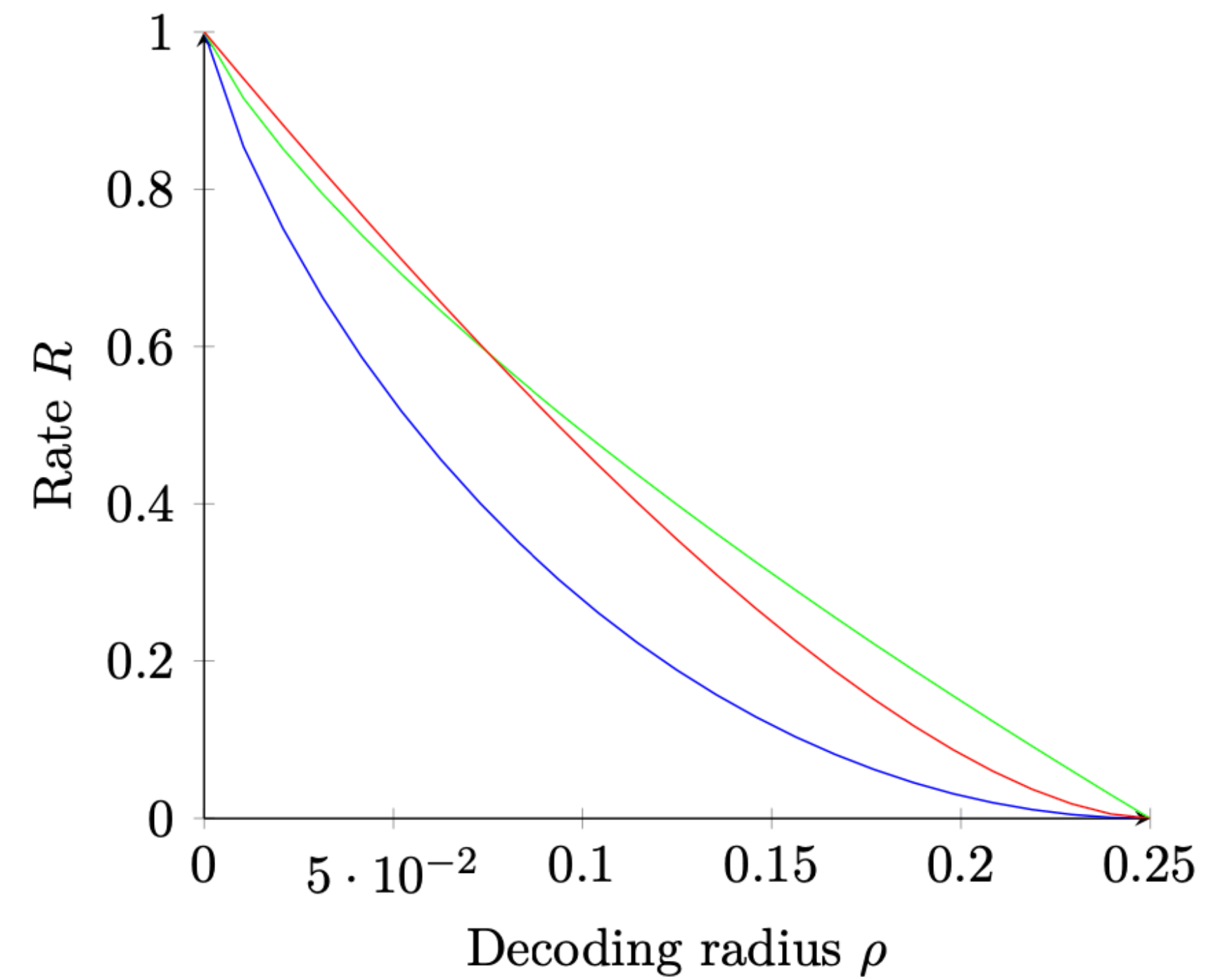
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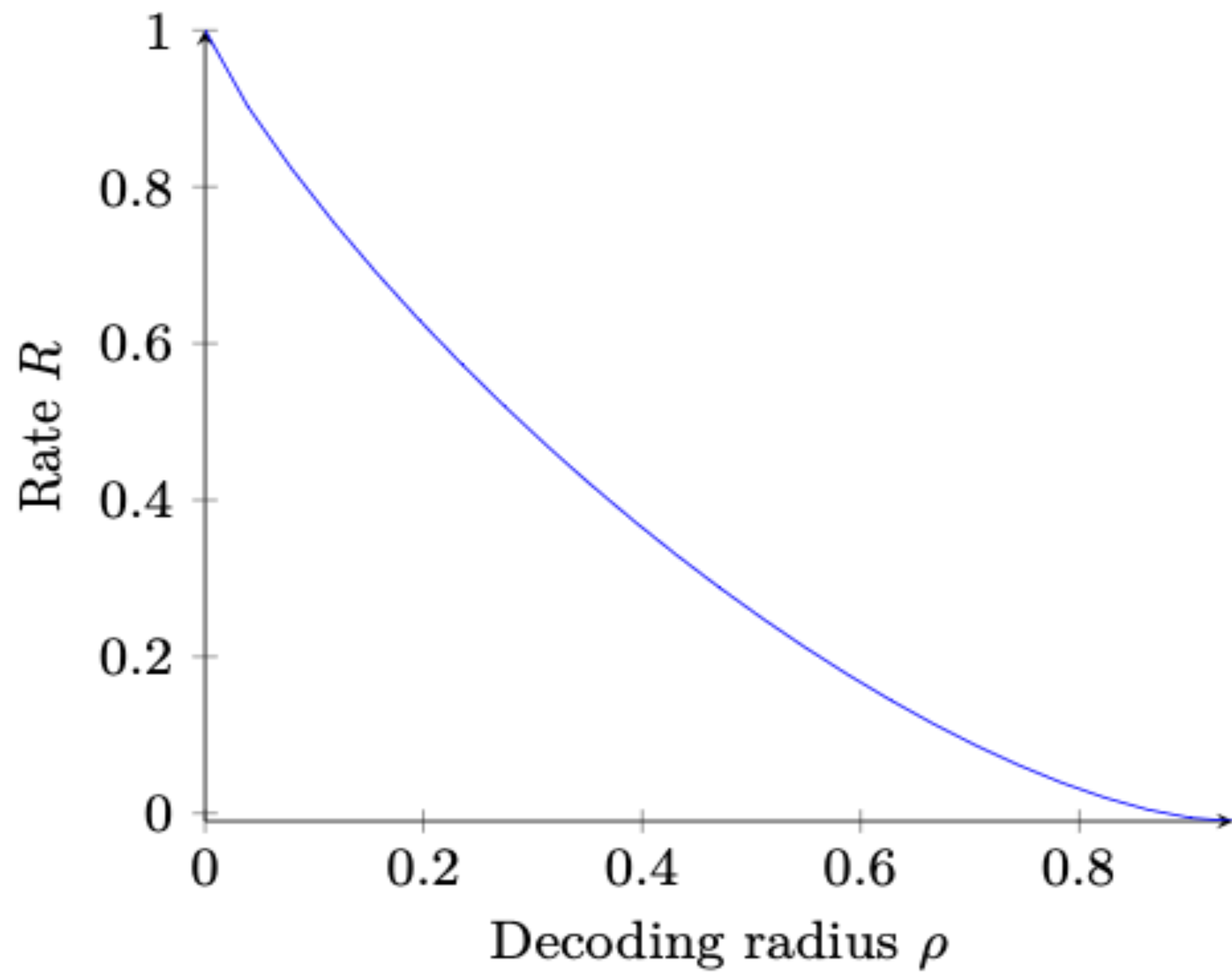
- Larger  $L, q$ : seems very difficult...



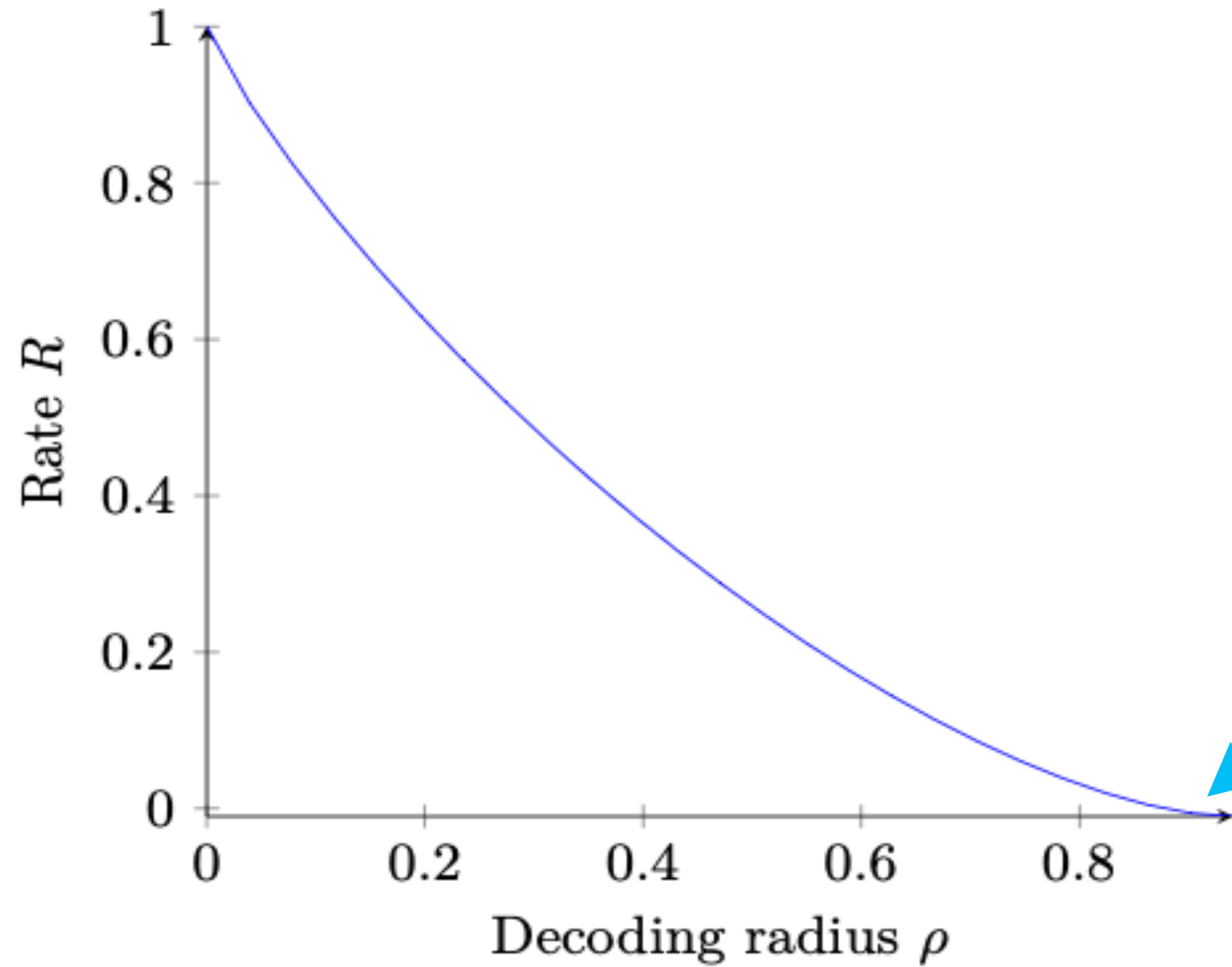


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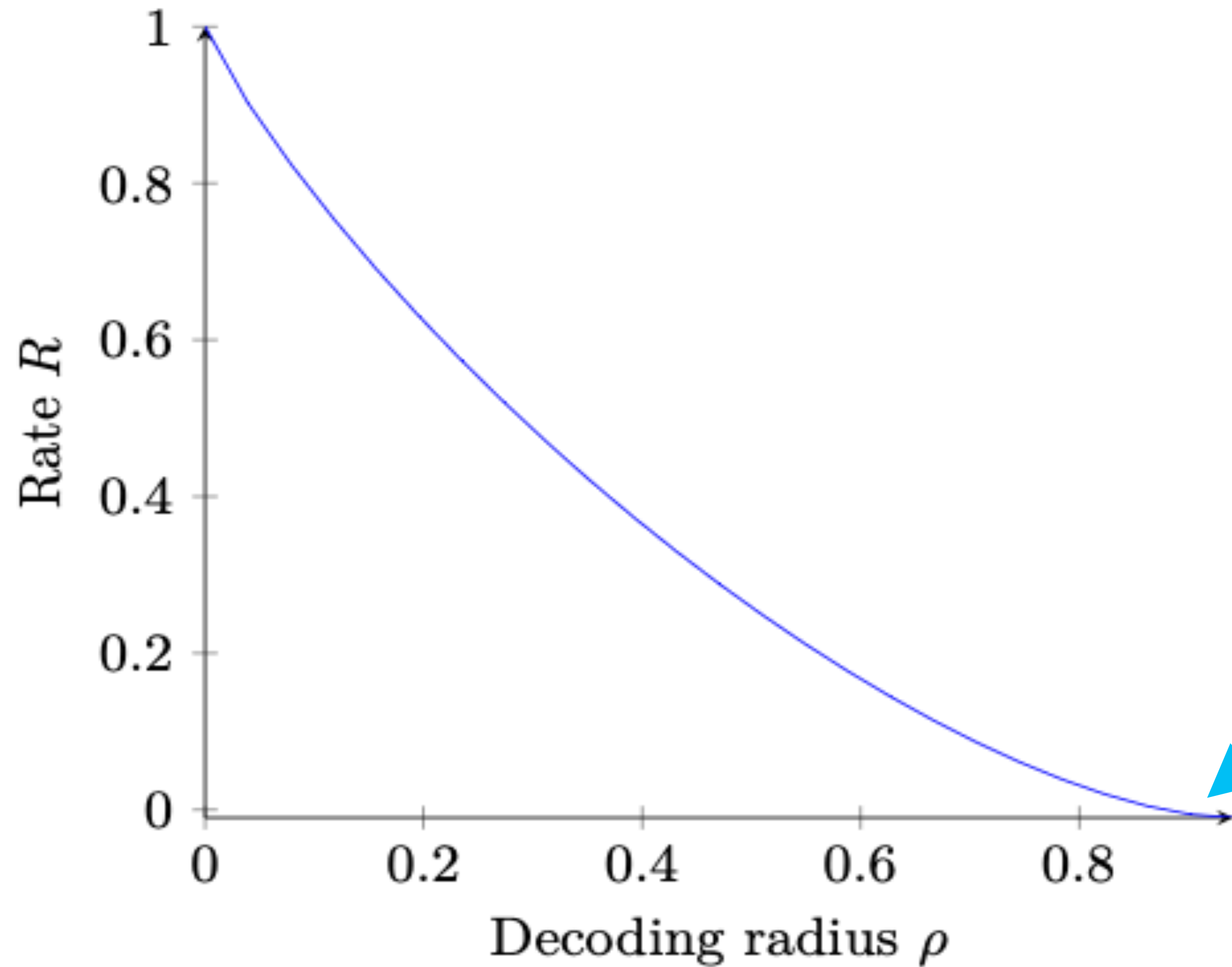


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Zero-rate threshold  
 $\rho_*(q, L)$

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$$\text{pl}(\text{■} \text{■} \text{■} \text{■} \text{■} \text{■} \text{■} \text{■} \text{■} \text{■}) = 4$$

- Blinovsky '05, '08 claims a proof.. but it's flawed

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Analogous result  
for list-recovery

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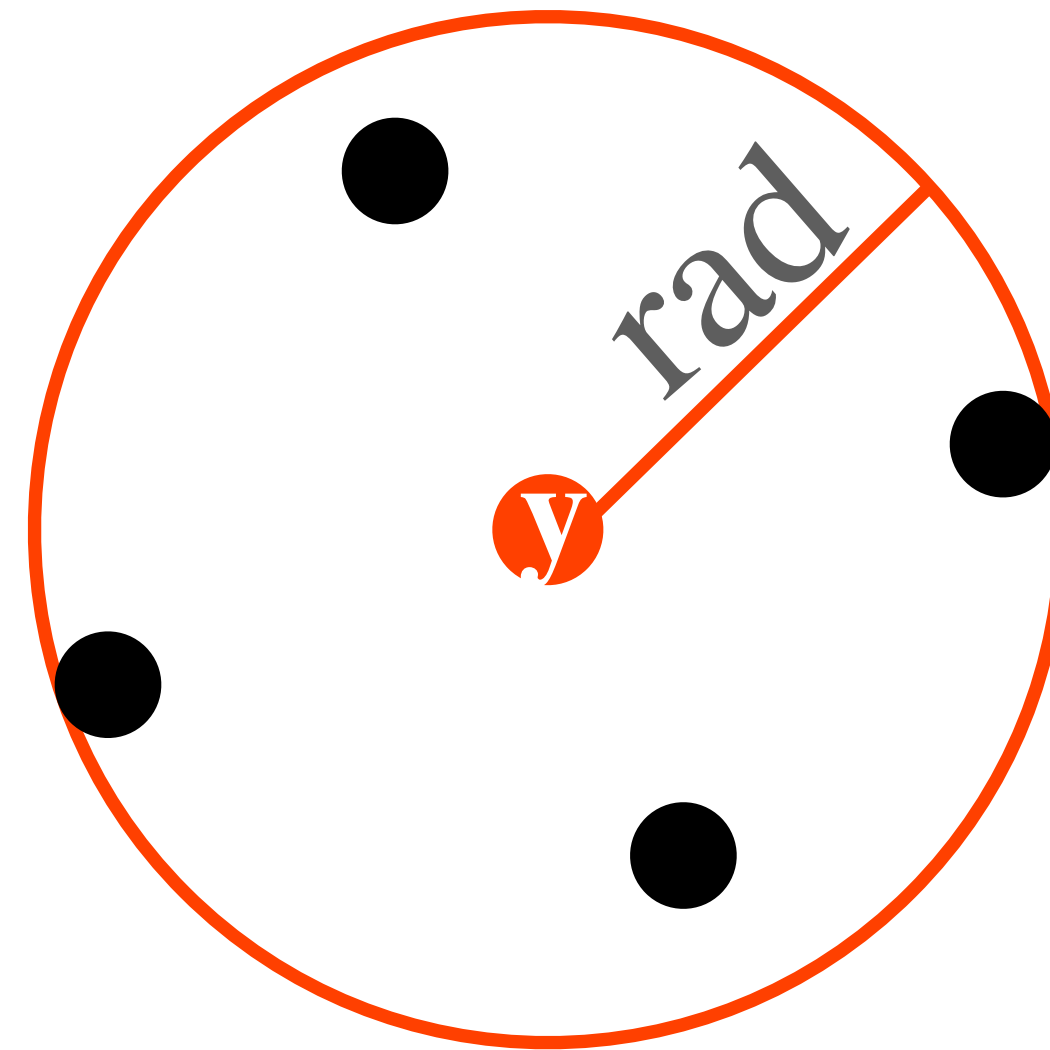
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# HOW TO BOUND A CODE

# RADIUS

- Given vectors  $\mathbf{x}_1, \dots, \mathbf{x}_L \in [q]^n$ , their *radius* is

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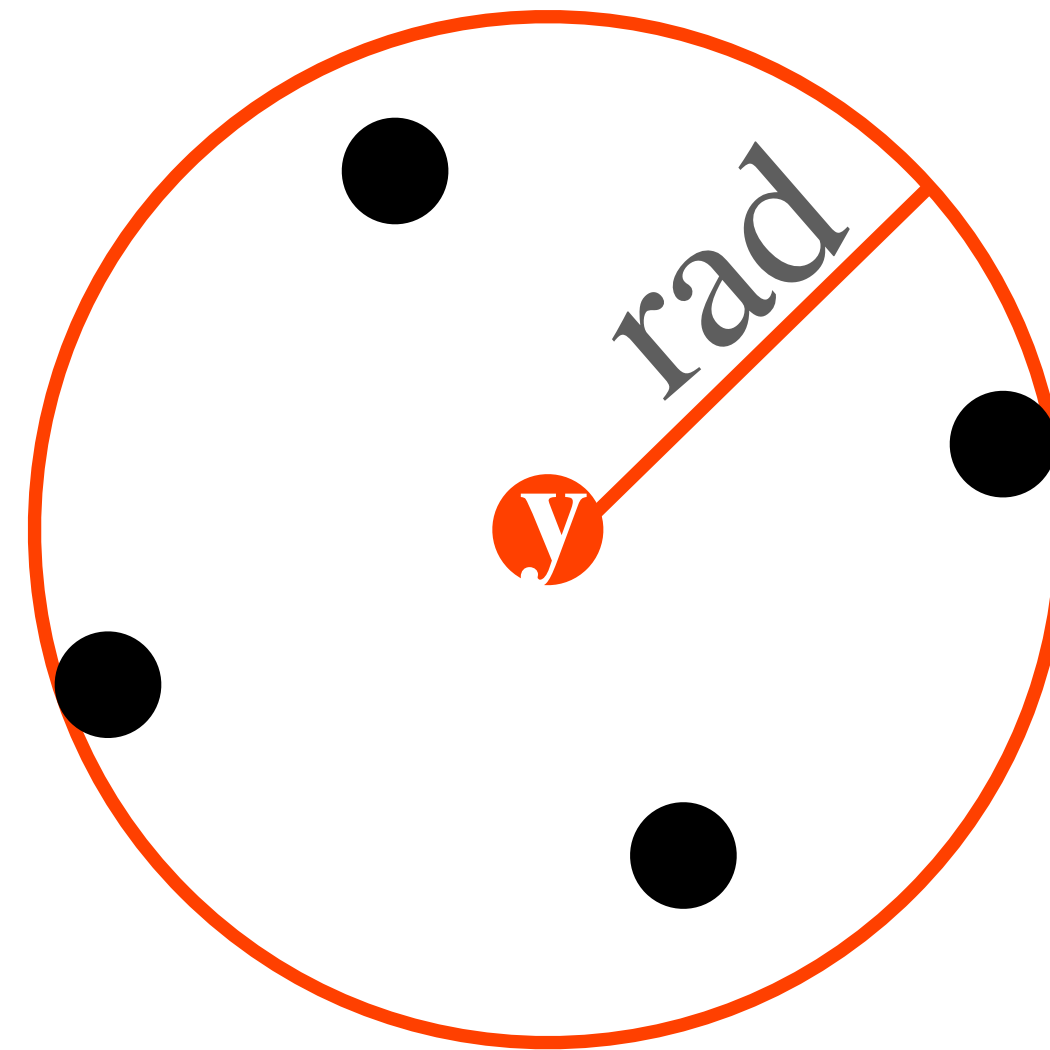


# RADIUS

Radius of smallest ball  
containing  $\{\mathbf{x}_1, \dots, \mathbf{x}_L\}$

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# AVERAGE-RADIUS

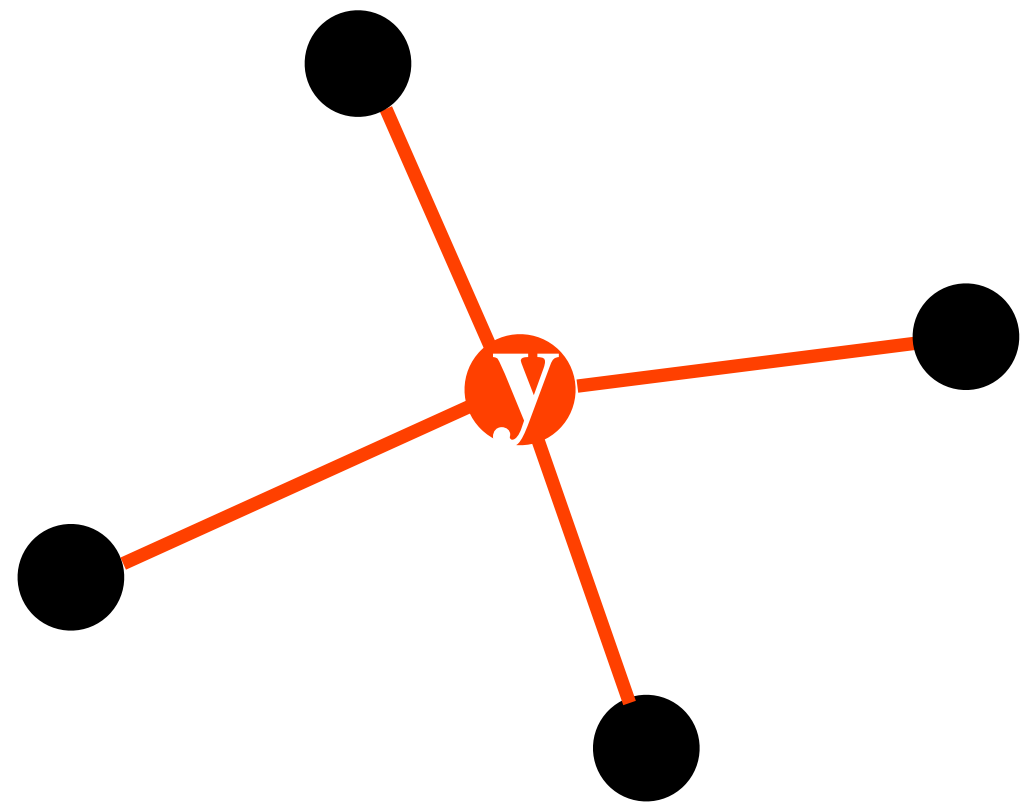
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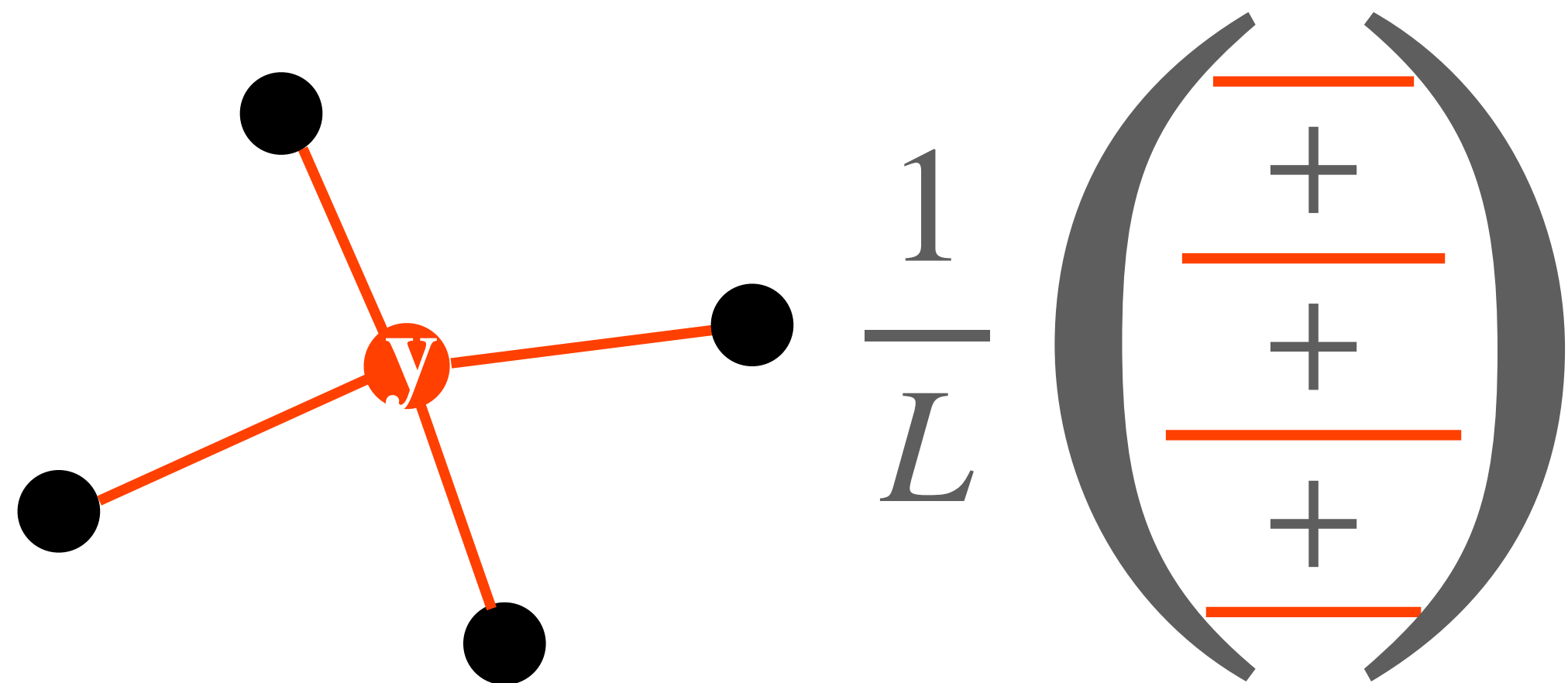
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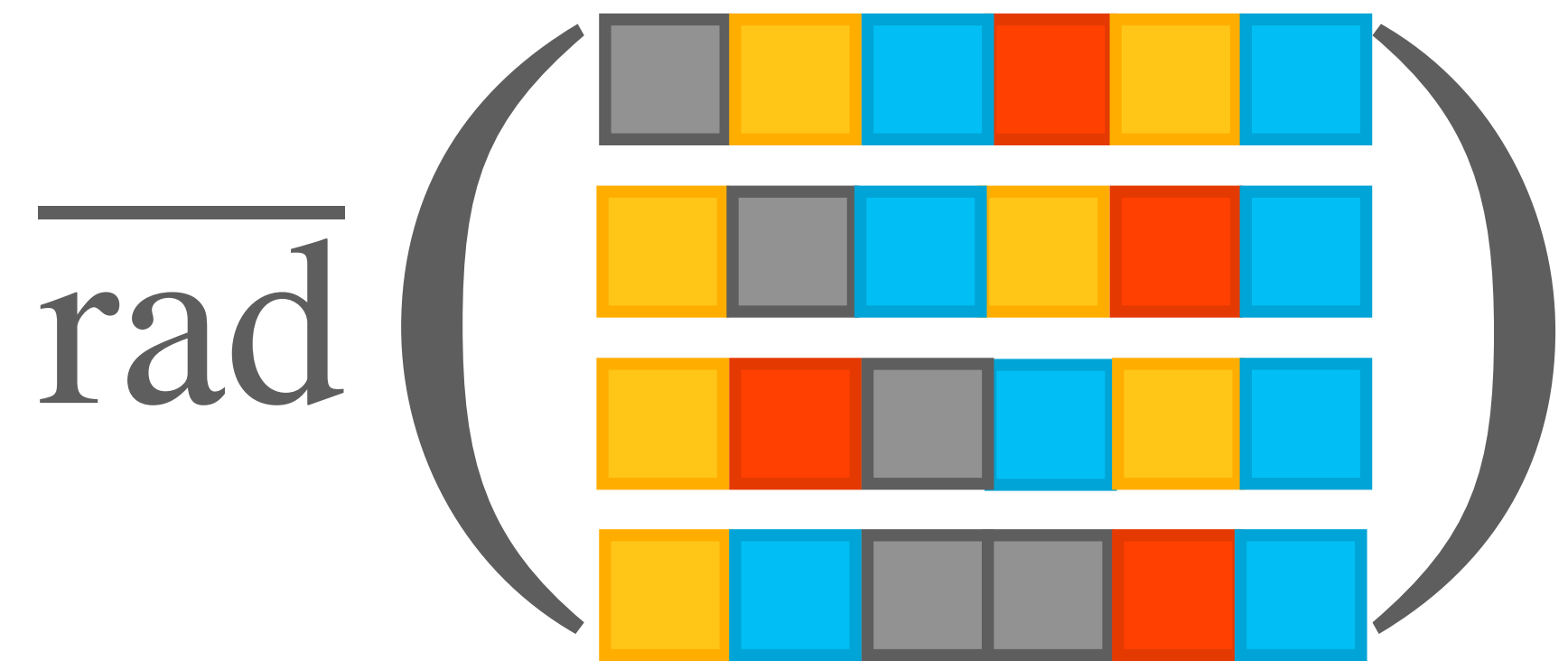
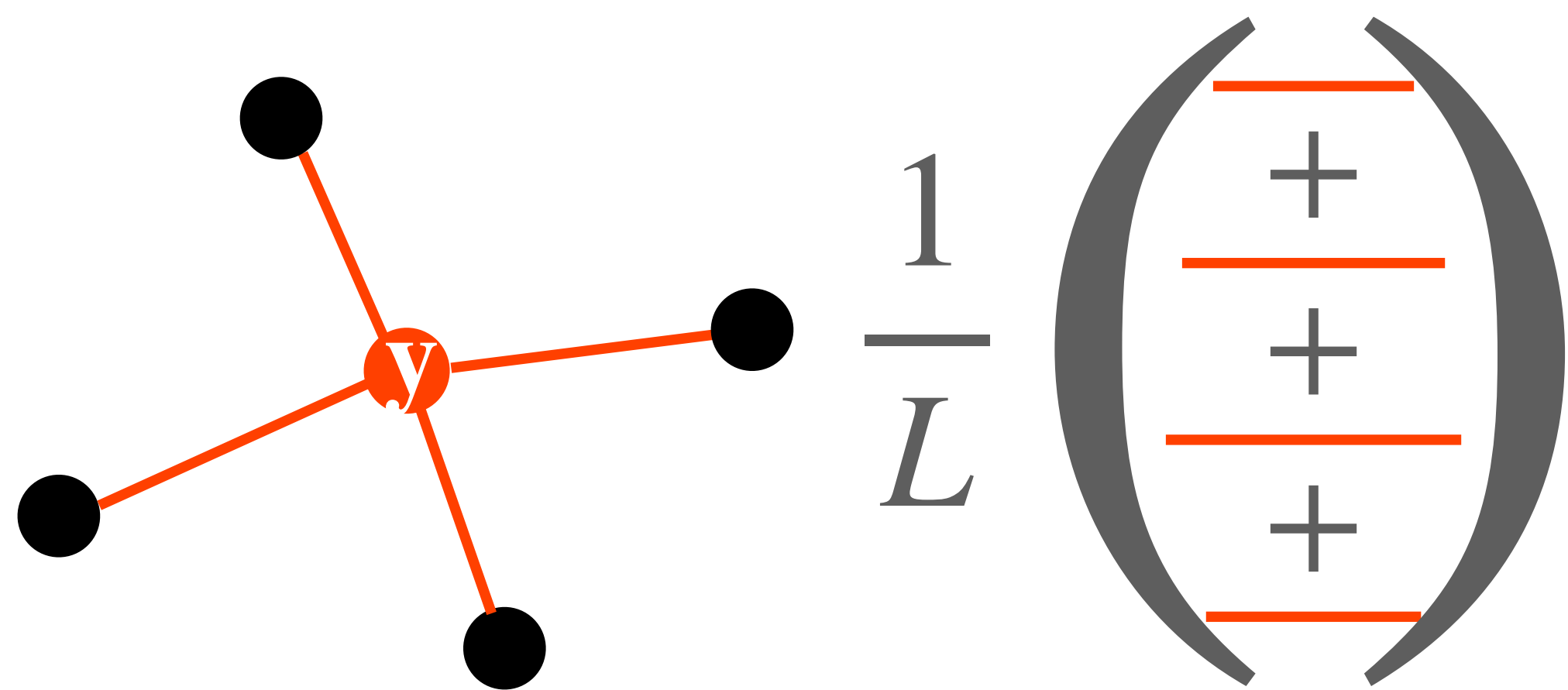
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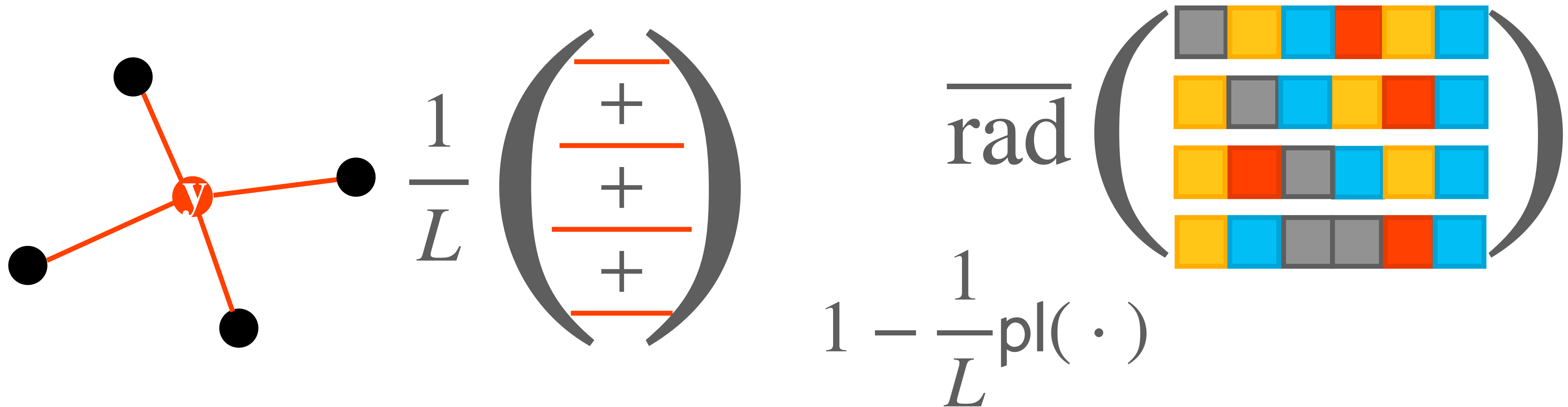
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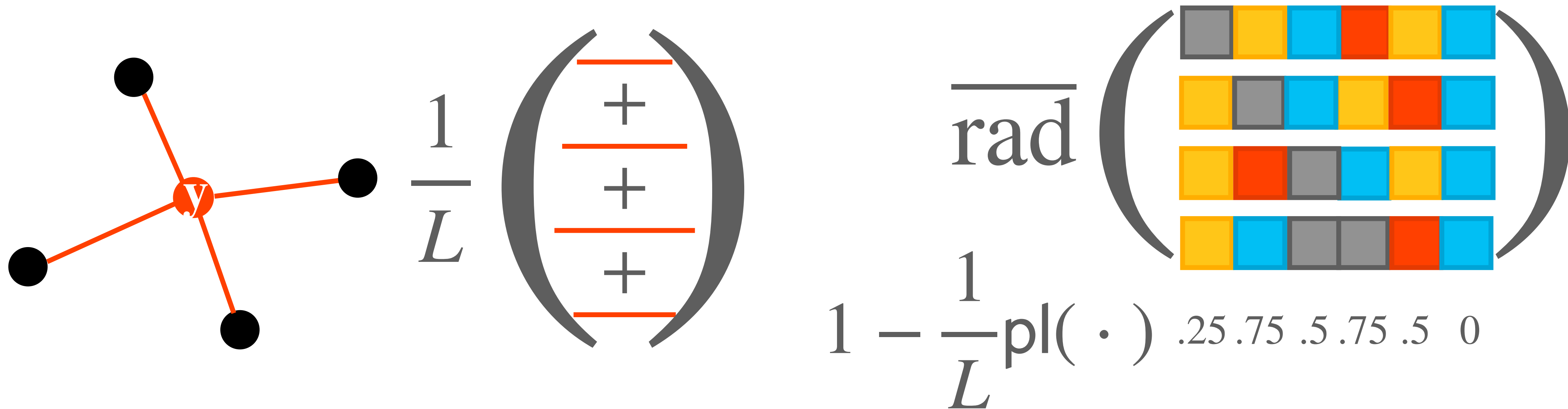
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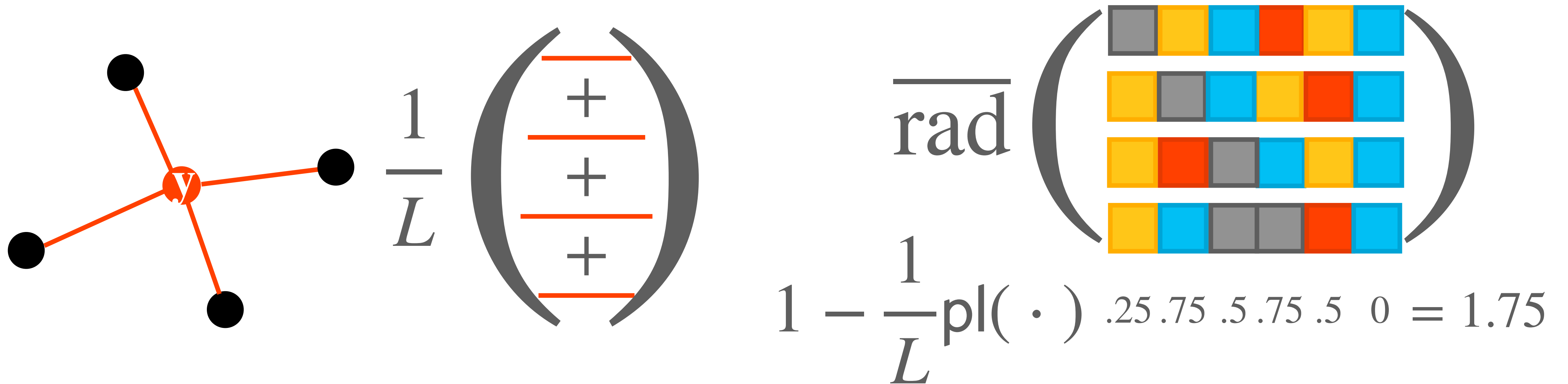
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Ramsey-theory...

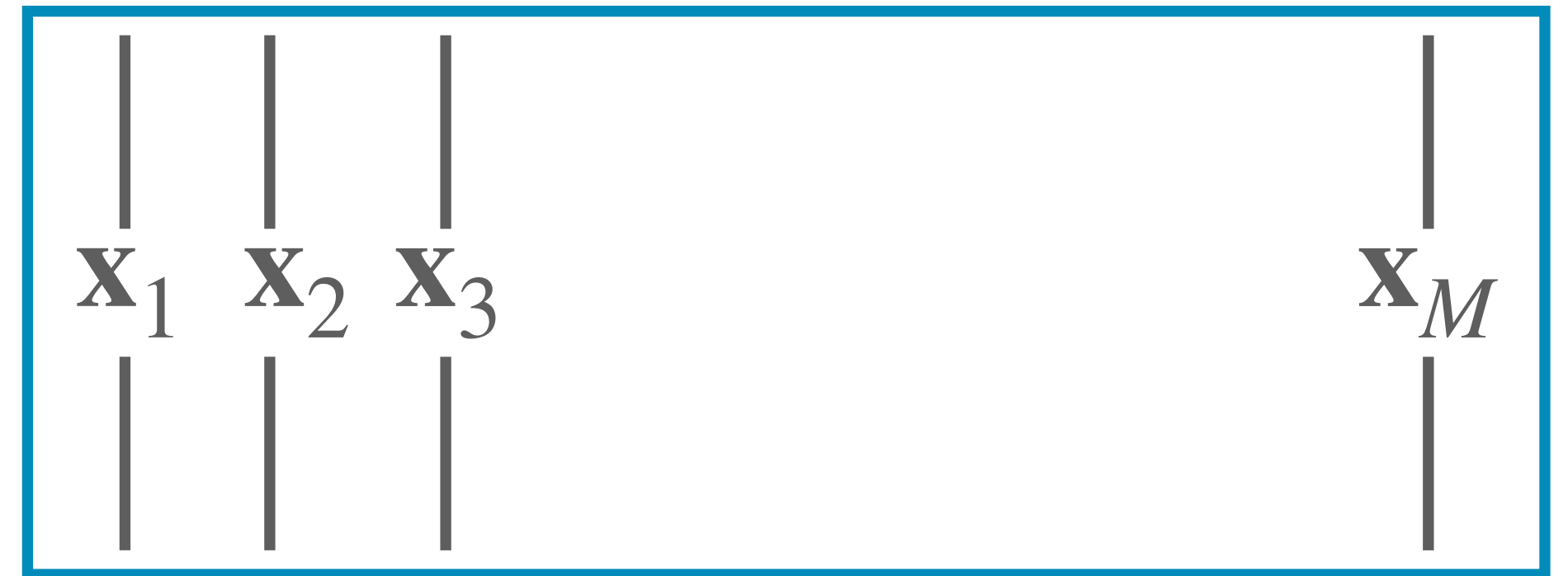
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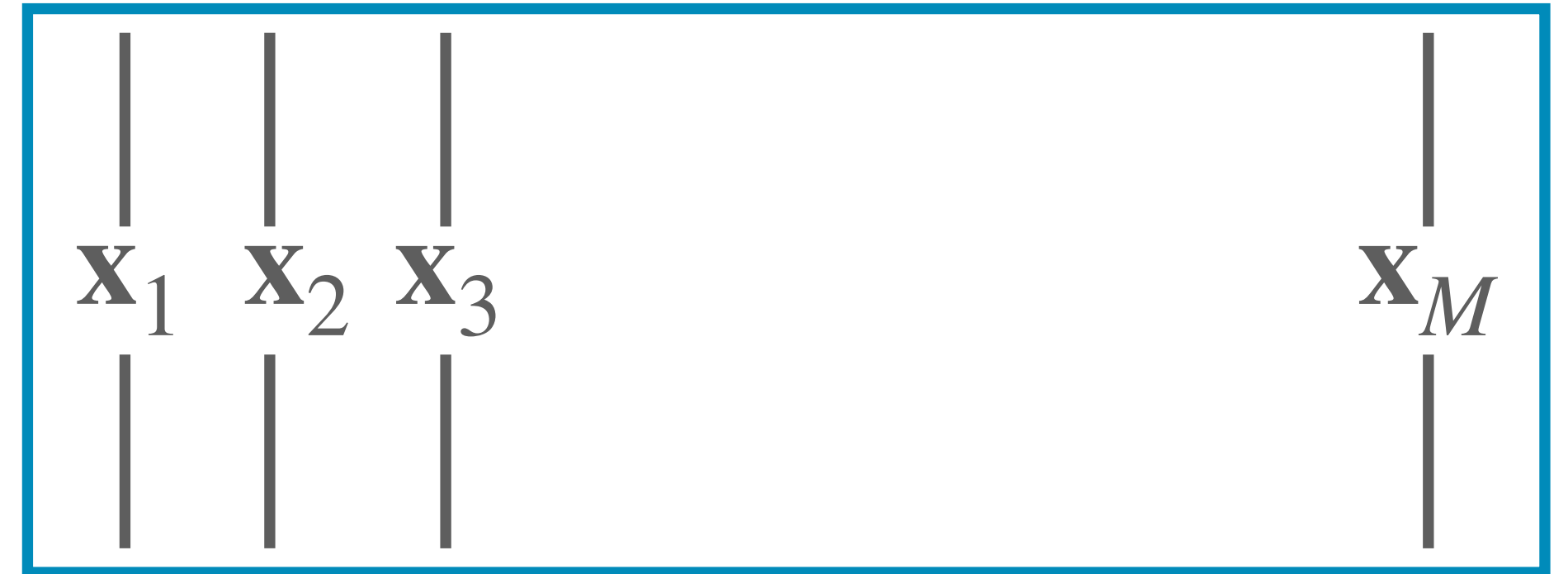
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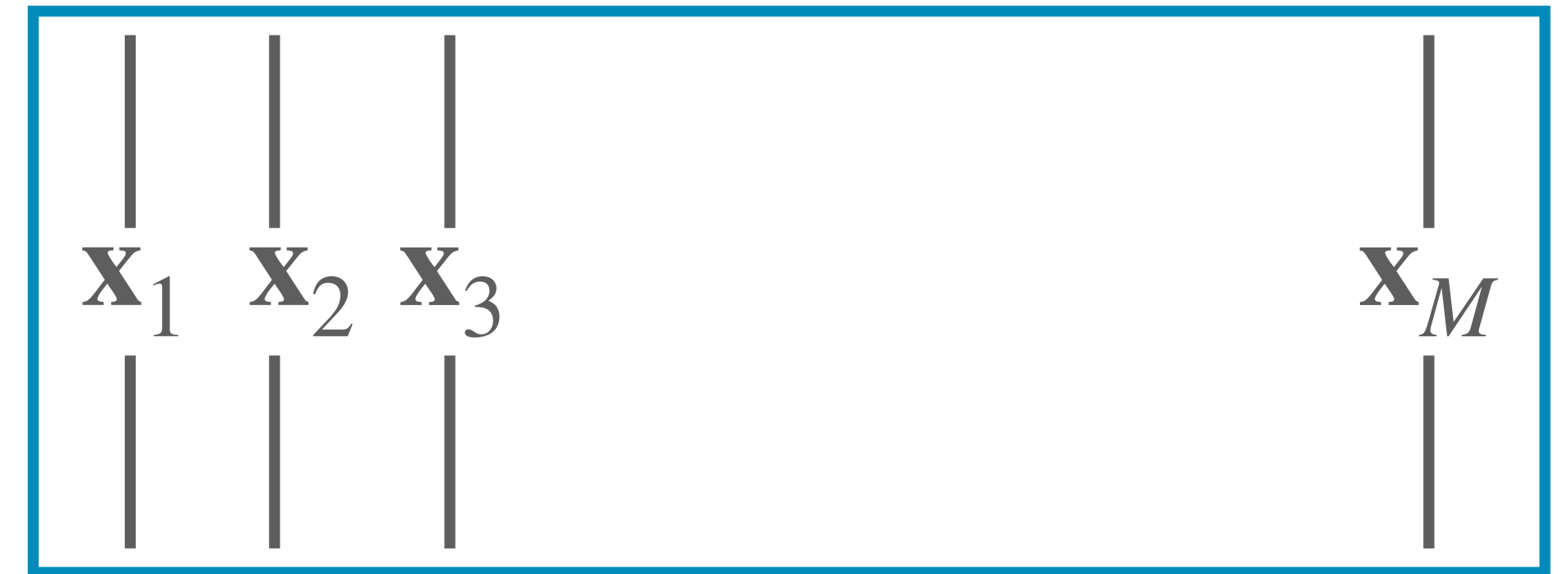


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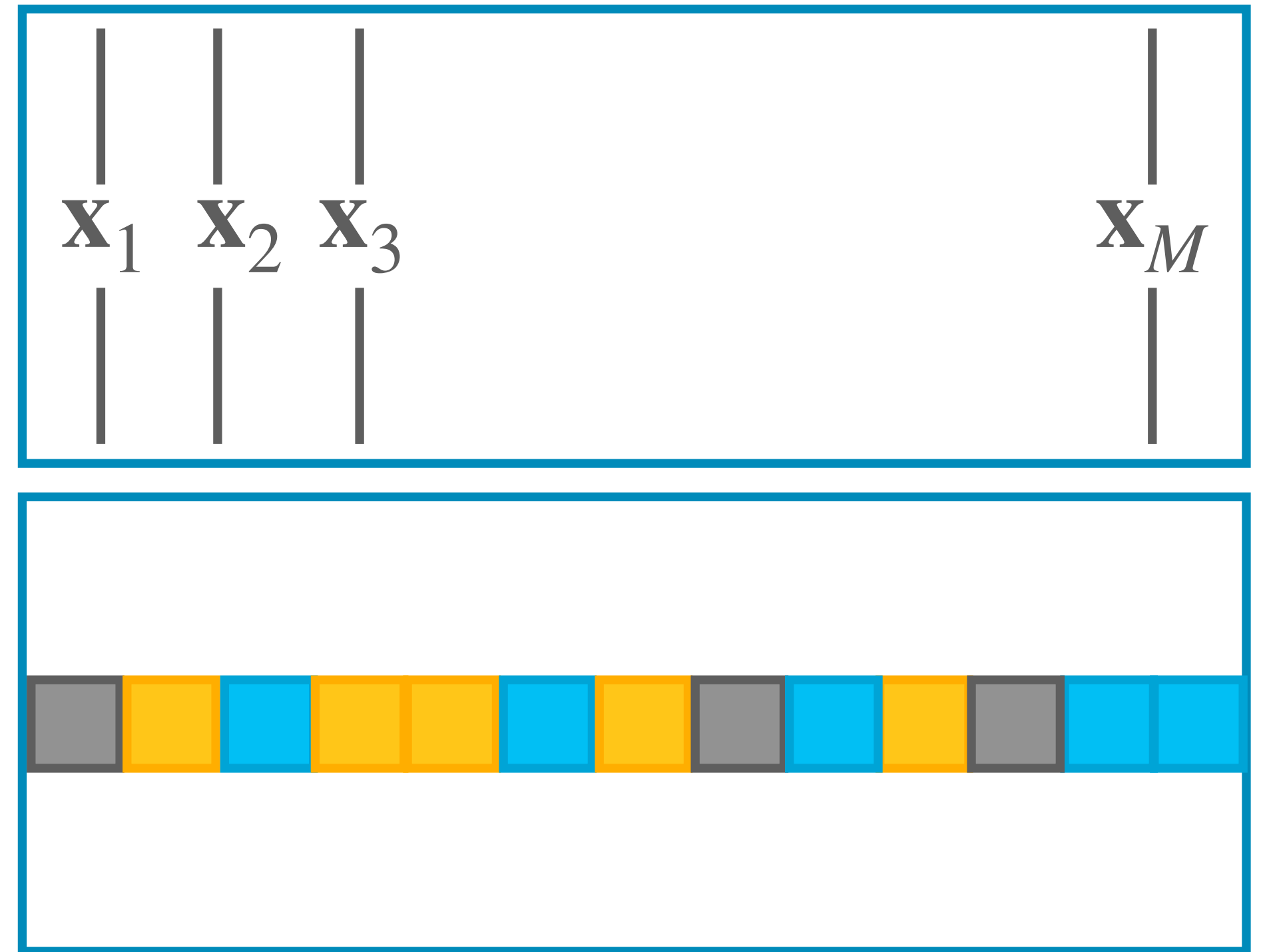
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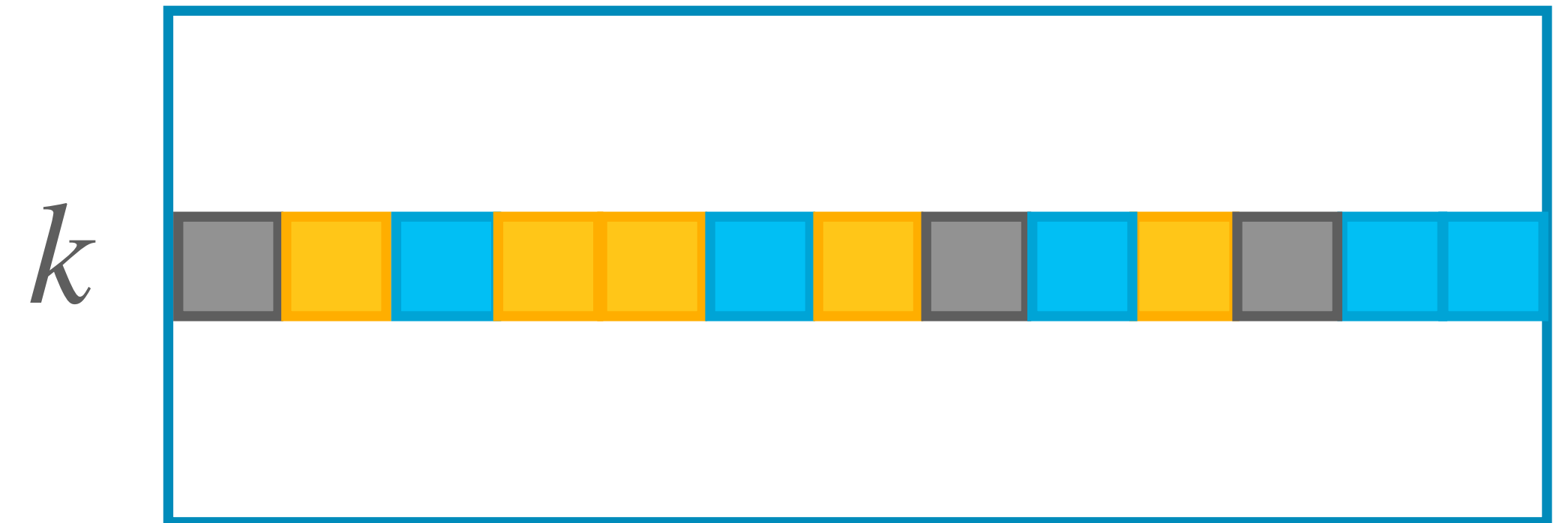


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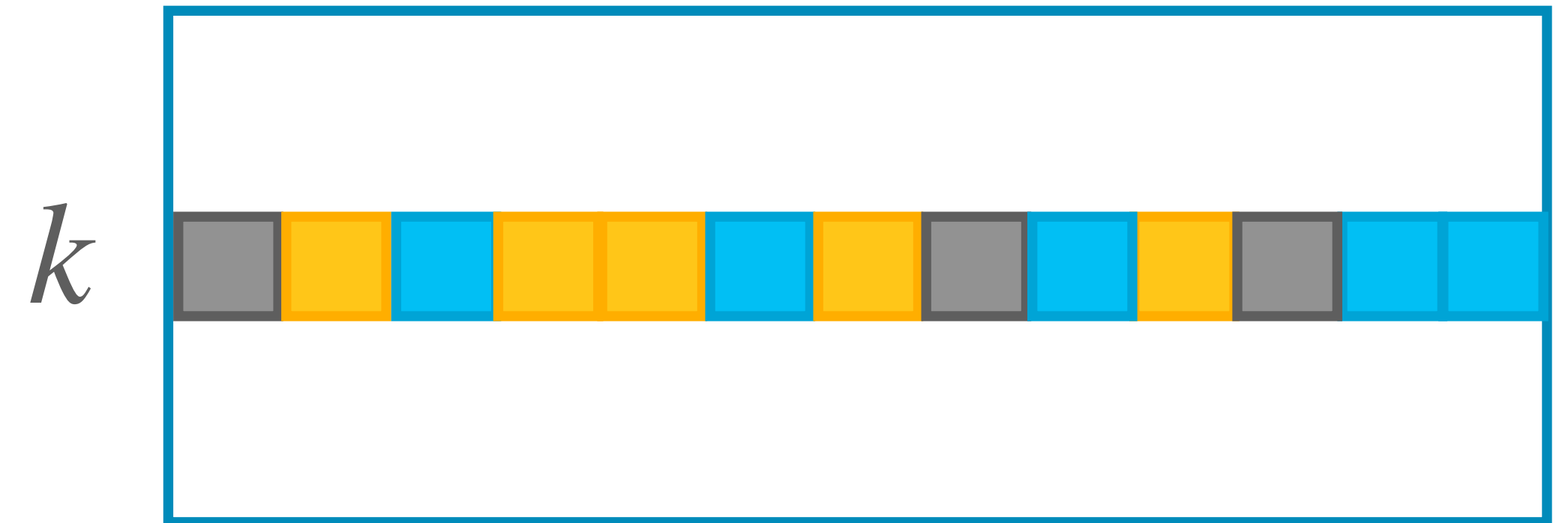


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$$P_k(\blacksquare) = 3/13$$

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**Contradiction!**

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**PROVING THAT INEQUALITY**

# INEQUALITY TO PROVE

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— Define function on prob. dists.  $P \in \Delta([q]) = \left\{ Q \in \mathbb{R}_{\geq 0}^q : \sum_i Q_i = 1 \right\}$

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# INEQUALITY TO PROVE

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- Want to show:  $f$  is **minimized on uniform distribution**

# SCHUR CONVEXITY



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- Well-studied class of functions minimized on uniform distribution:  
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$\mathbf{x}$	<table border="1"><tr><td>1/3</td><td>1/3</td><td>1/3</td></tr></table>	1/3	1/3	1/3	$\mathbf{y}$	<table border="1"><tr><td>1/5</td><td>1/2</td><td>3/10</td></tr></table>	1/5	1/2	3/10
1/3	1/3	1/3							
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$\mathbf{x}^\downarrow$	<table border="1"><tr><td>1/3</td><td>1/3</td><td>1/3</td></tr></table> 1/3   2/3   1	1/3	1/3	1/3	$\mathbf{y}^\downarrow$	<table border="1"><tr><td>1/2</td><td>3/10</td><td>1/5</td></tr></table> 1/2   8/10   1	1/2	3/10	1/5
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	1/3	2/3	1			1/2	8/10	1

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$(1/q, \dots, 1/q)$  is majorized by every other probability vector

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— Can verify this holds for

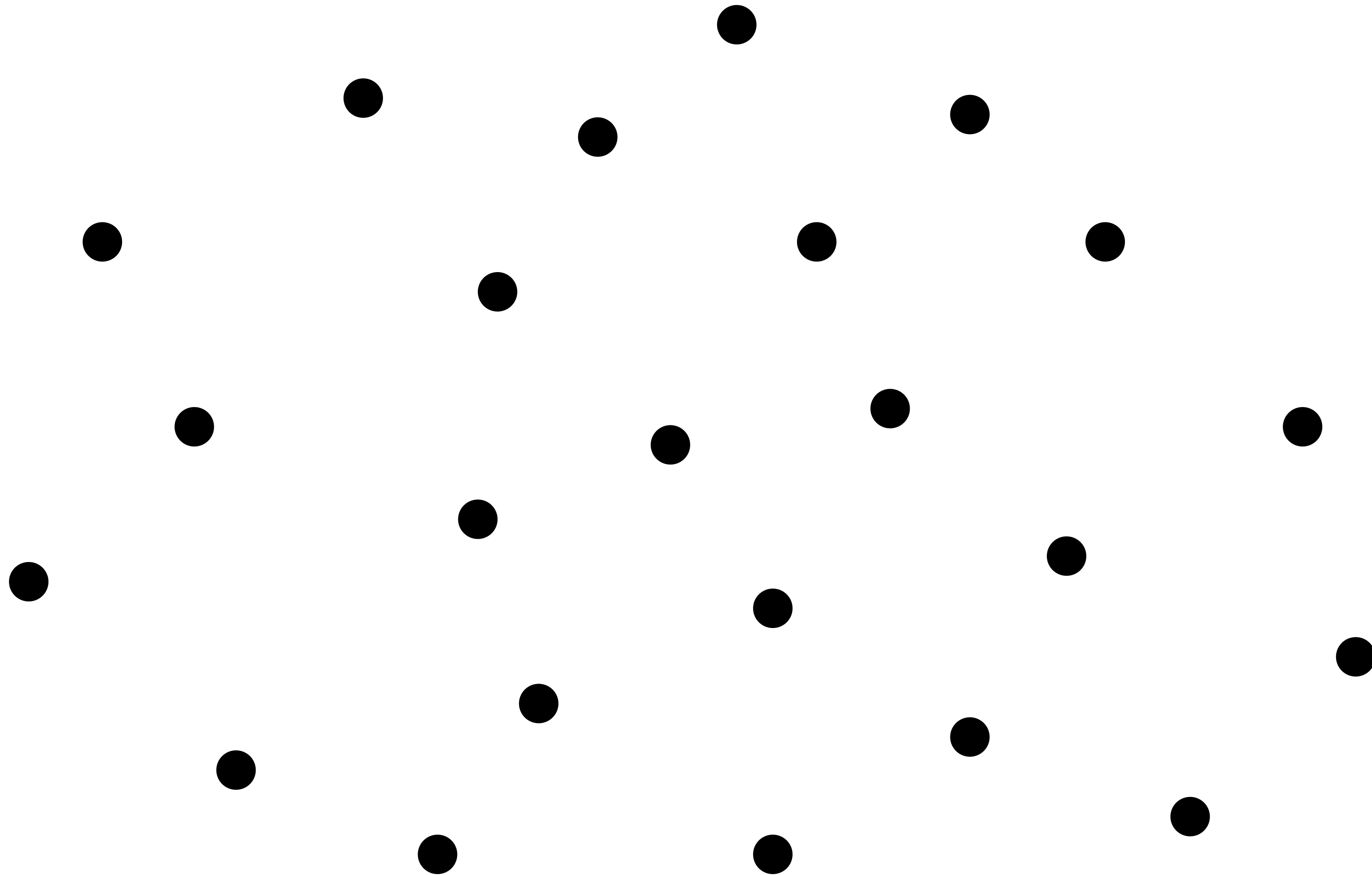
$$f_{q,L}(P) = \mathbb{E}_{(X_1, \dots, X_L) \sim P^{\otimes L}} [\text{pl}(X_1, \dots, X_L)] = \sum_{(a_0, \dots, a_{q-1})} \binom{L}{a_0, \dots, a_{q-1}} \max\{a_0, \dots, a_{q-1}\} \prod_{i \in [q]} P_i^{a_i}$$

where sum runs over all  $(a_0, \dots, a_{q-1}) \in \mathbb{N}_{\geq 0}^q$  s.t.  $\sum_i a_i = L$

**ELIAS-BASSALYGO-LIKE BOUND**

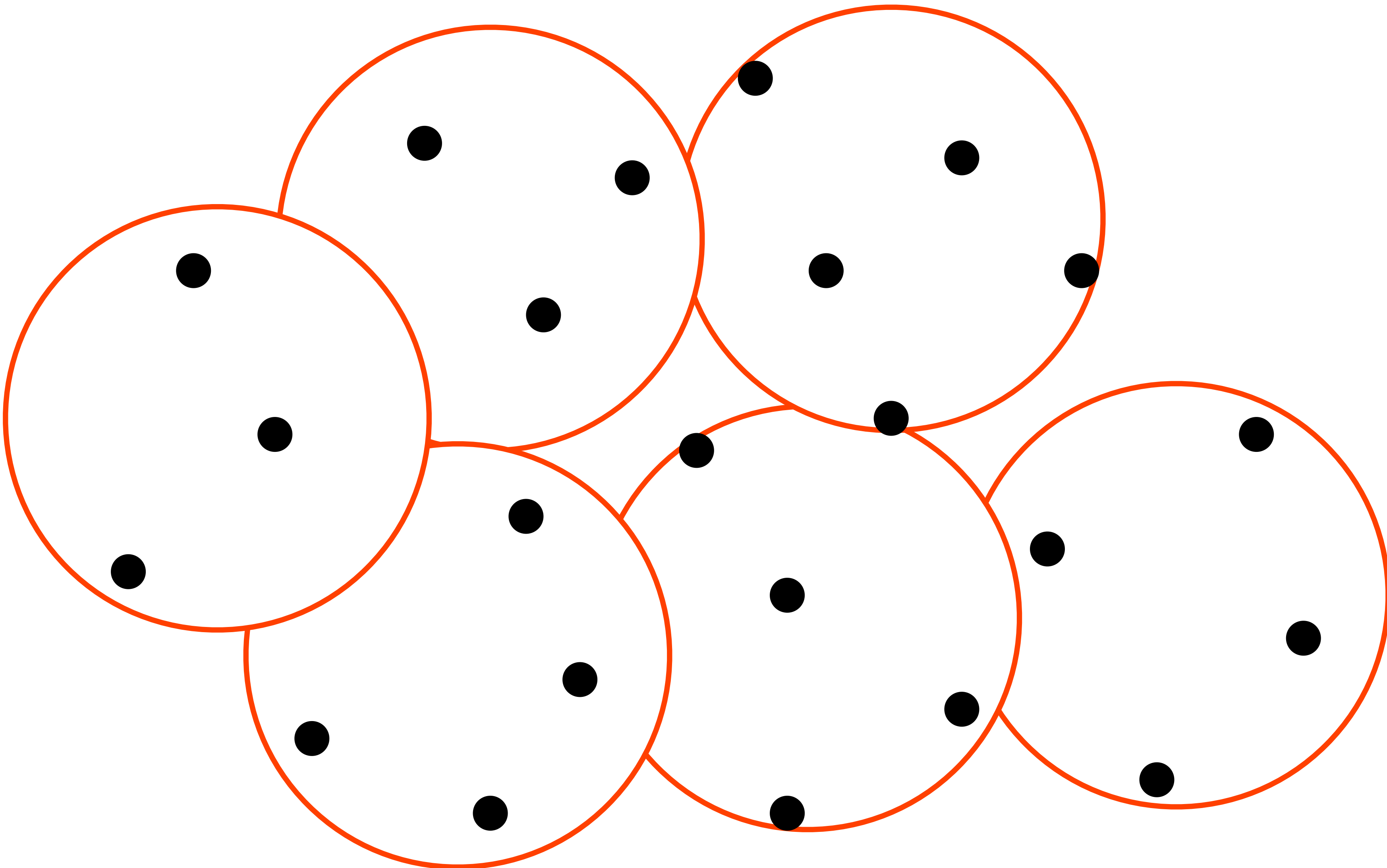
# RATE-DECODING RADIUS TRADEOFFS

[Bassalygo'65,Elias]



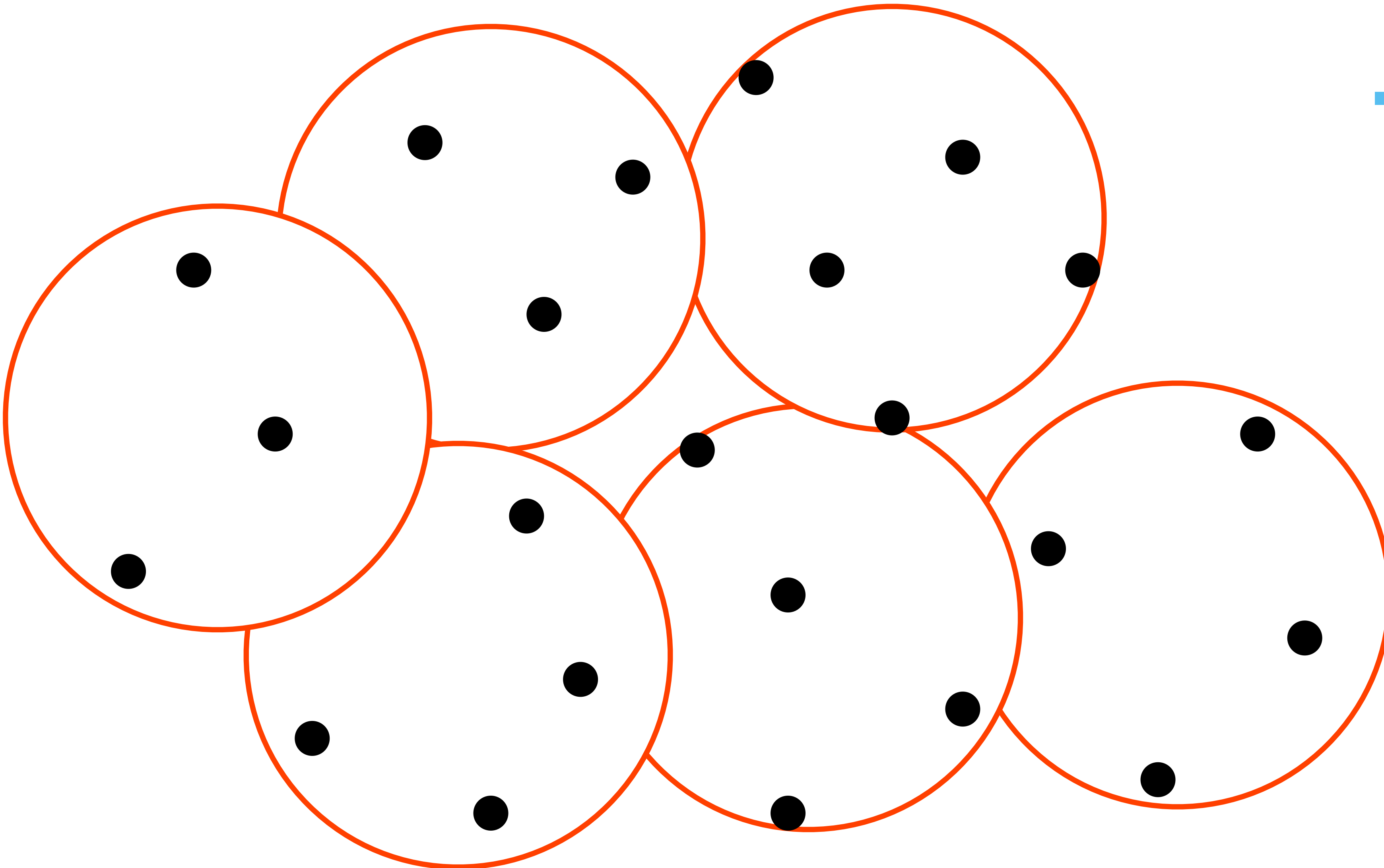
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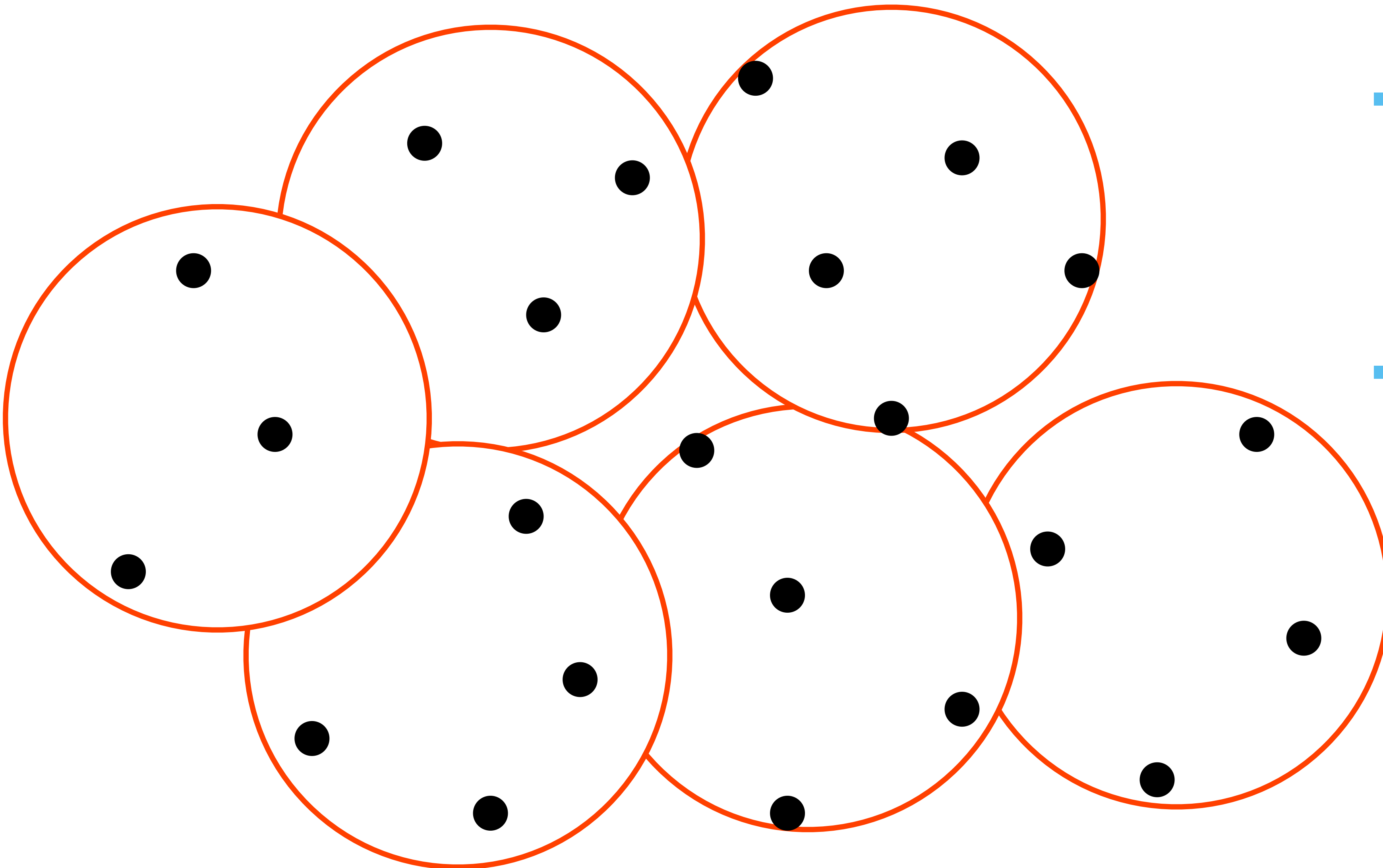
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- Use Plotkin bound on each Hamming ball subcode

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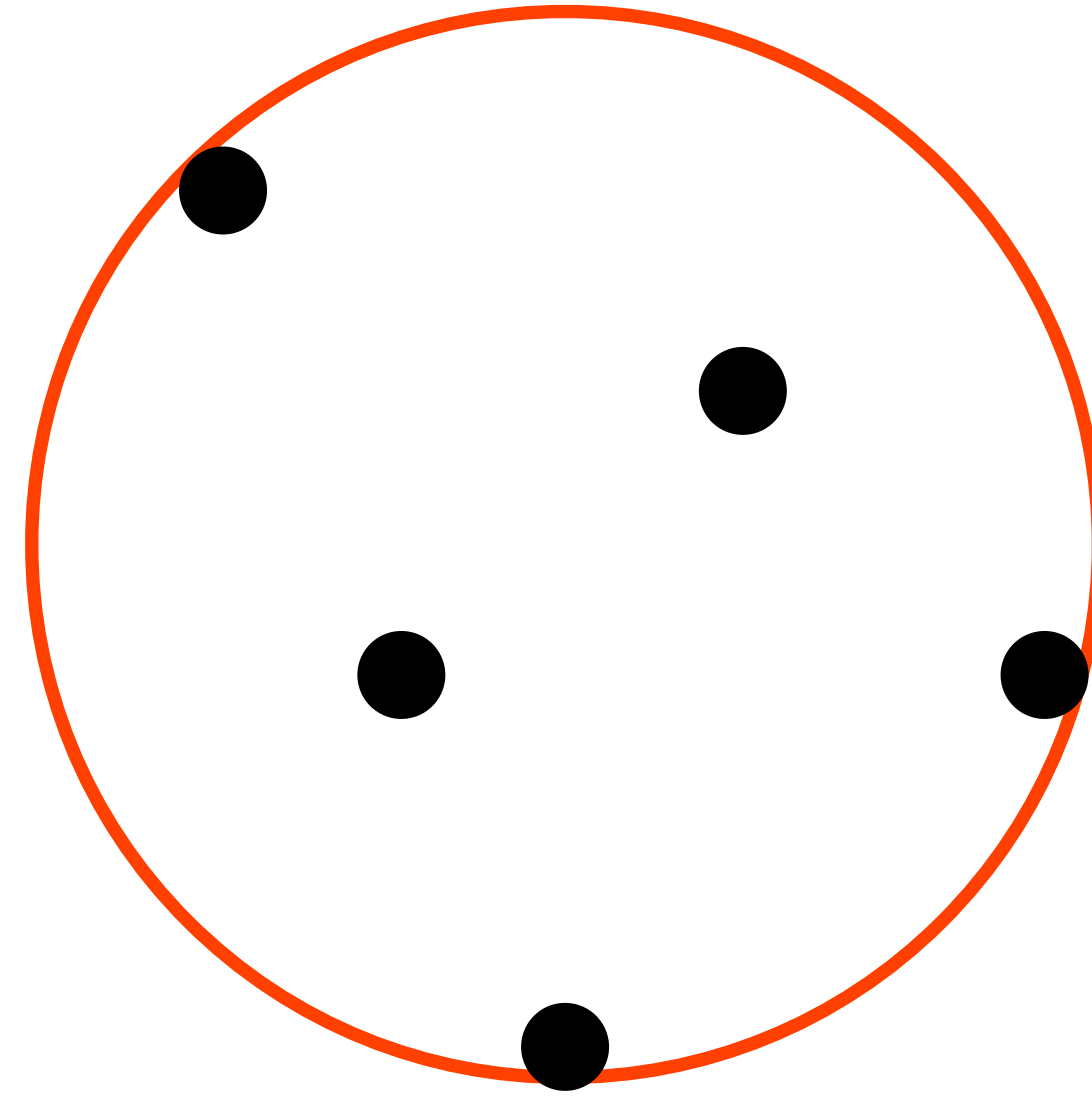


- Use Plotkin bound on each Hamming ball subcode
- Use fact that you *don't need too many* Hamming balls to cover  $[q]^n$



# NEED TO USE WEIGHT CONSTRAINT!

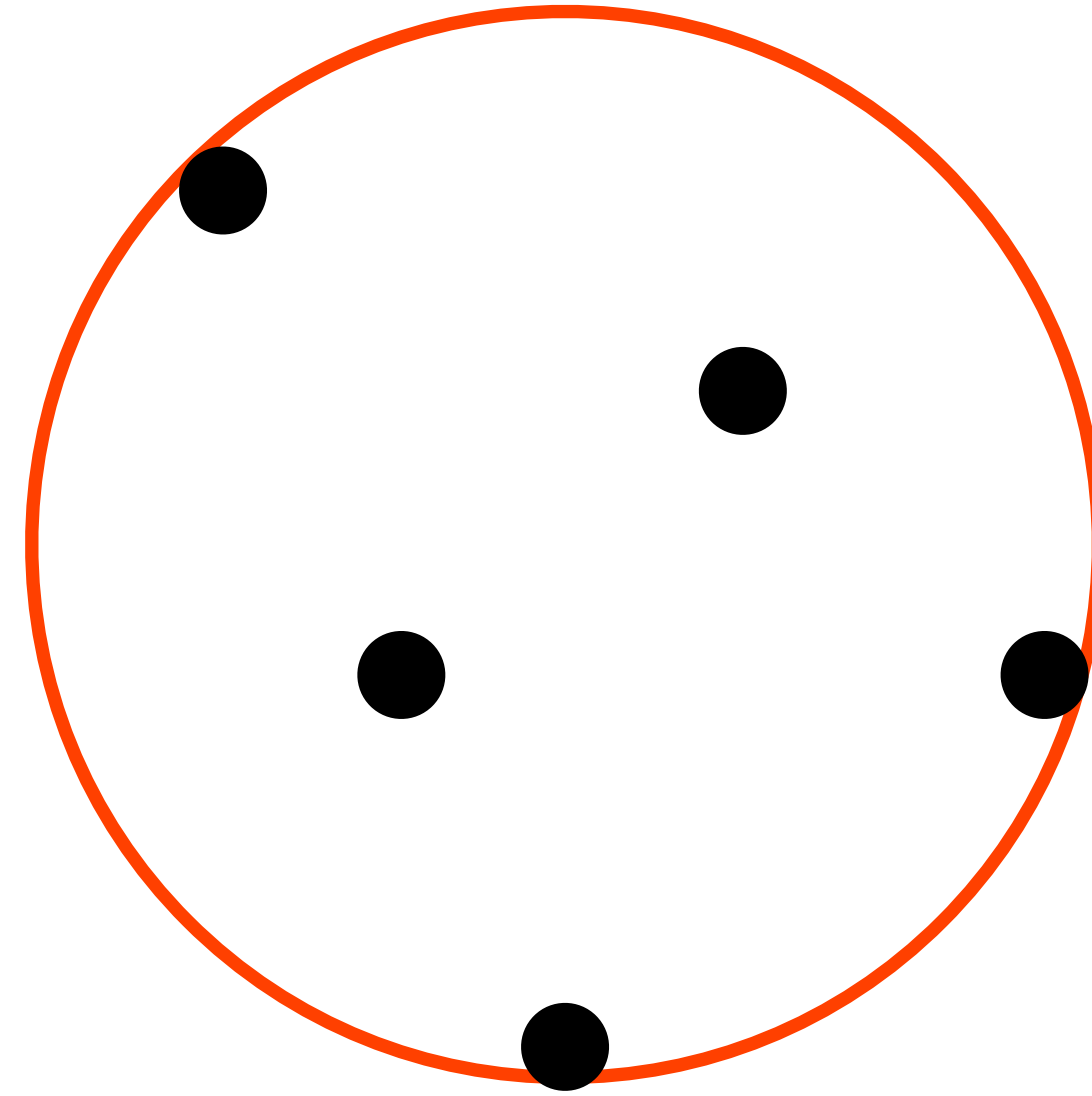
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is not only list-decodable, but is code **with bounded weight**

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**Want: Plotkin bound for list-decodable  
 $\mathcal{C}$  with weight constraint**

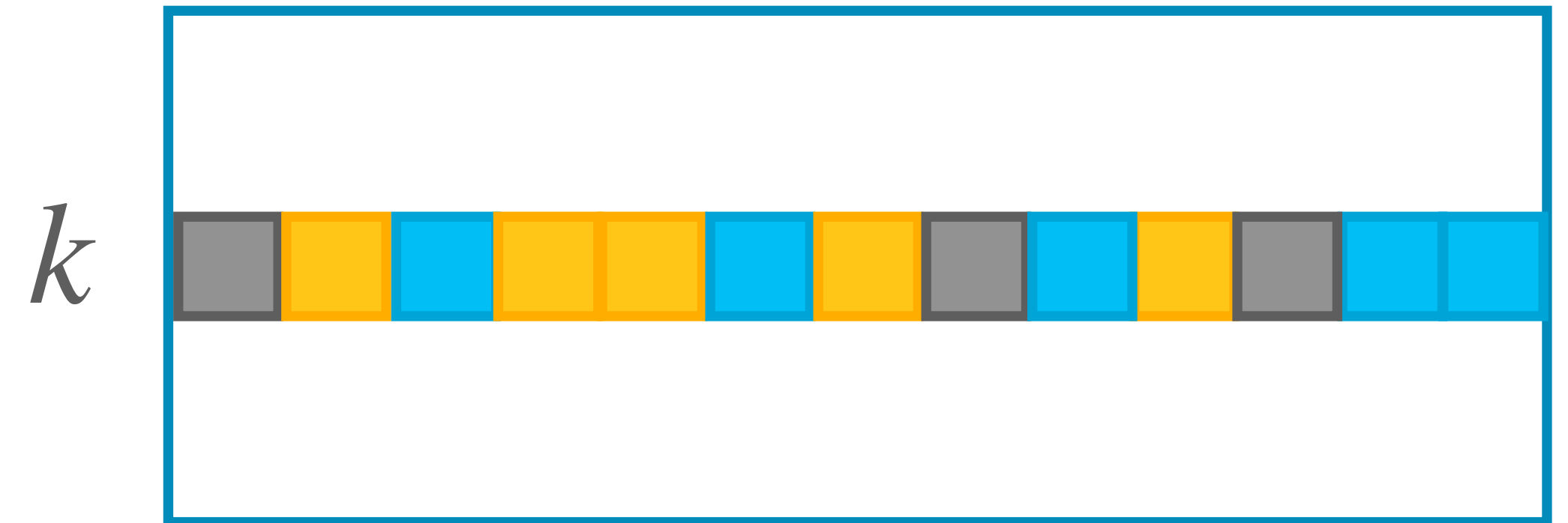
$$\text{wt}(\mathbf{x}) \leq w \quad \forall \mathbf{x} \in \mathcal{C}$$

# LET'S RETURN HERE

$$\Phi = \frac{1}{M^L} \sum_{(\mathbf{x}_1, \dots, \mathbf{x}_L) \in \mathcal{C}^L} \overline{\text{rad}}(\mathbf{x}_1, \dots, \mathbf{x}_L)$$

$$= \dots = \sum_{k=1}^n \sum_{(y_1, \dots, y_L) \in [q]^L} \left( \prod_{j=1}^L P_k(y_j) \right) \left( 1 - \frac{1}{L} \text{pl}(y_1, \dots, y_L) \right)$$

$$P_k(y) = \frac{1}{M} \sum_{\mathbf{x} \in \mathcal{C}} 1_{\{x_k = y\}}$$



$$P_k(\blacksquare) = 3/13$$

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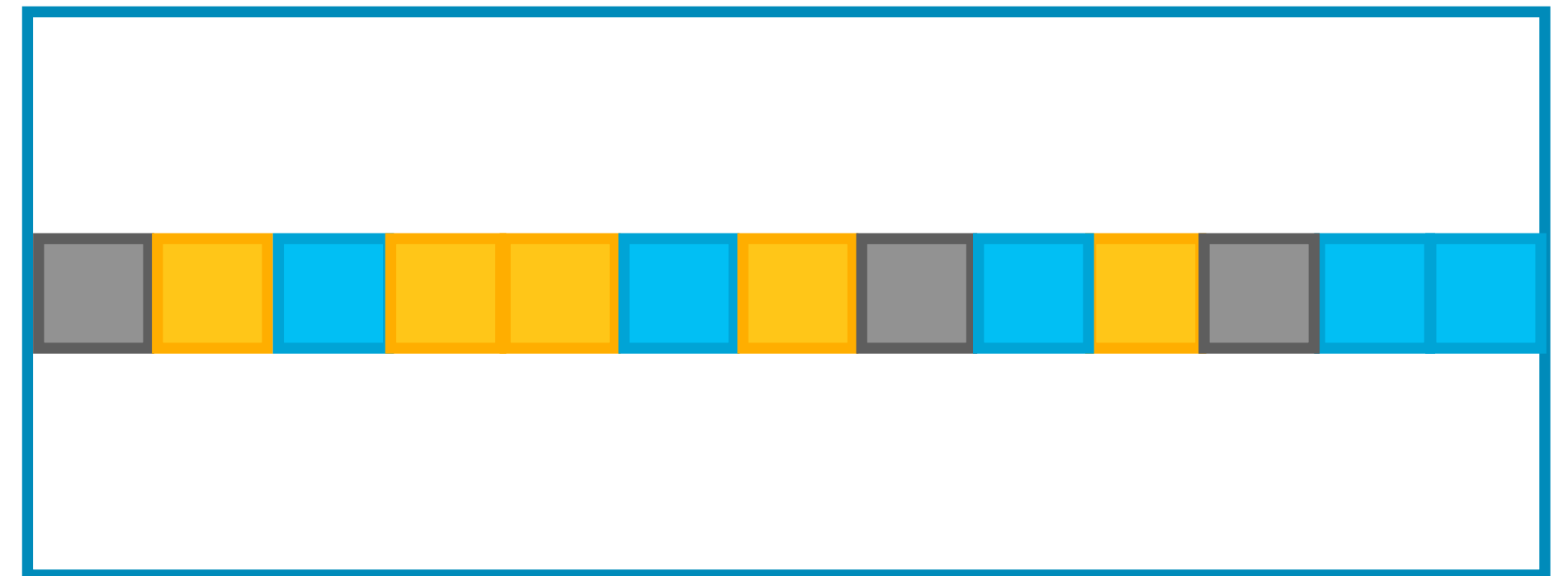
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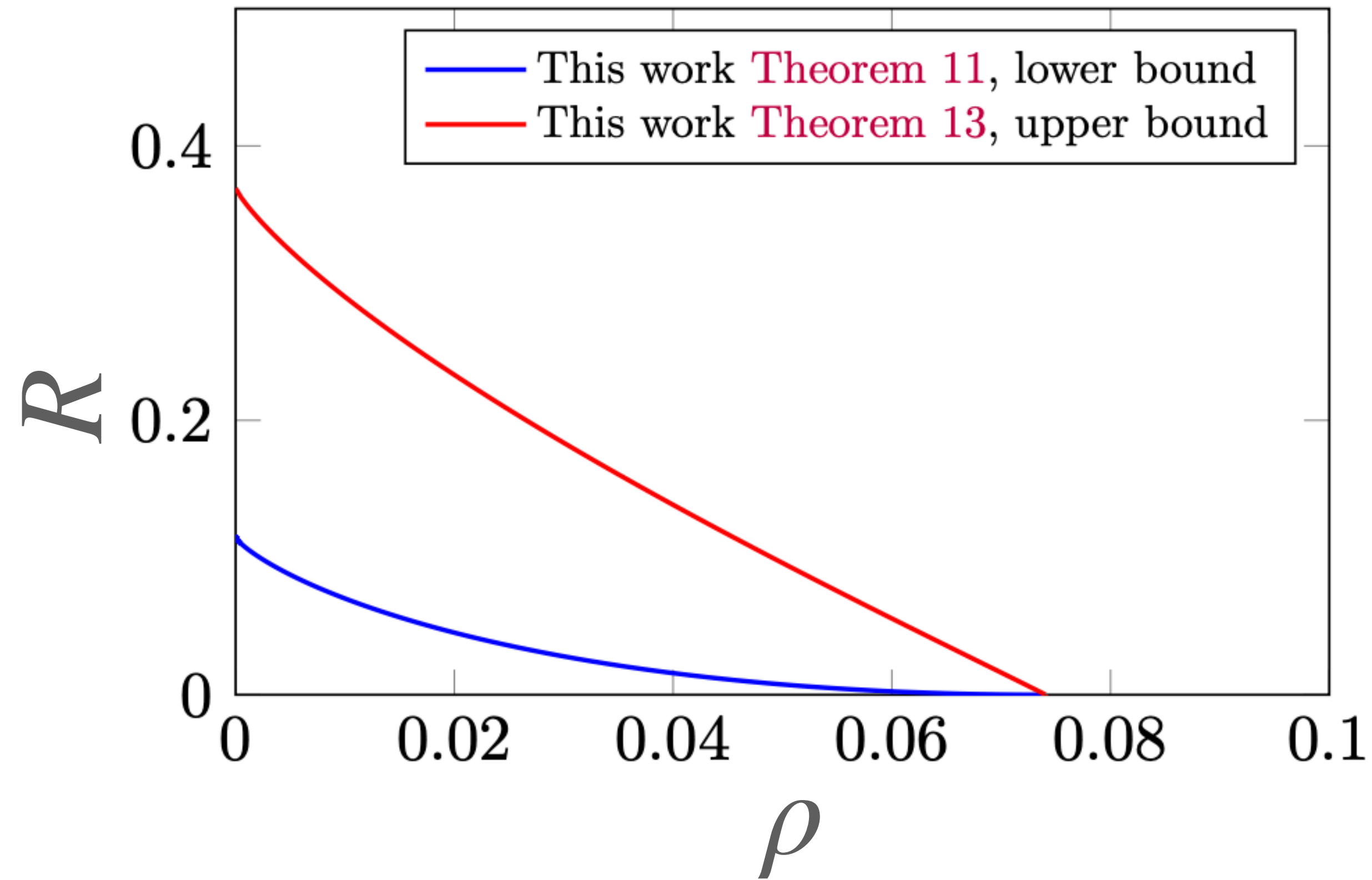
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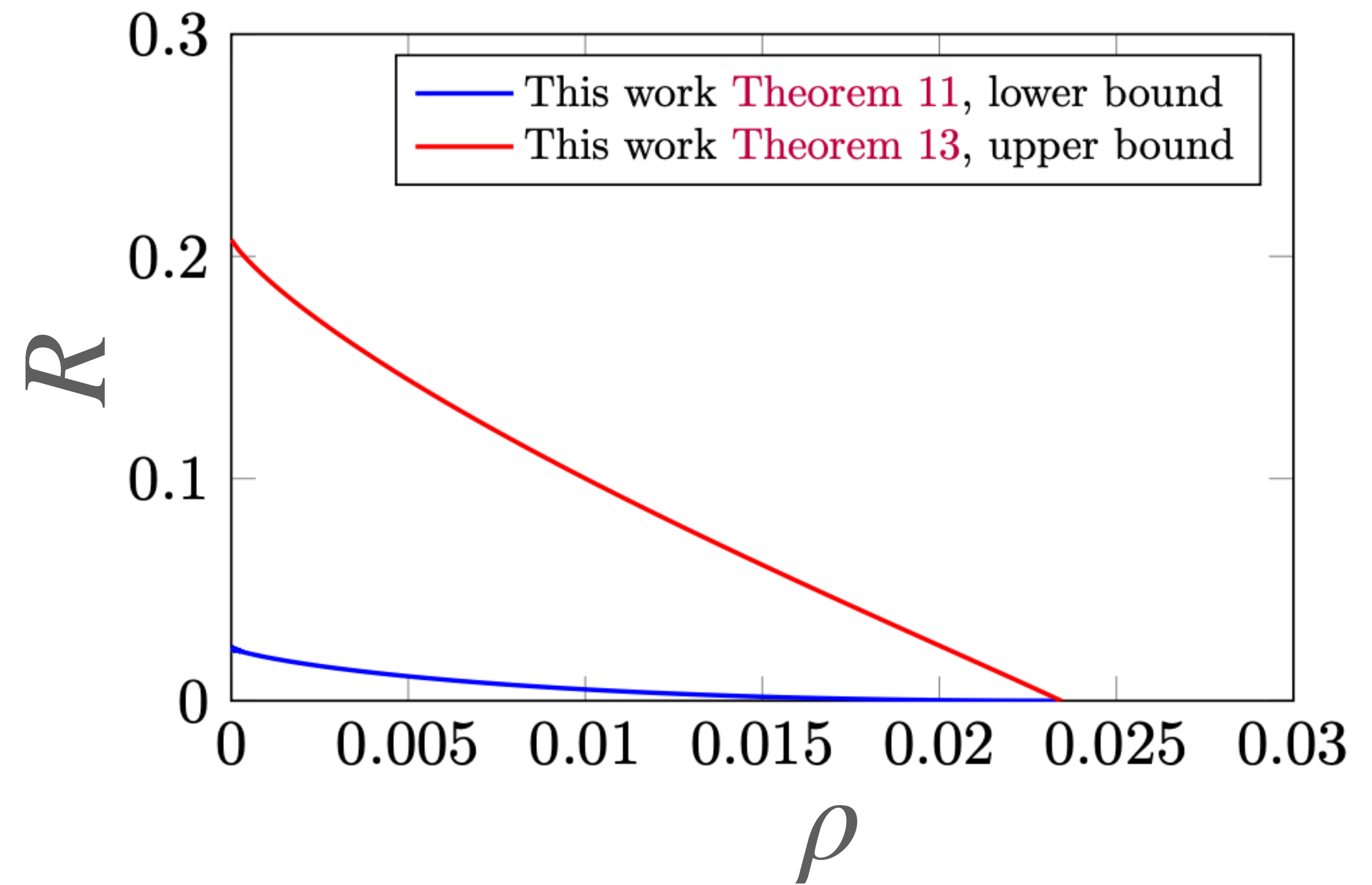
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# SOME PLOTS

$q = 3, L = 3$



$q = 3, L = 4$



**CONCLUSION**

# RECAP

Analogous result  
for list-recovery

From this: derive  
new upper bounds  
on rate of  $(\rho, L)$ -list-  
decodable codes

- For all  $q \geq 2$  and  $L \geq 2$ , we prove the equality

$$\rho_*(q, L) = 1 - \frac{1}{L} \mathbb{E}_{(X_1, \dots, X_L) \sim \text{Unif}([q]^L)} [\text{pl}(X_1, \dots, X_L)]$$

- Two parts:

Possibility Result:  $\exists$  positive  
rate  $(\rho_* - \epsilon, L)$ -list-dec.  
codes (standard random  
code argument)

Impossibility Result: If  
 $\mathcal{C} \subseteq [q]^n$  is  $(\rho_* + \epsilon, L)$ -list-dec.,  
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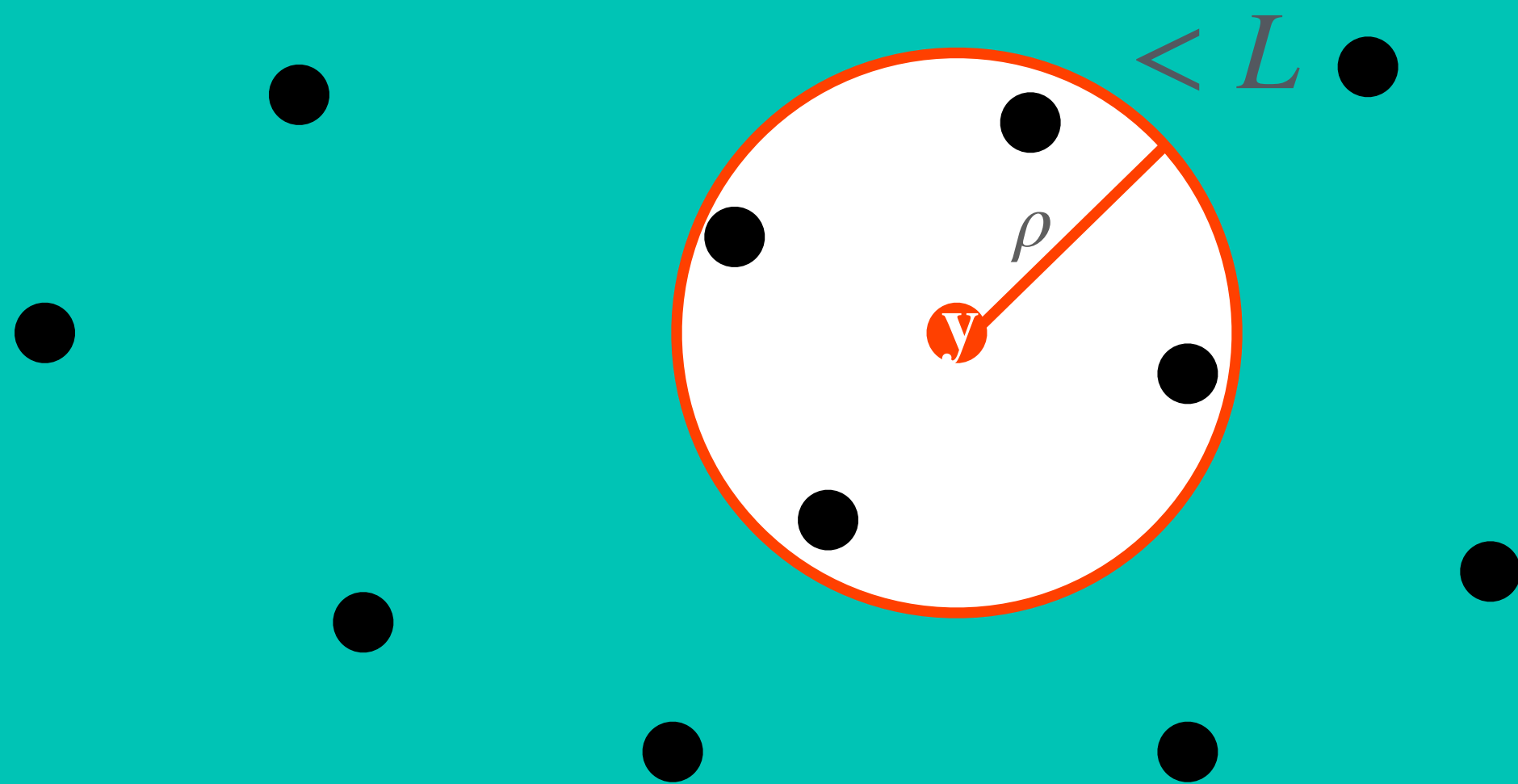
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Question: explicit bound à la  
[Alon-Bukh-Polyanskiy'18]?

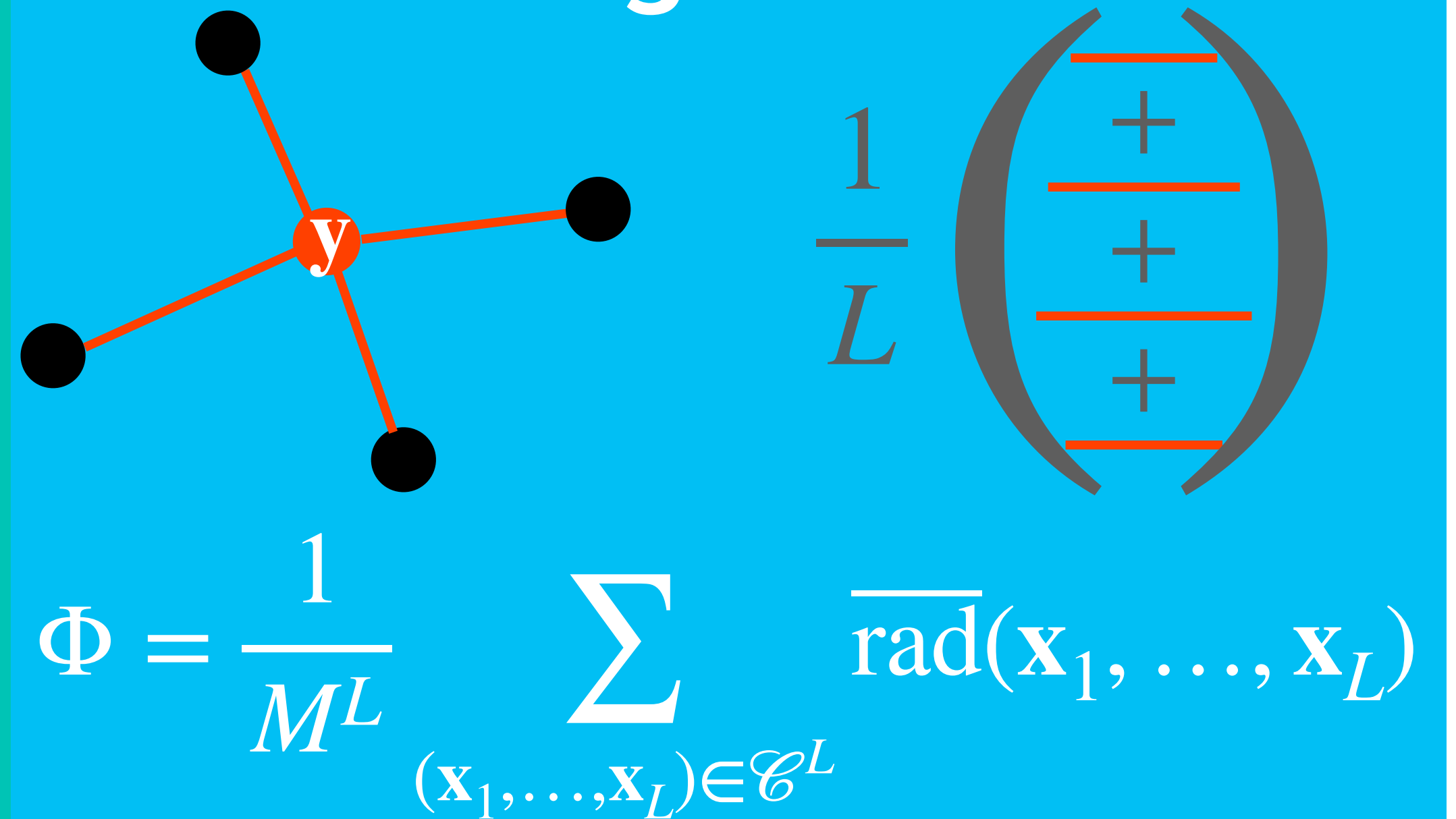
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## List-decodable codes



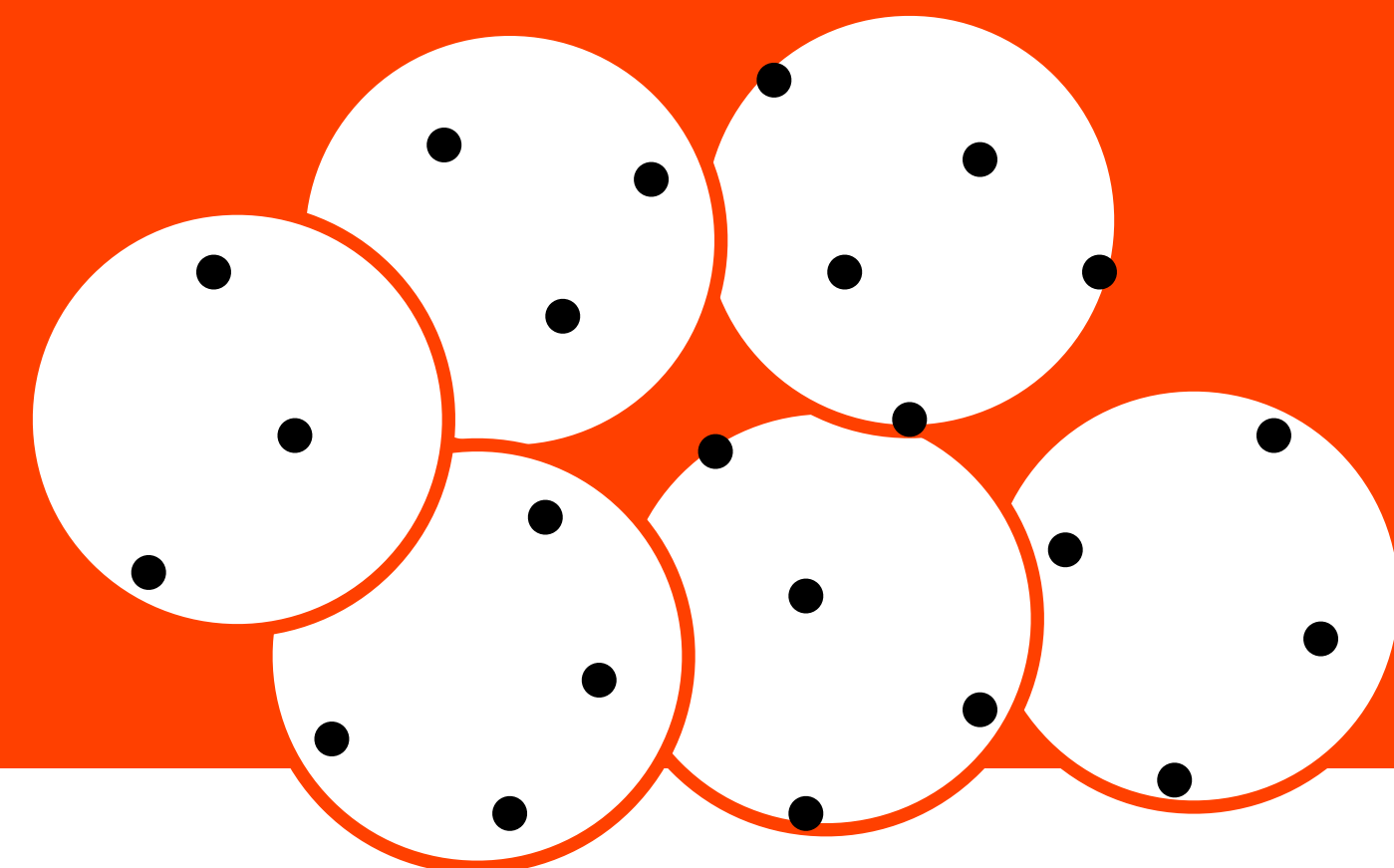
## Average-Radius



## Schur-Convexity

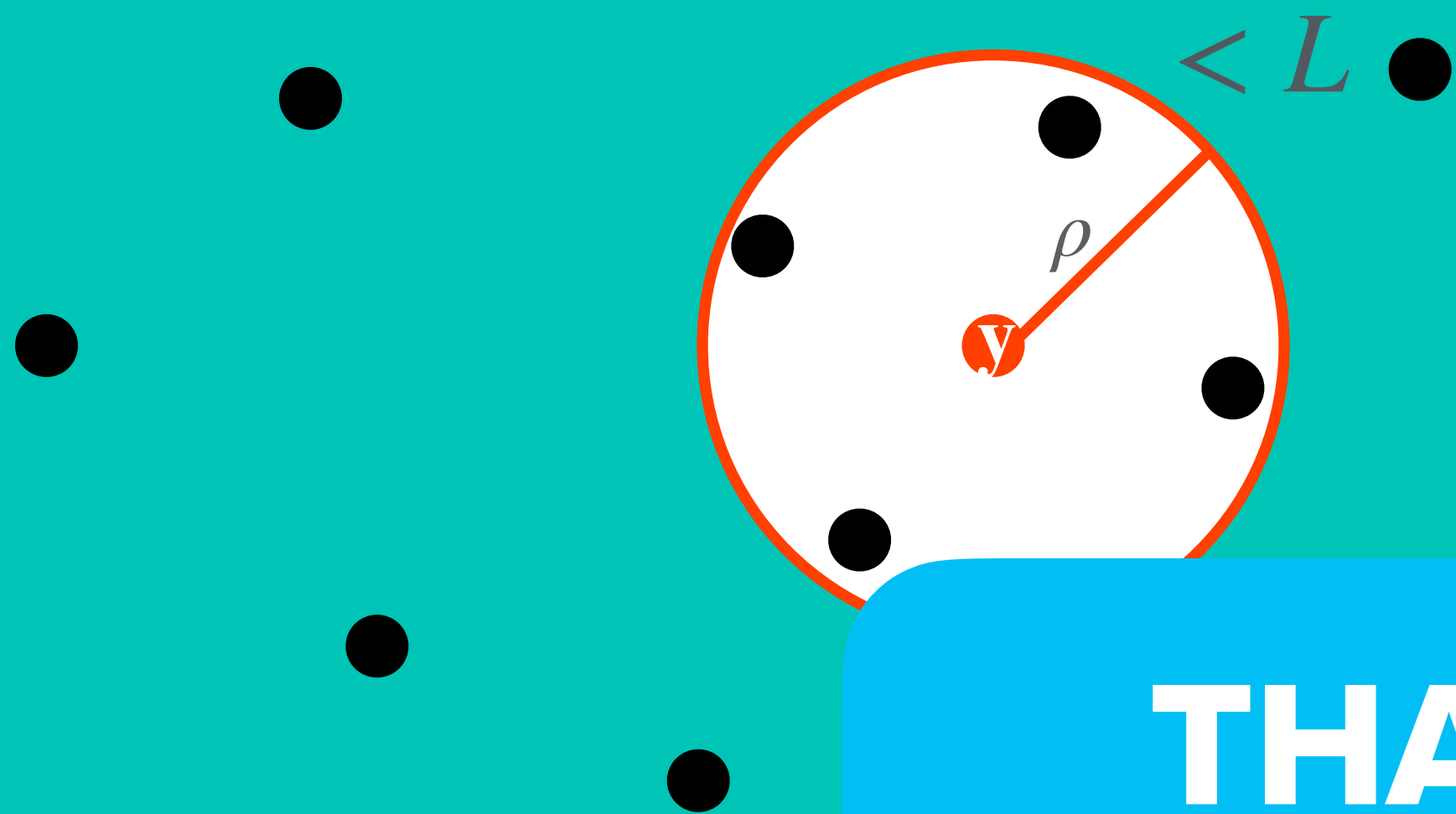
$$(P_i - P_j) \left( \frac{\partial f_{q,L}}{\partial P_i}(P) - \frac{\partial f_{q,L}}{\partial P_j}(P) \right) \geq 0$$

## Elias-Bassalygo Bound

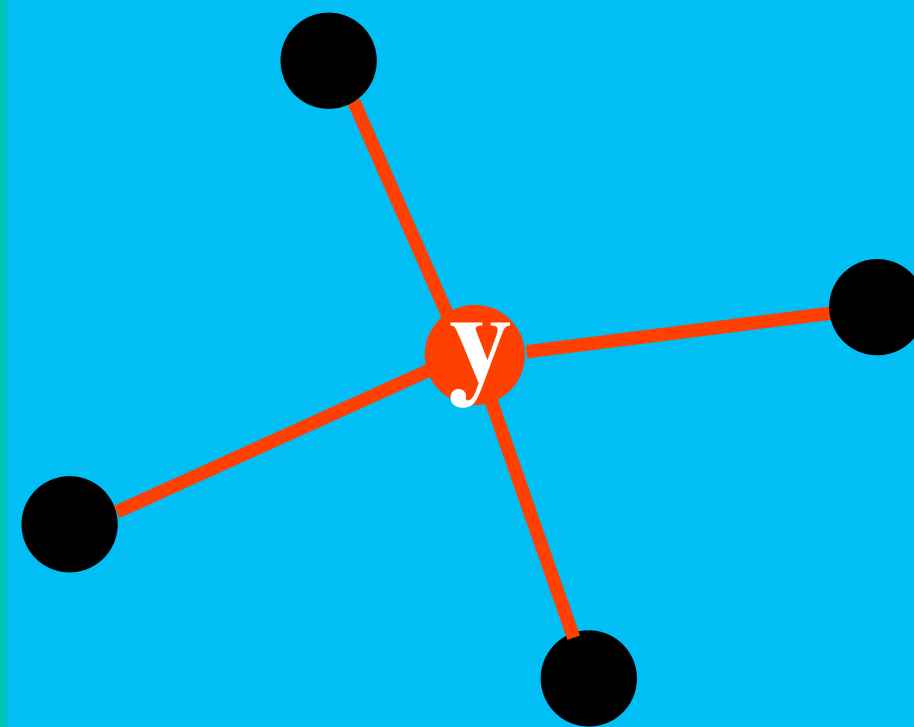




## List-decodable codes



## Average-Radius



$$\frac{1}{L} \left( \begin{array}{c} \text{---} \\ + \\ \text{---} \\ + \\ \text{---} \\ + \\ \text{---} \end{array} \right)$$

**THANK YOU!**  
**QUESTIONS?**

$$\overline{\text{rad}}(\mathbf{x}_1, \dots, \mathbf{x}_L)$$

$$\in \mathcal{C}^L$$

## Schur-Convex

$$(P_i - P_j) \left( \frac{\partial f_{q,L}}{\partial P_i}(P) - \frac{\partial f_{q,L}}{\partial P_j}(P) \right) \geq 0$$

## Altygo Bound

