

Sieving for shortest lattice vectors using nearest neighbor techniques

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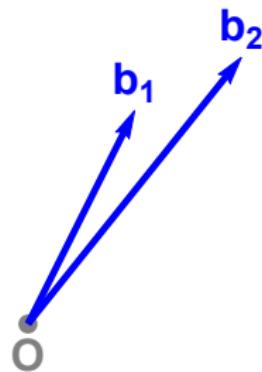
Lattices

What is a lattice?



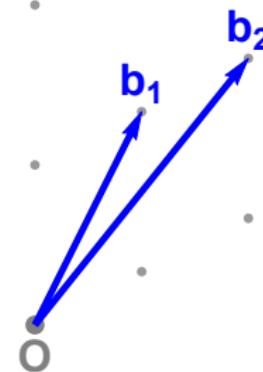
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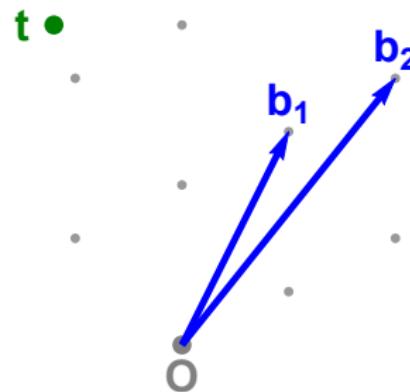
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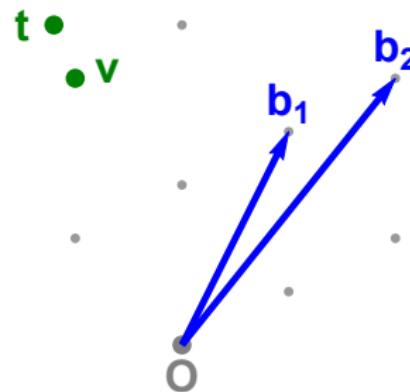
Lattices

Closest Vector Problem (CVP)



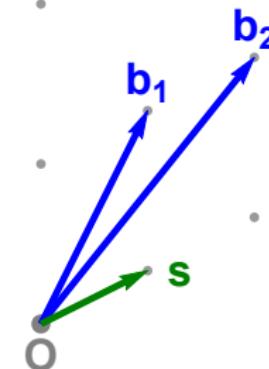
Lattices

Closest Vector Problem (CVP)



Lattices

Shortest Vector Problem (SVP)



Lattices

Applications

- “Constructive cryptography”: Lattice-based cryptosystems
 - ▶ Based on hard lattice problems (SVP, CVP, LWE, SIS)
 - ▶ NTRU cryptosystem [HPS98, ..., HPSSWZ15]
 - ▶ HIMMO key pre-distribution scheme [GMGPGRST14]
 - ▶ Fully Homomorphic Encryption [Gen09, ..., CM15]
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- “Destructive cryptography”: Lattice cryptanalysis
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How hard are hard lattice problems such as SVP?

Lattices

Exact SVP algorithms

	Algorithm	$\log_2(\text{Time})$	$\log_2(\text{Space})$
Provable SVP	Enumeration [Poh81, Kan83, ..., MW15]	$\Omega(n \log n)$	$O(\log n)$
	AKS-sieve [AKS01, NV08, MV10, HPS11]	$3.398n$	$1.985n$
	ListSieve [MV10, MDB14]	$3.199n$	$1.327n$
	AKS-sieve-birthday [PS09, HPS11]	$2.648n$	$1.324n$
	ListSieve-birthday [PS09]	$2.465n$	$1.233n$
	Voronoi cell algorithm [AEVZ02, MV10b]	$2.000n$	$1.000n$
Heuristic SVP	Discrete Gaussians [ADRS15, ADS15, Ste16]	$1.000n$	$1.000n$
	Nguyen-Vidick sieve [NV08]	$0.415n$	$0.208n$
	GaussSieve [MV10, ..., IKMT14, BNvdP14]	$0.415n?$	$0.208n$
	Two-level sieve [WLTB11]	$0.384n$	$0.256n$
	Three-level sieve [ZPH13]	$0.3778n$	$0.283n$
	Overlattice sieve [BGJ14]	$0.3774n$	$0.293n$
	Hyperplane LSH [Laa15, MLB15]	$0.337n$	$0.208n$
	May and Ozerov's NNS method [BGJ15]	$0.311n$	$0.208n$
	Spherical LSH [LdW15]	$0.298n$	$0.208n$
	Cross-polytope LSH [BL15]	$0.298n$	$0.208n$
	Spherical filtering [BDGL16]	$0.293n$	$0.208n$

Nguyen-Vidick sieve

O

Nguyen-Vidick sieve

1. Sample a list L of random lattice vectors



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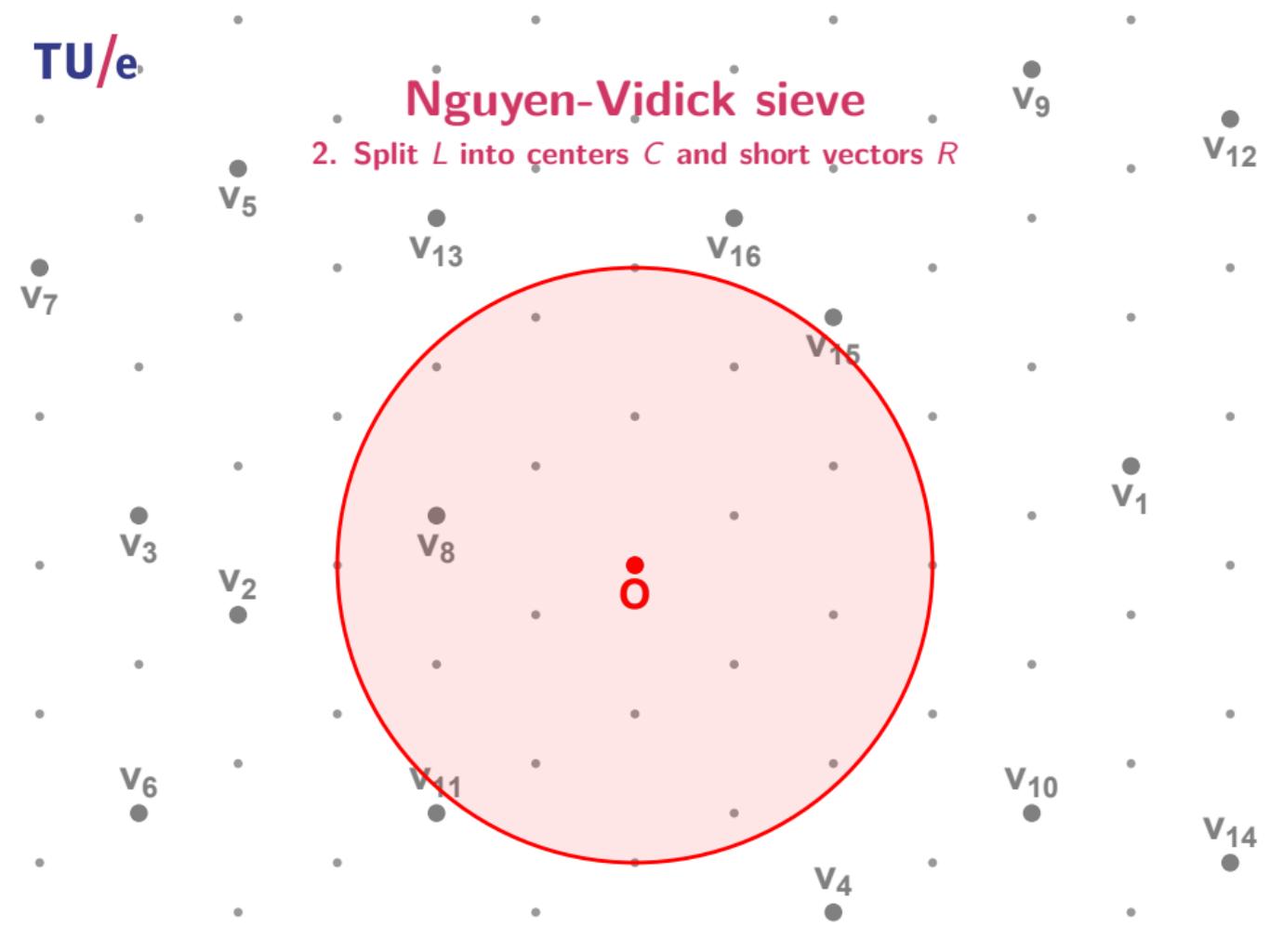
Nguyen-Vidick sieve

2. Split L into centers C and short vectors R



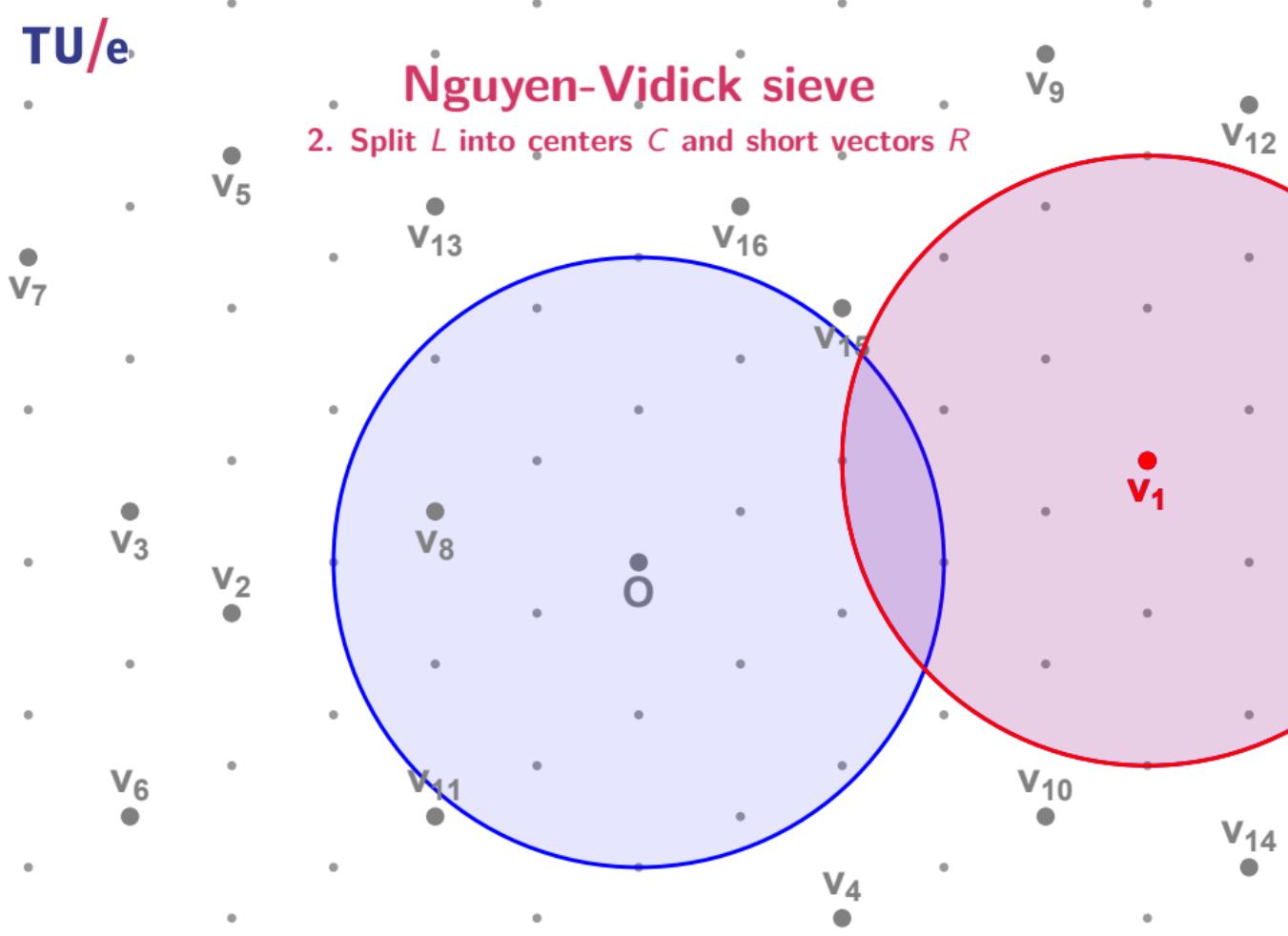
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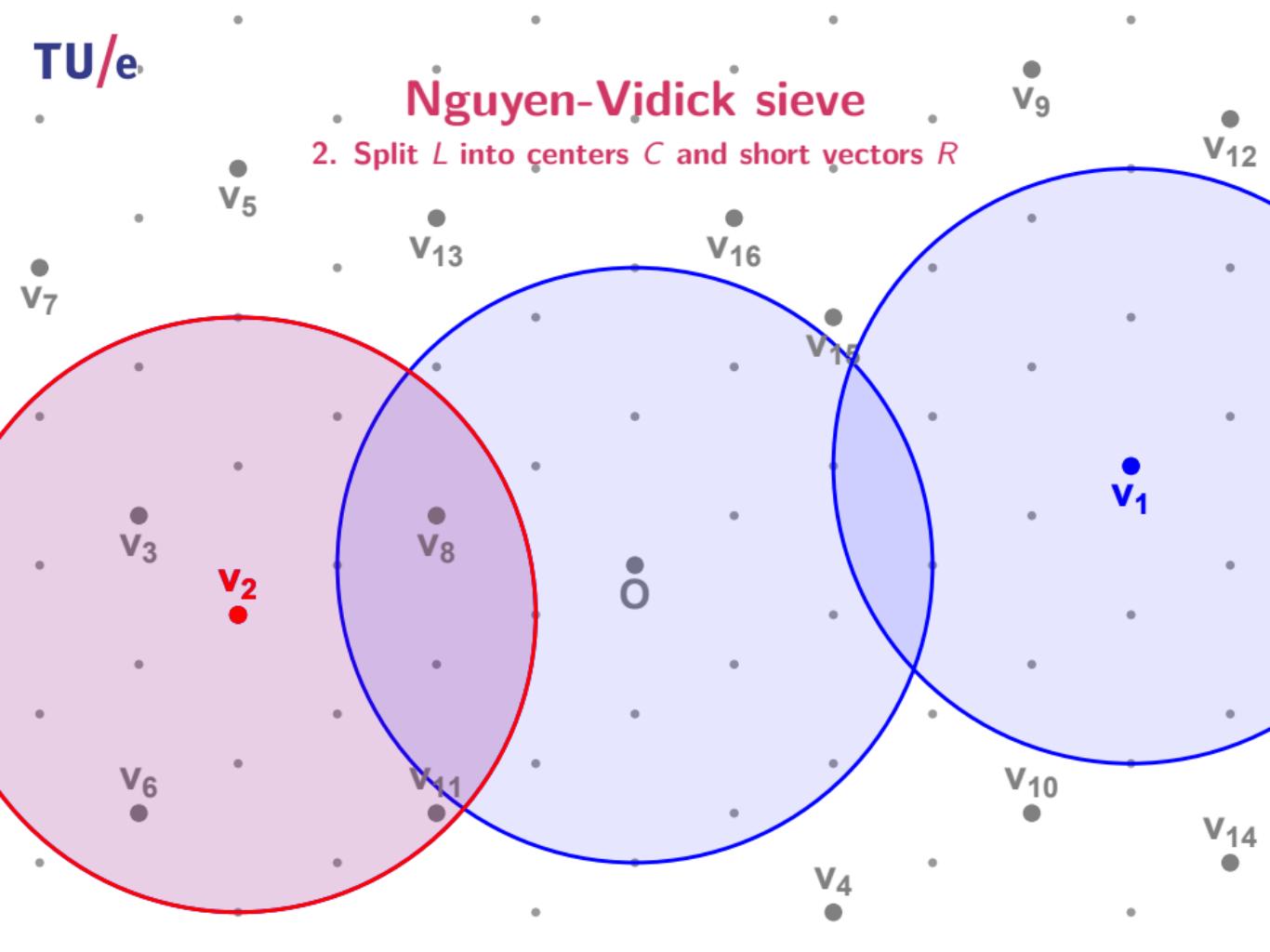
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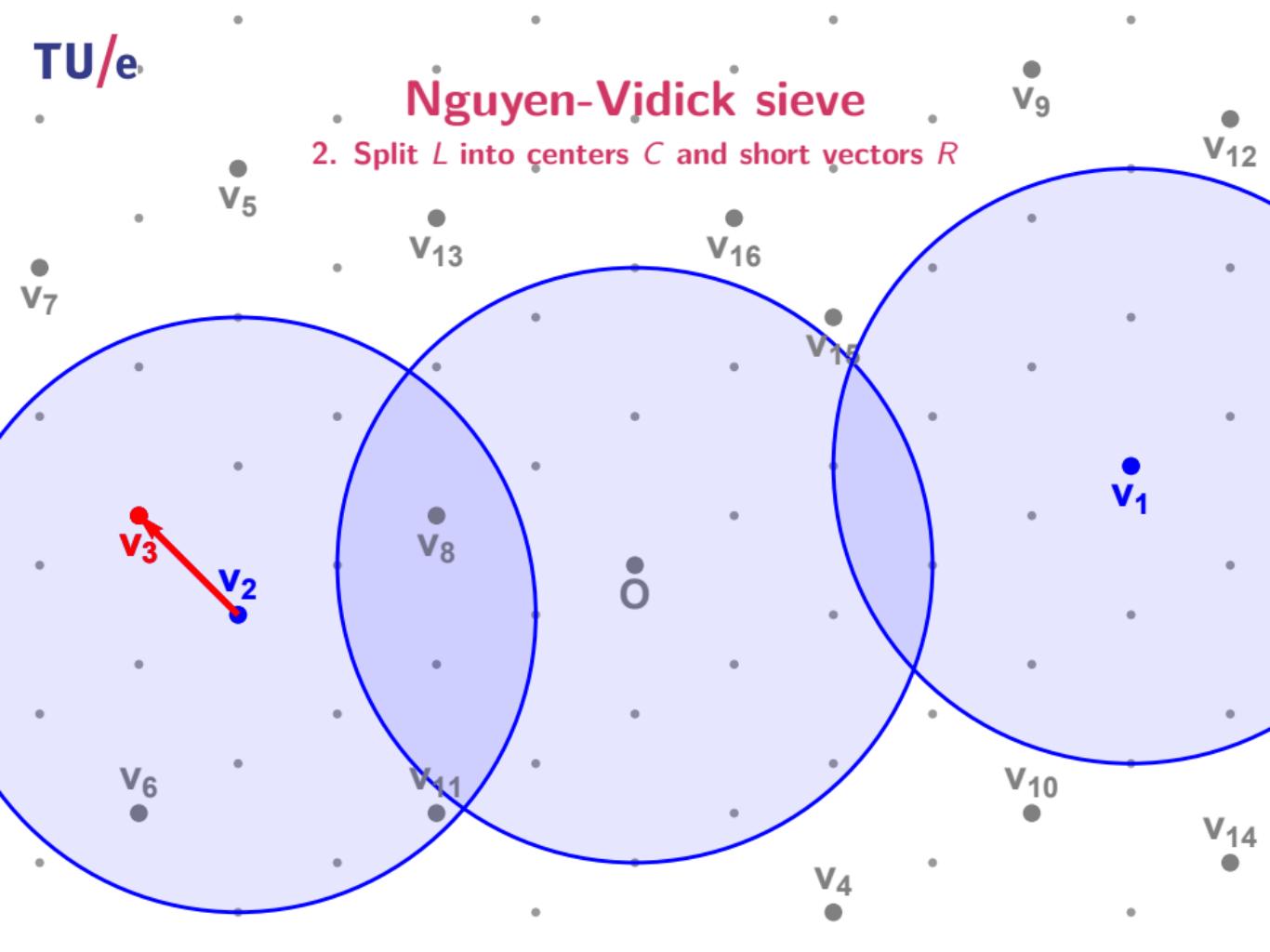
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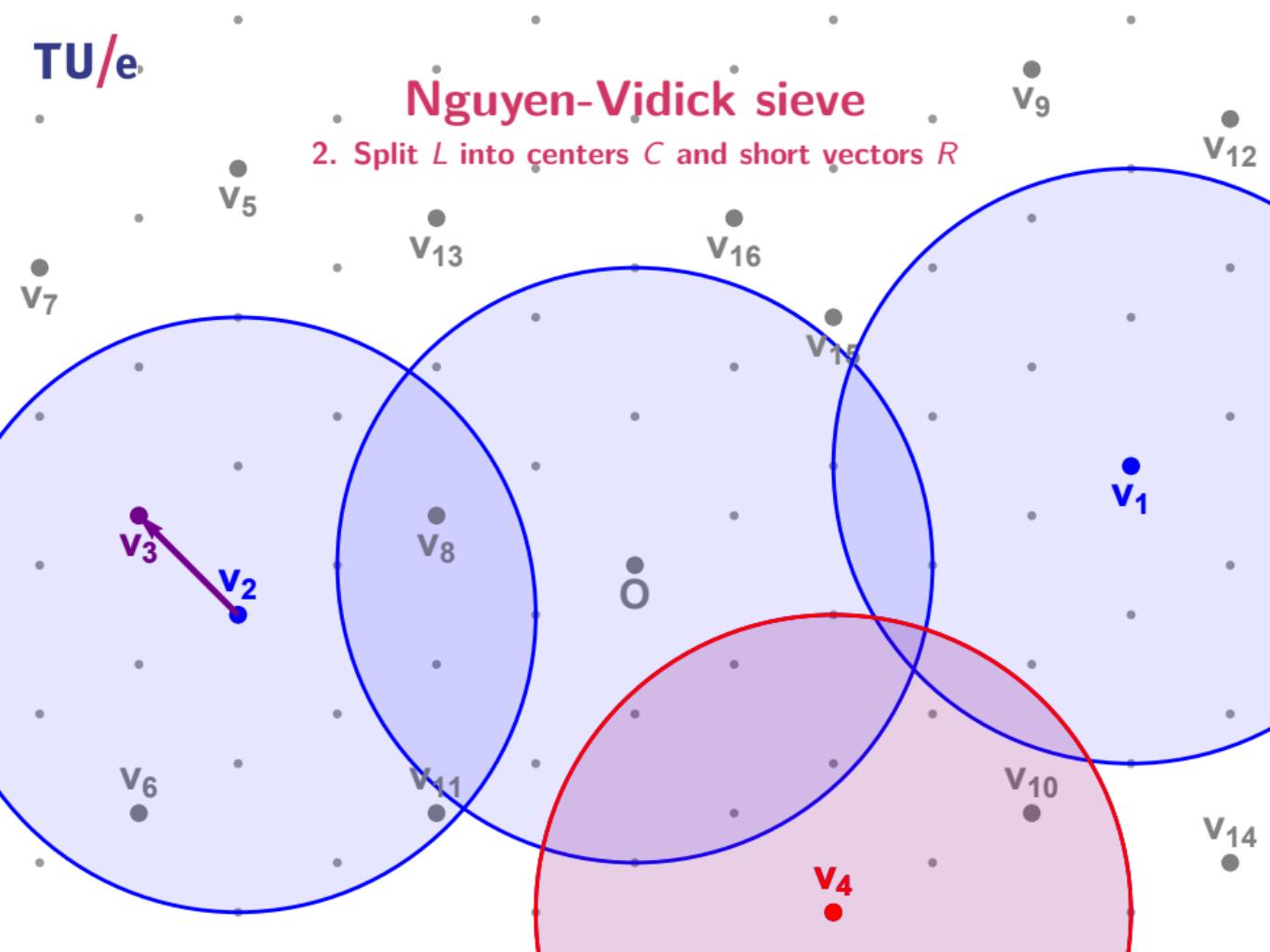
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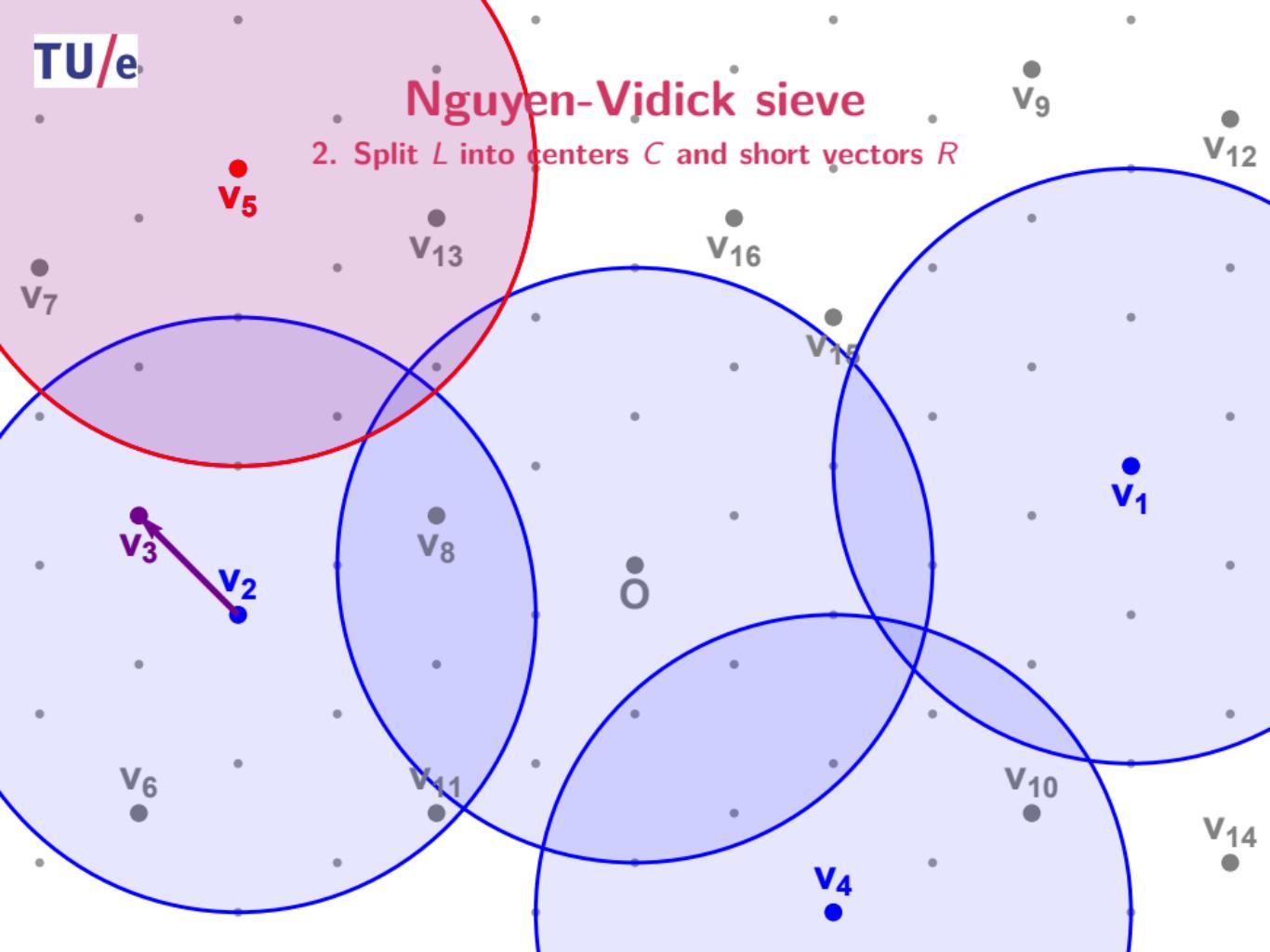
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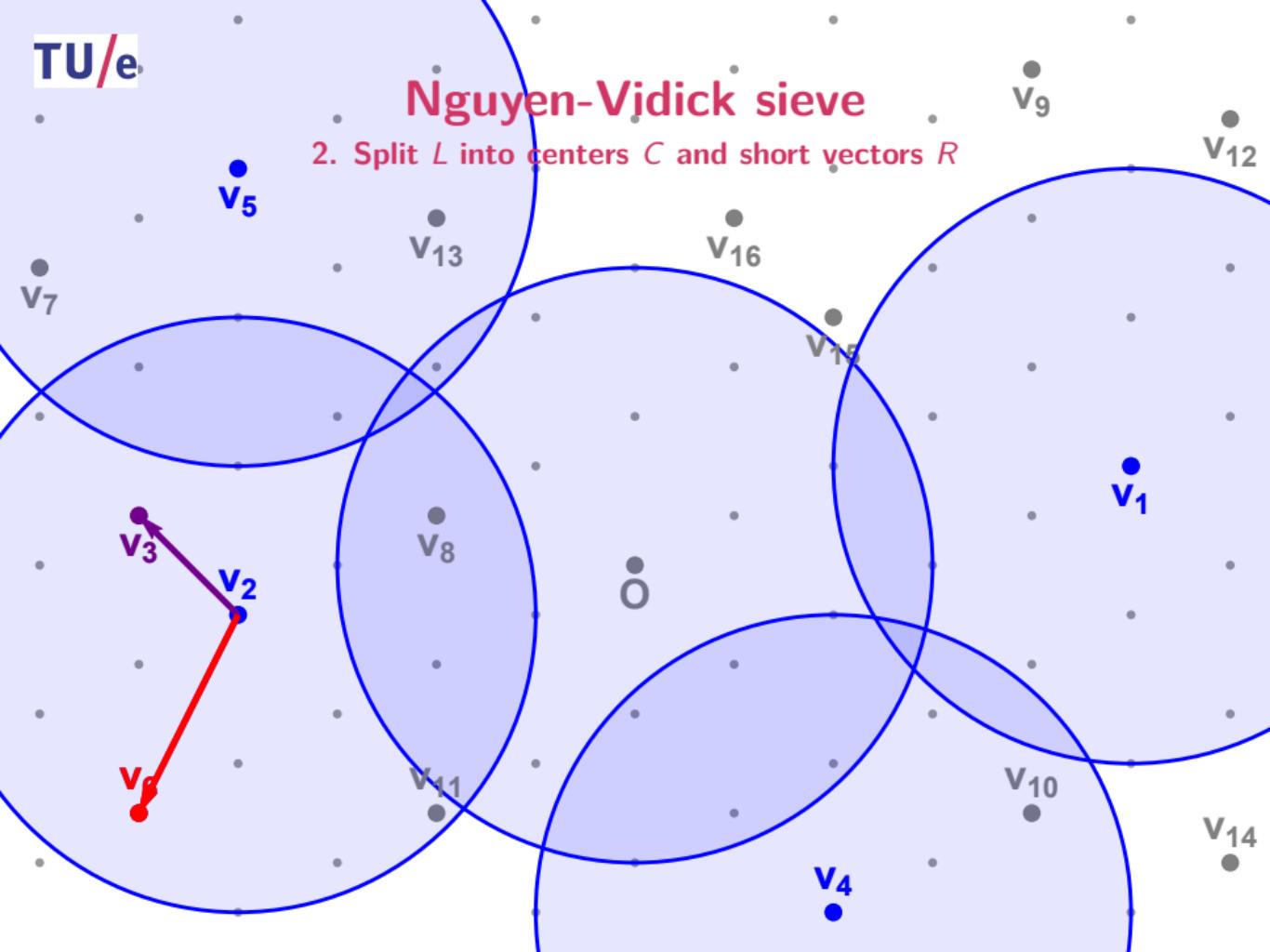
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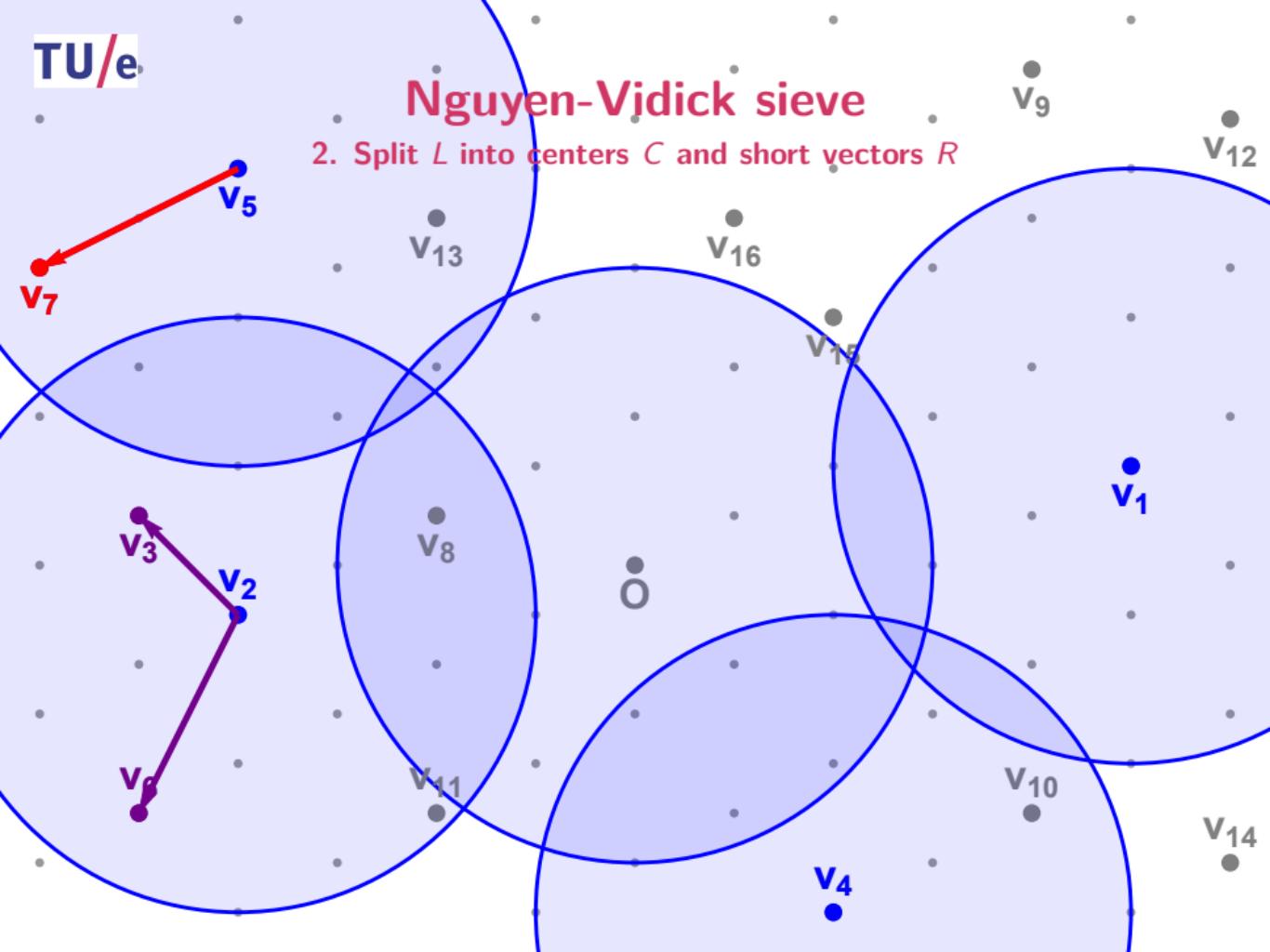
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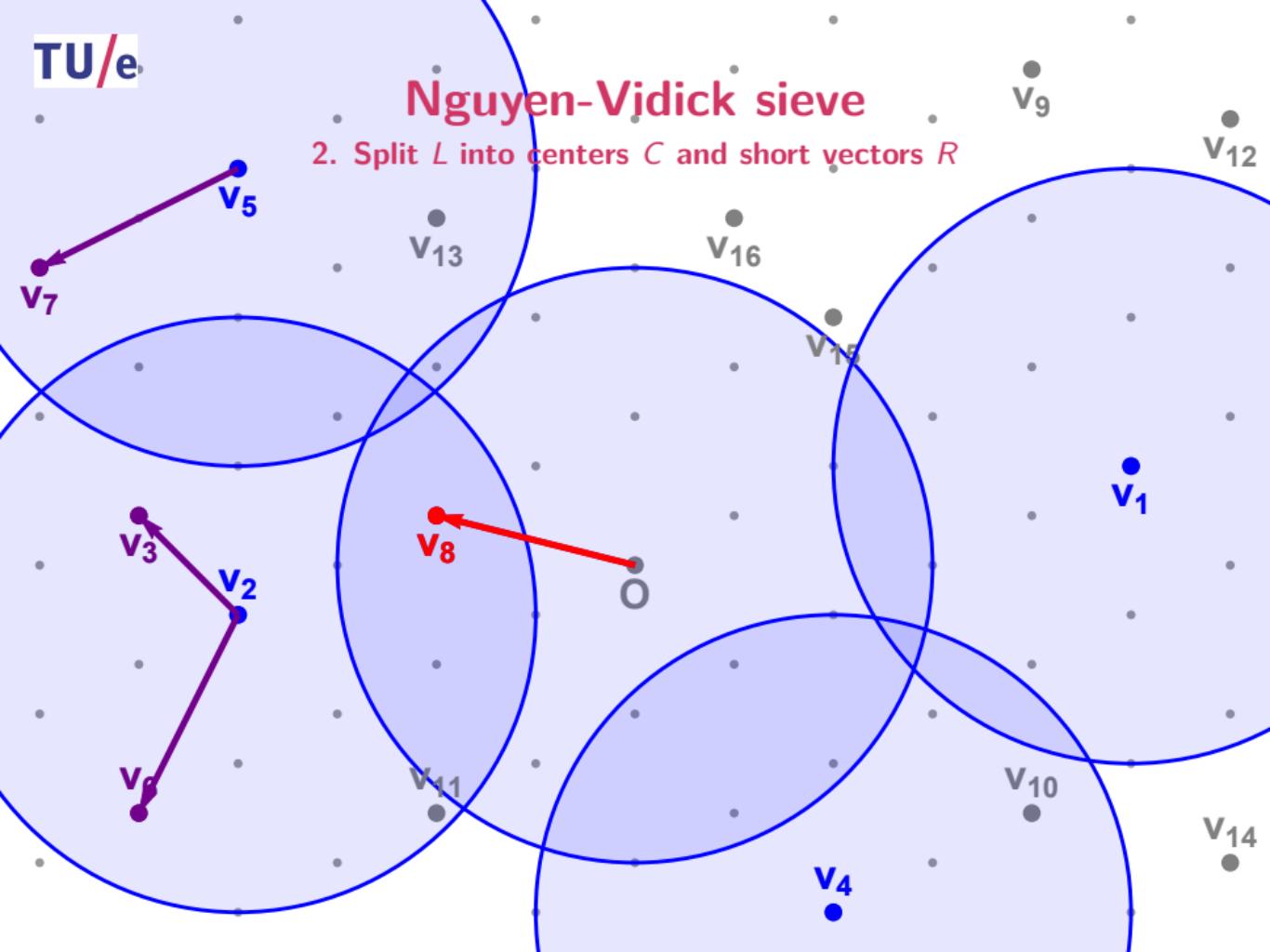
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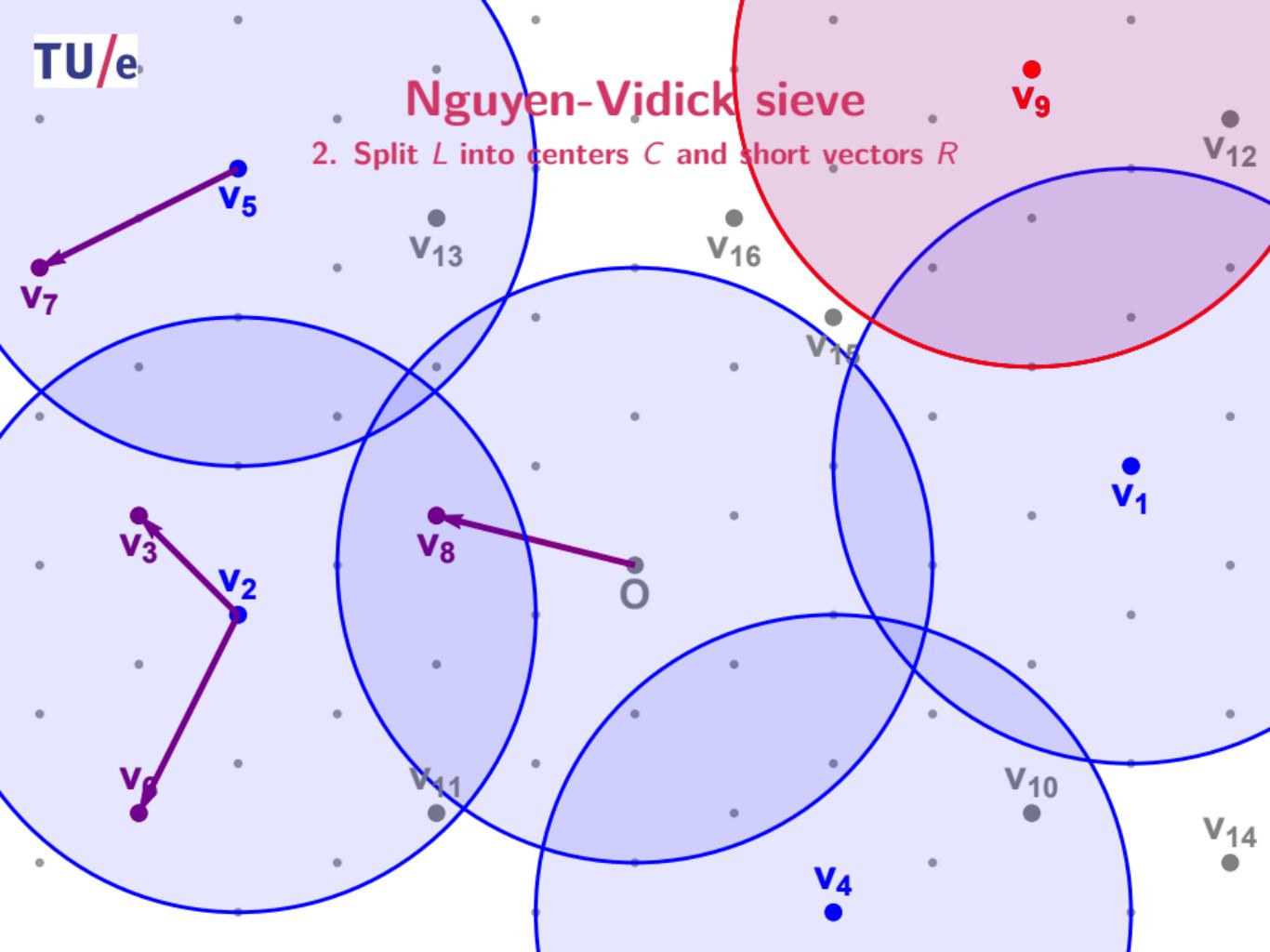
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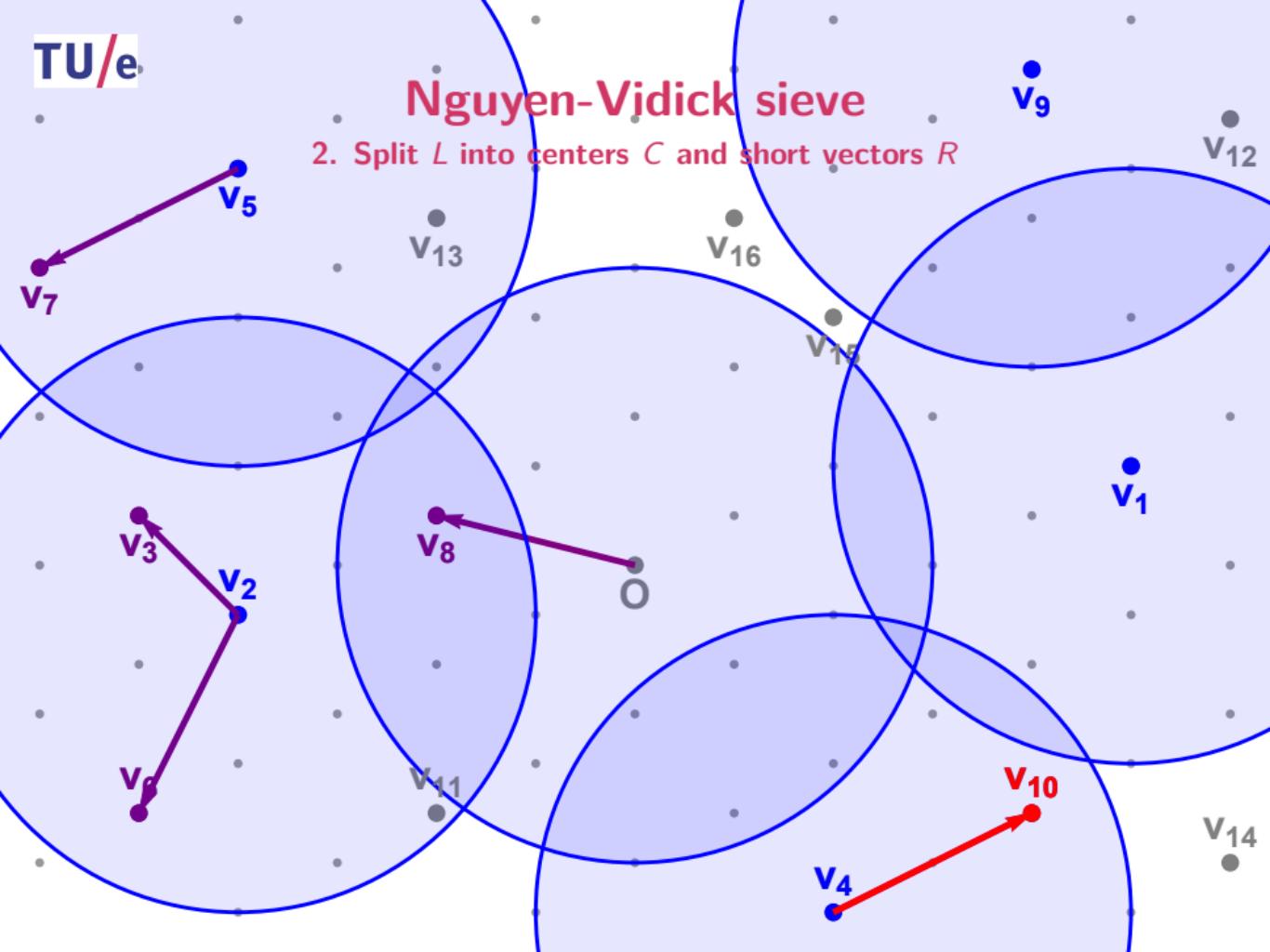
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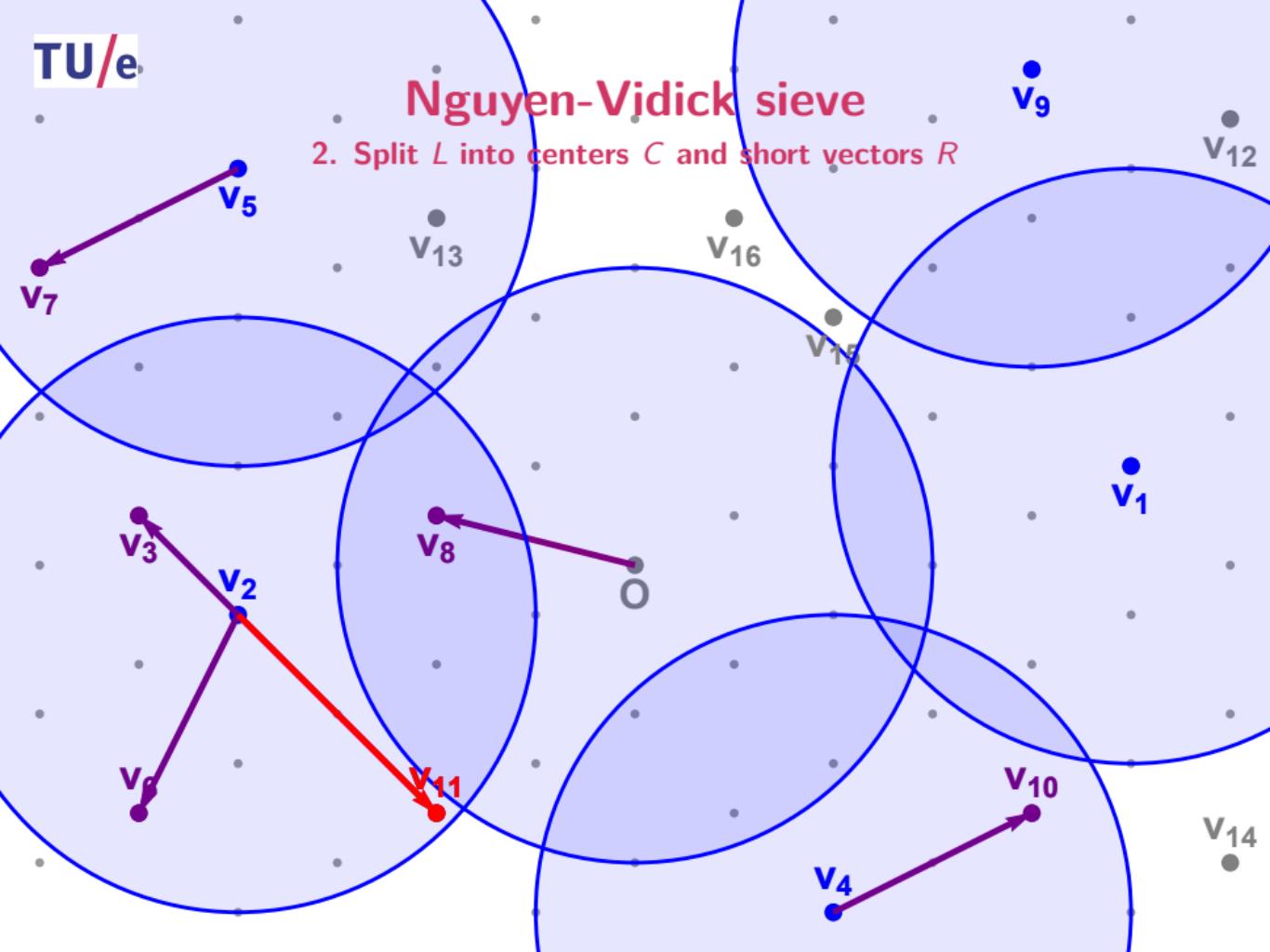
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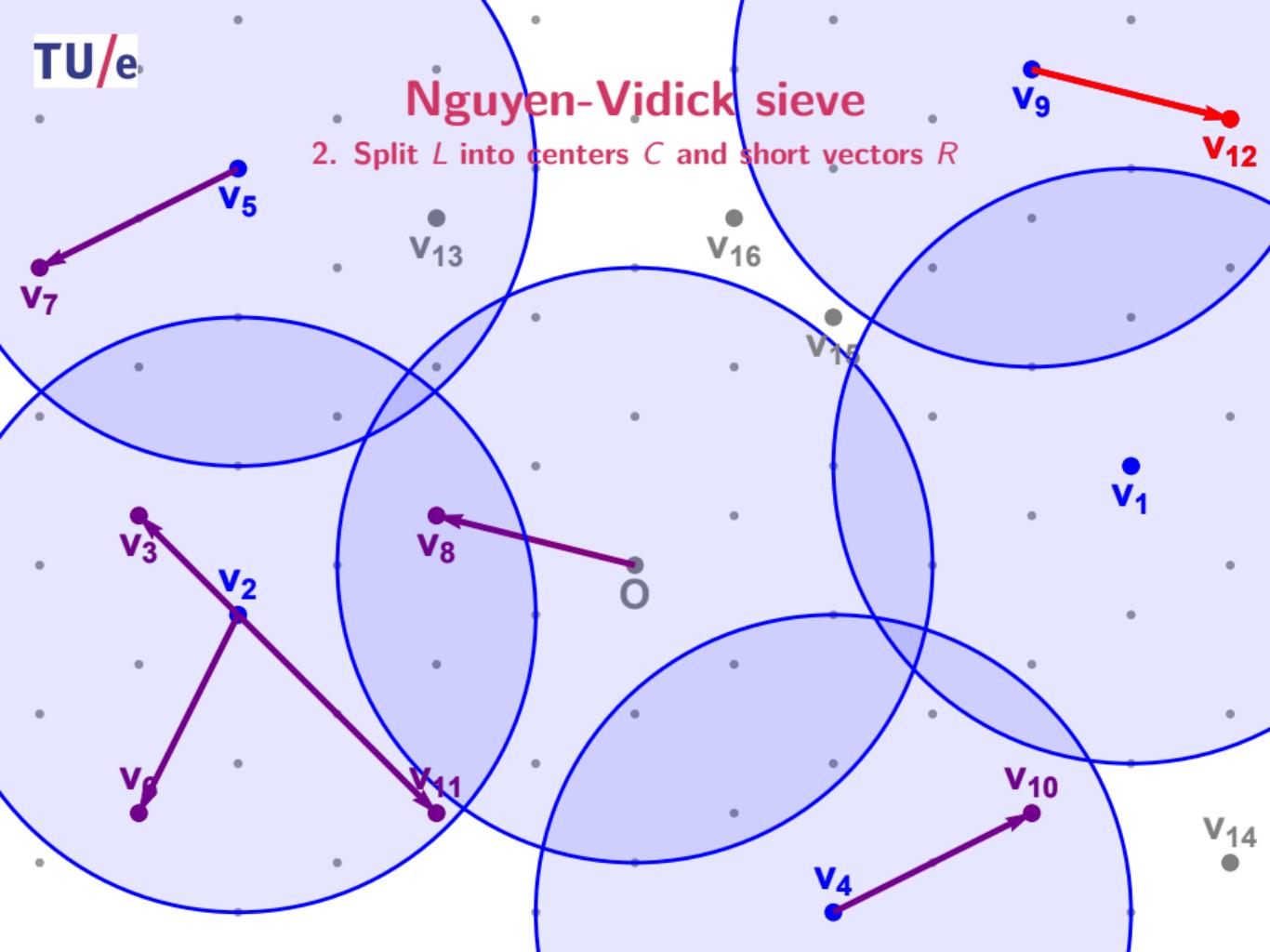
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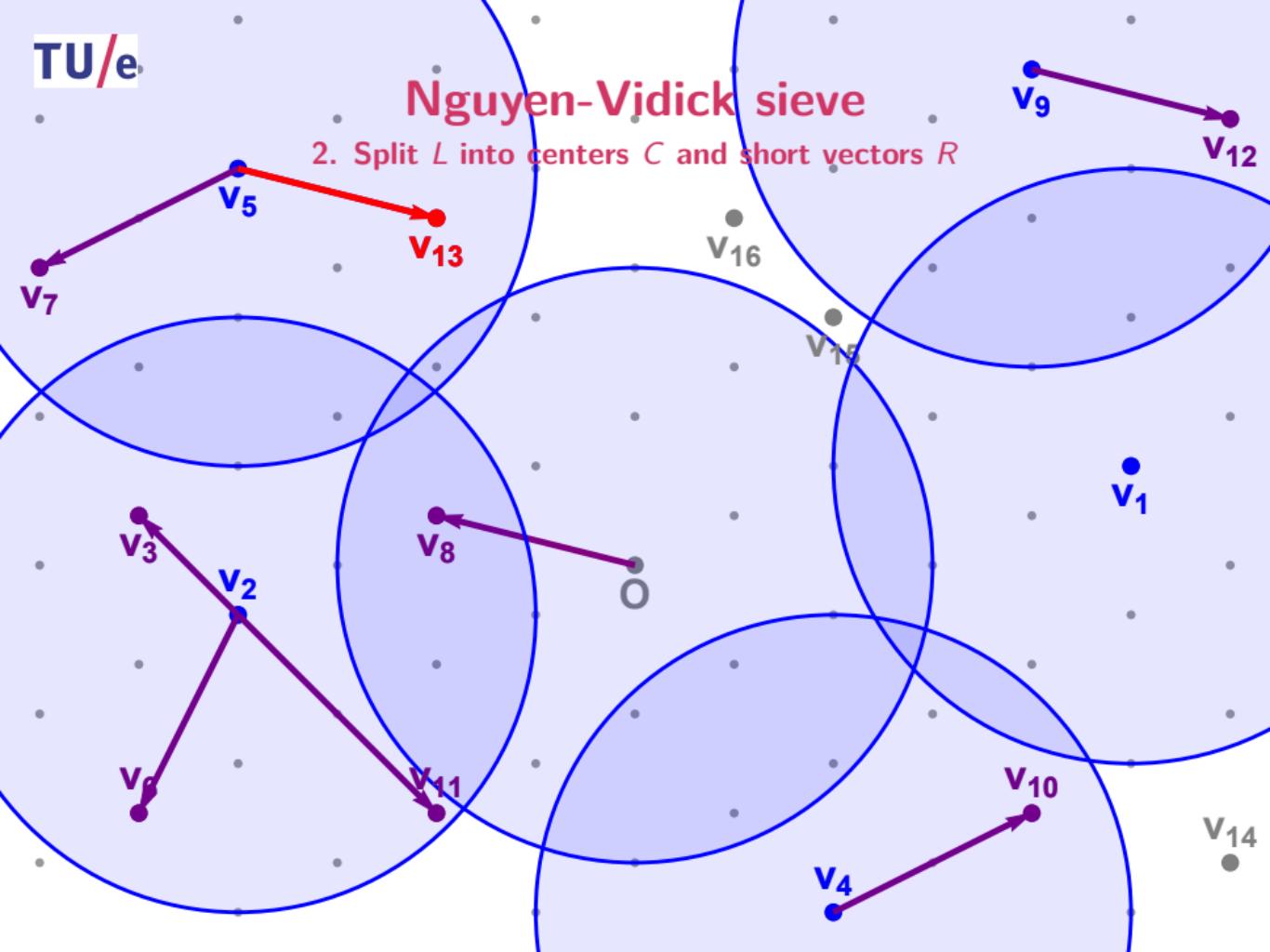
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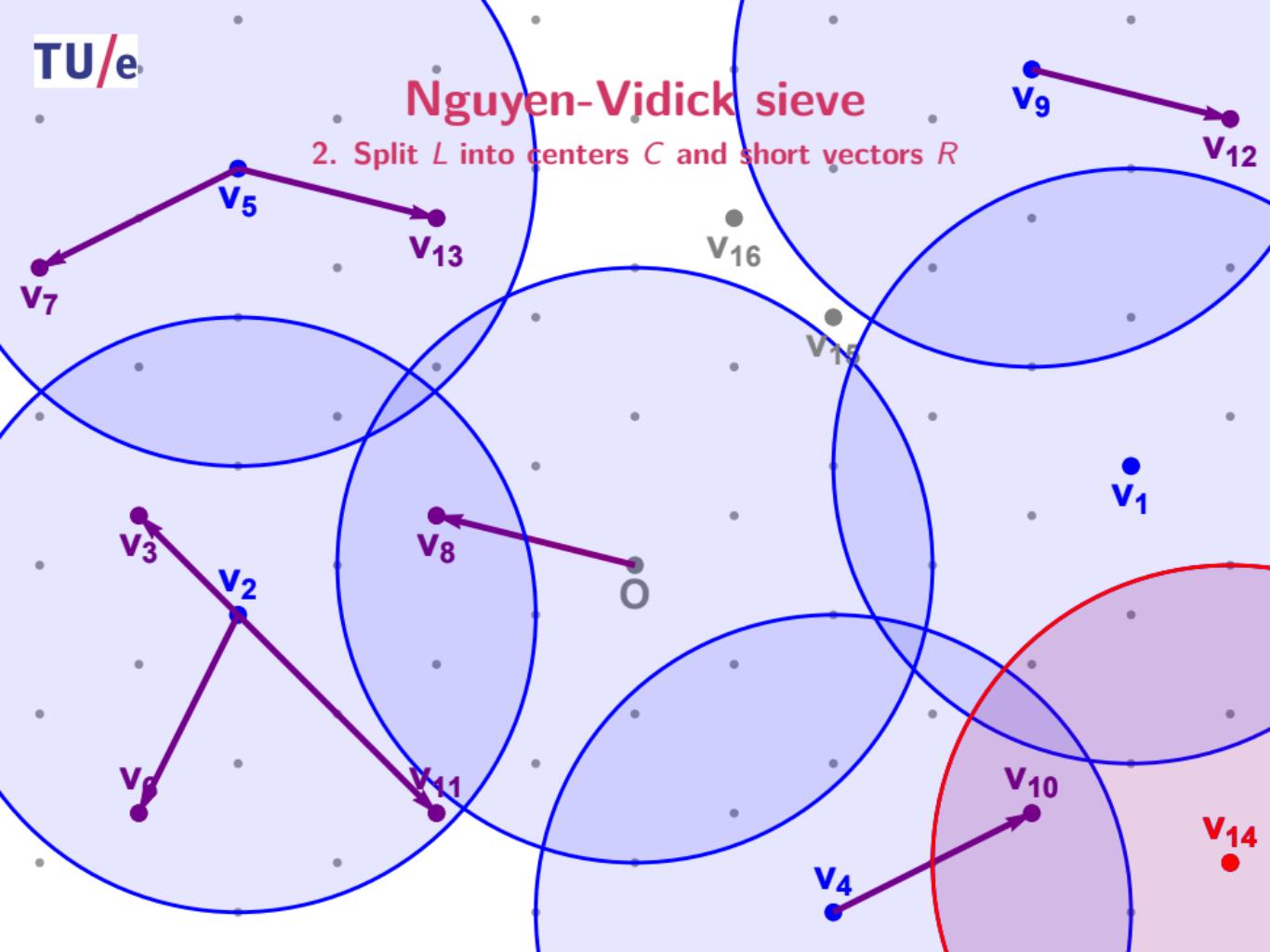
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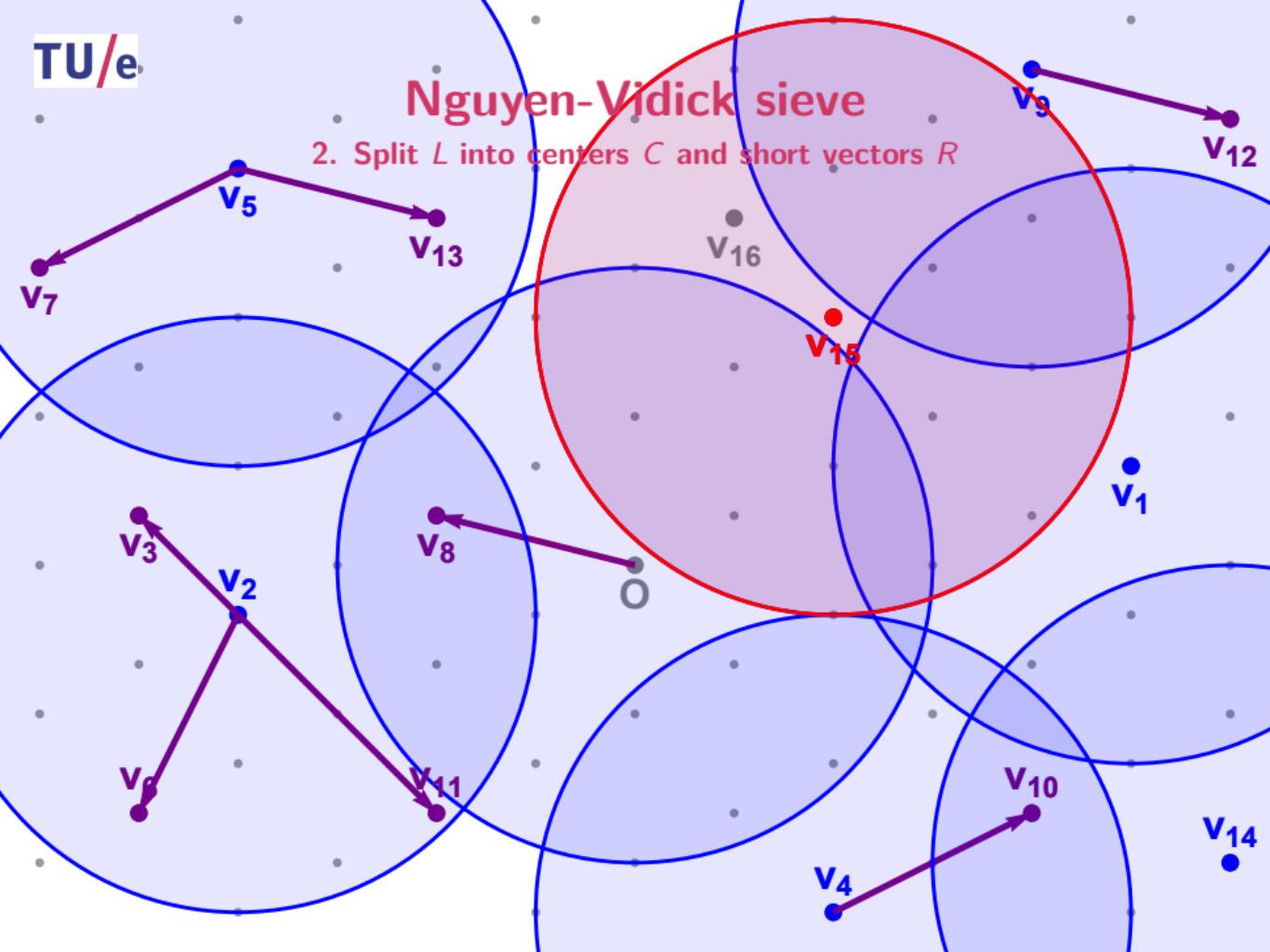
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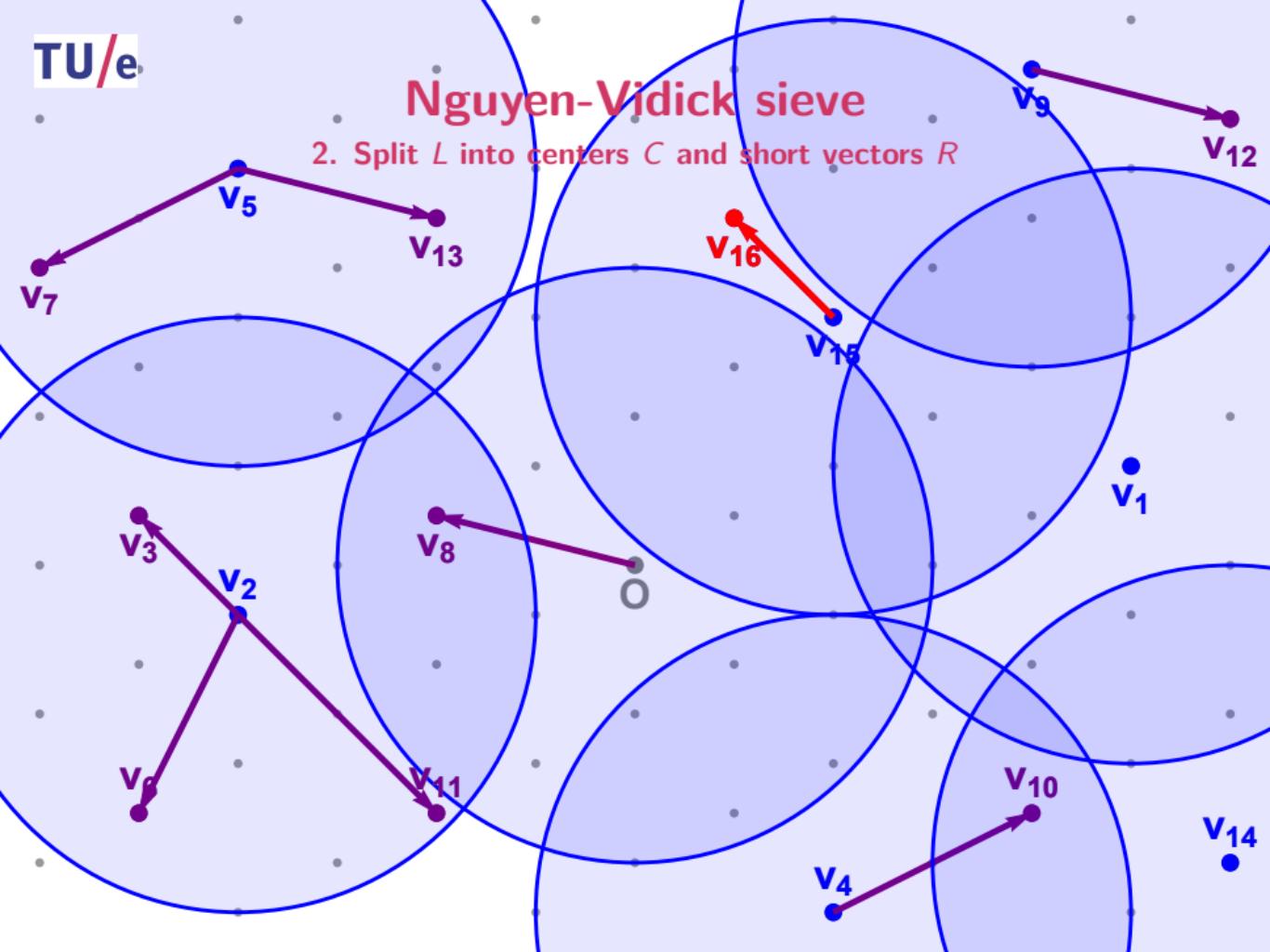
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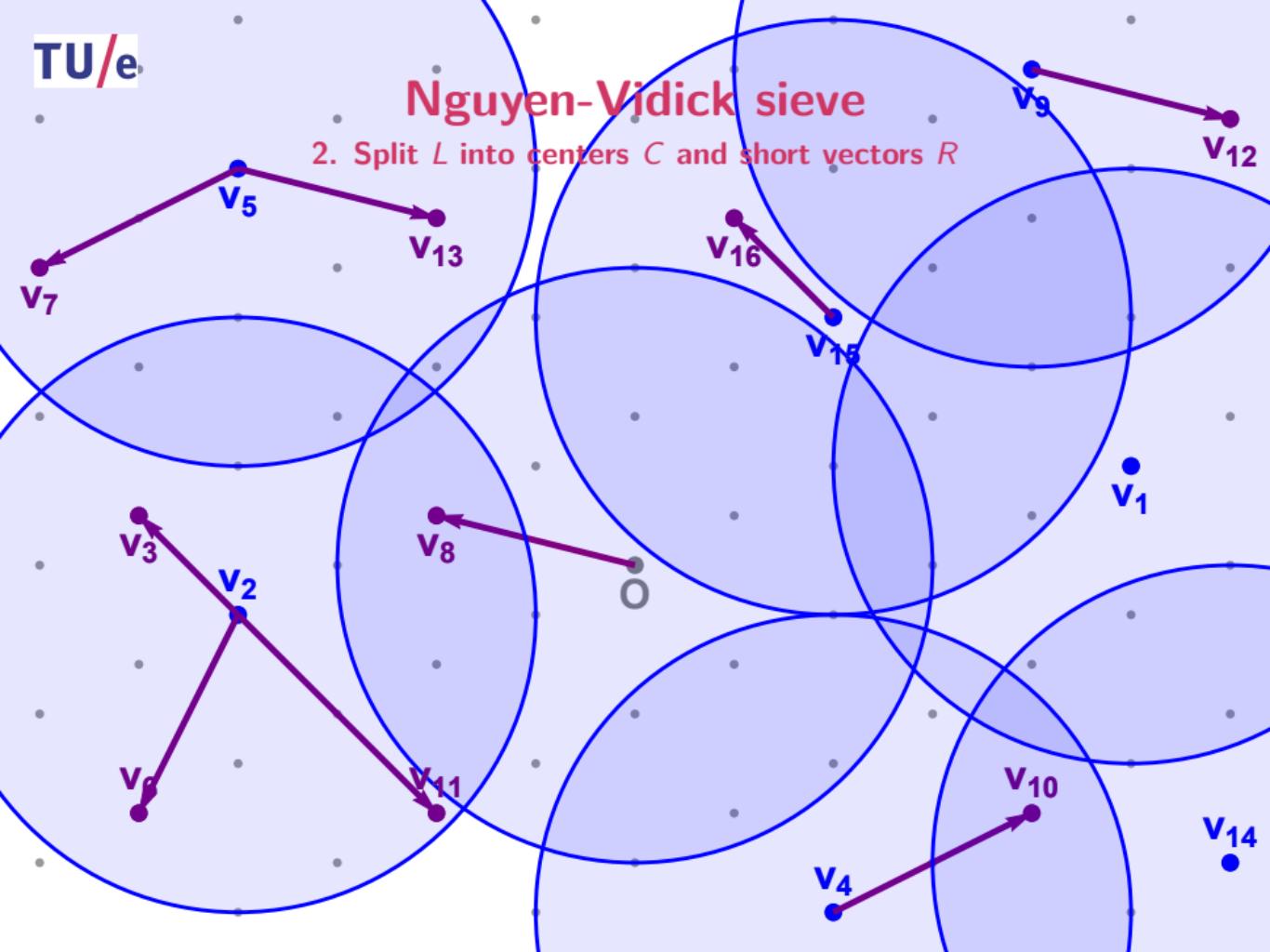
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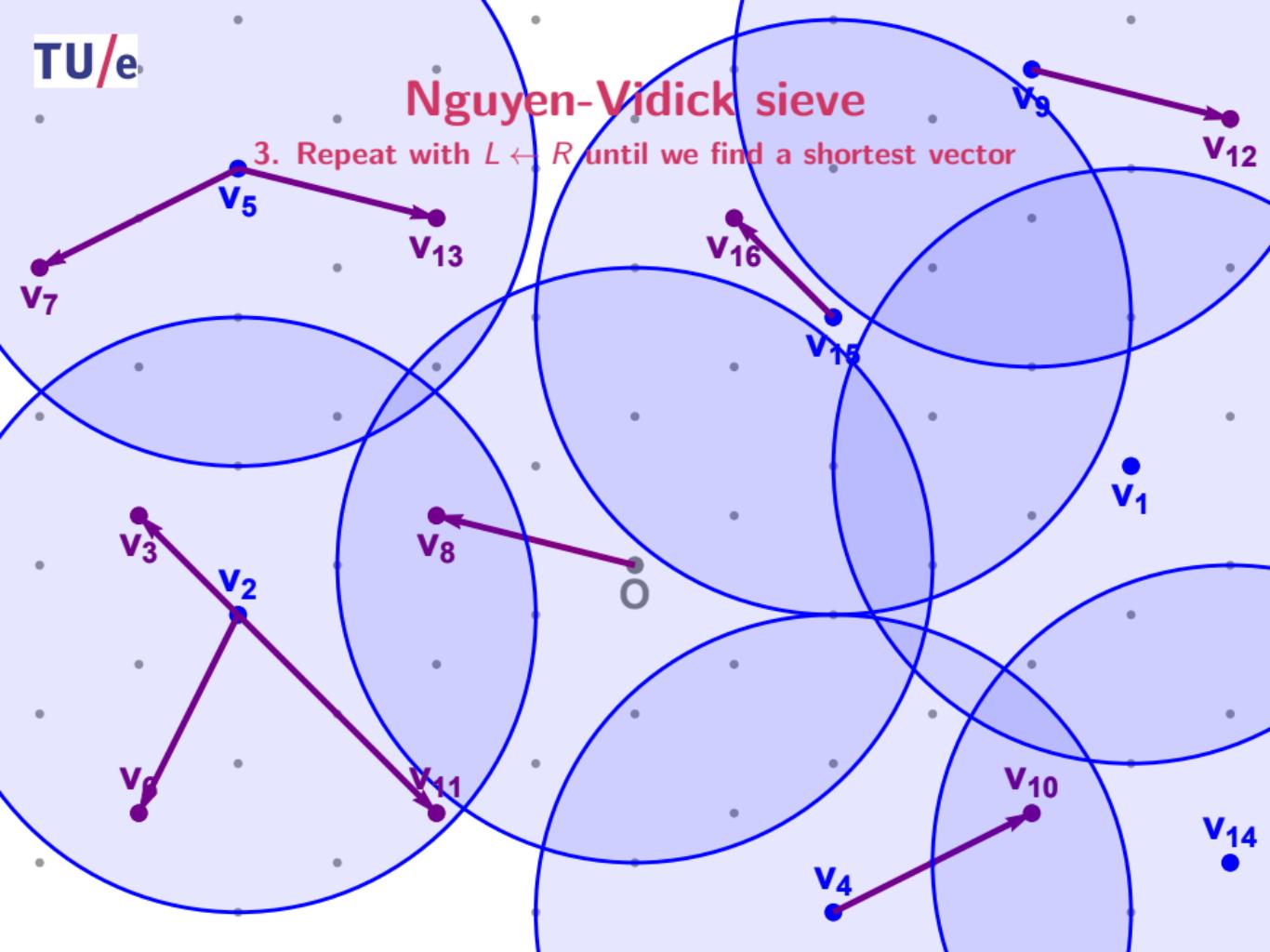
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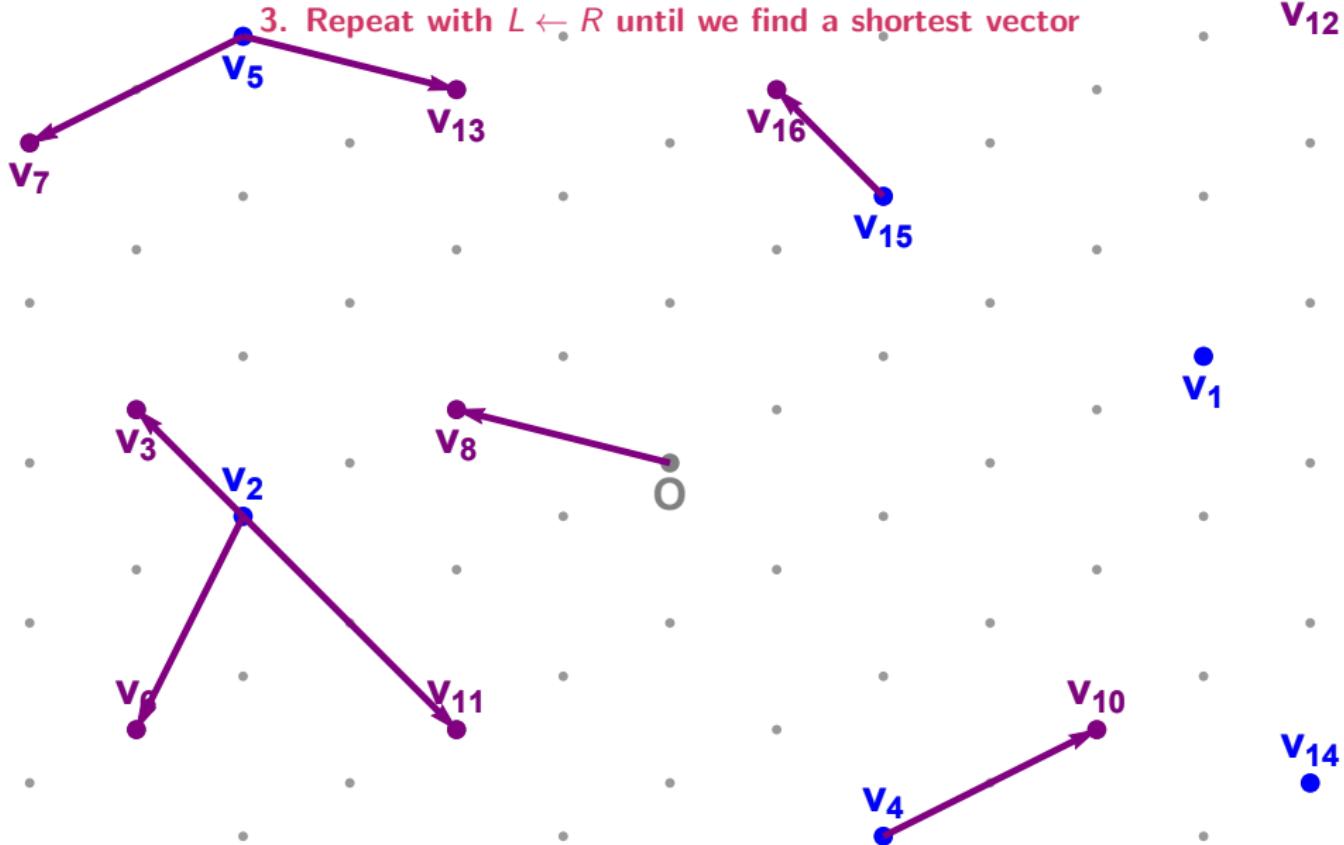
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3. Repeat with $L \leftarrow R$ until we find a shortest vector



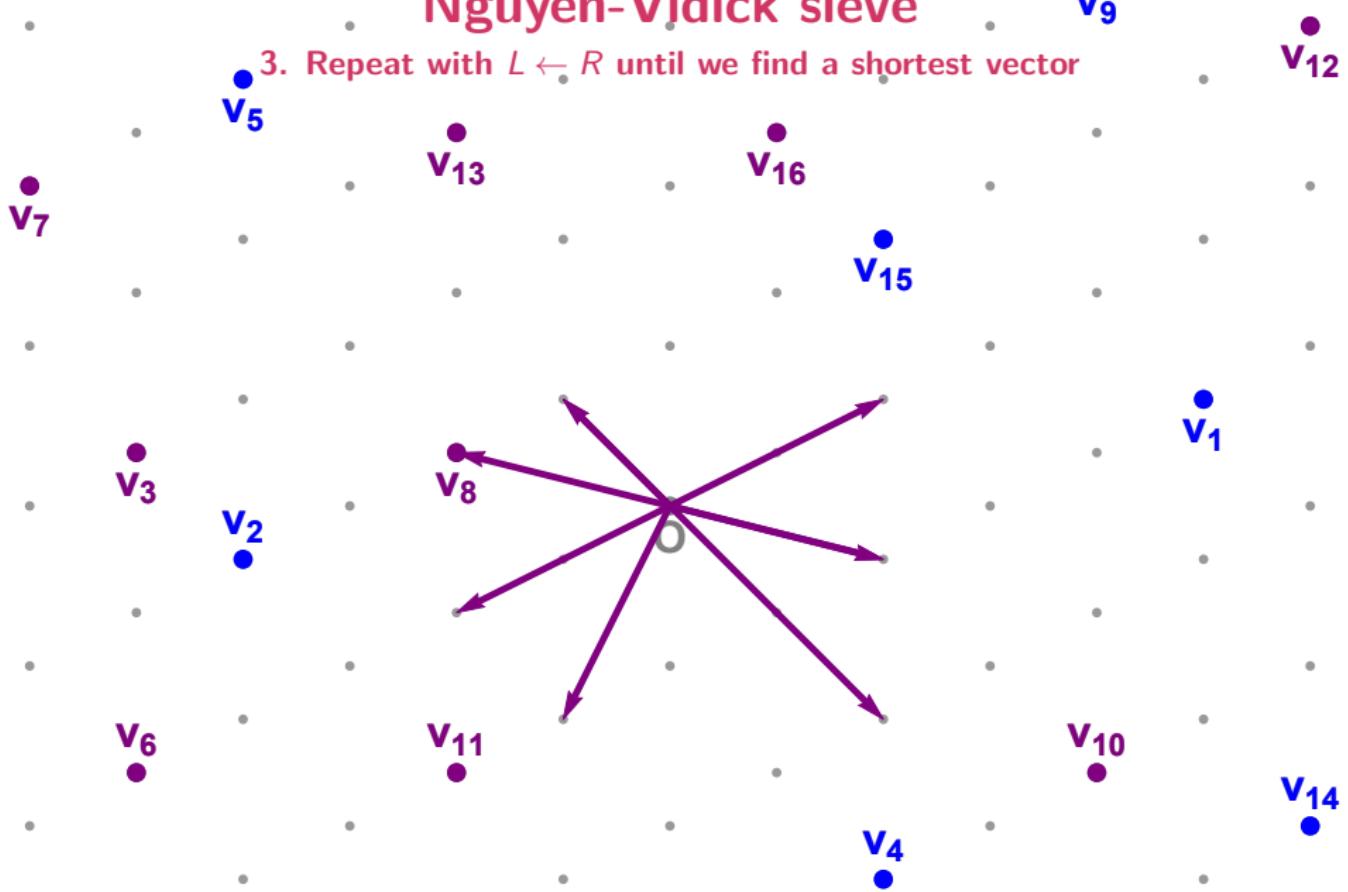
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Nguyen-Vidick sieve

Overview



Nguyen-Vidick sieve

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- Space complexity: $\sqrt{4/3}^n \approx 2^{0.21n+o(n)}$ vectors
 - ▶ Need $\sqrt{4/3}^n$ vectors to cover all corners of \mathbb{R}^n

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- Space complexity: $\sqrt{4/3}^n \approx 2^{0.21n+o(n)}$ vectors
 - ▶ Need $\sqrt{4/3}^n$ vectors to cover all corners of \mathbb{R}^n
- Time complexity: $(4/3)^n \approx 2^{0.42n+o(n)}$
 - ▶ Comparing a target vector to all centers: $2^{0.21n+o(n)}$
 - ▶ Repeating this for each list vector: $2^{0.21n+o(n)}$
 - ▶ Repeating the whole sieving procedure: $\text{poly}(n)$

Nguyen-Vidick sieve

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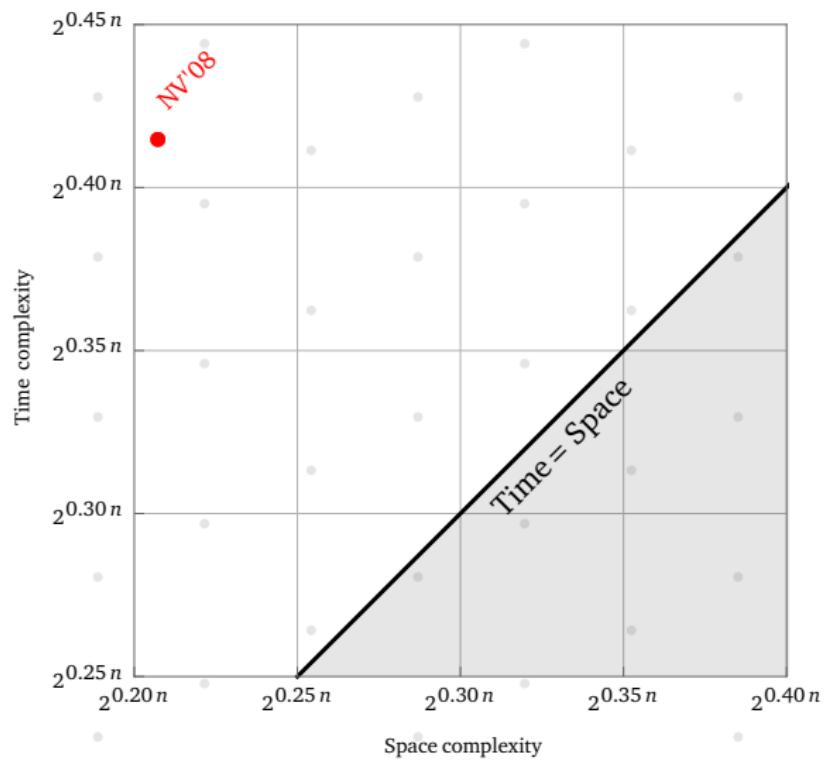
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Heuristic (Nguyen and Vidick, J. Math. Crypt. '08)

The NV-sieve runs in time $2^{0.42n+o(n)}$ and space $2^{0.21n+o(n)}$.

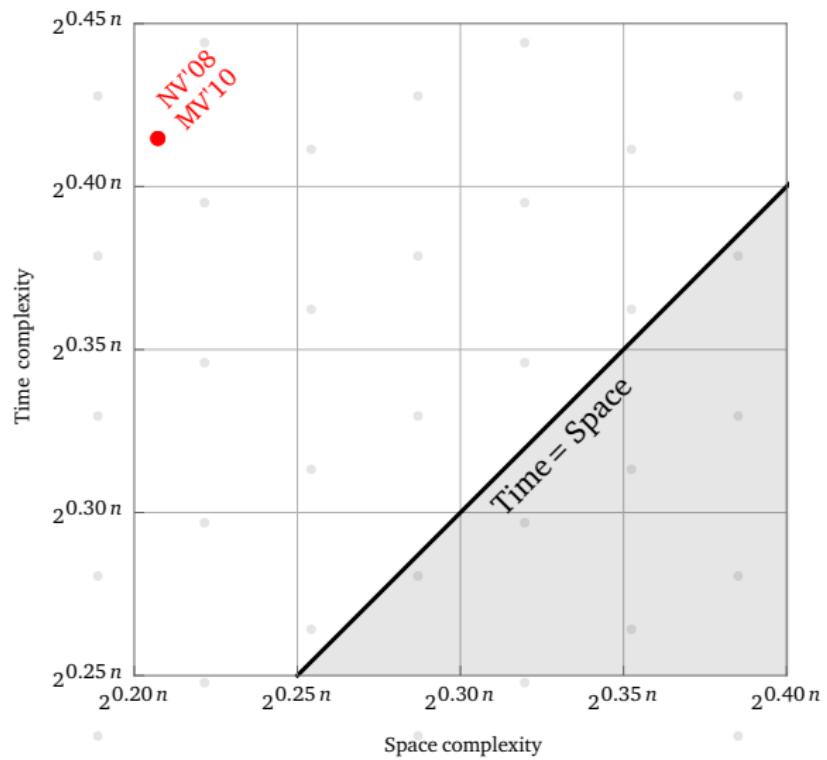
Nguyen-Vidick sieve

Space/time trade-off



GaussSieve

Space/time trade-off



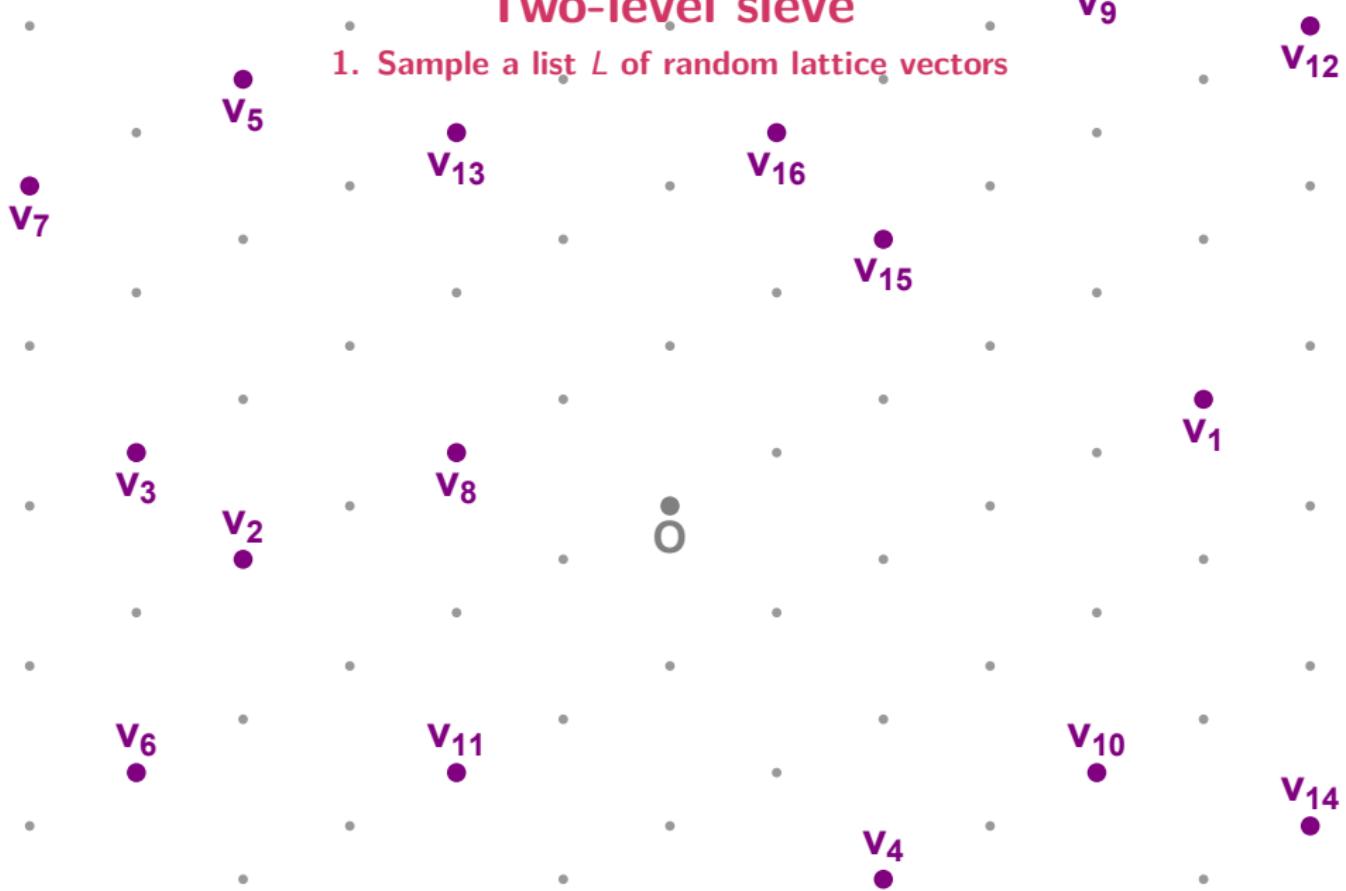
Two-level sieve

1. Sample a list L of random lattice vectors



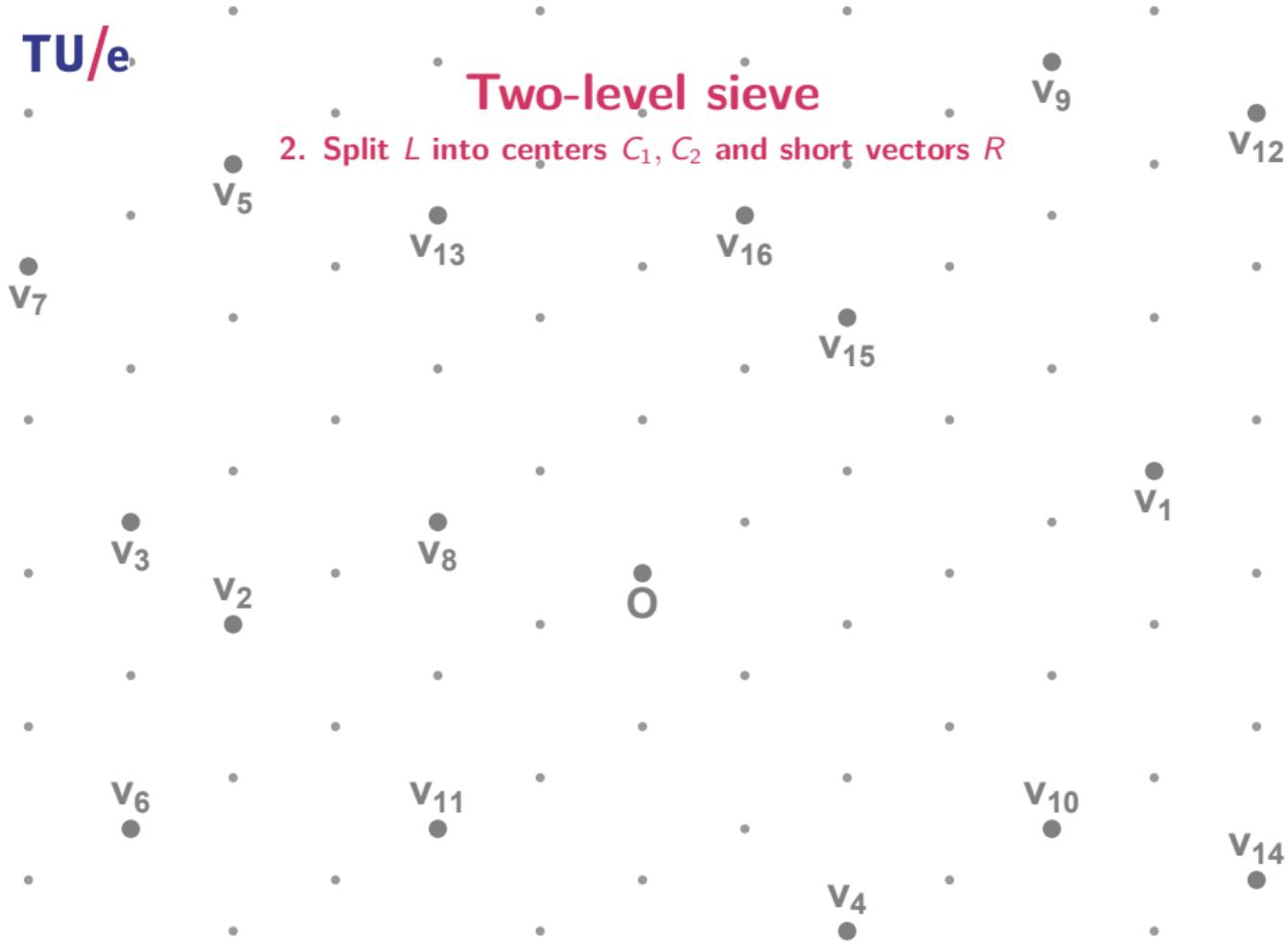
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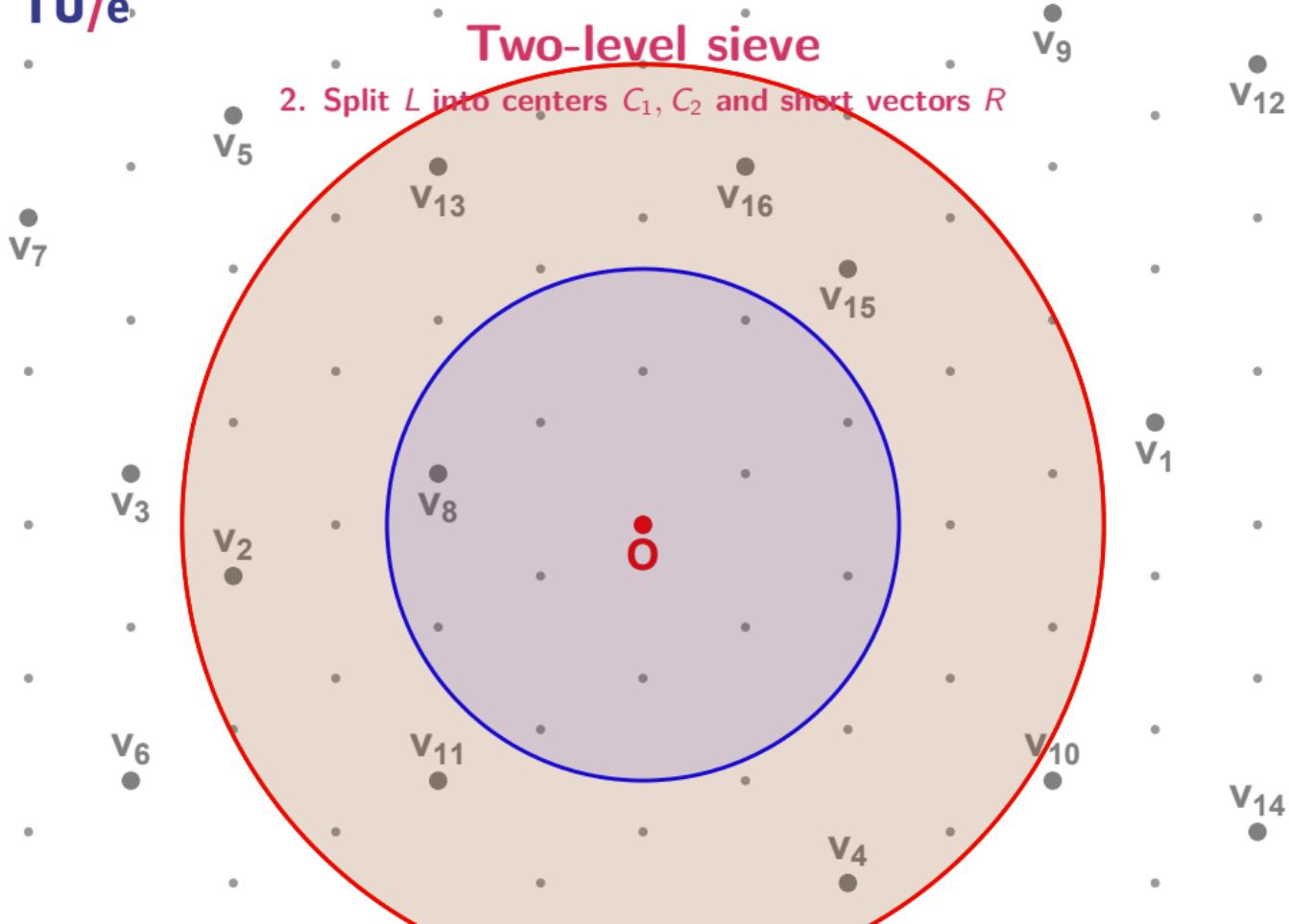
Two-level sieve

2. Split L into centers C_1, C_2 and short vectors R



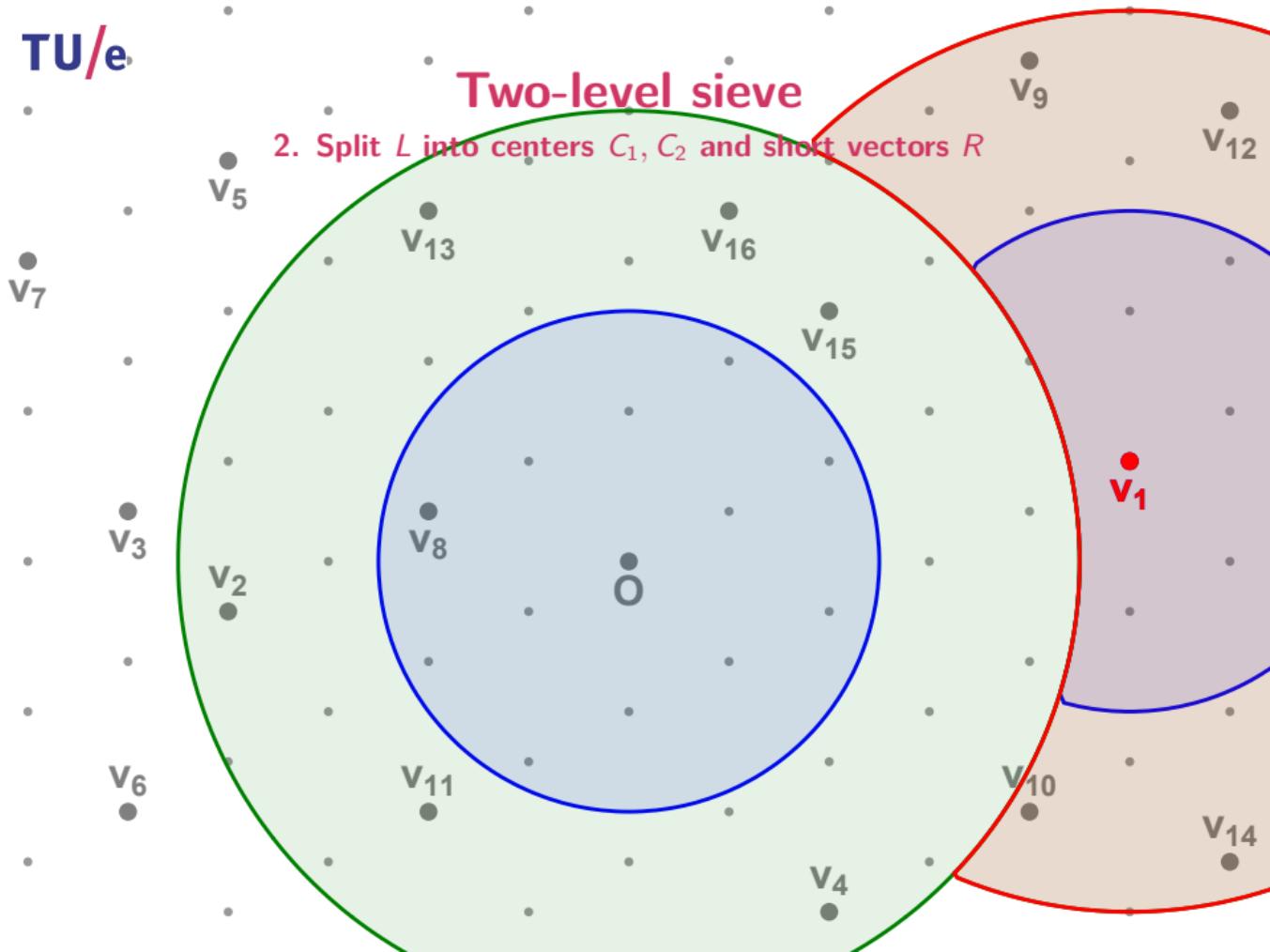
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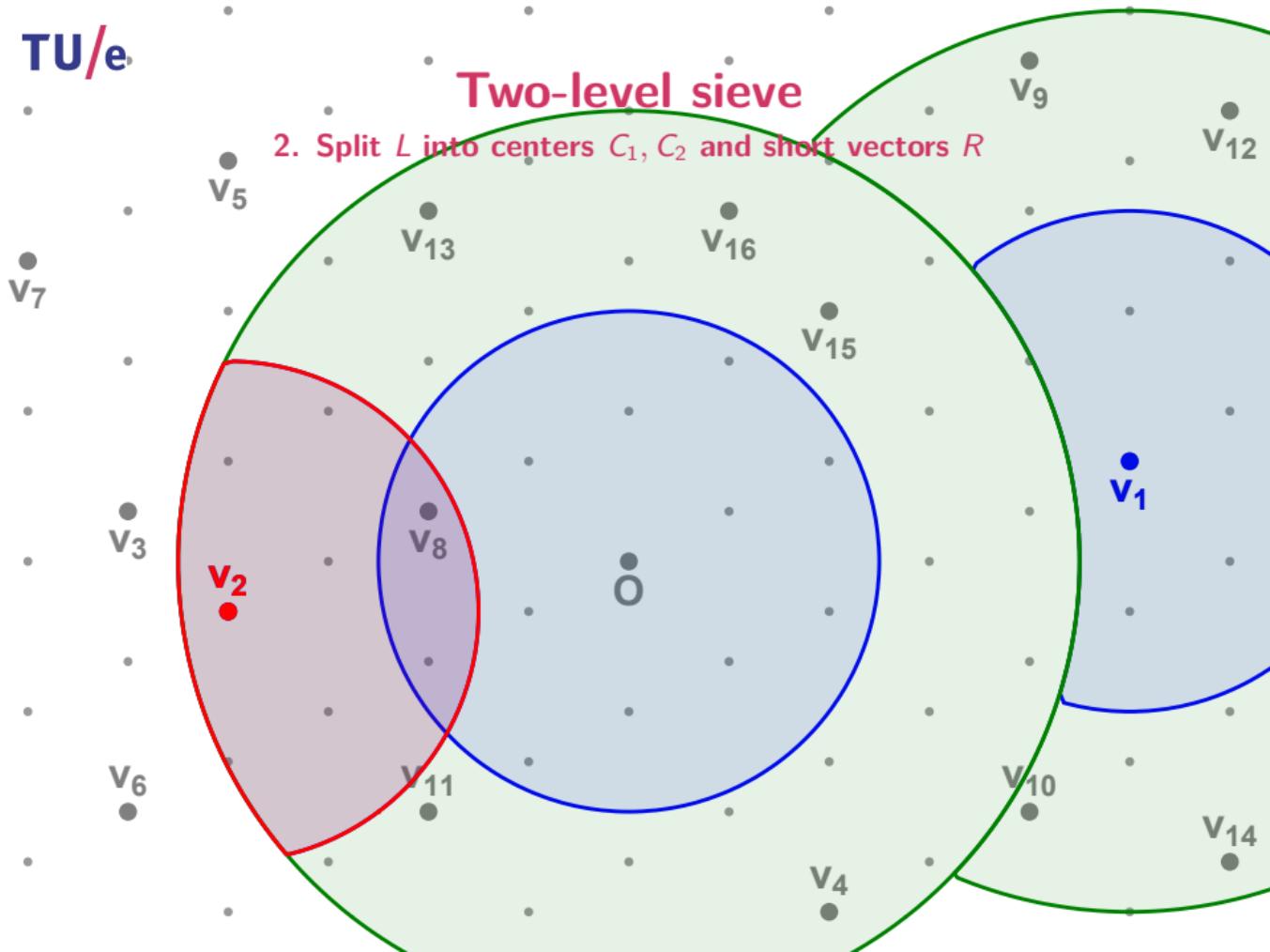
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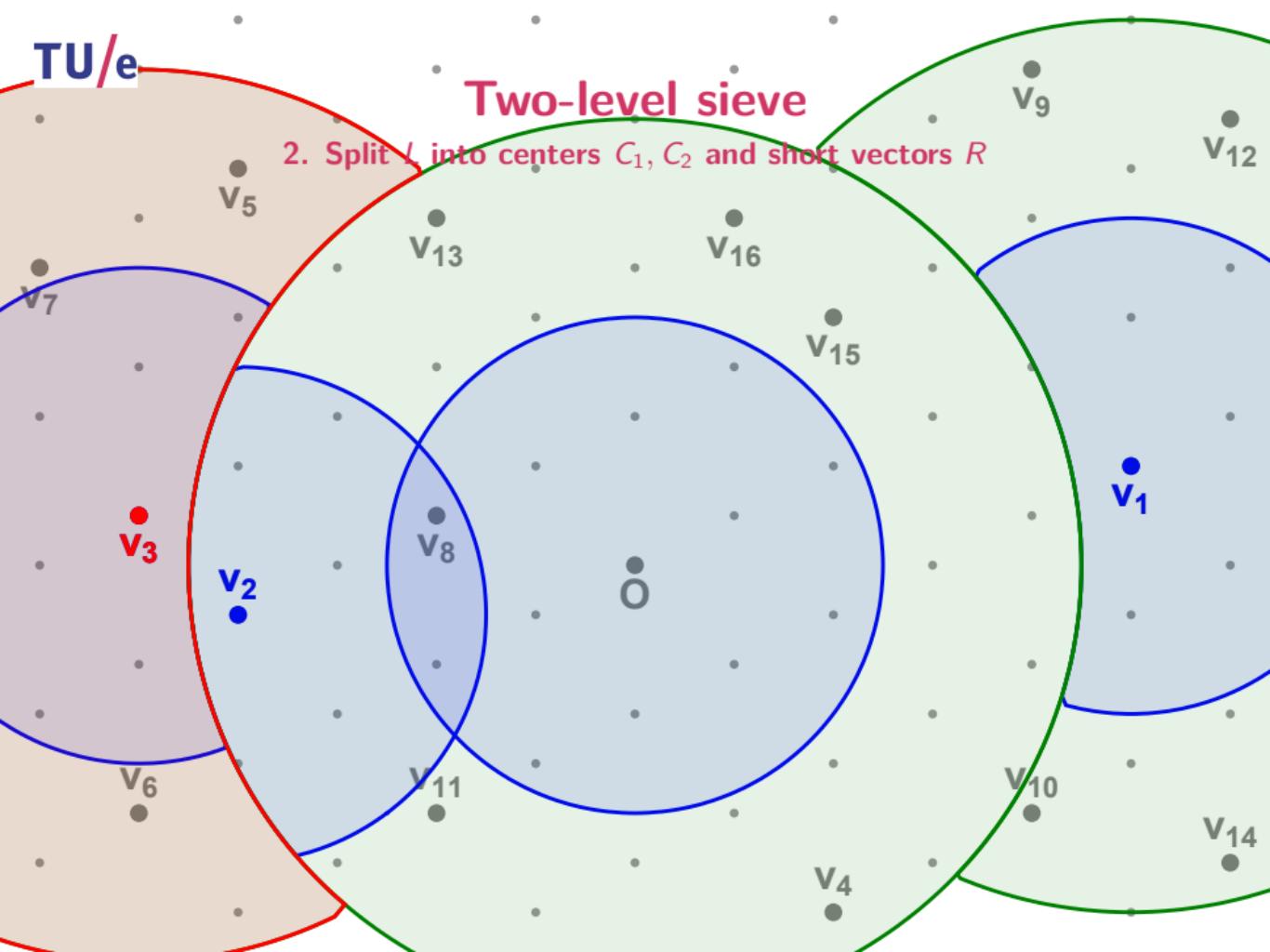
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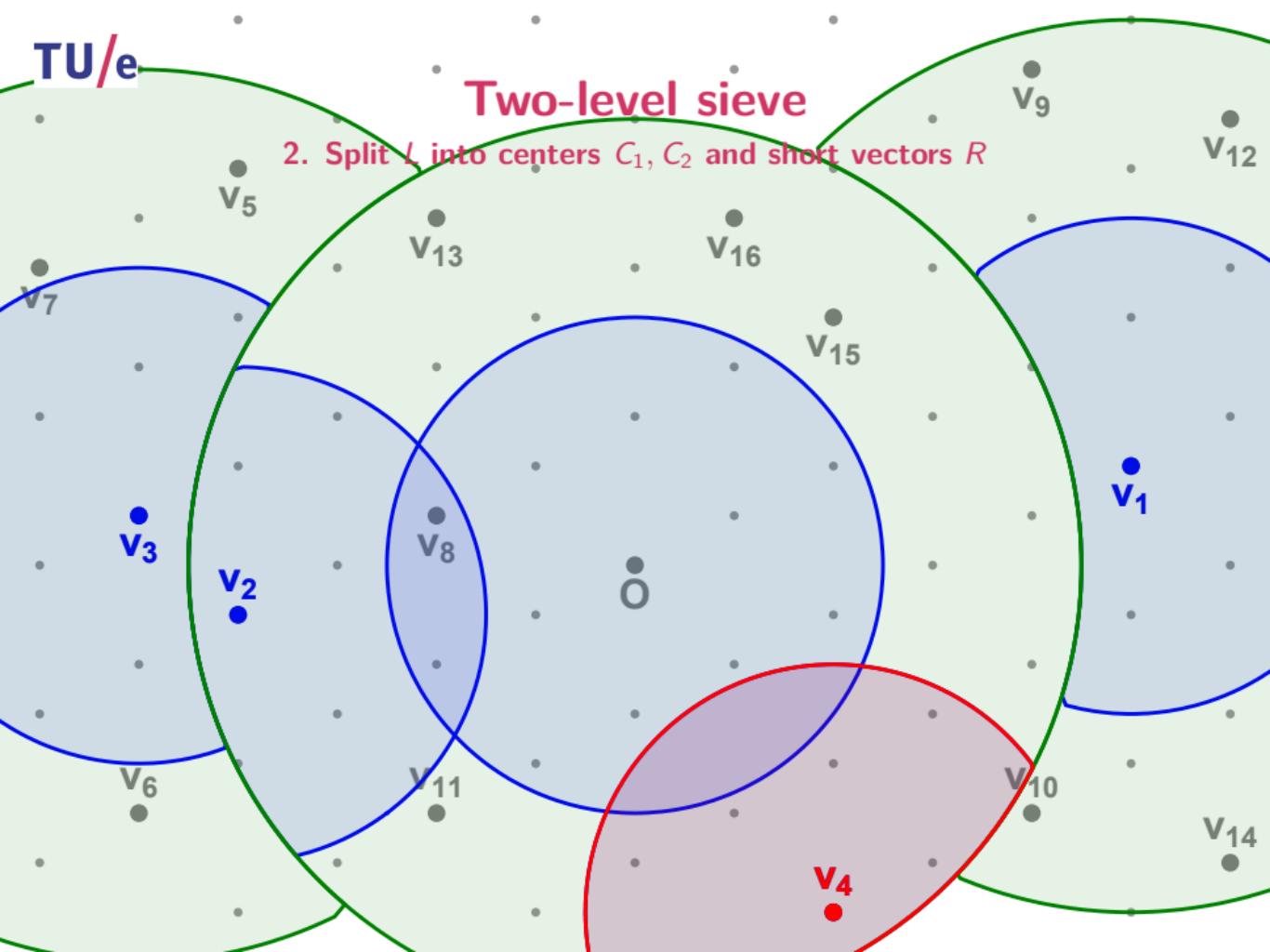
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2. Split \mathcal{V} into centers C_1, C_2 and short vectors R



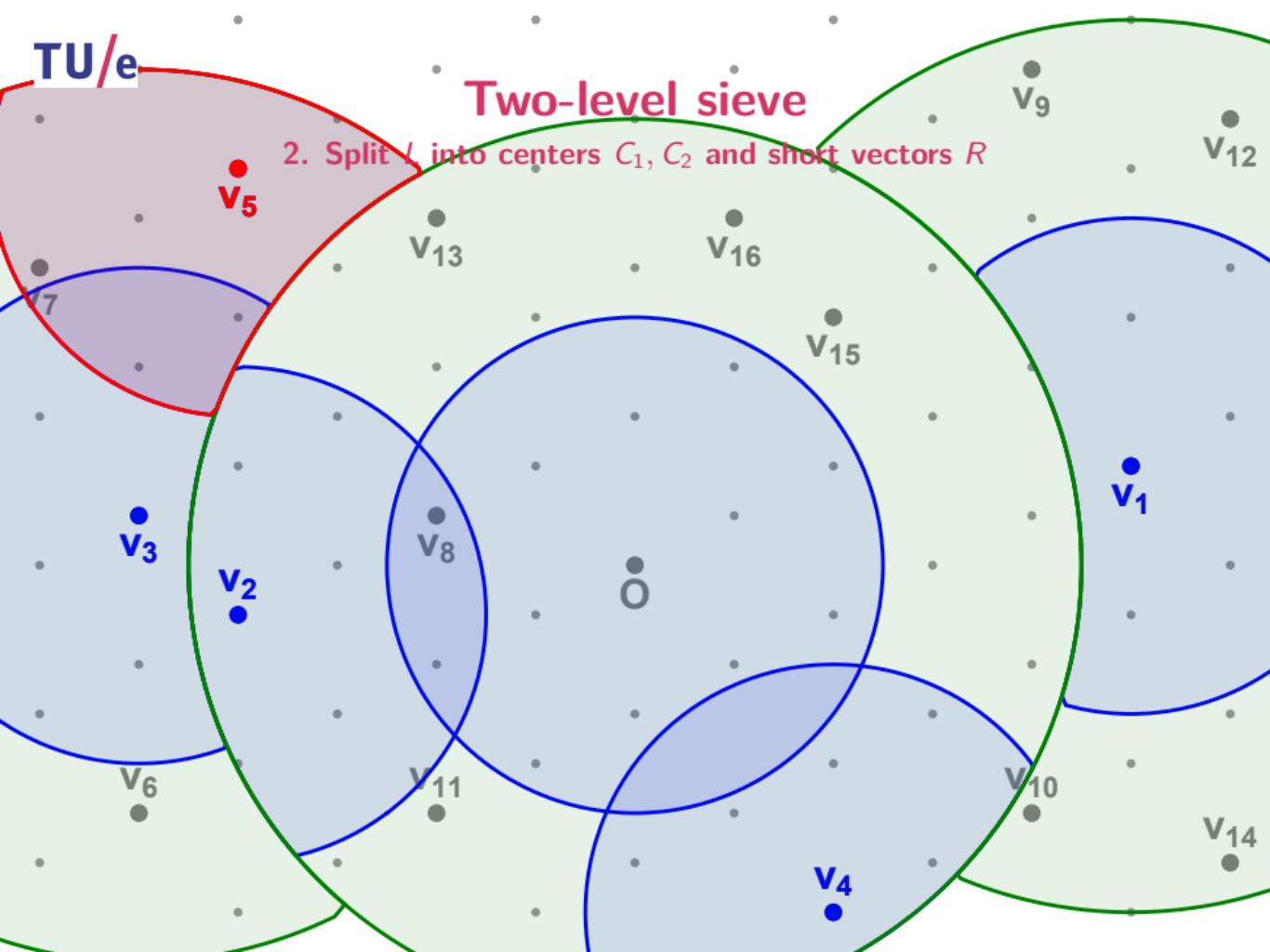
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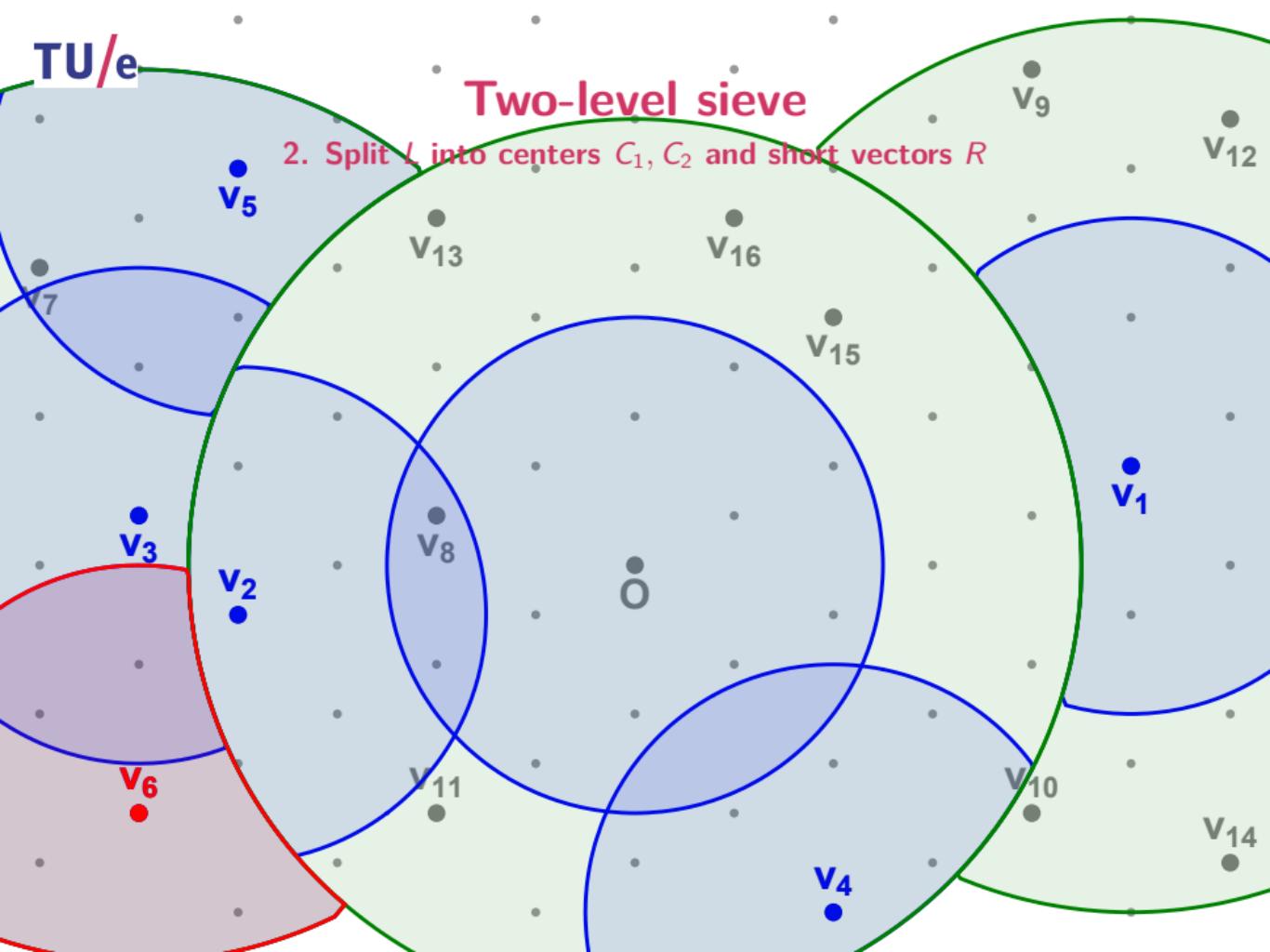
Two-level sieve

2. Split V into centers C_1, C_2 and short vectors R



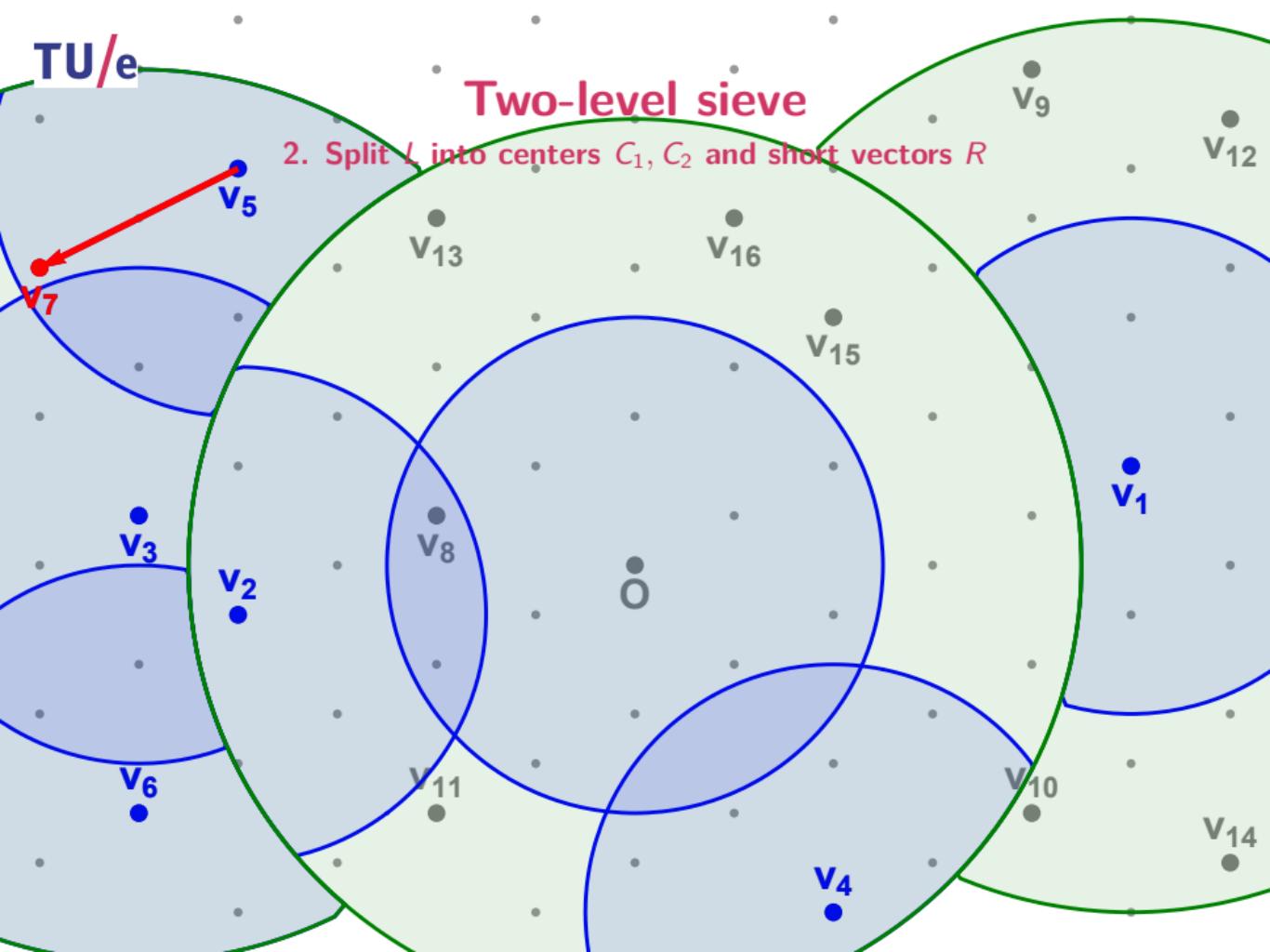
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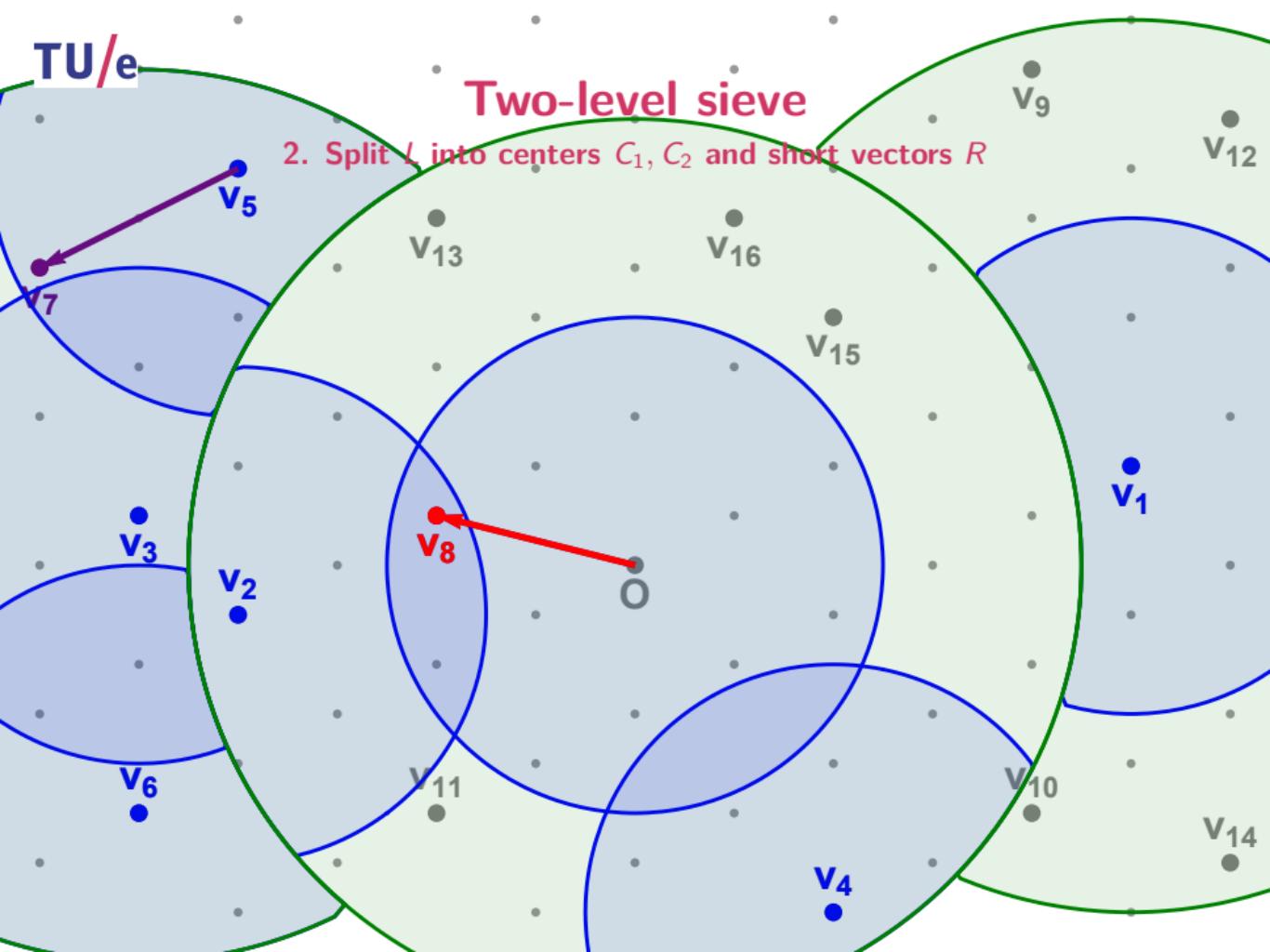
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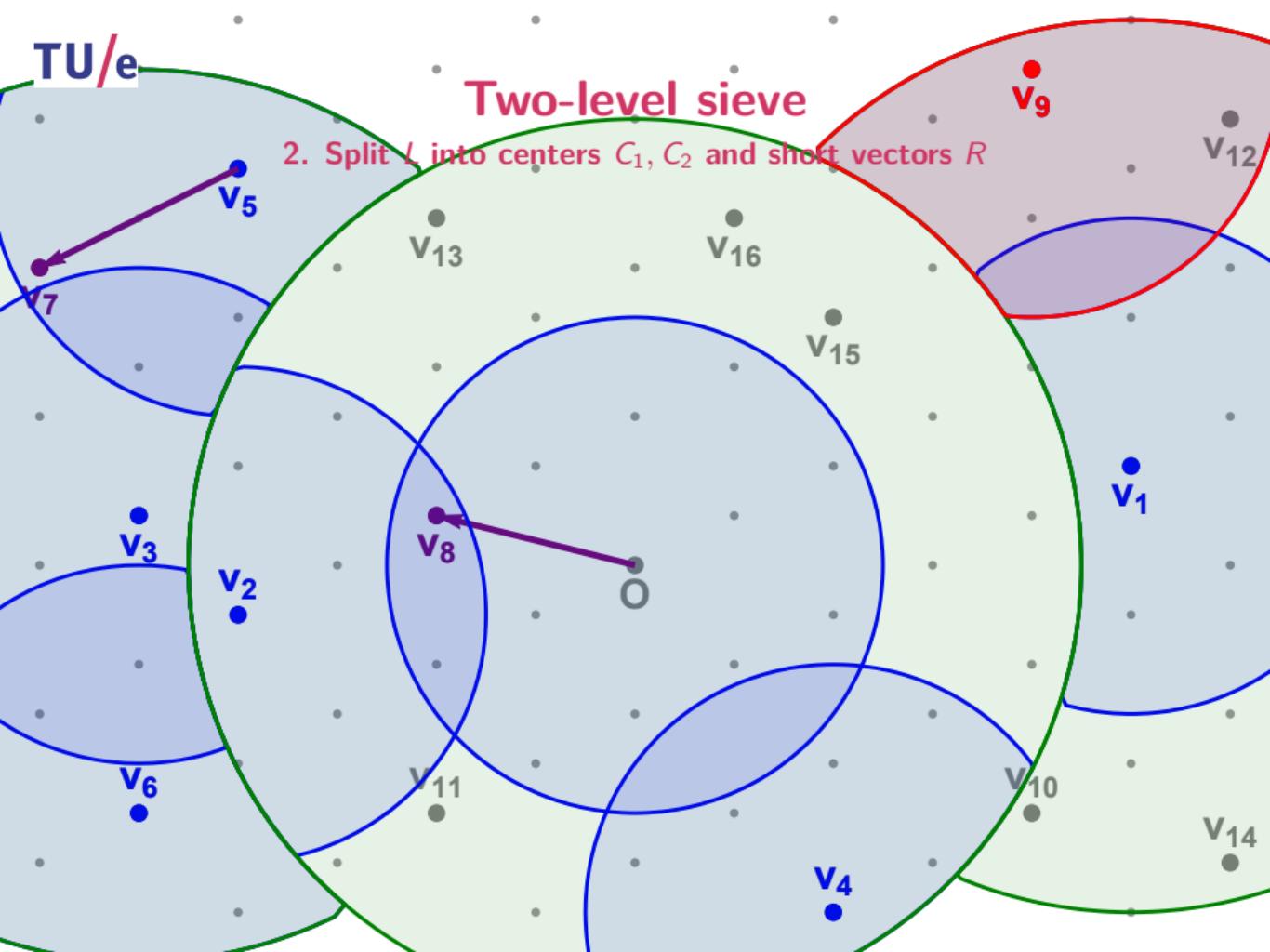
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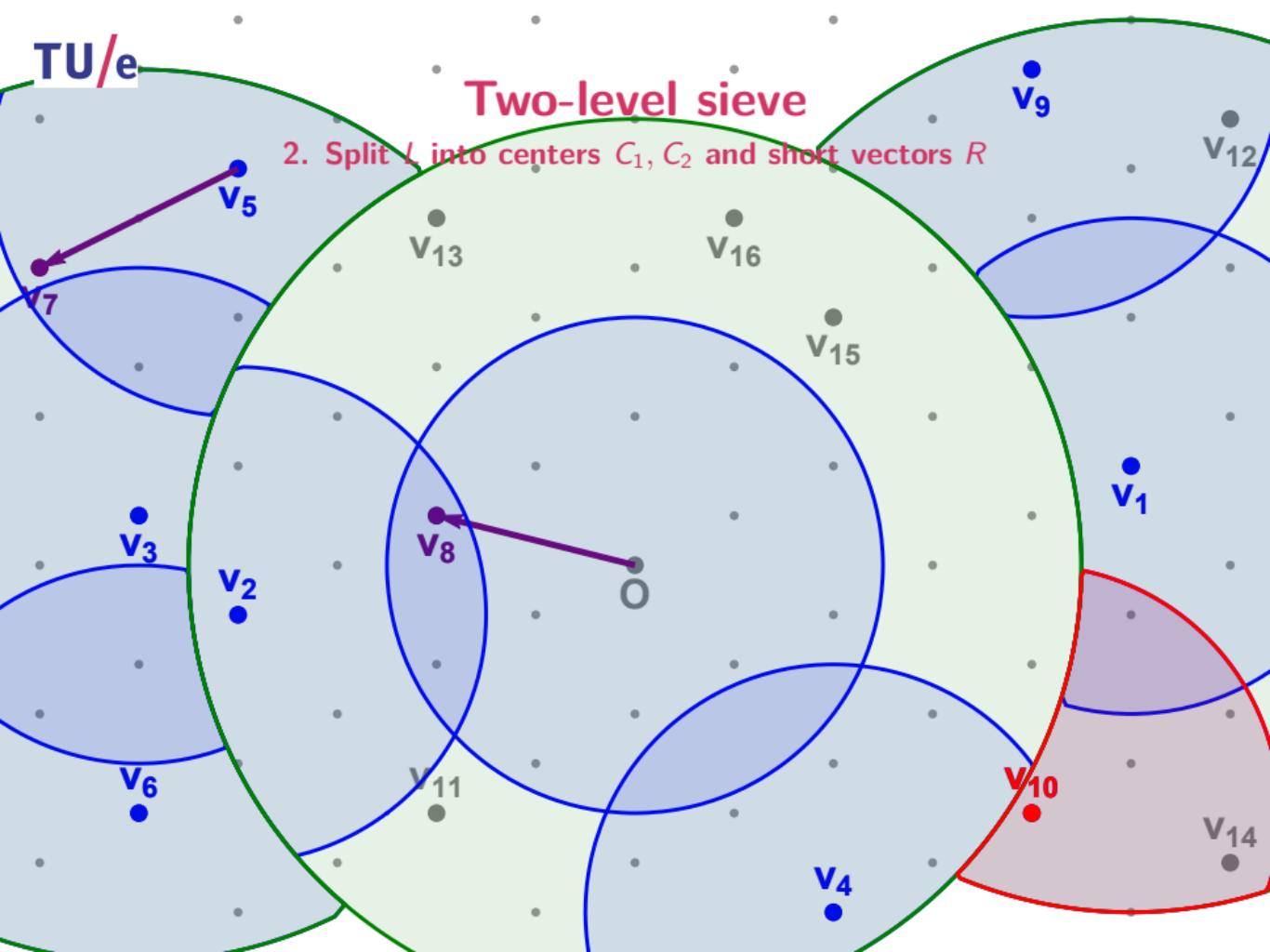
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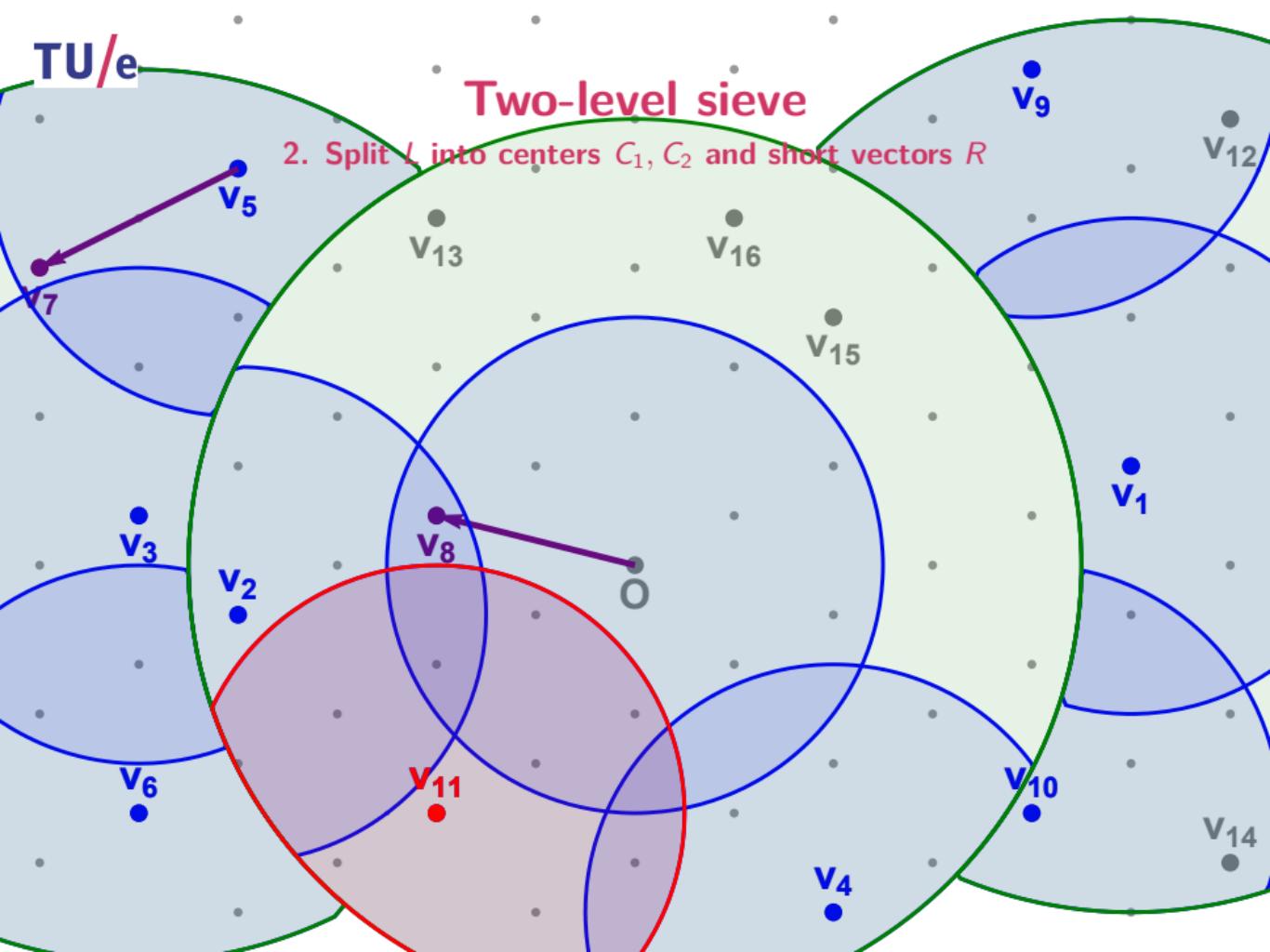
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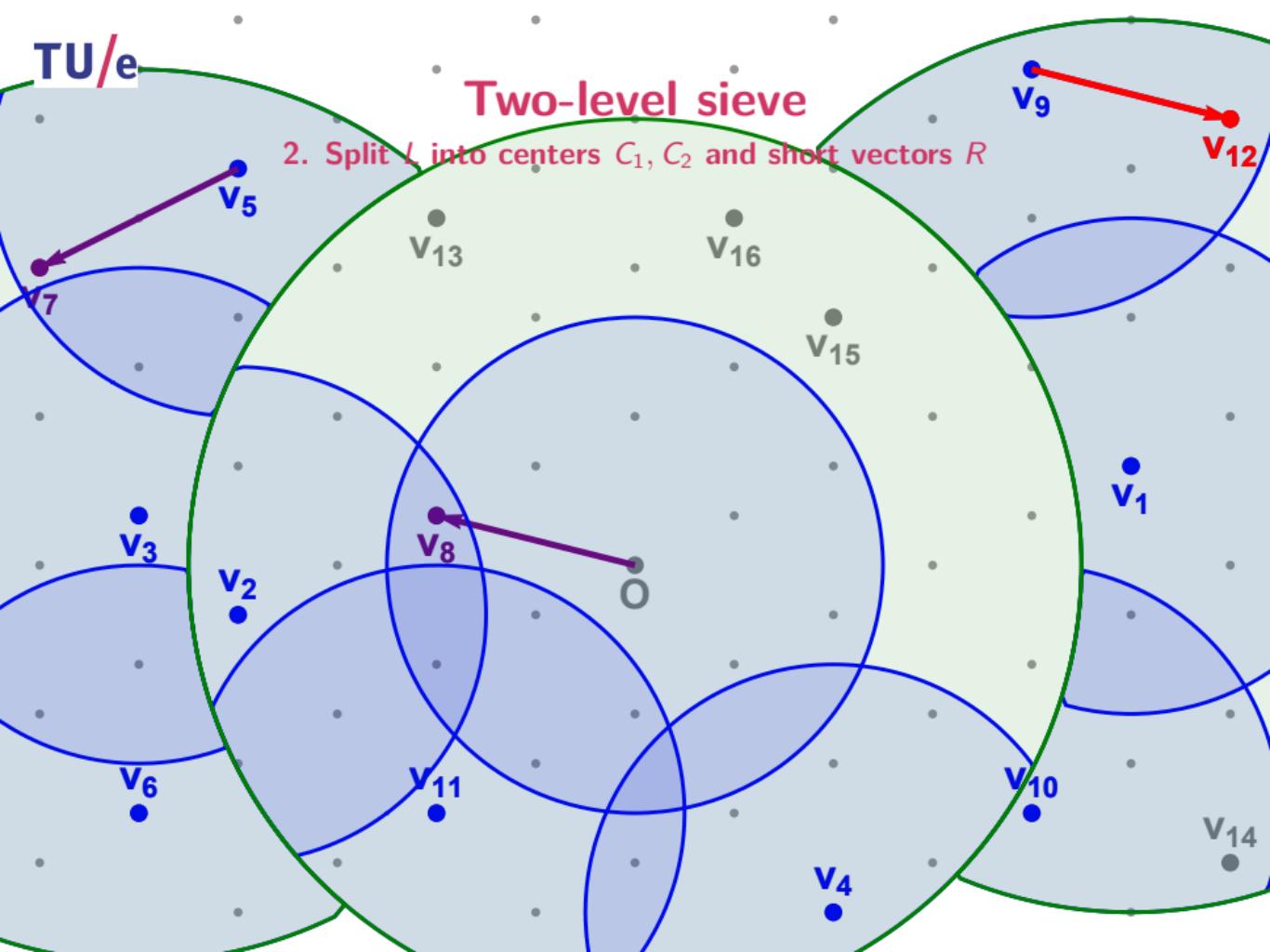
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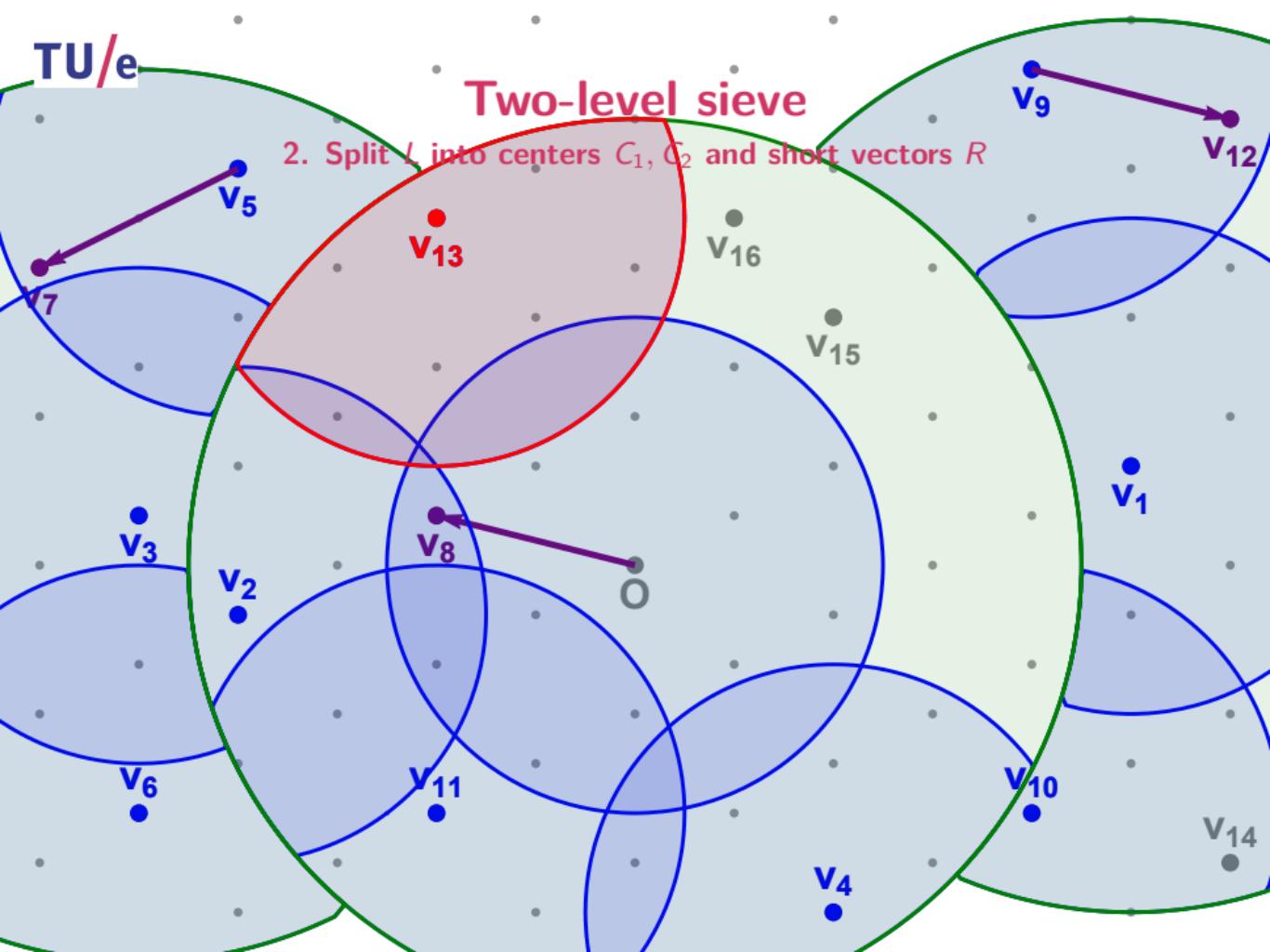
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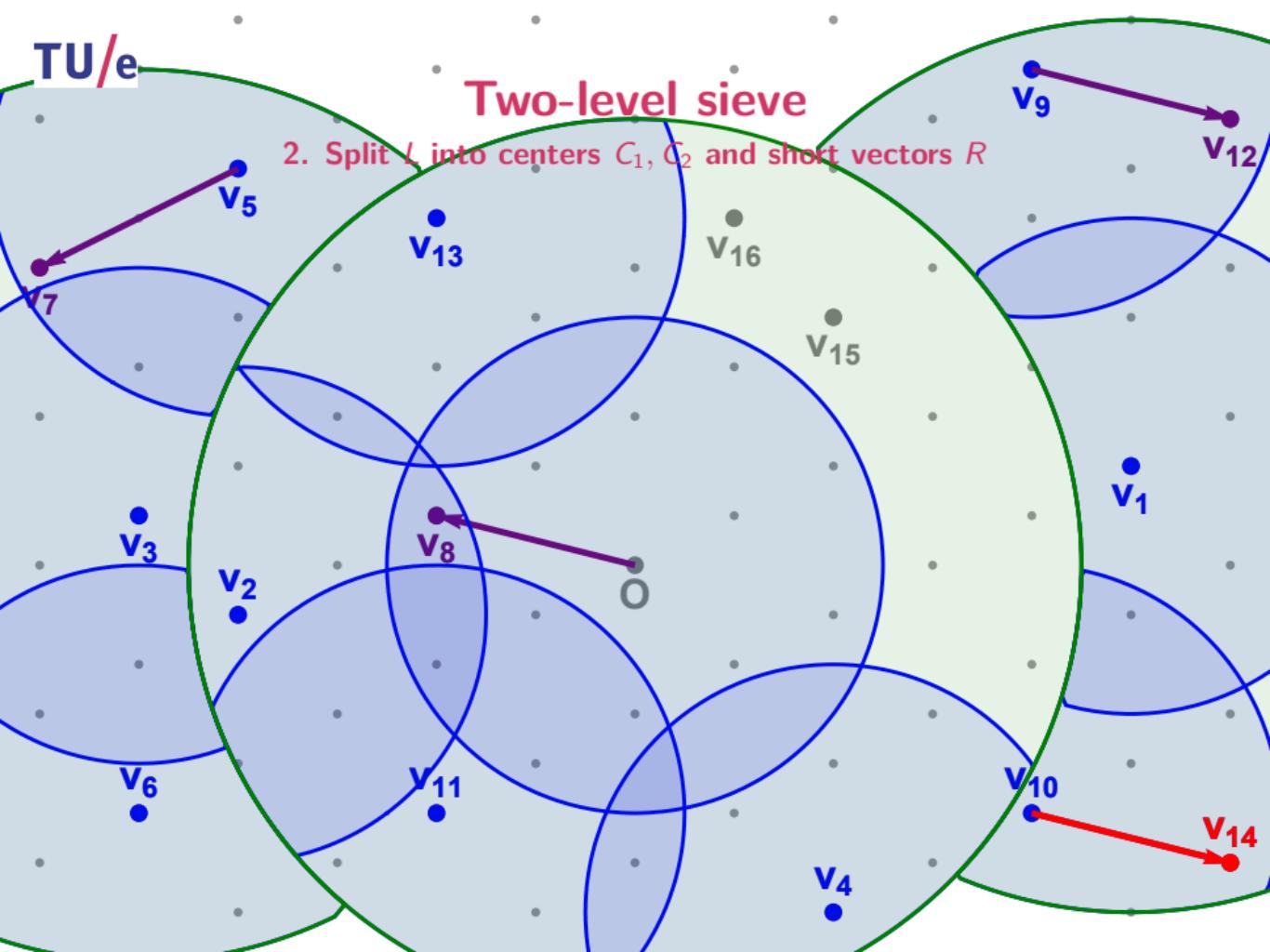
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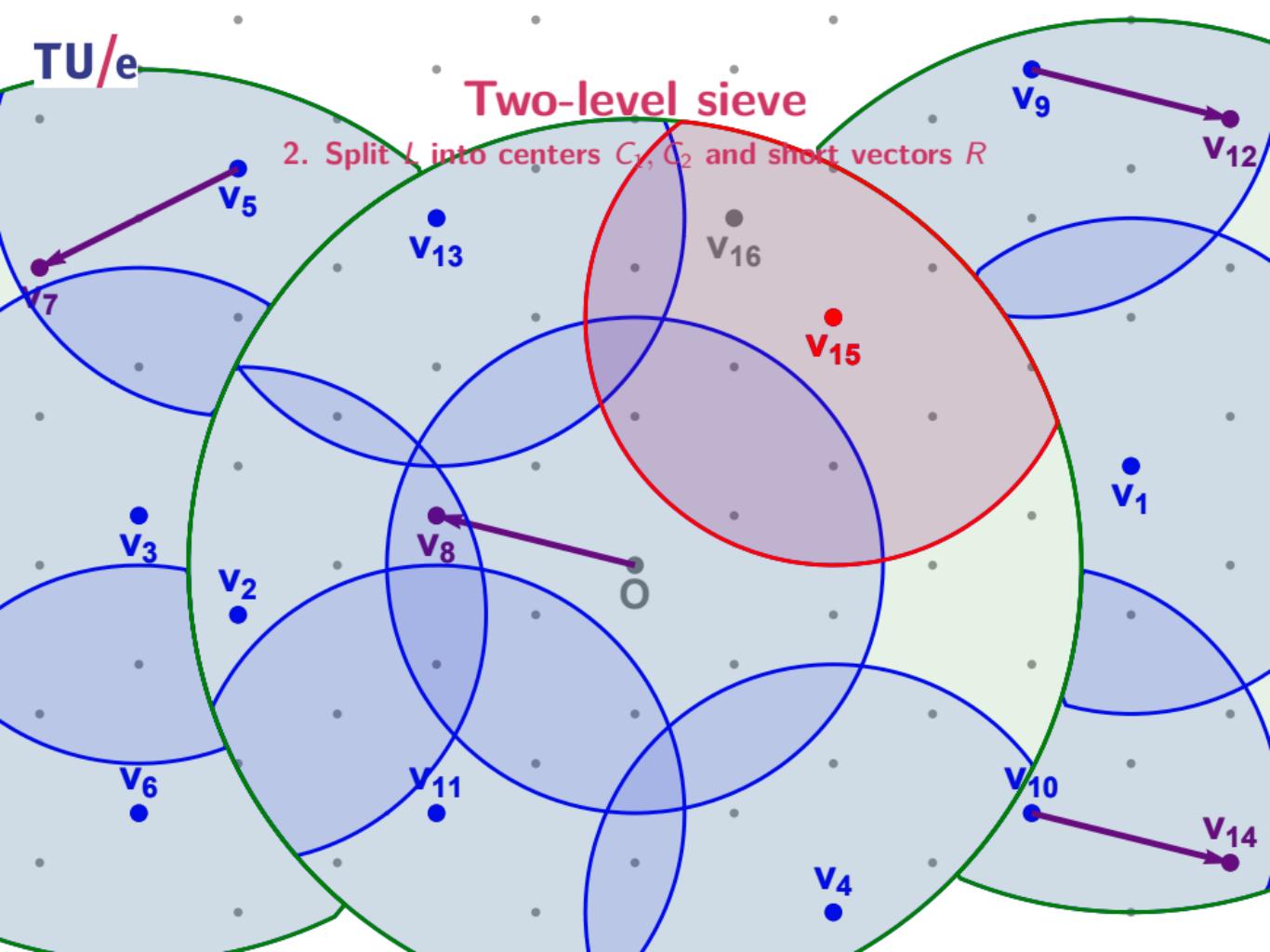
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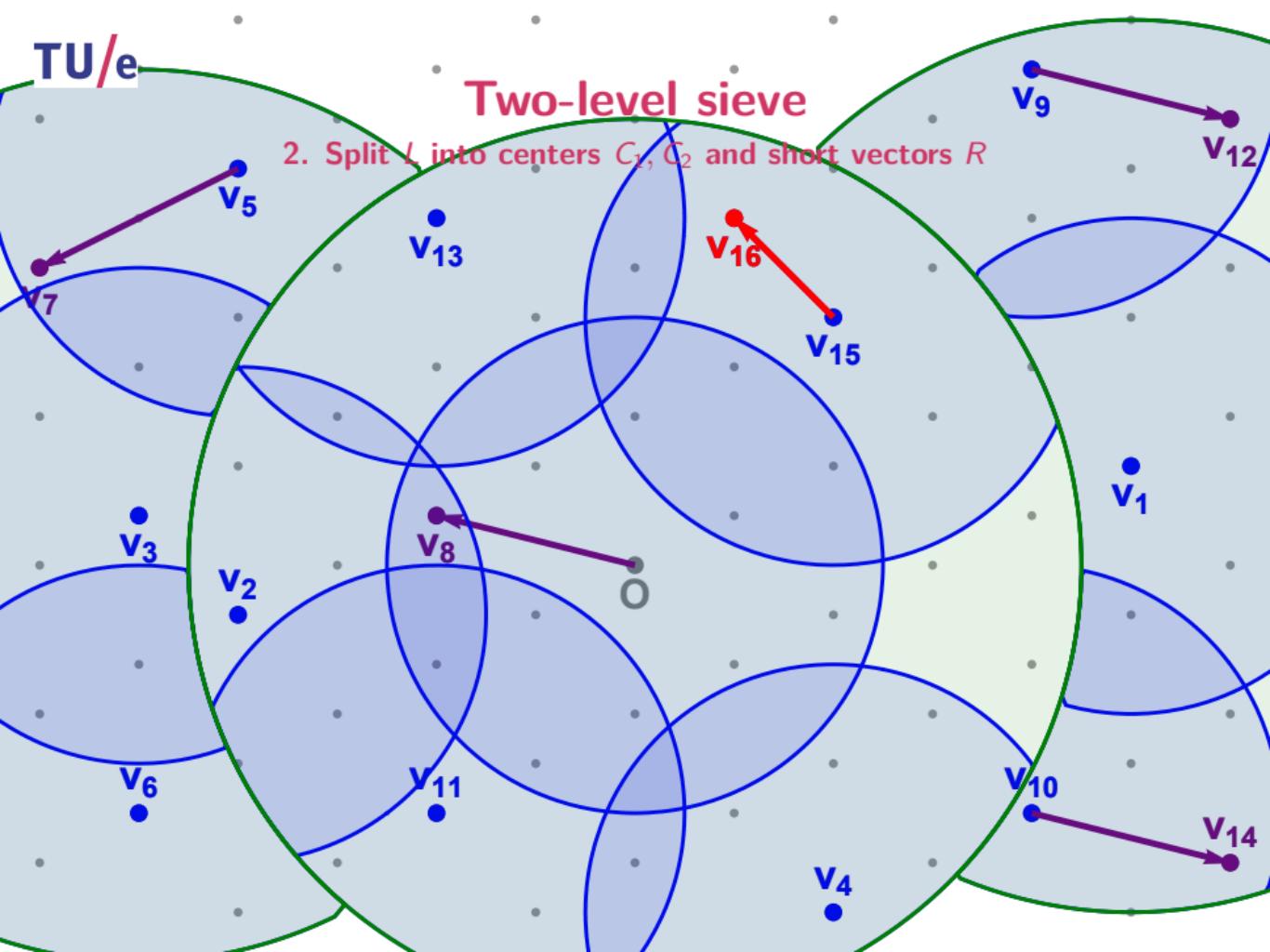
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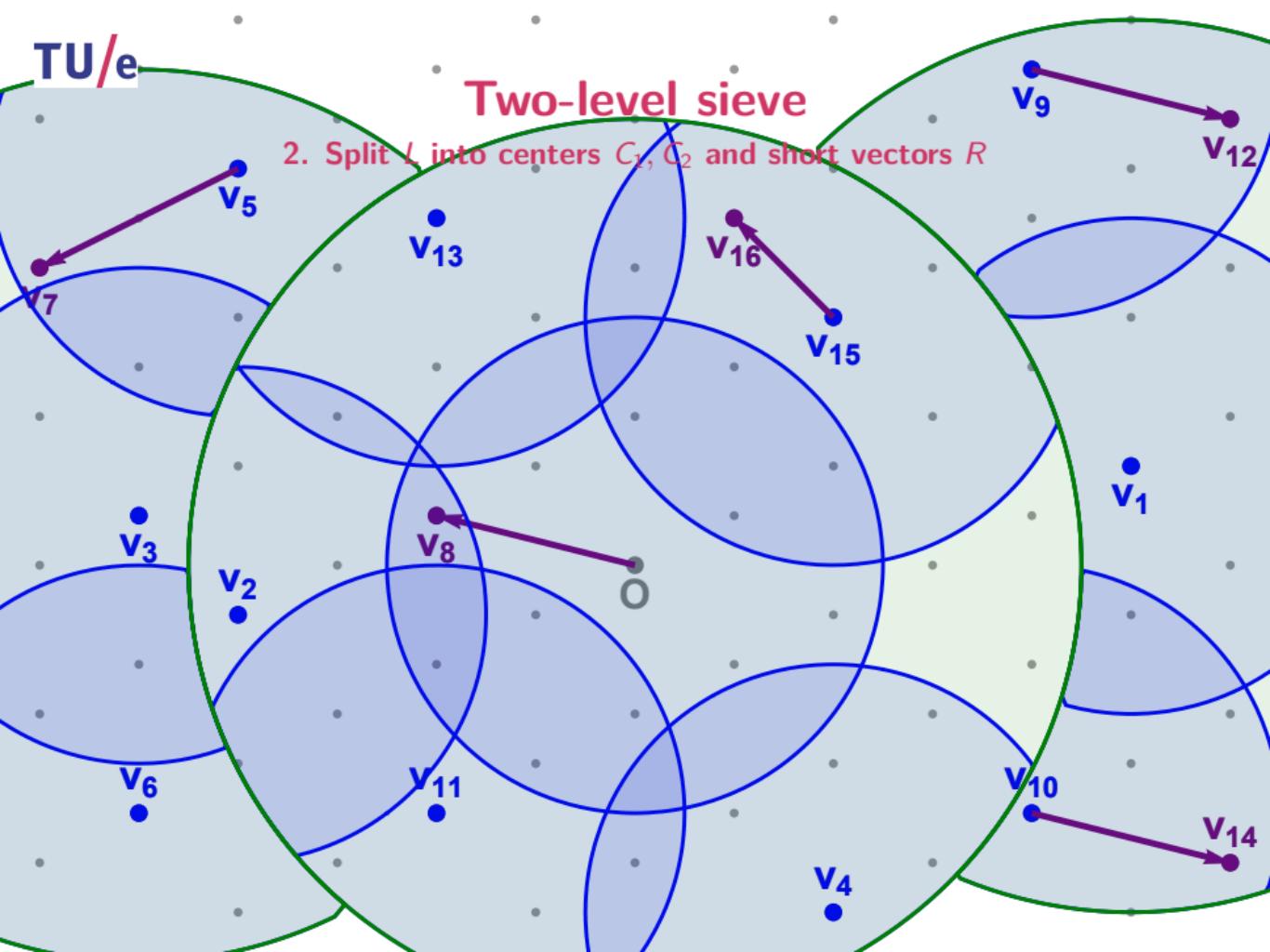
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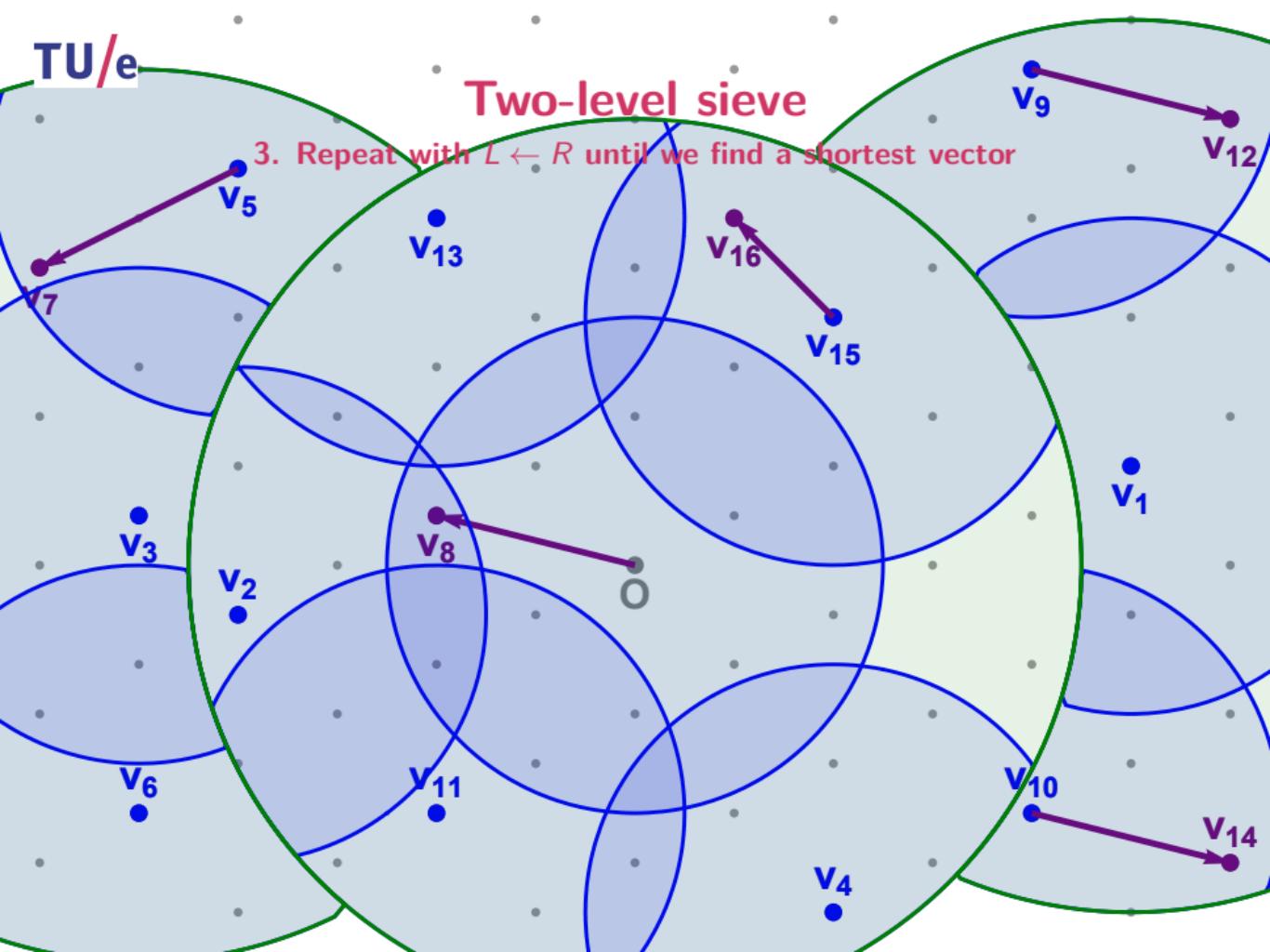
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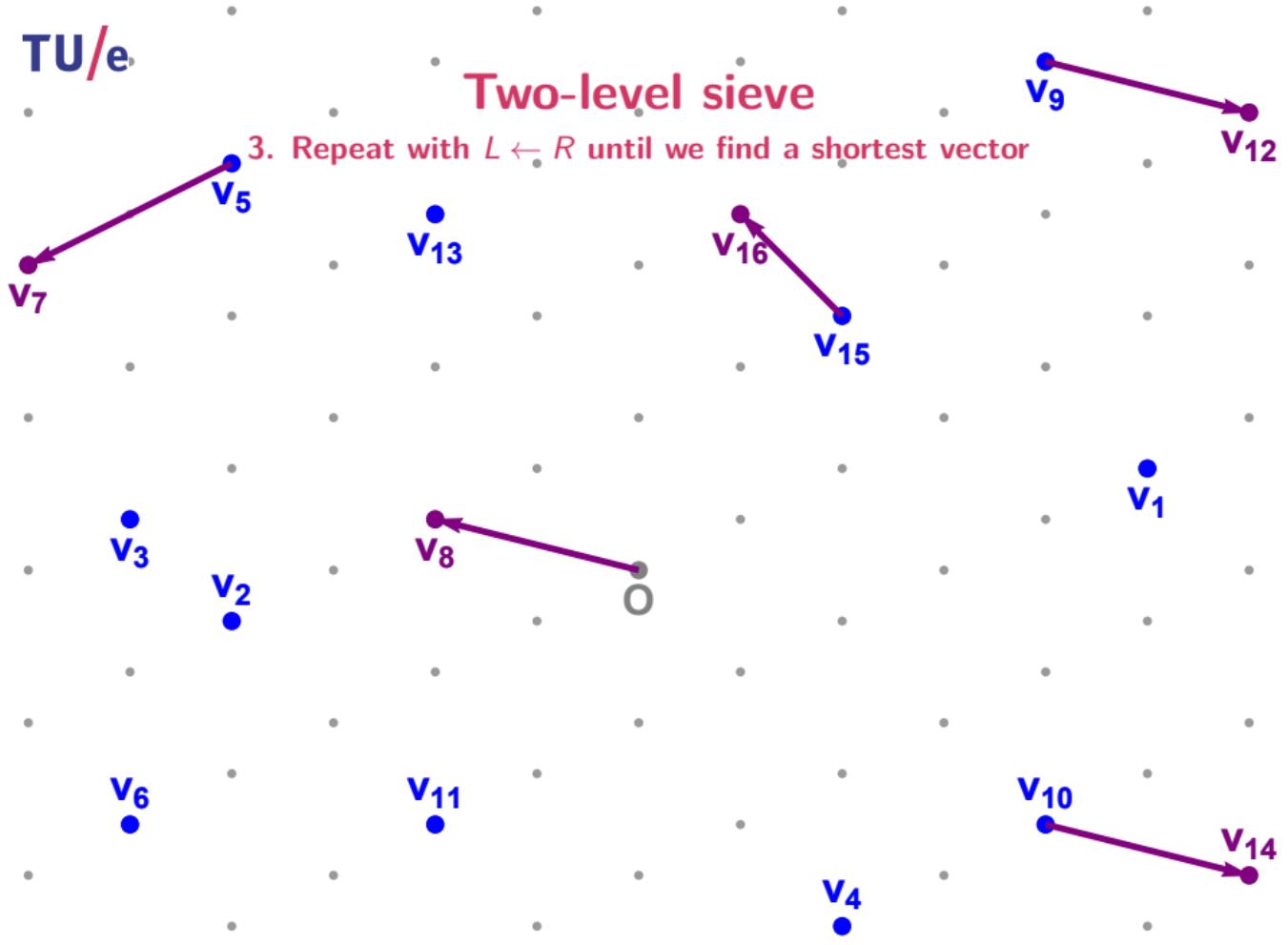
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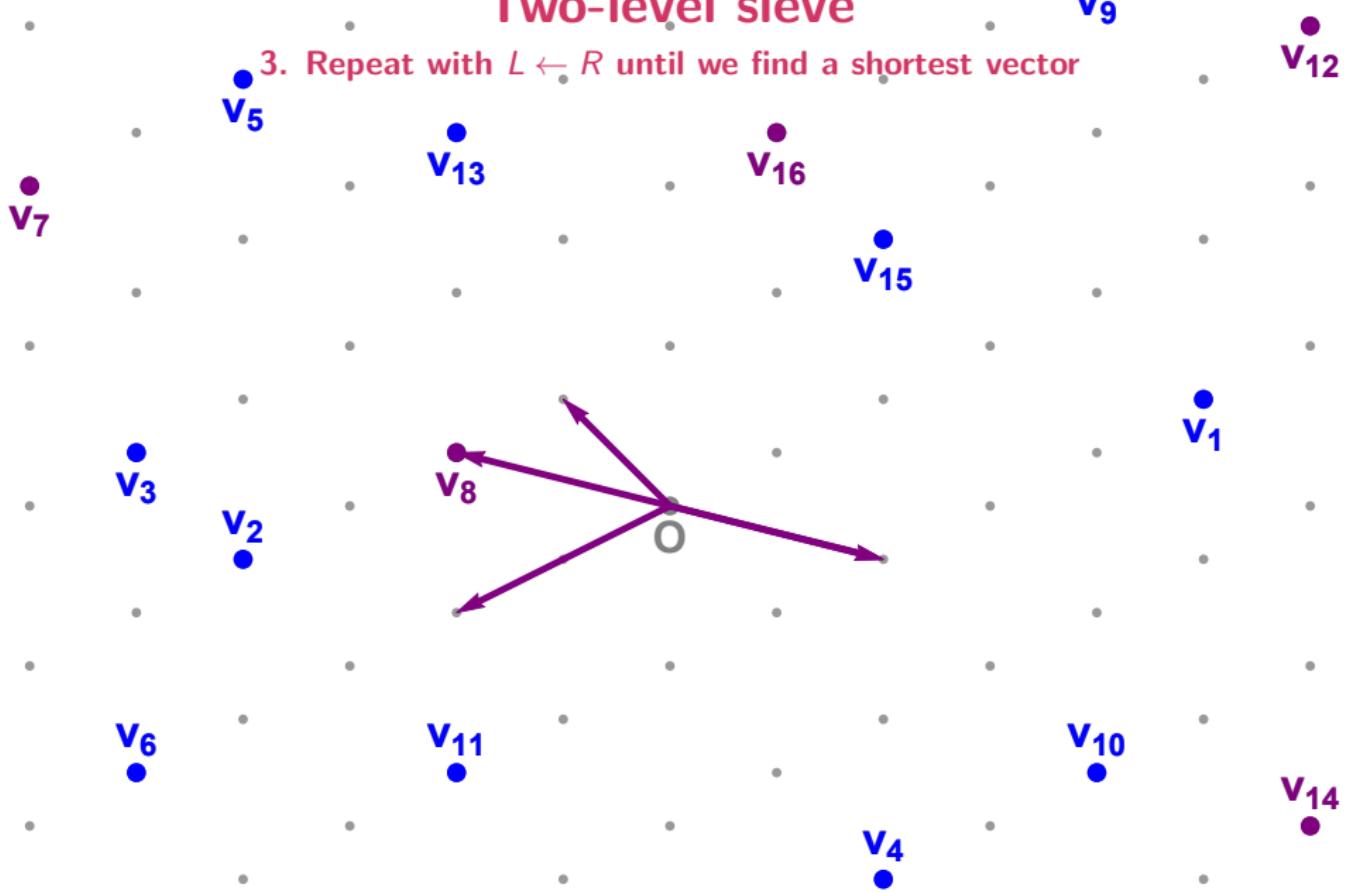
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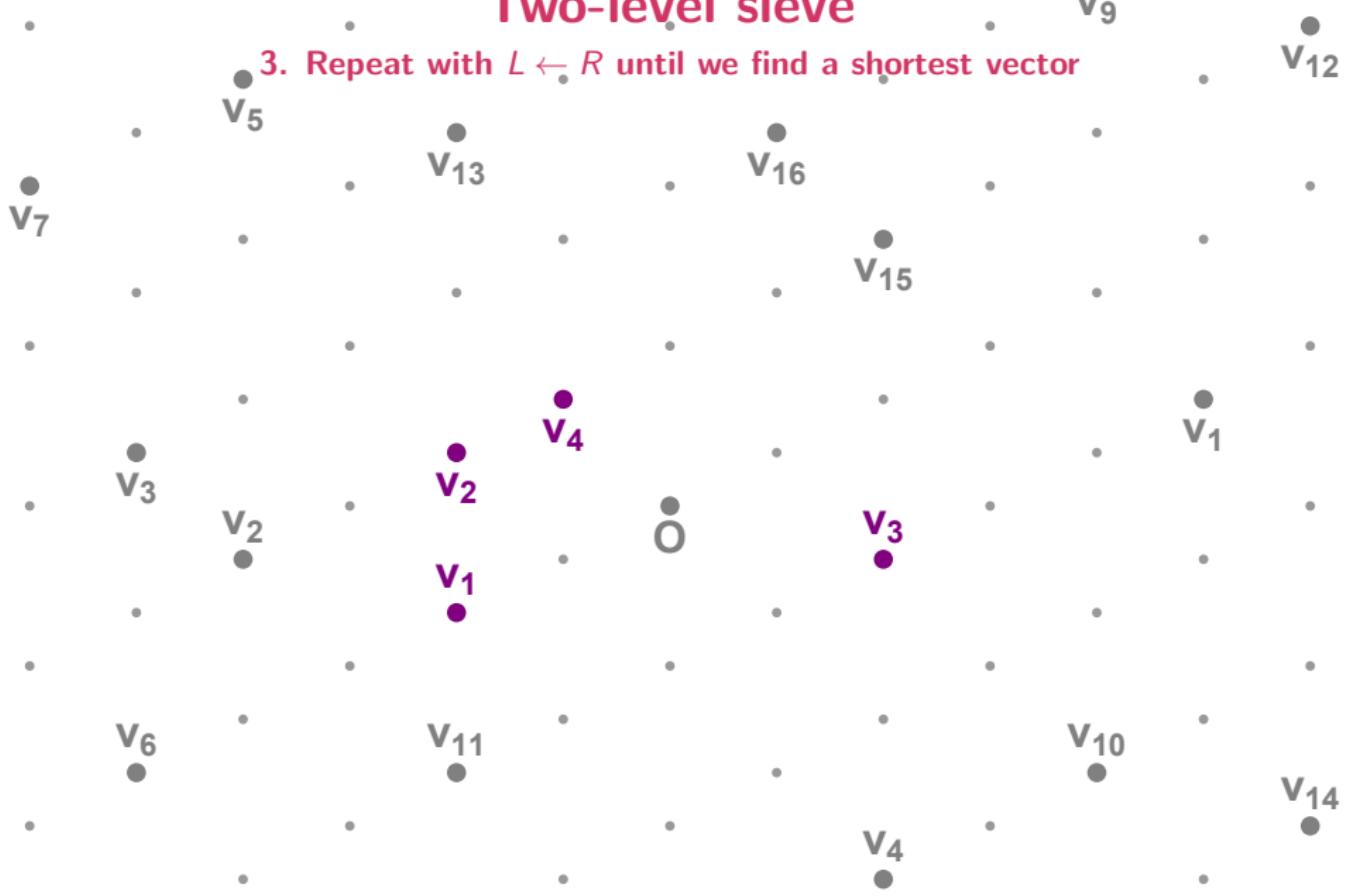
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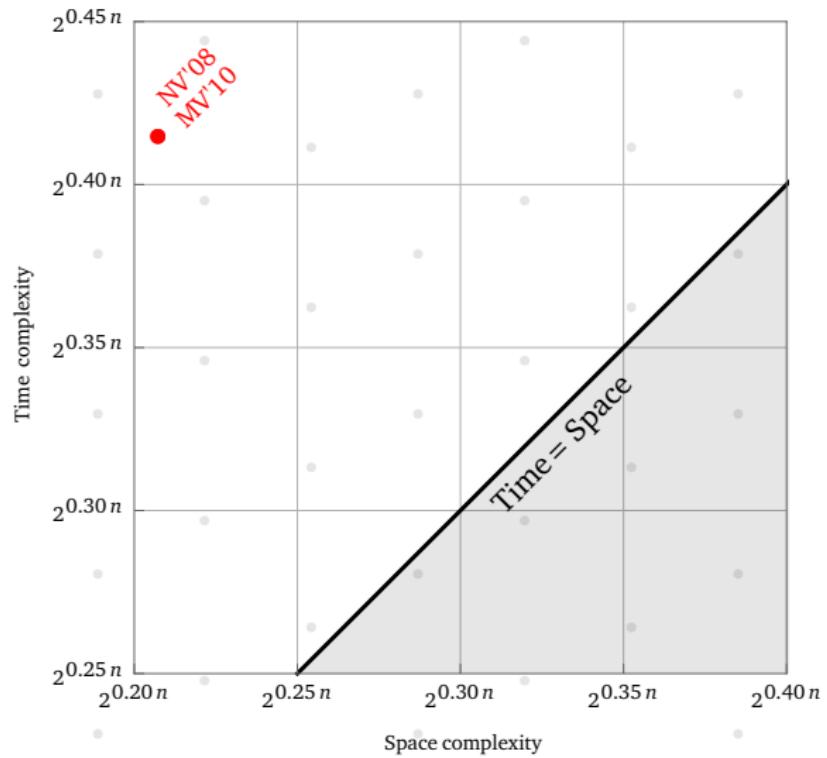
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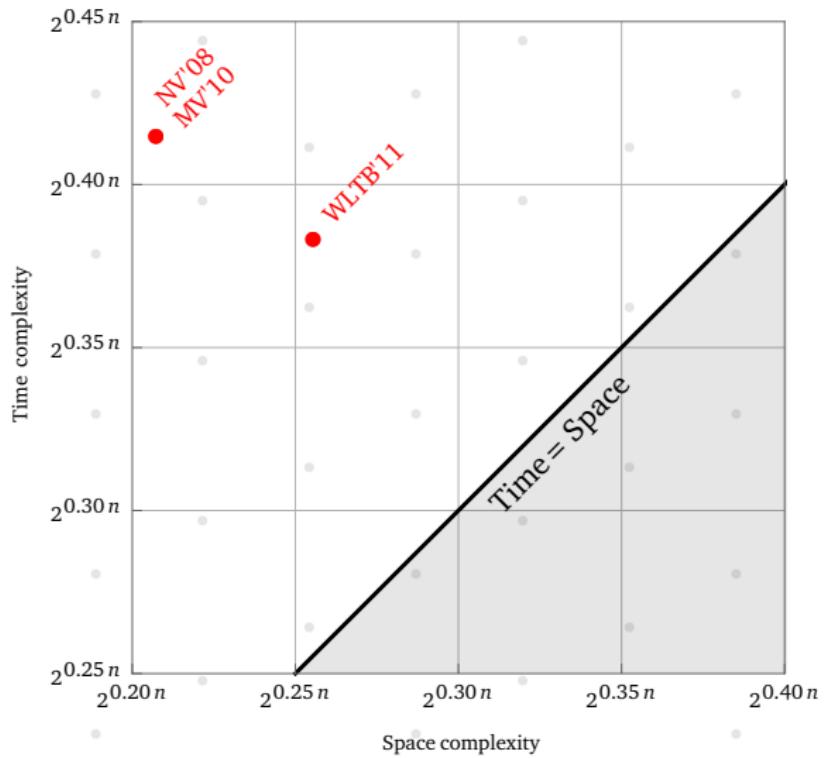
Two-level sieve

Space/time trade-off



Two-level sieve

Space/time trade-off



Three-level sieve

Overview

Heuristic (Nguyen and Vidick, J. Math. Crypt. '08)

The one-level sieve runs in time $2^{0.4150n}$ and space $2^{0.2075n}$.

Three-level sieve

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The one-level sieve runs in time $2^{0.4150n}$ and space $2^{0.2075n}$.

Heuristic (Wang et al., ASIACCS'11)

The two-level sieve runs in time $2^{0.3836n}$ and space $2^{0.2557n}$.

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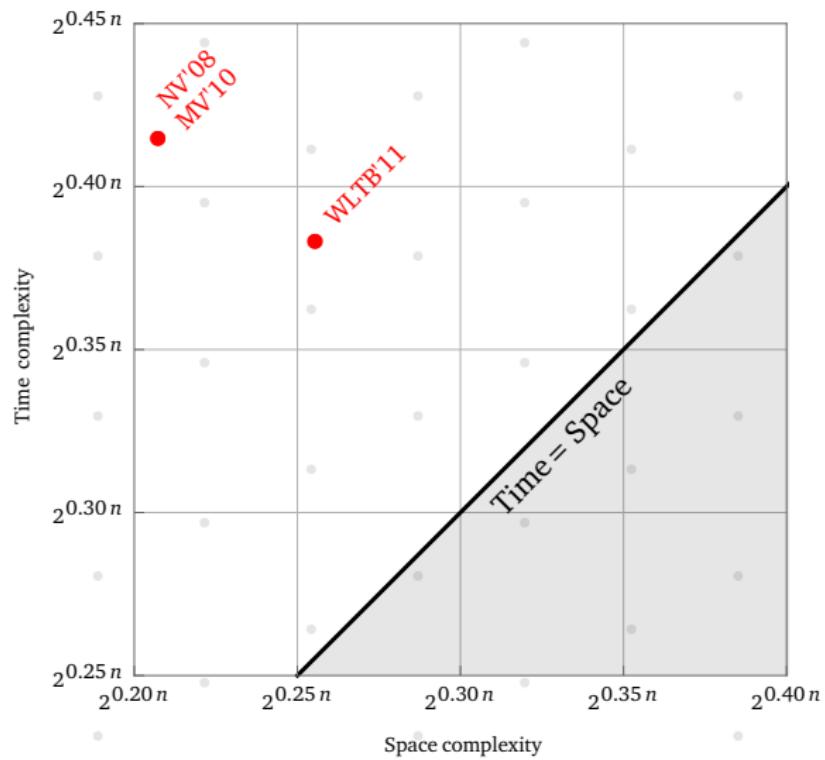
The three-level sieve runs in time $2^{0.3778n}$ and space $2^{0.2833n}$.

Conjecture

The four-level sieve runs in time $2^{0.3774n}$ and space $2^{0.2925n}$, and higher-level sieves are not faster than this.

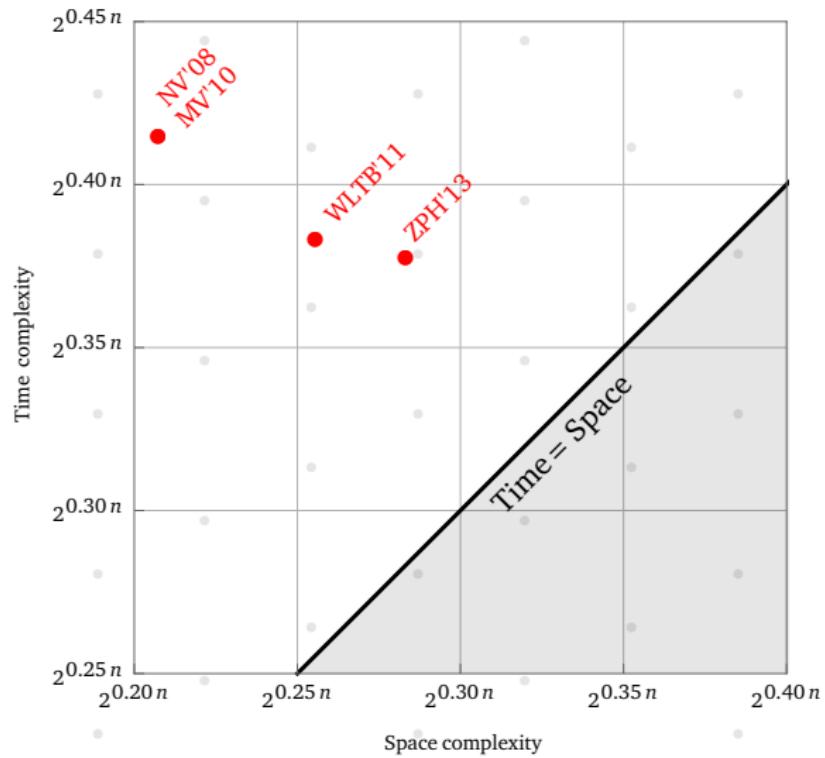
Three-level sieve

Space/time trade-off



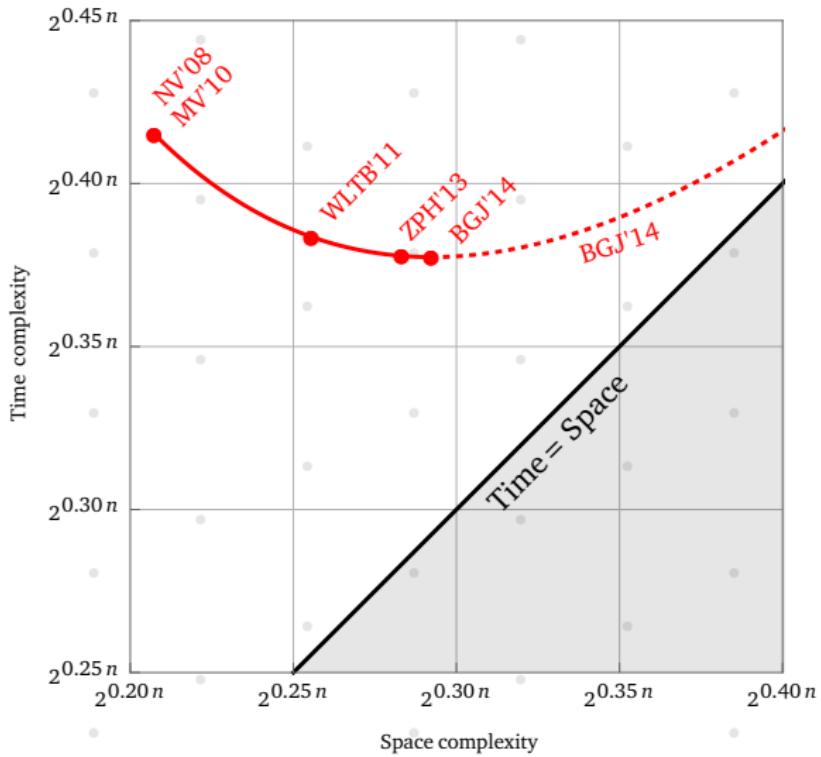
Three-level sieve

Space/time trade-off



Decomposition approach

Space/time trade-off



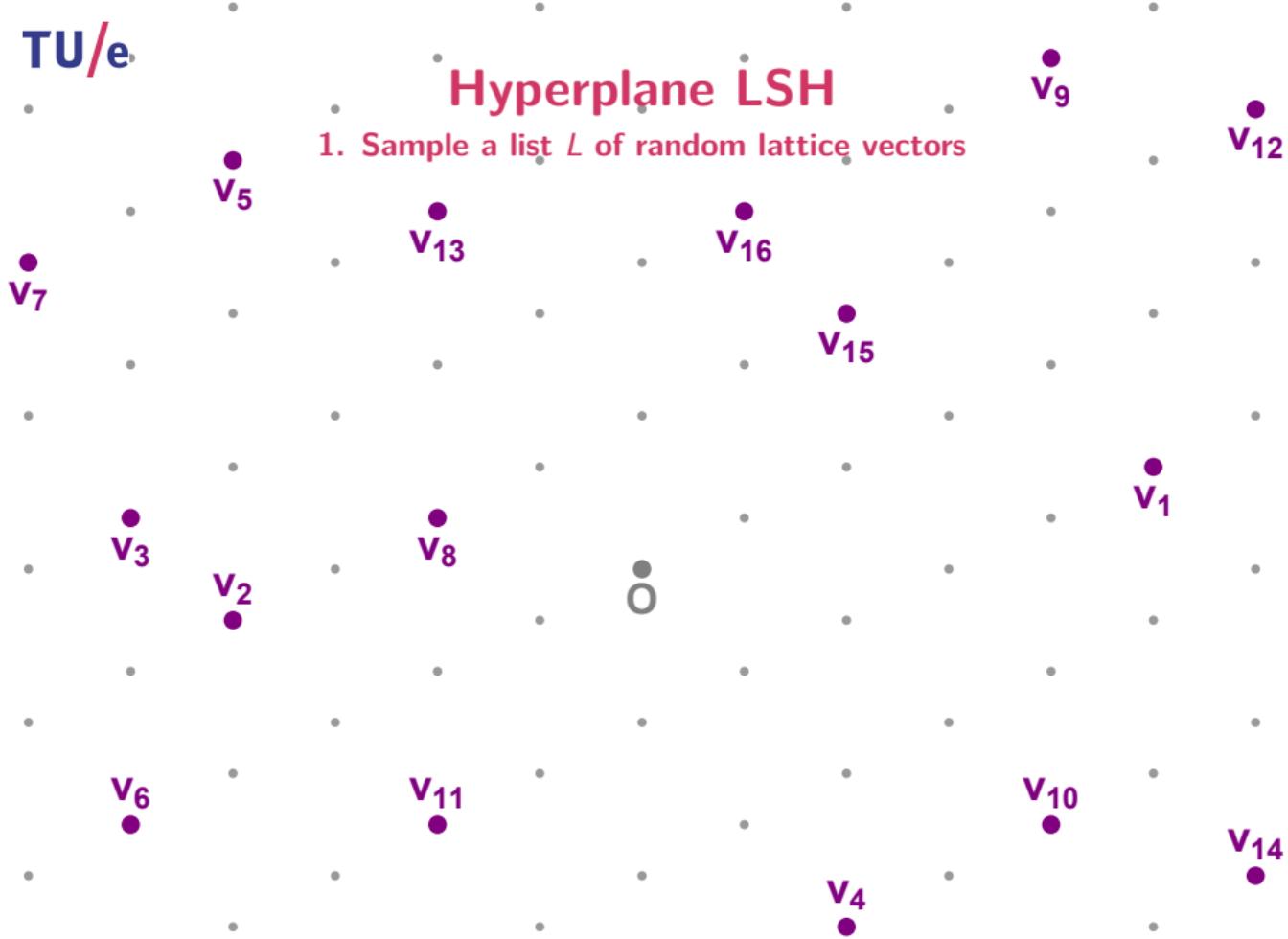
Hyperplane LSH

1. Sample a list L of random lattice vectors



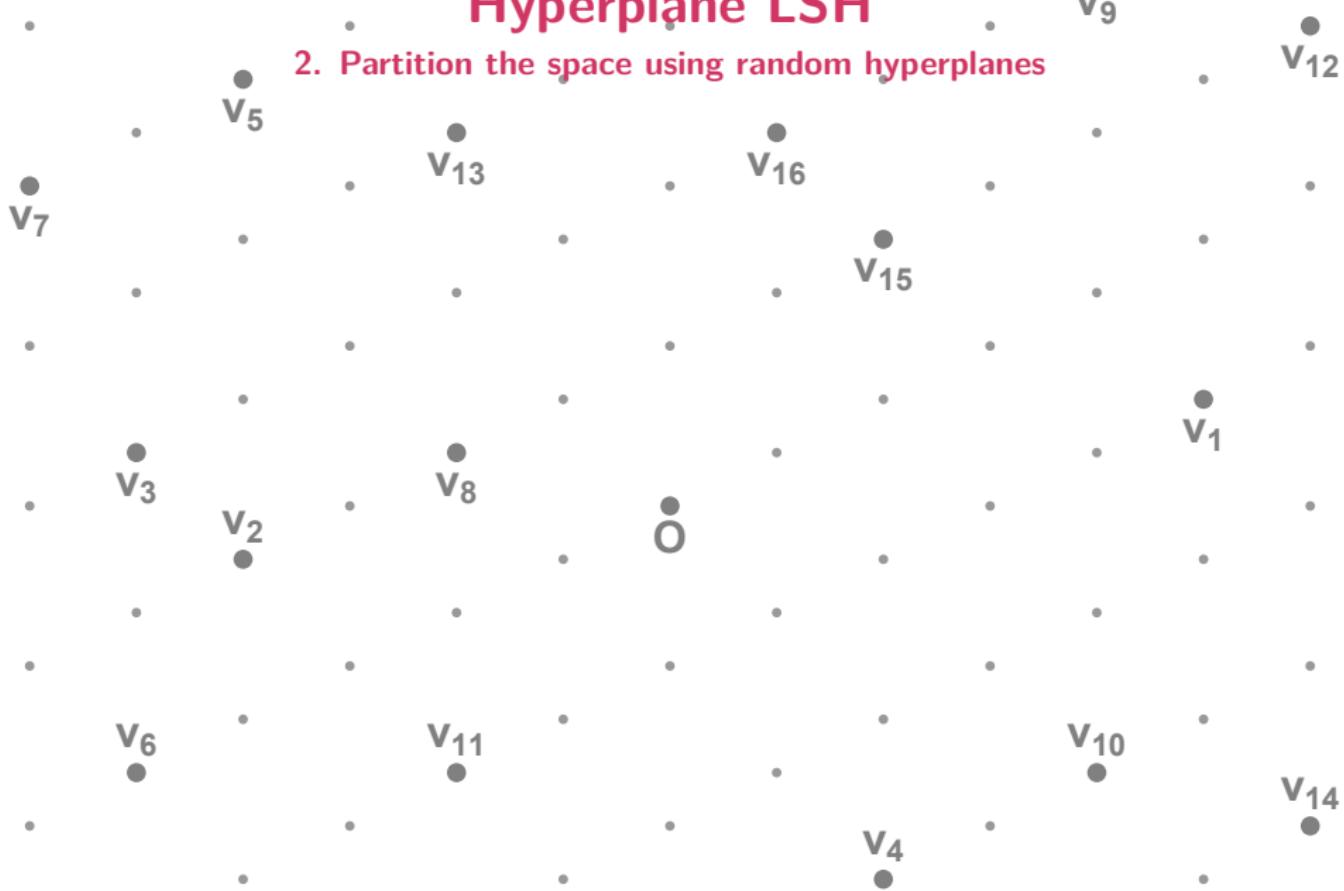
Hyperplane LSH

1. Sample a list L of random lattice vectors



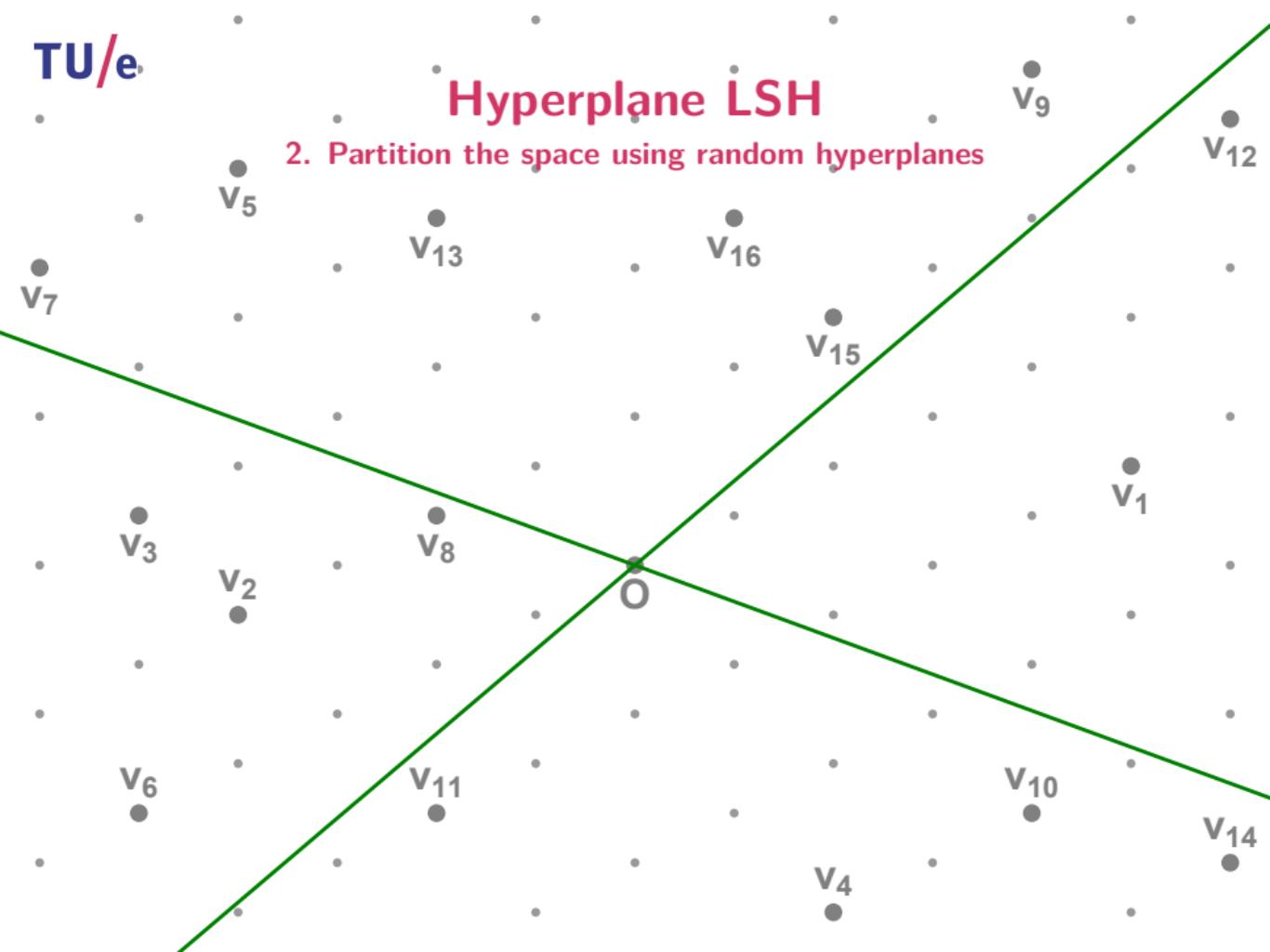
Hyperplane LSH

2. Partition the space using random hyperplanes



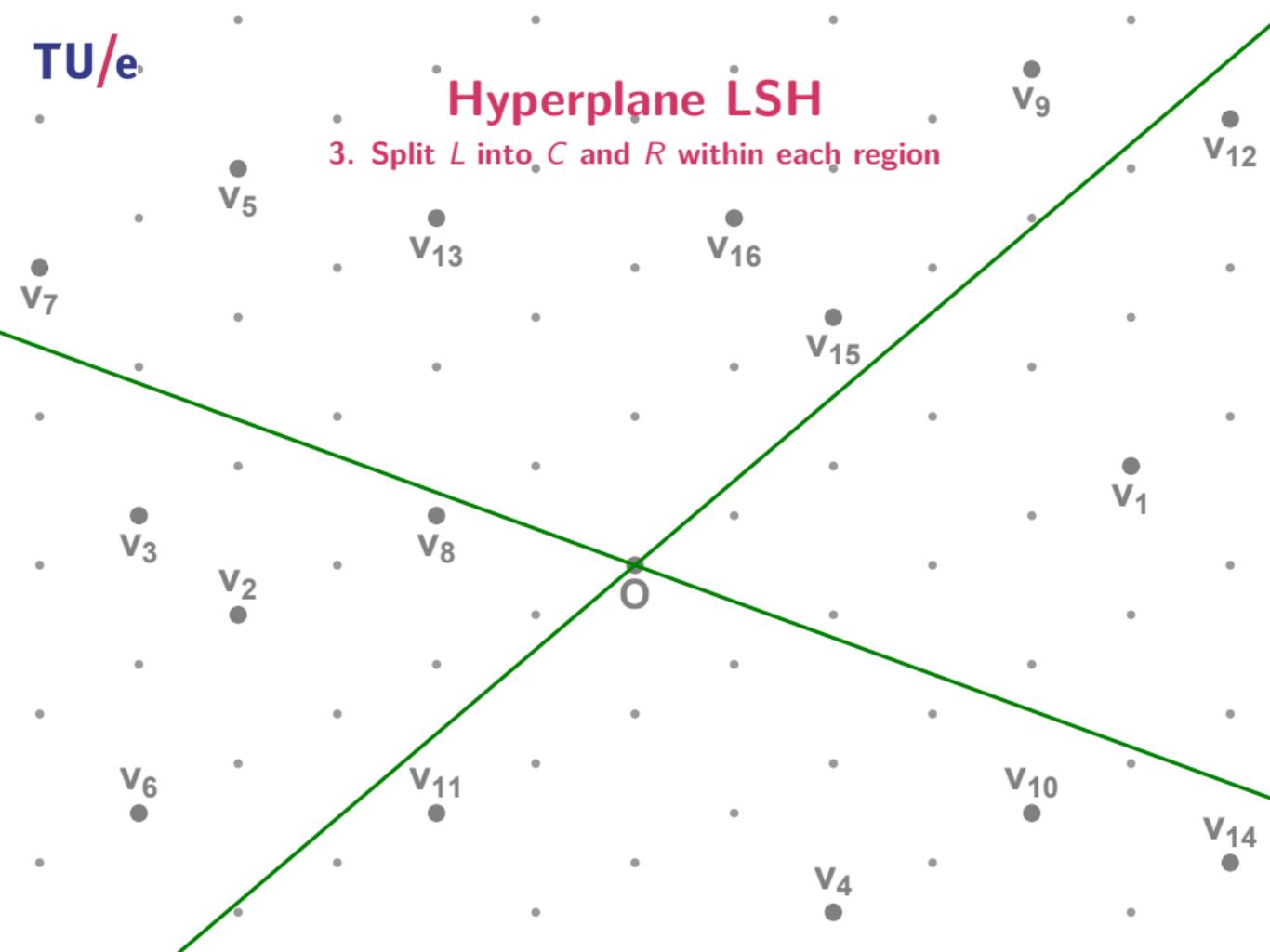
Hyperplane LSH

2. Partition the space using random hyperplanes



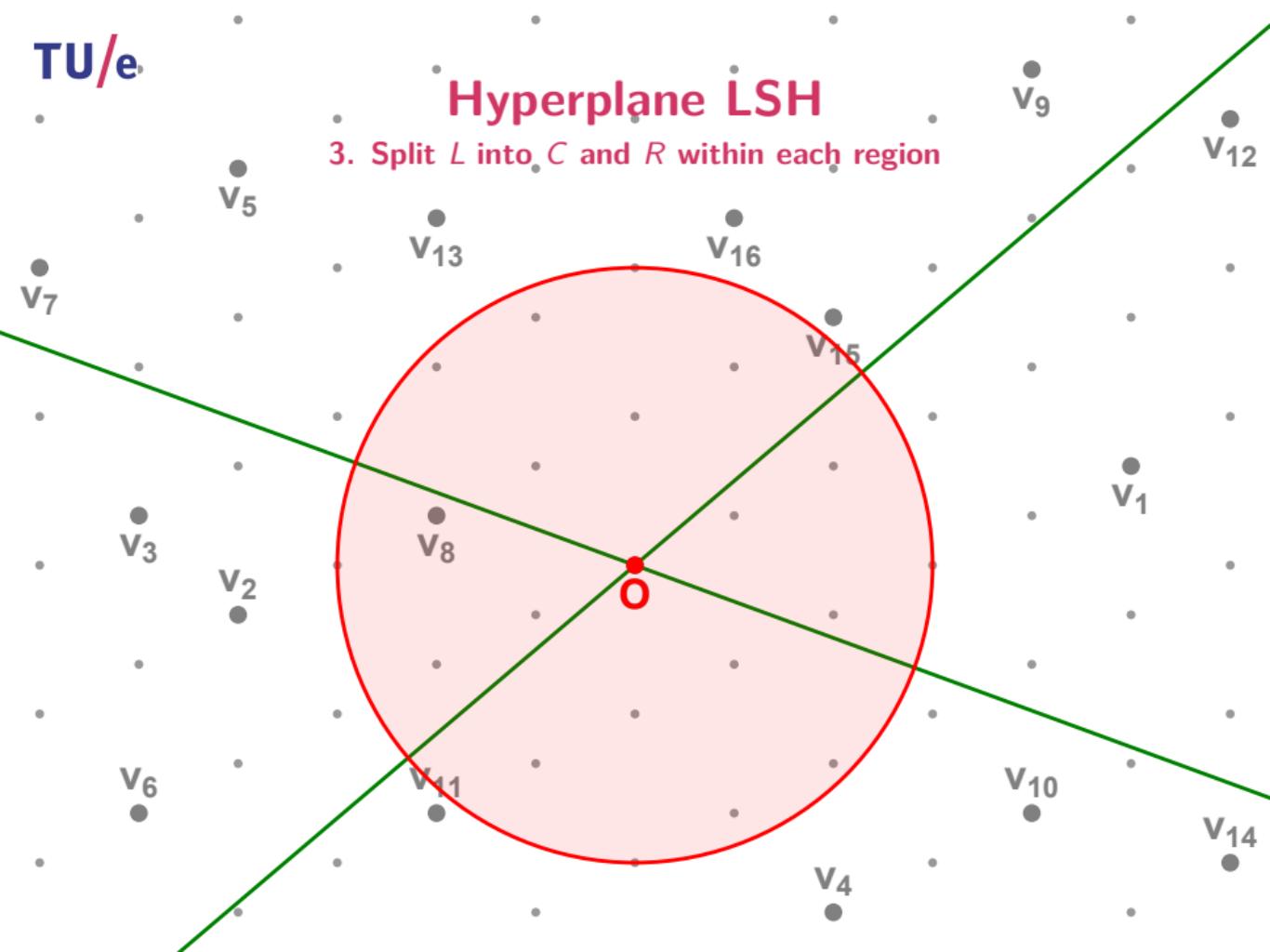
Hyperplane LSH

3. Split L into C and R within each region



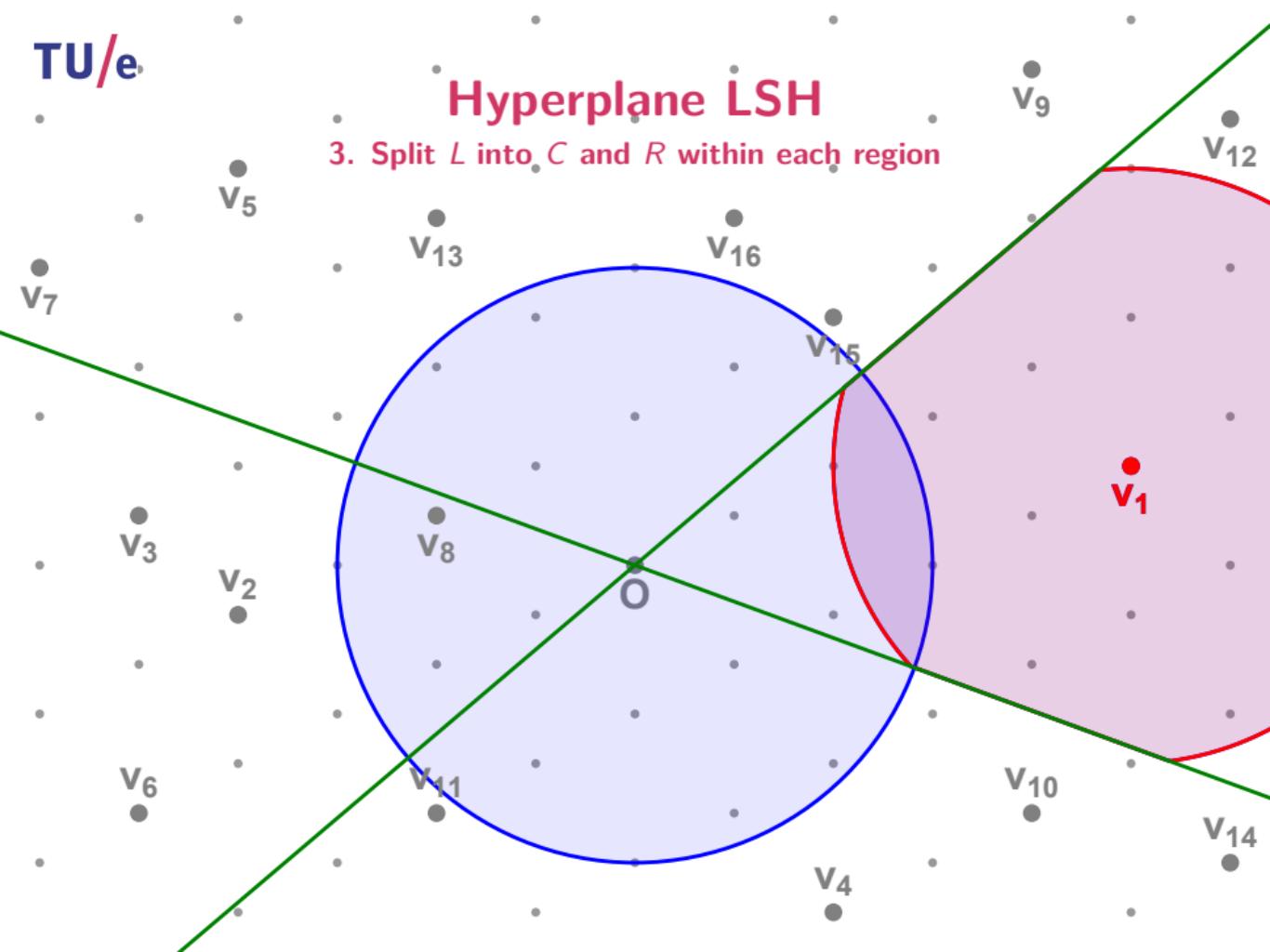
Hyperplane LSH

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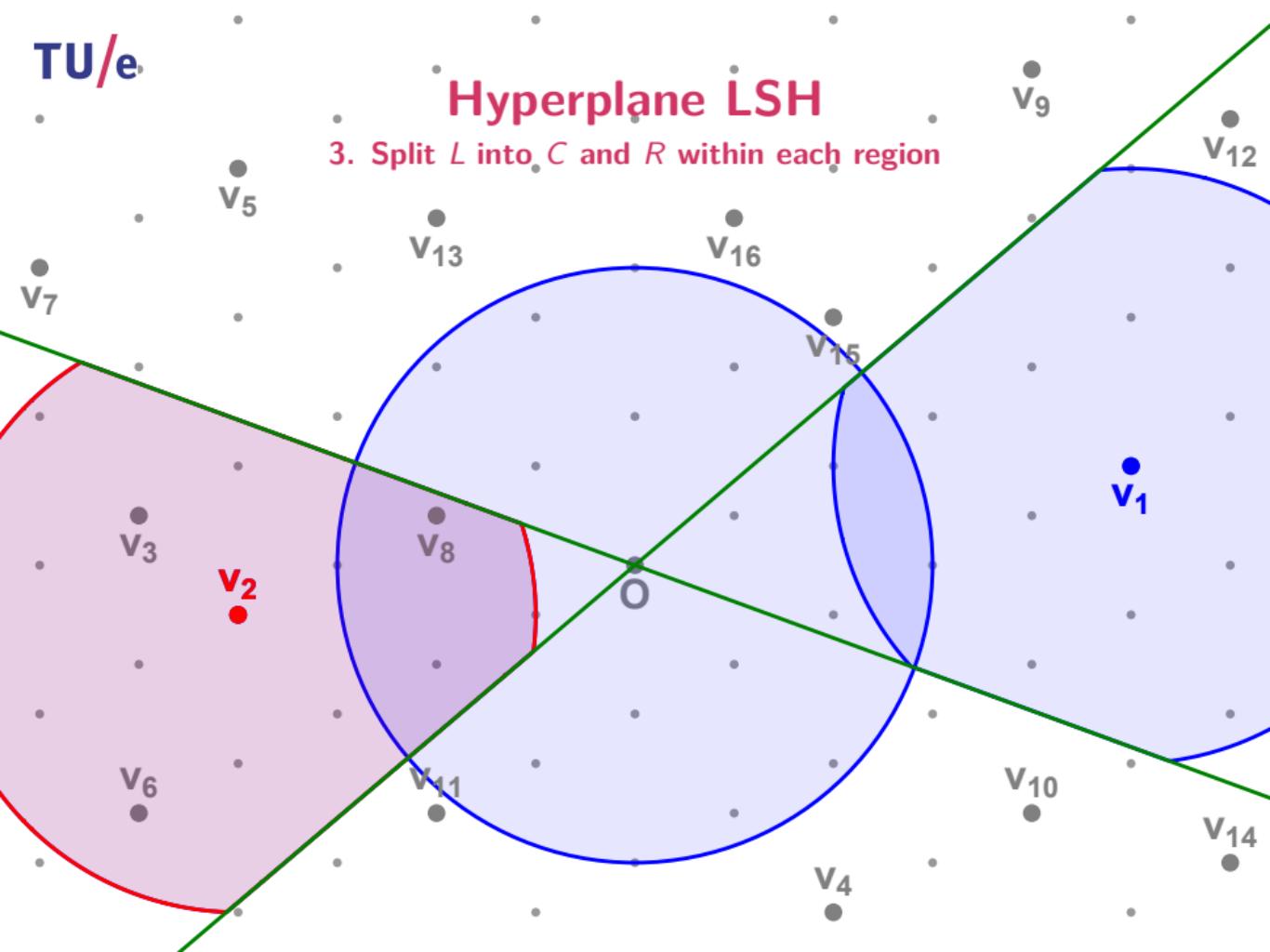
Hyperplane LSH

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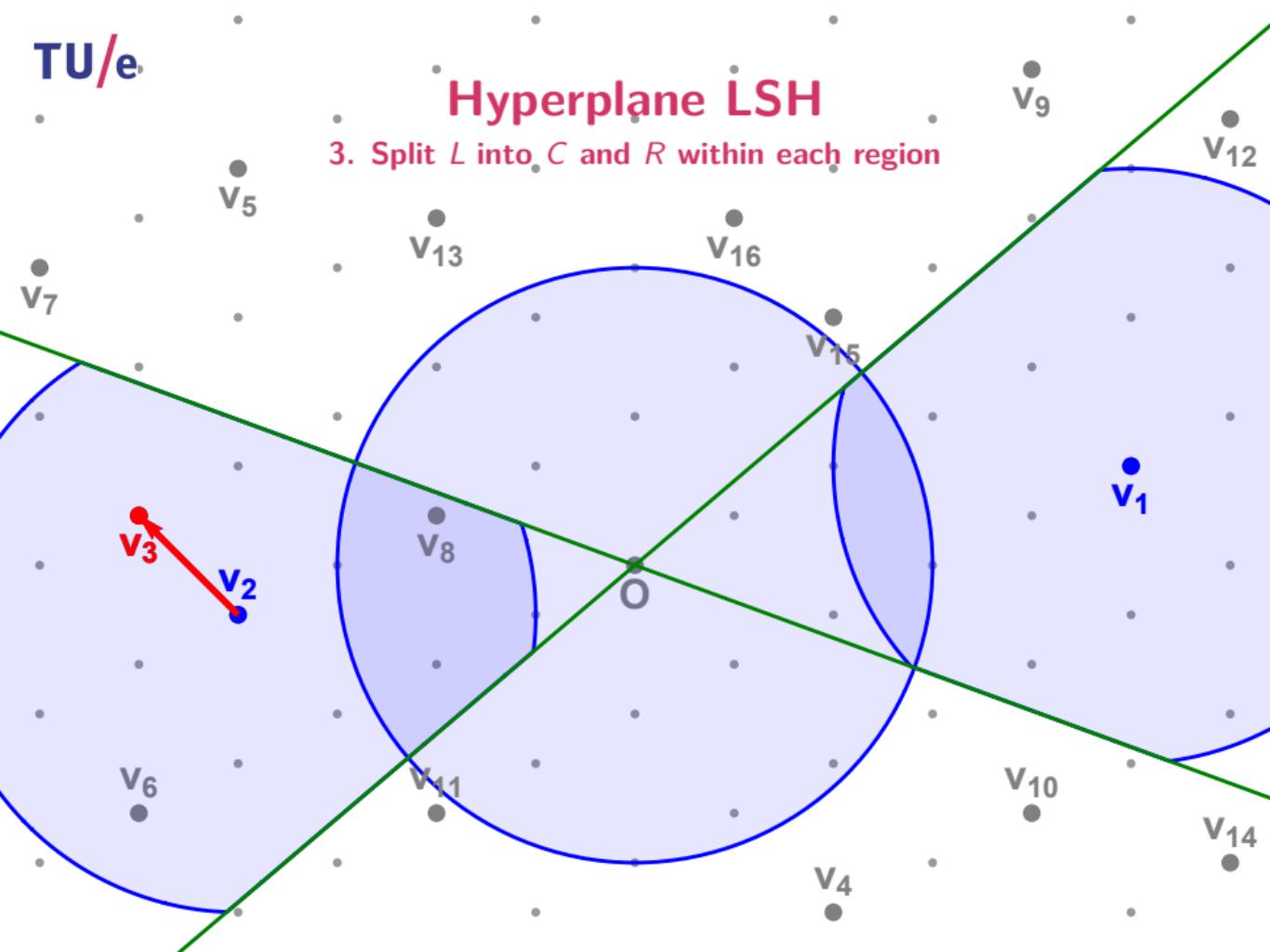
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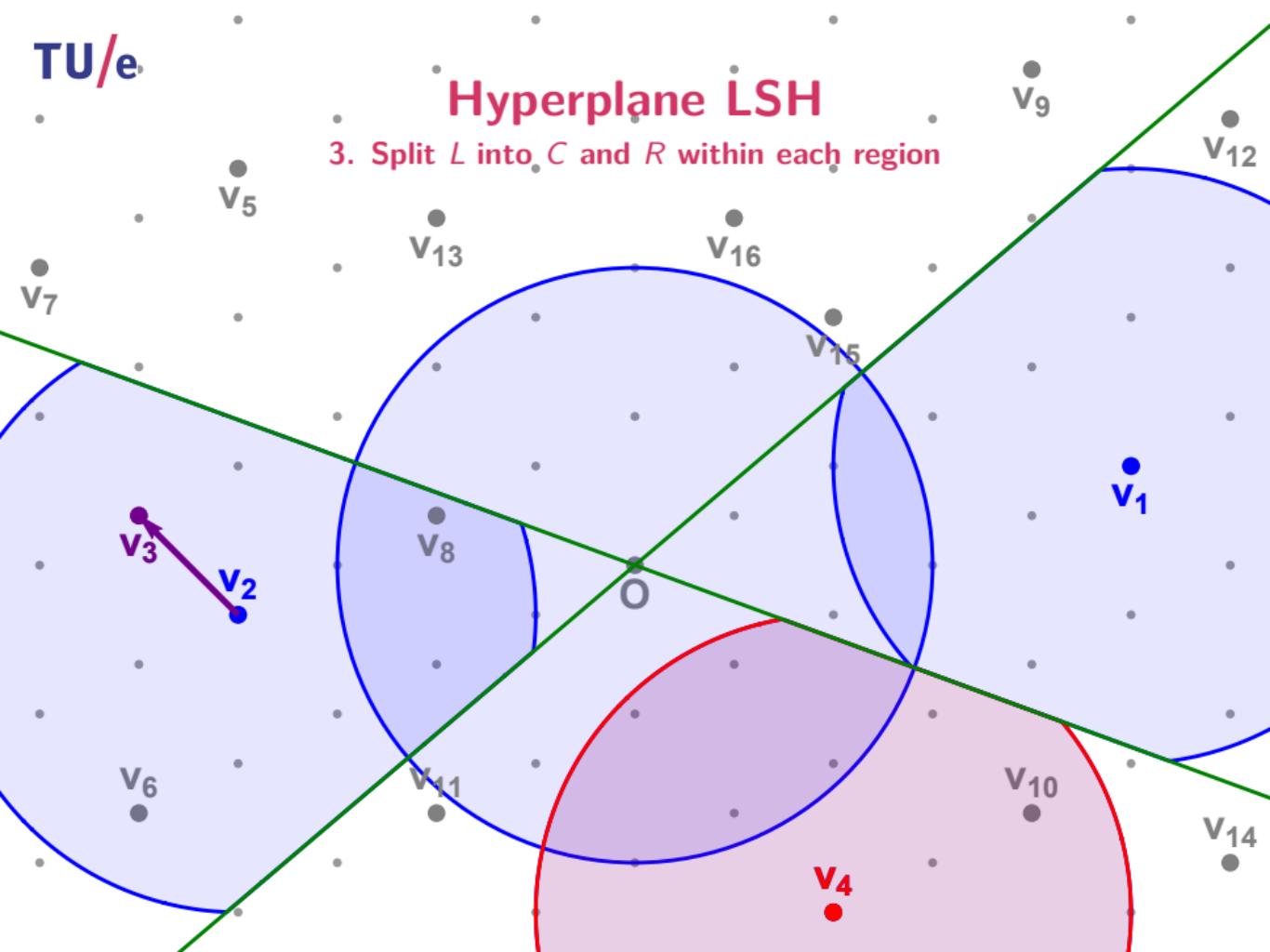
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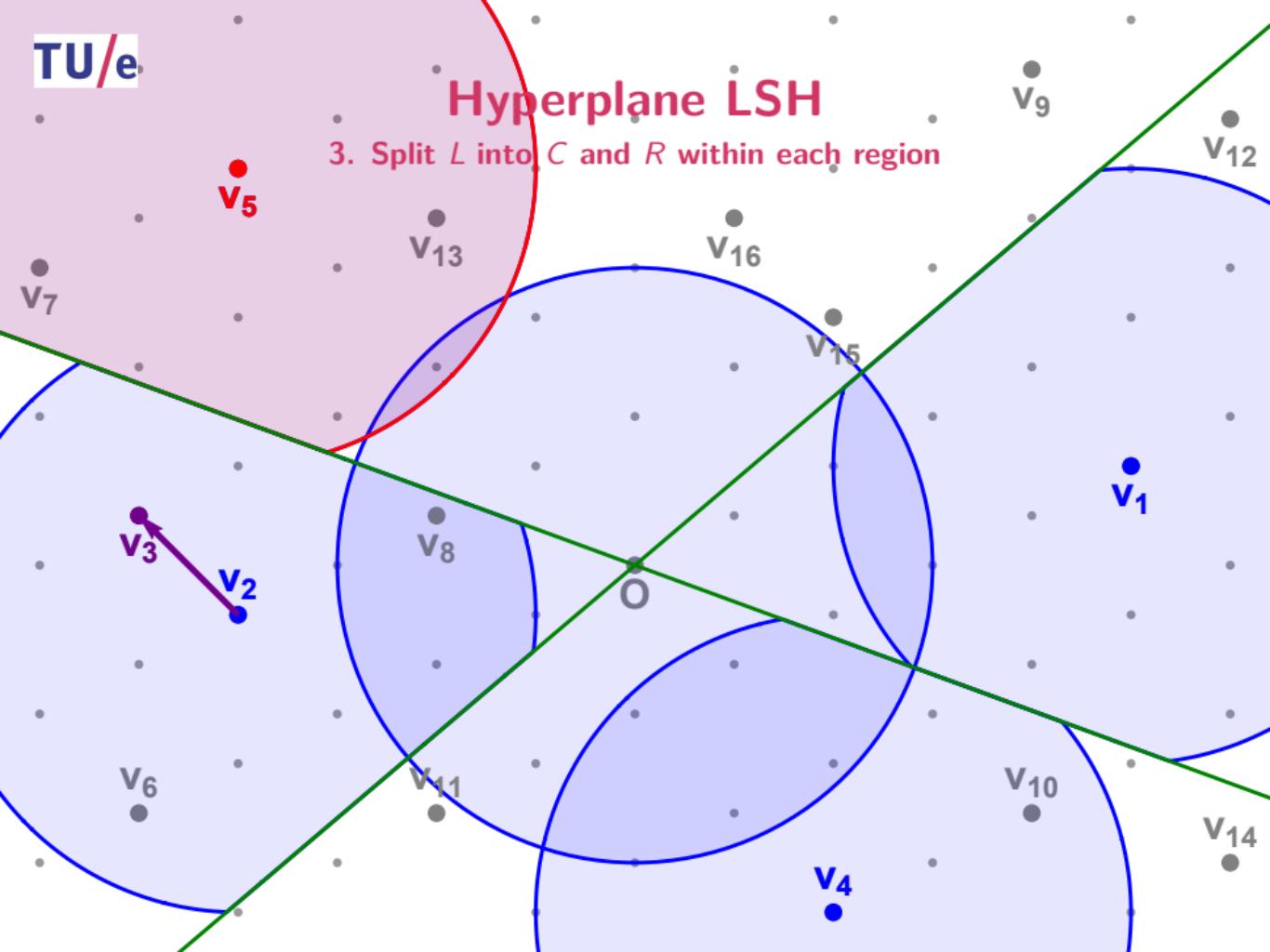
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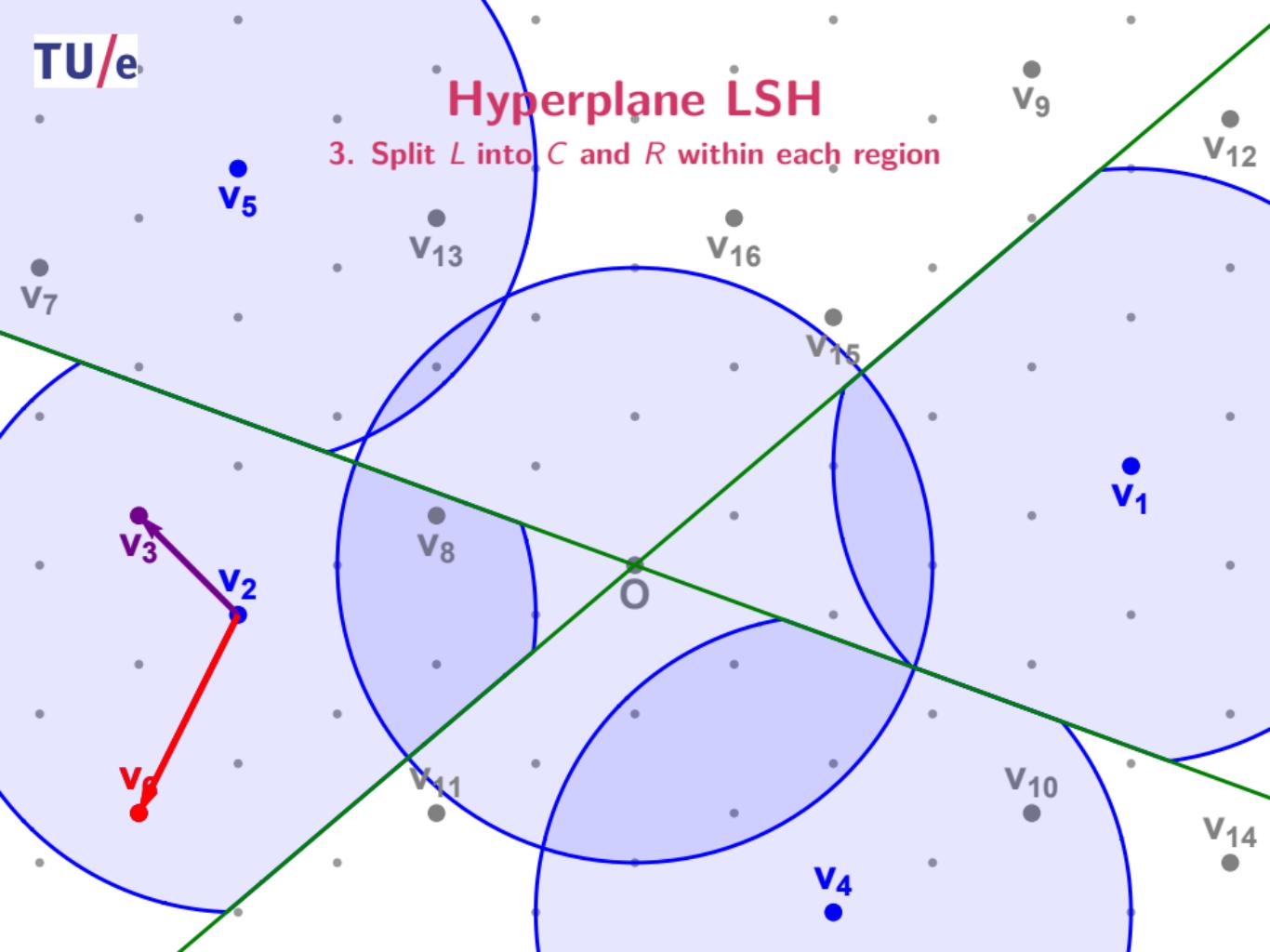
Hyperplane LSH

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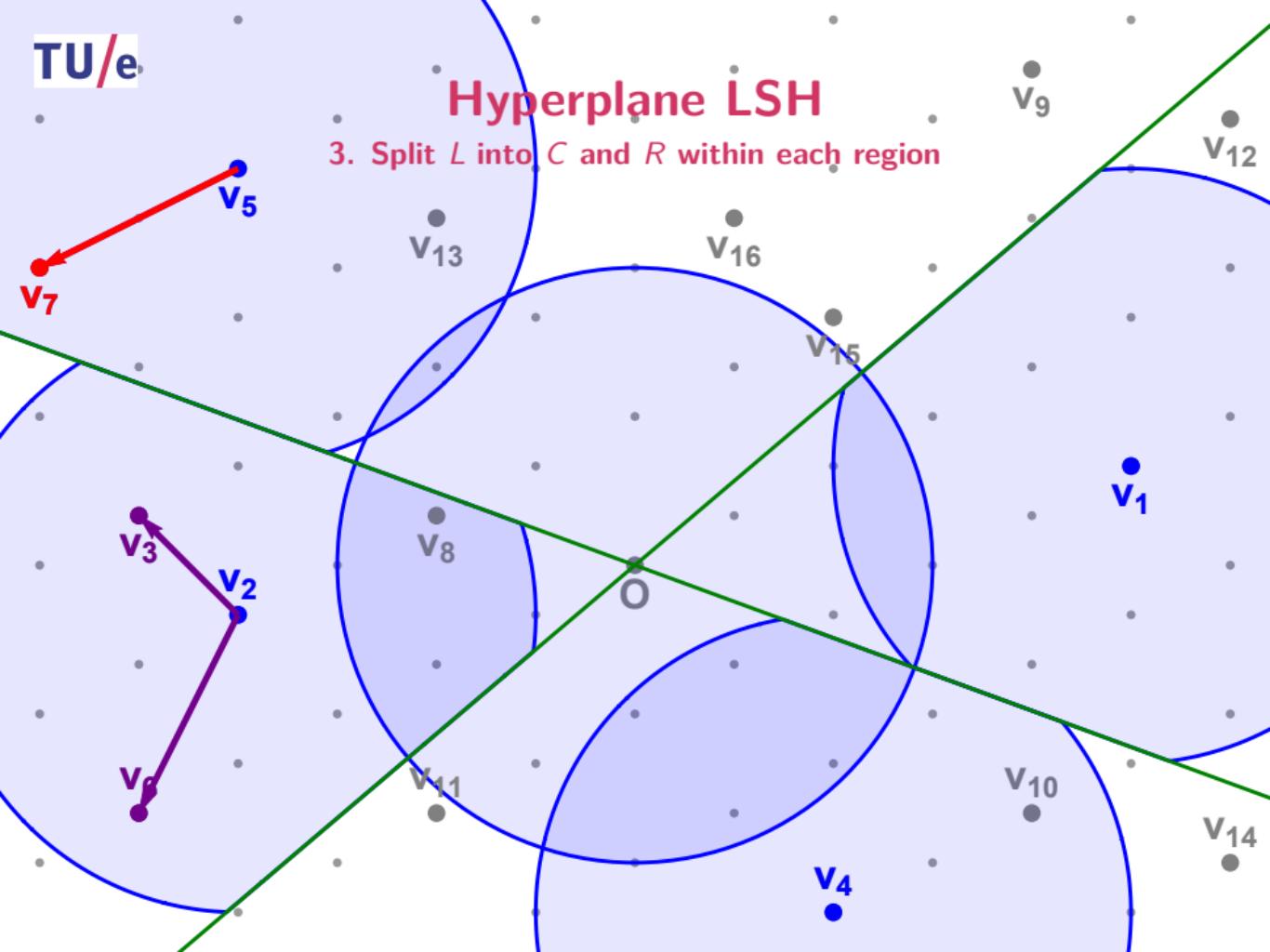
Hyperplane LSH

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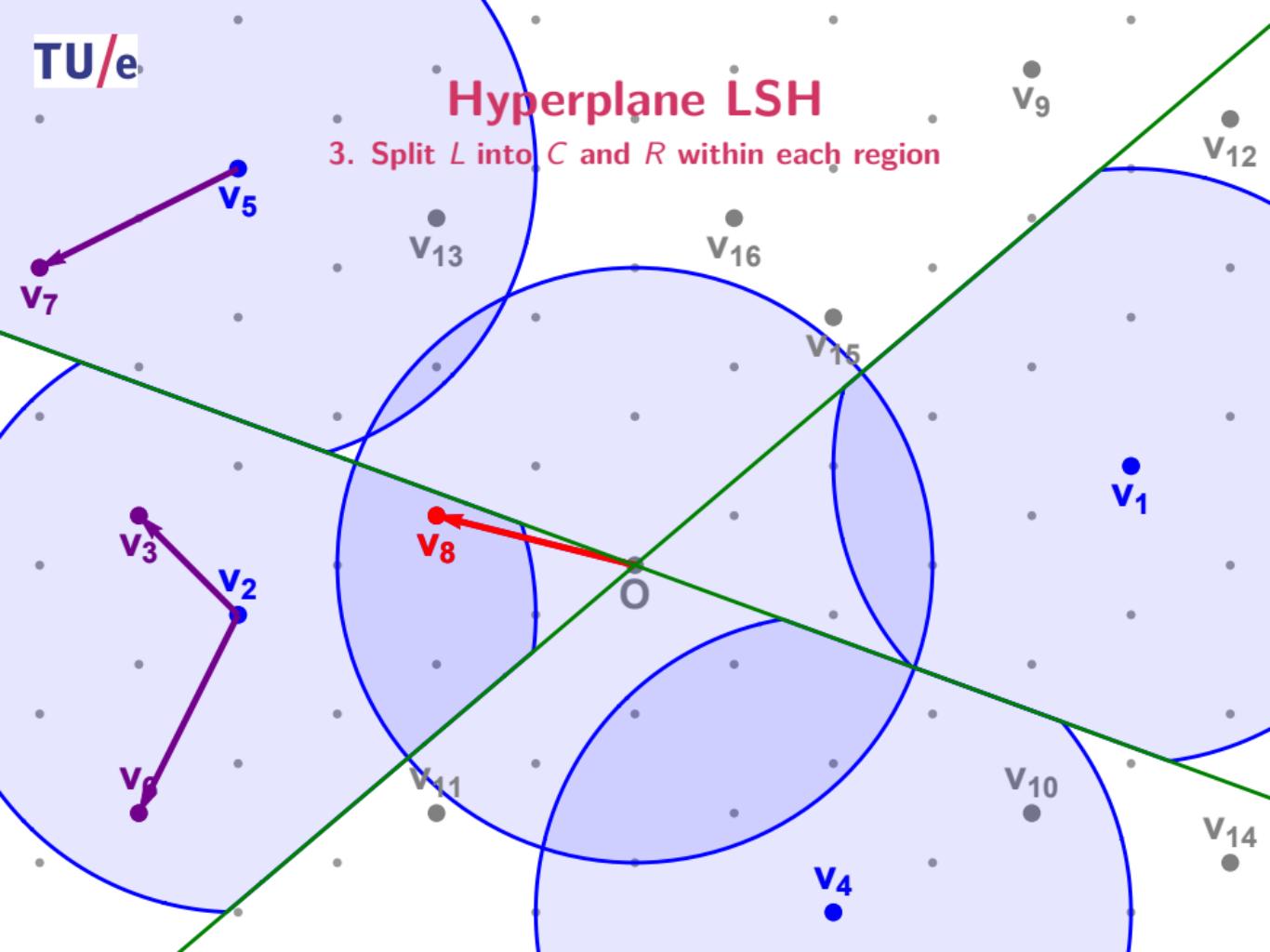
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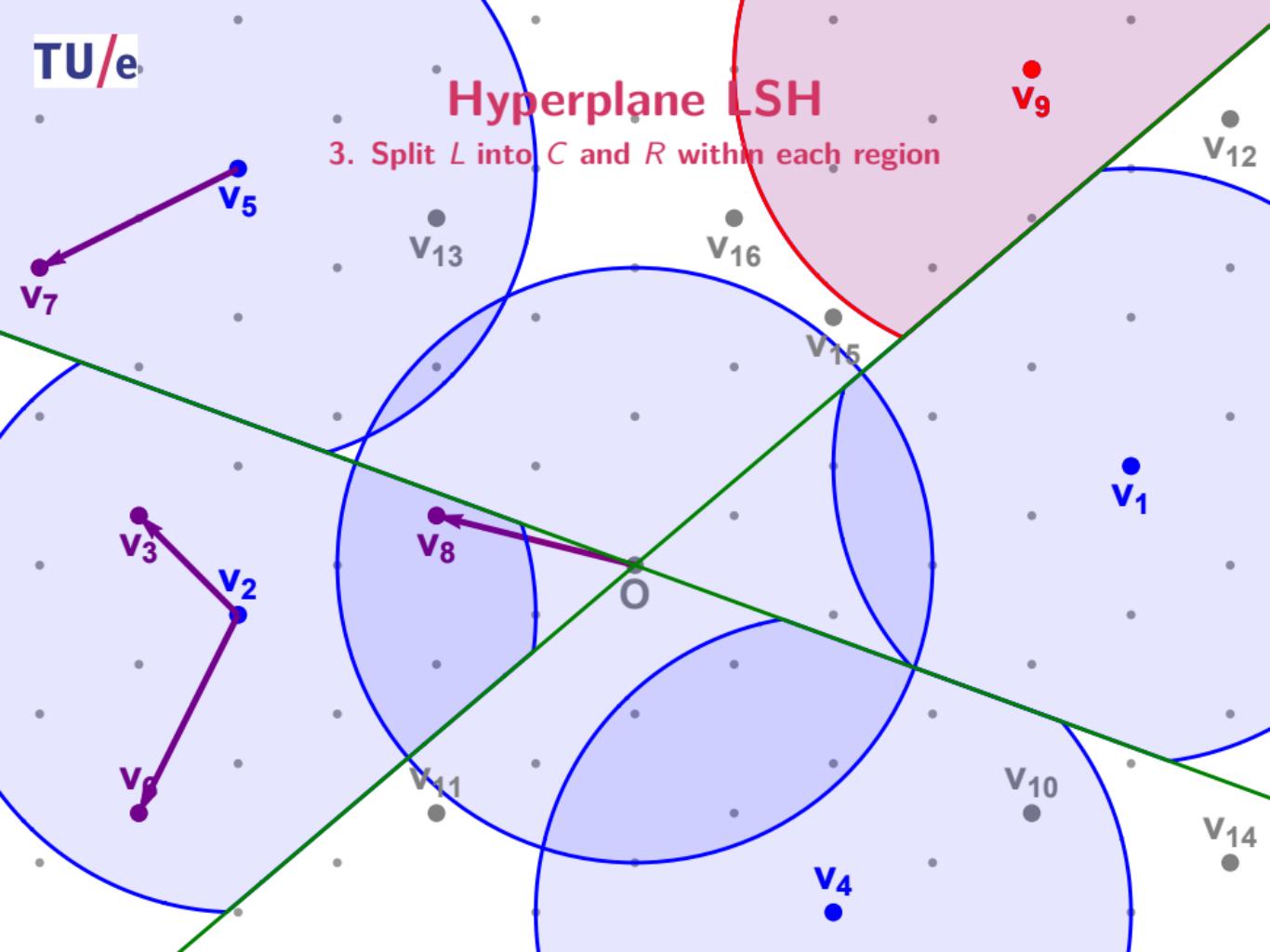
Hyperplane LSH

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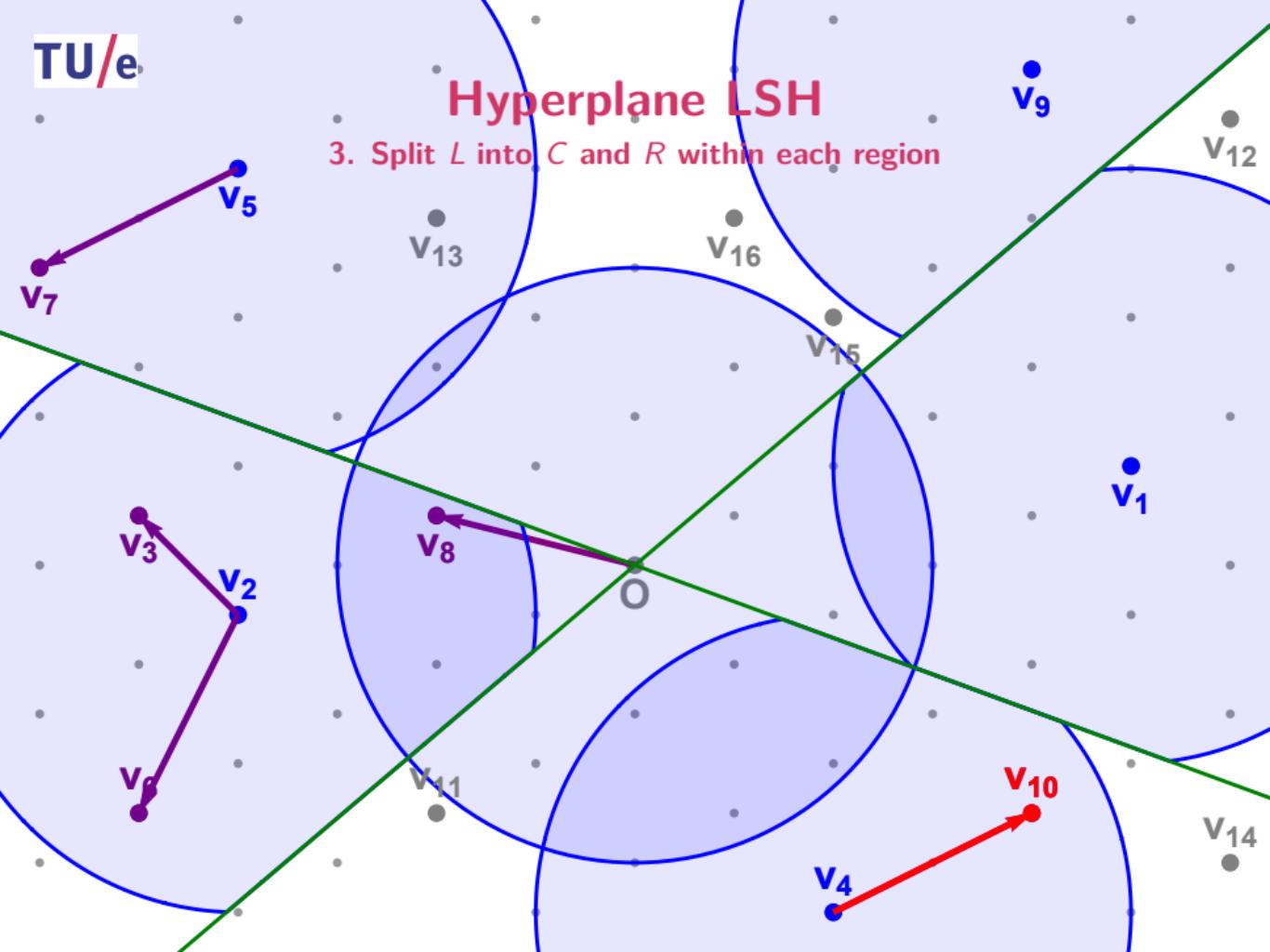
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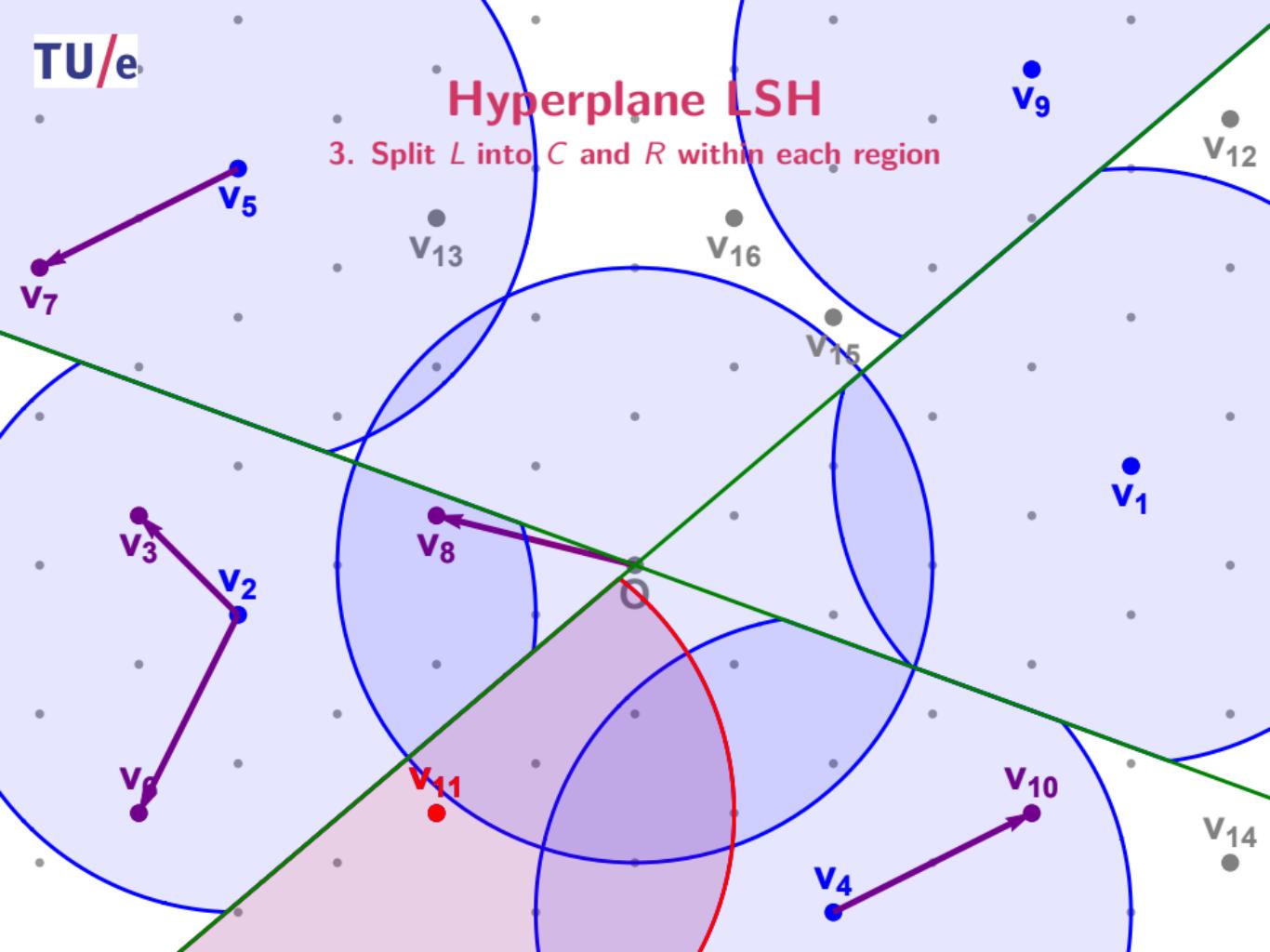
Hyperplane LSH

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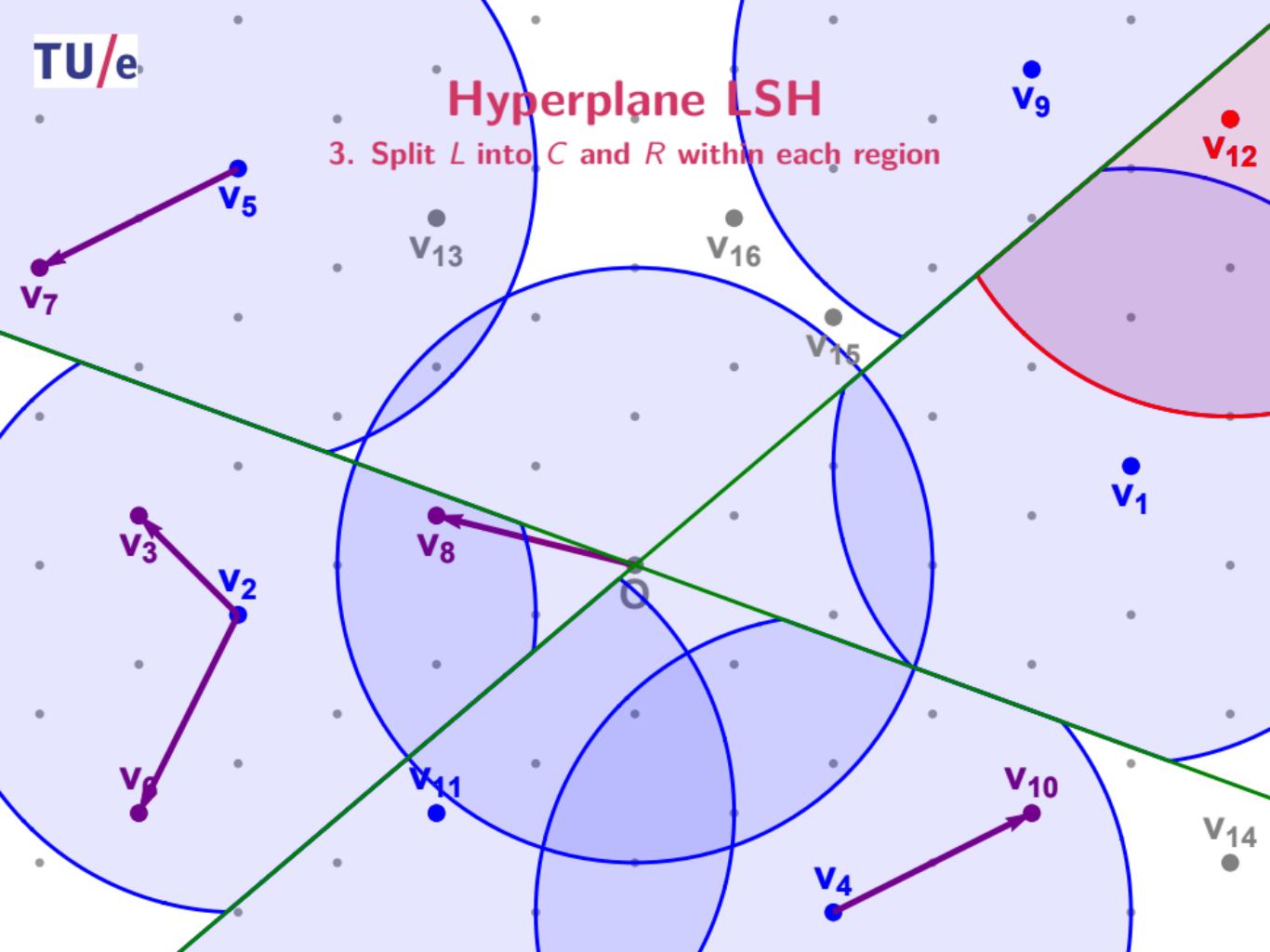
Hyperplane LSH

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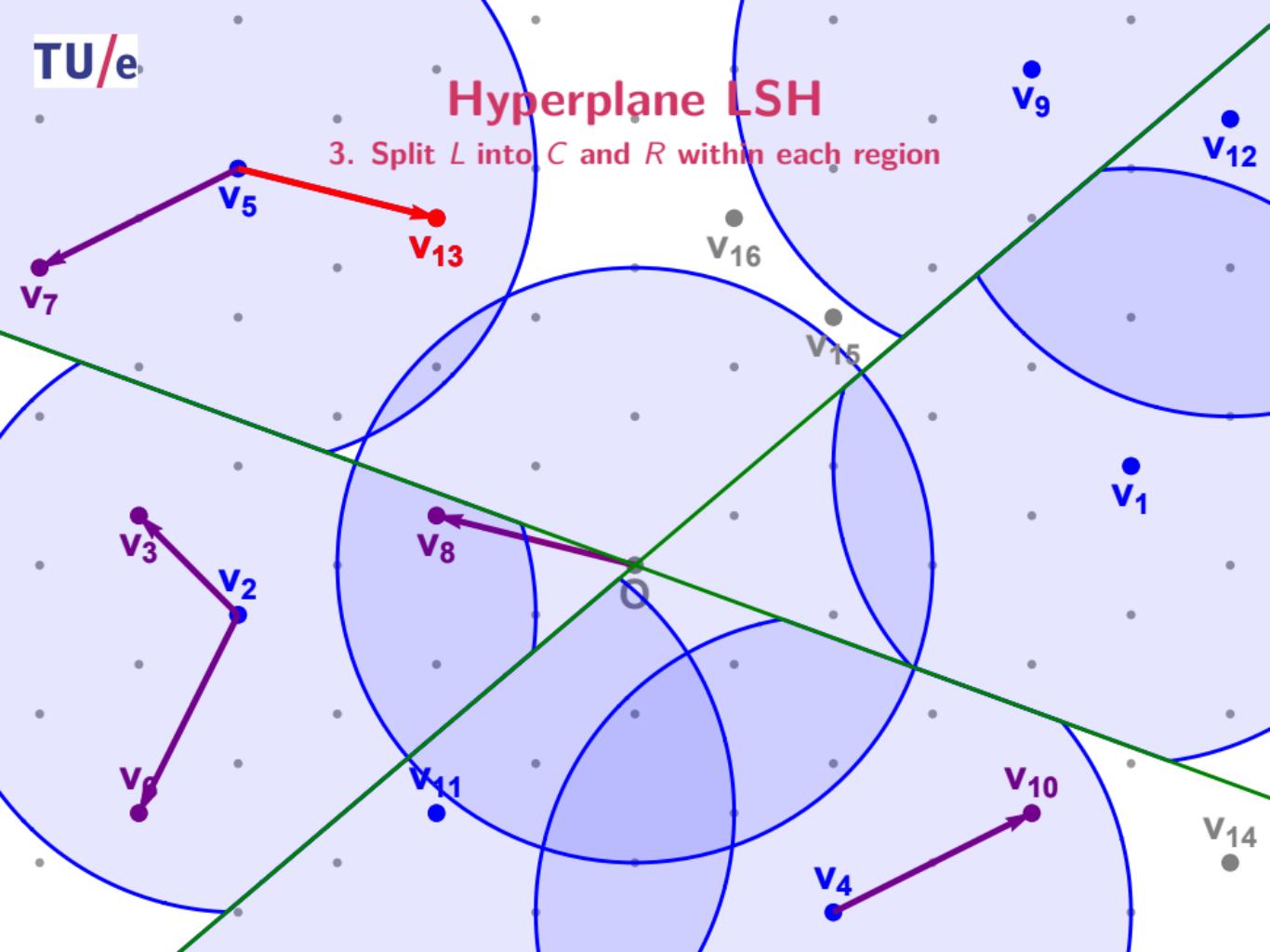
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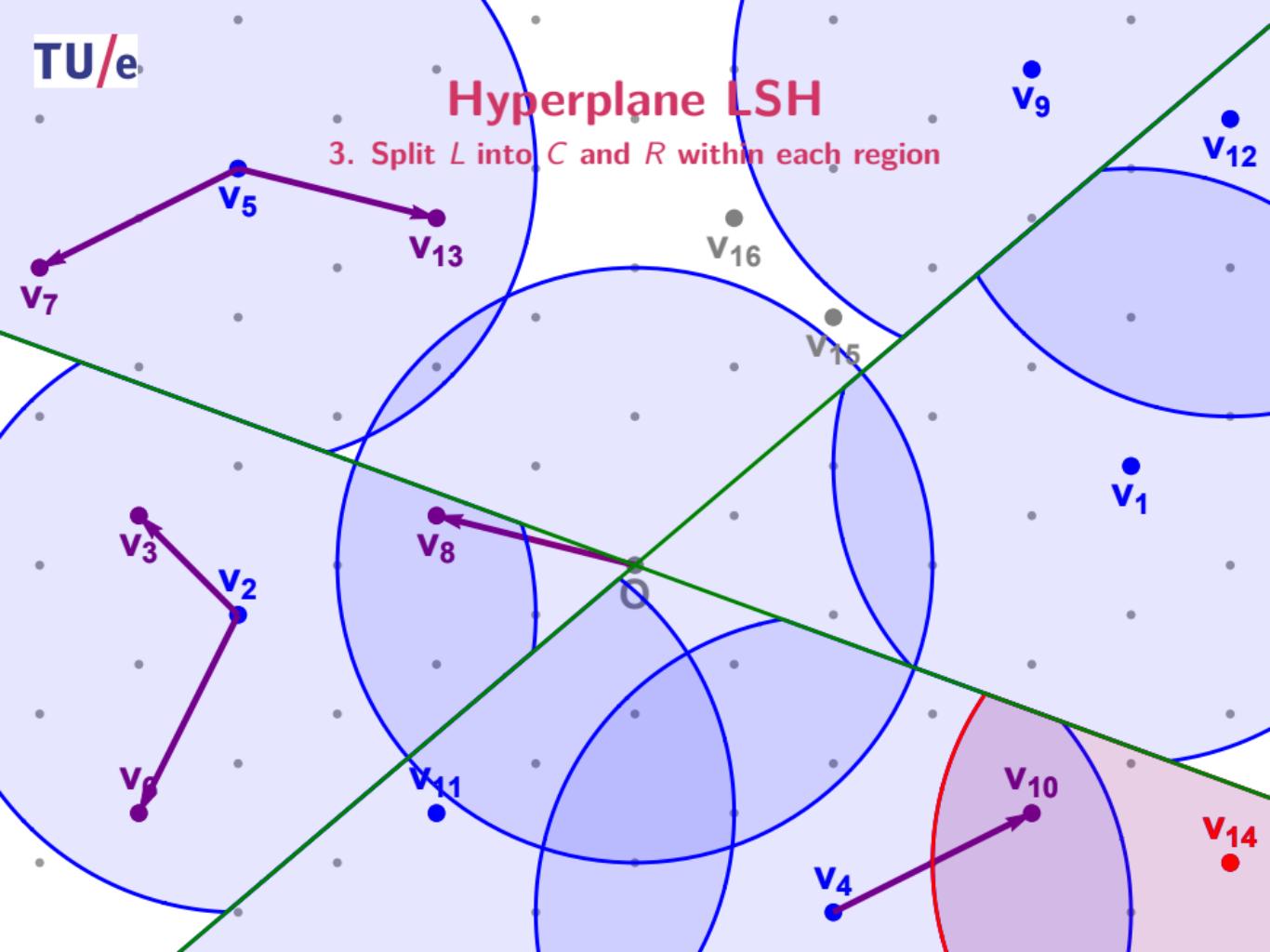
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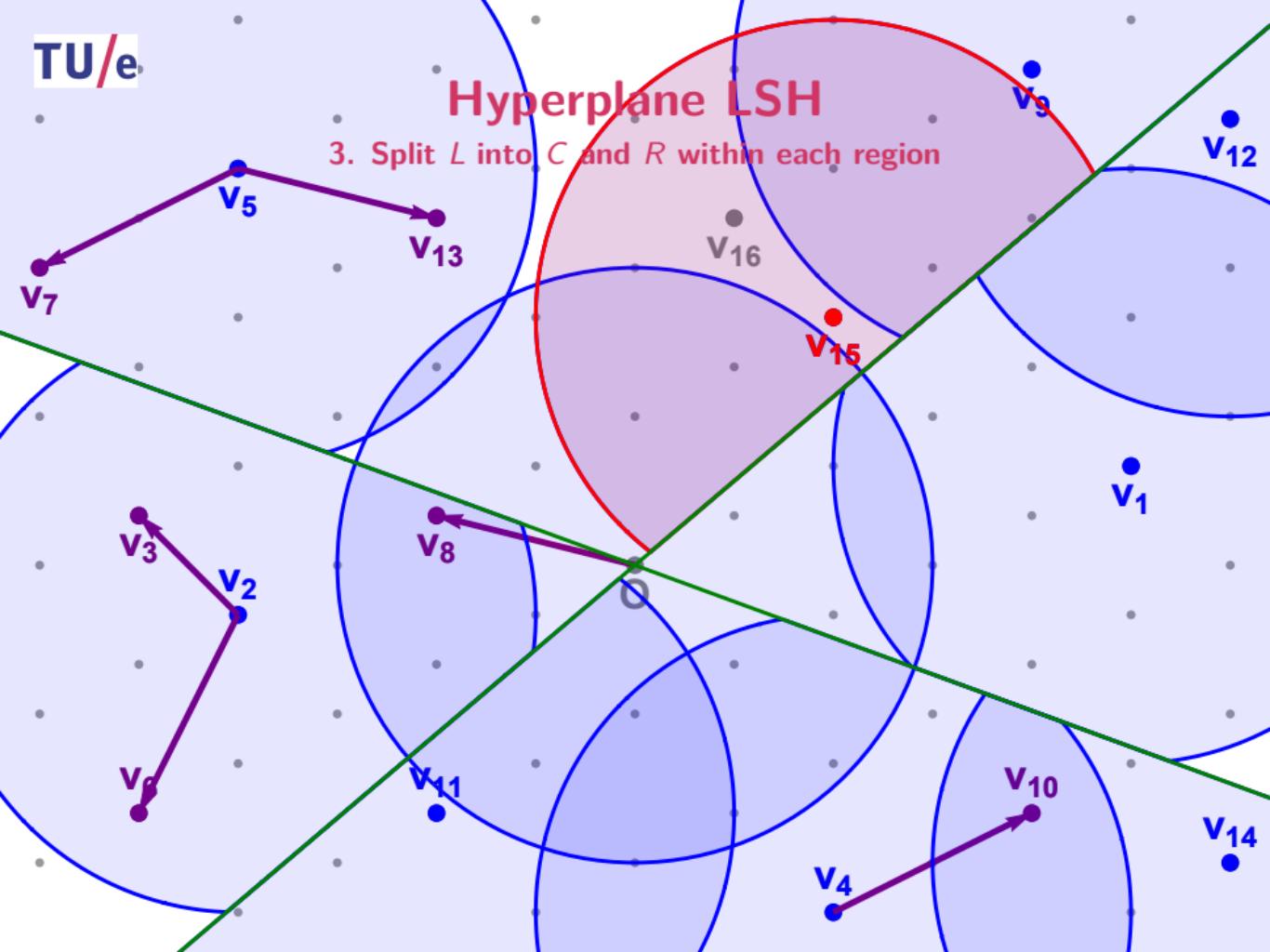
Hyperplane LSH

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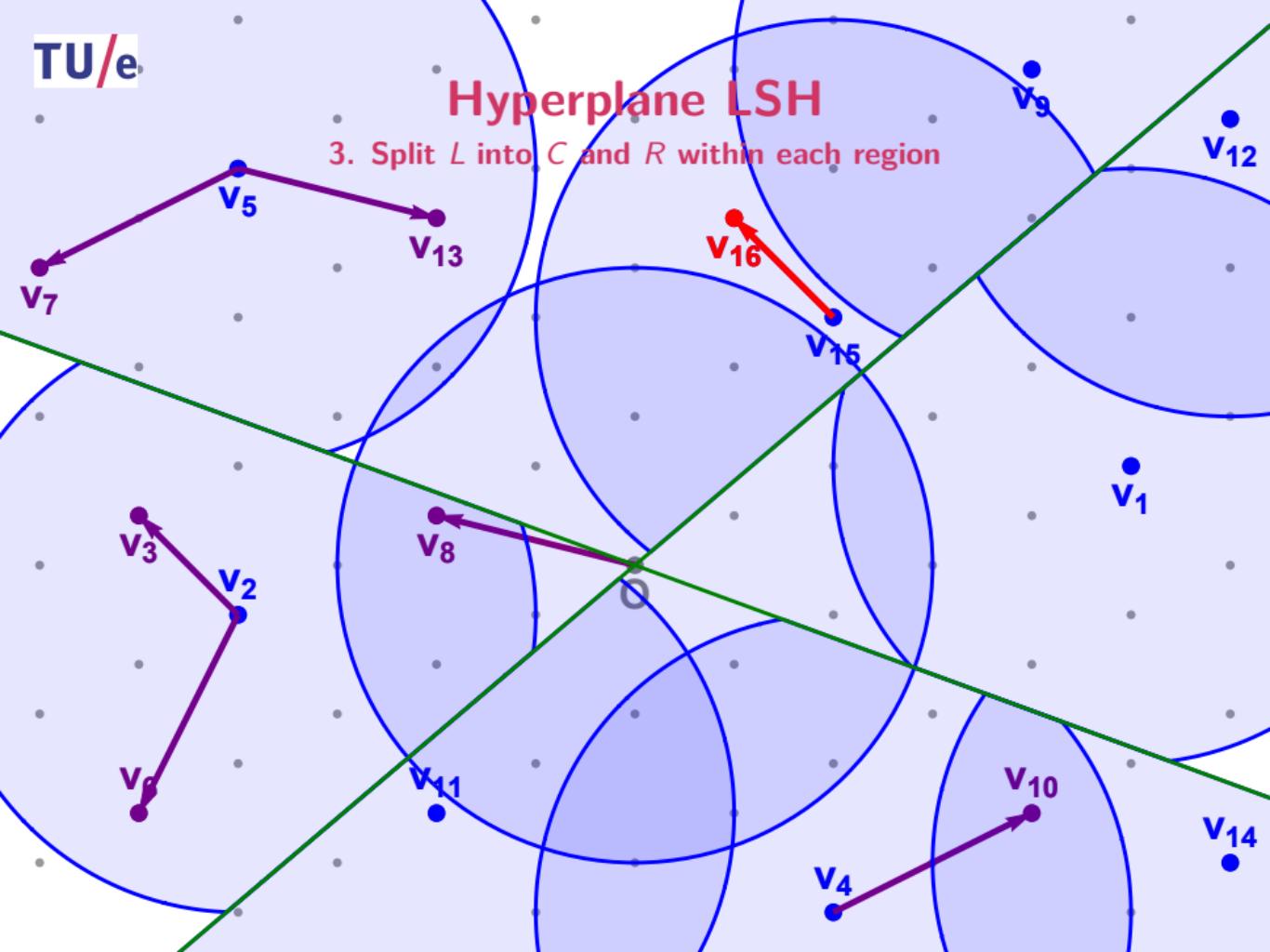
Hyperplane LSH

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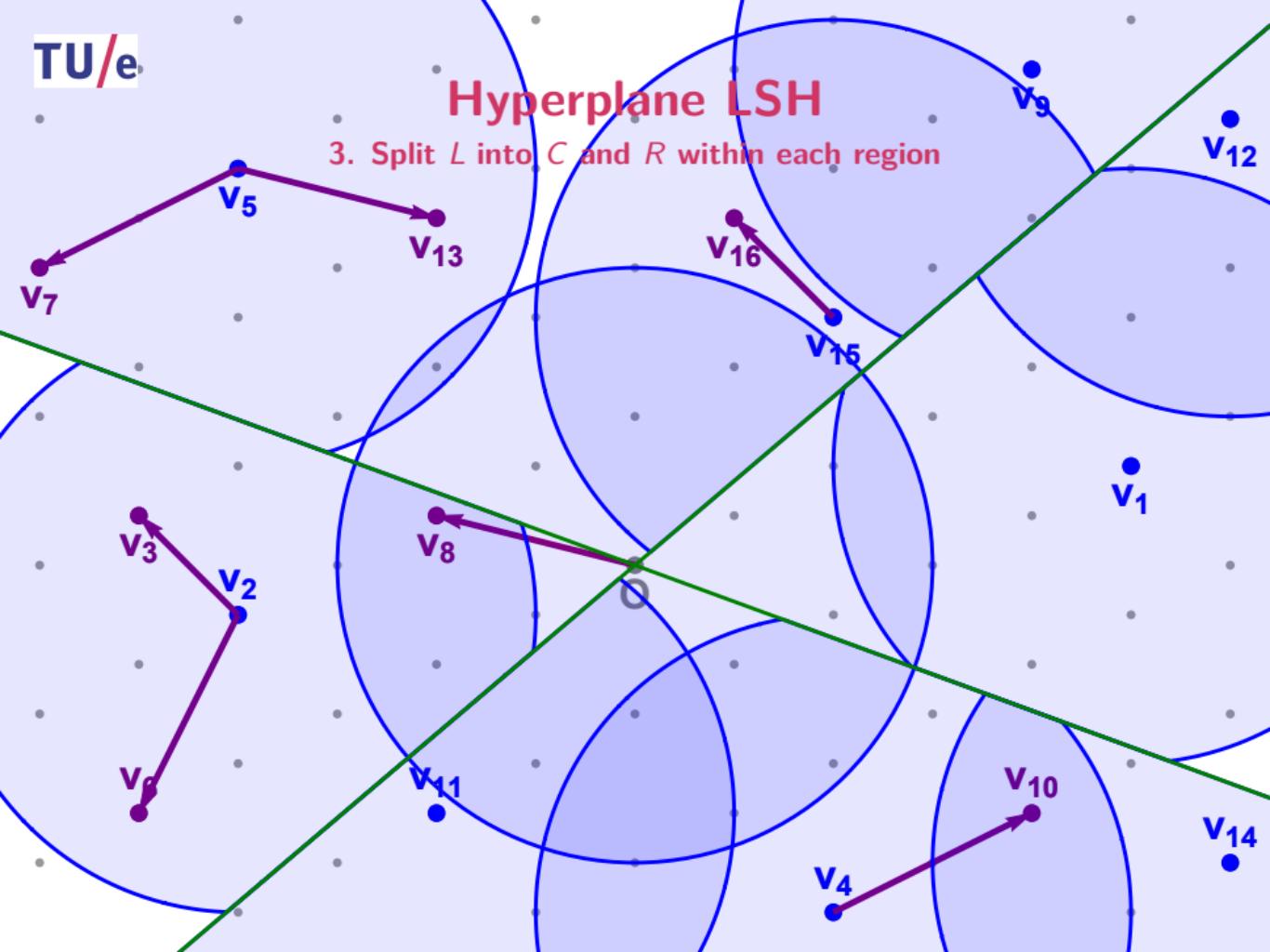
Hyperplane LSH

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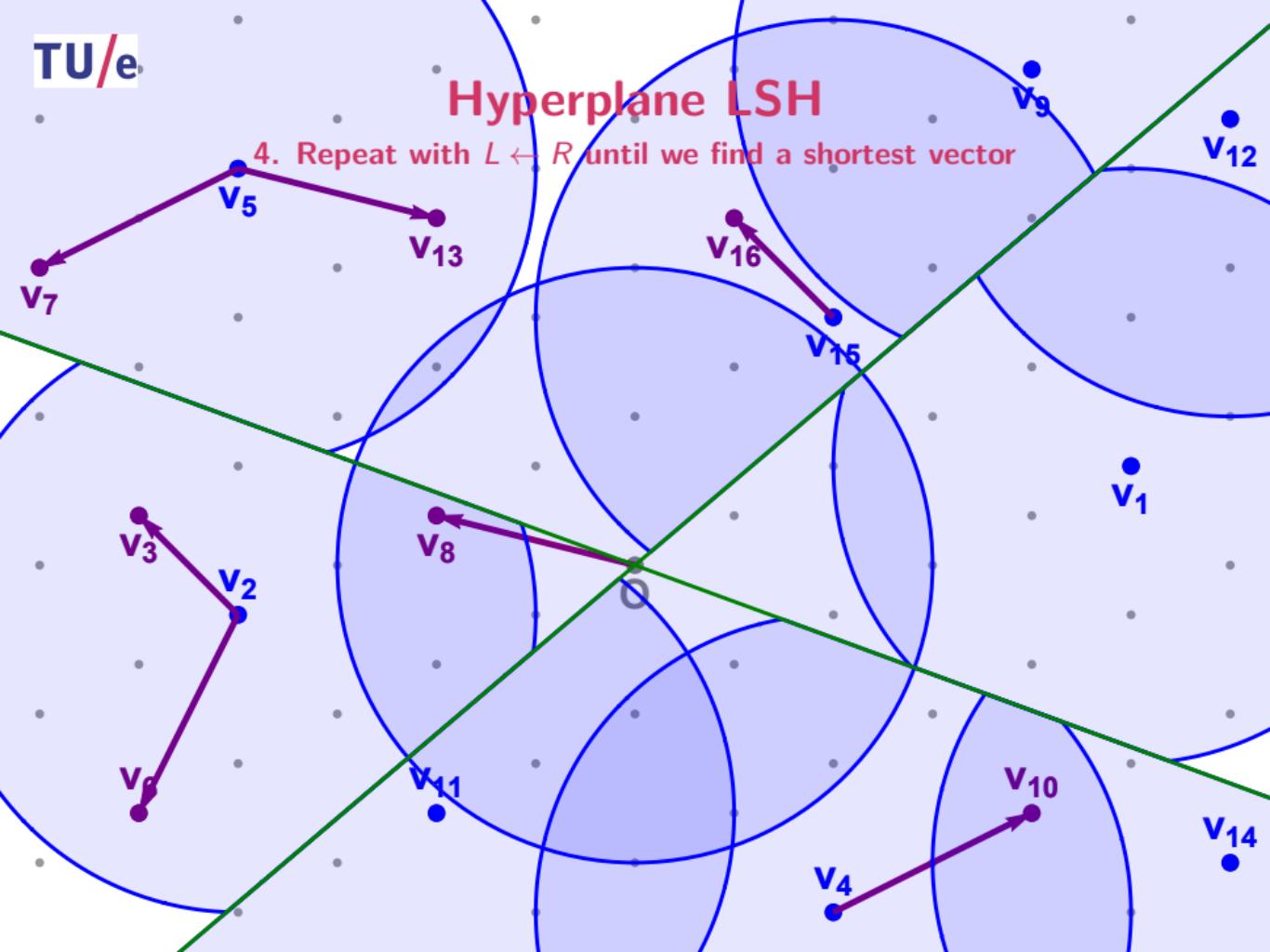
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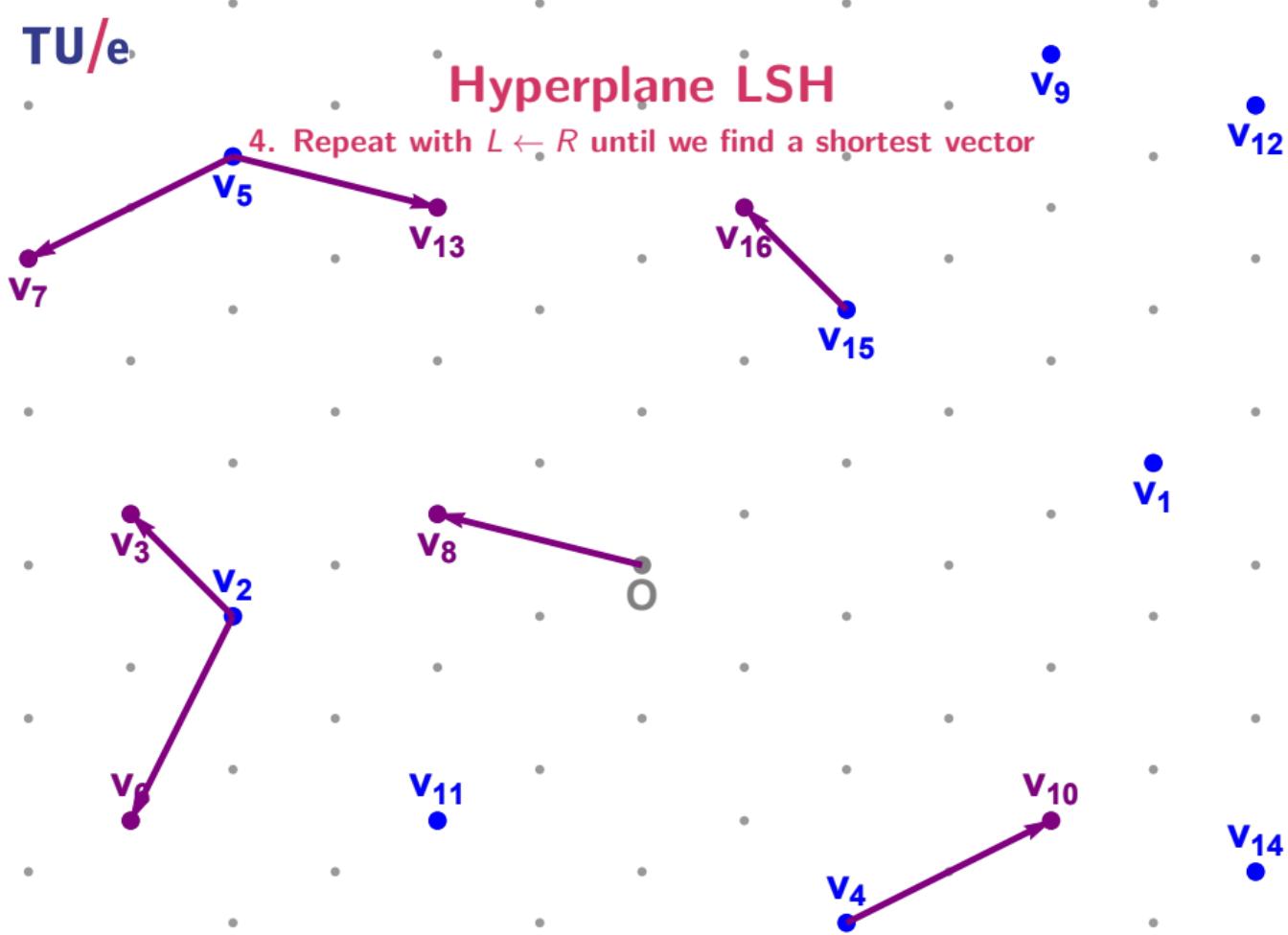
Hyperplane LSH

4. Repeat with $L \leftarrow R$ until we find a shortest vector



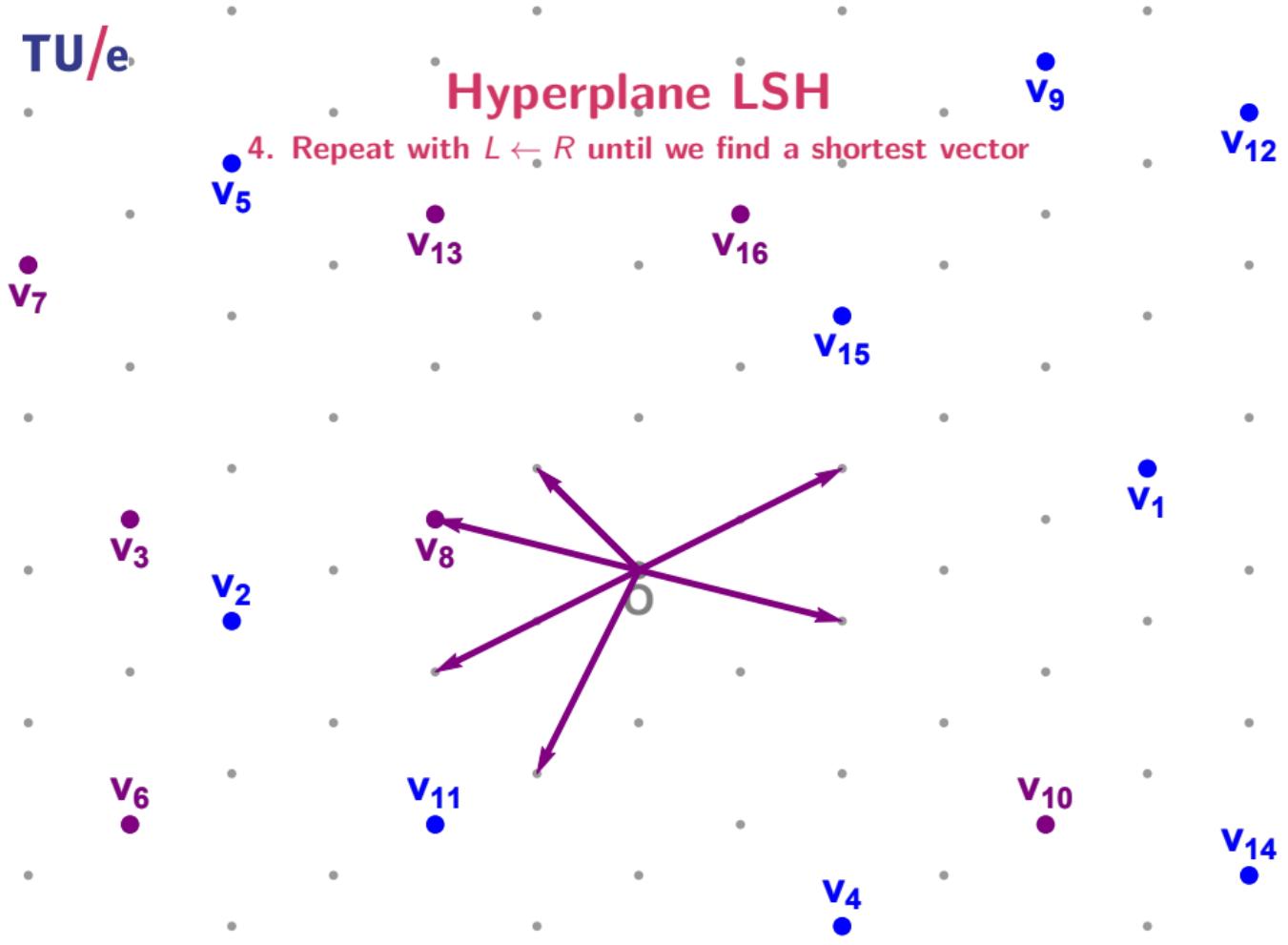
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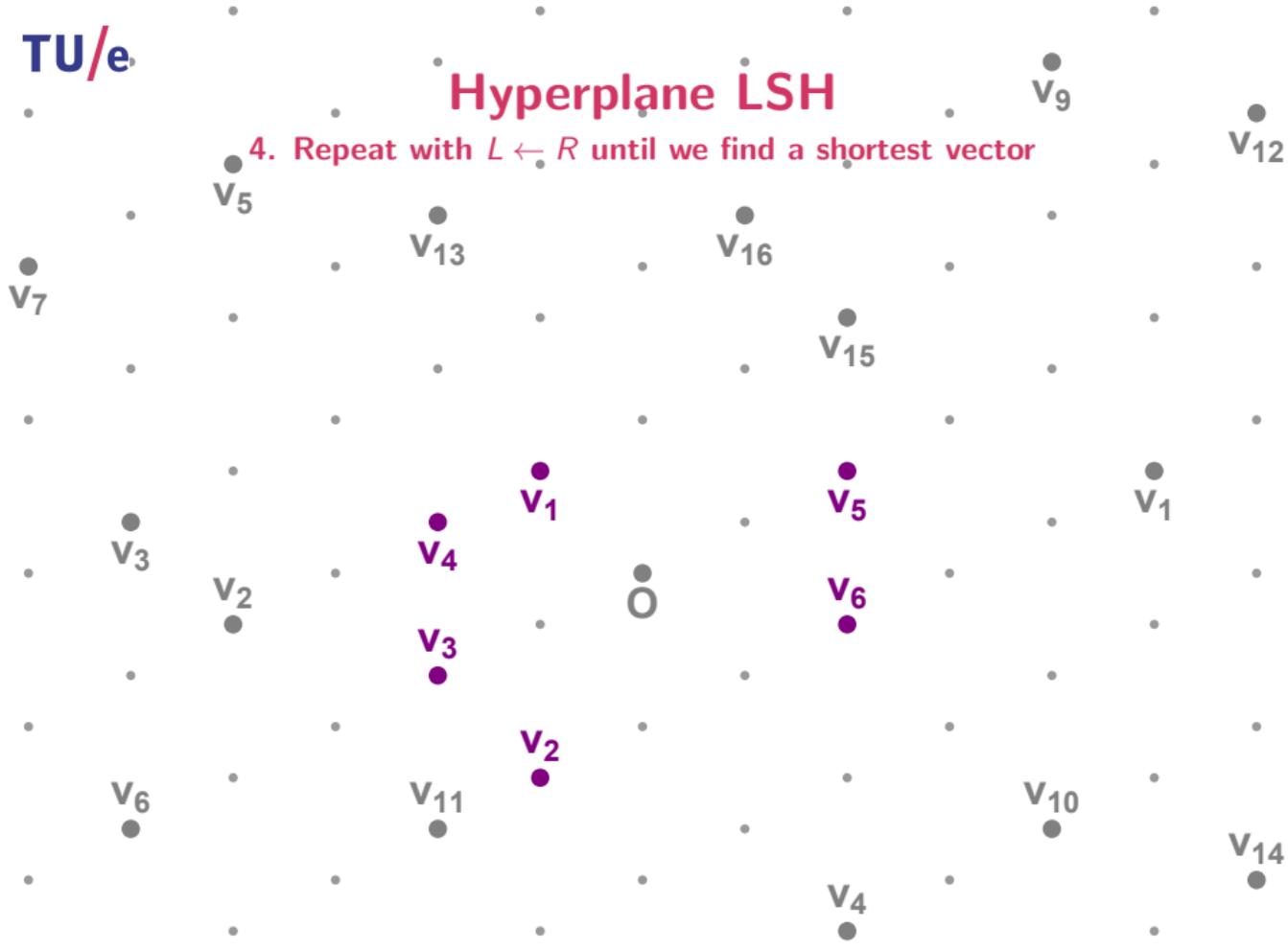
Hyperplane LSH

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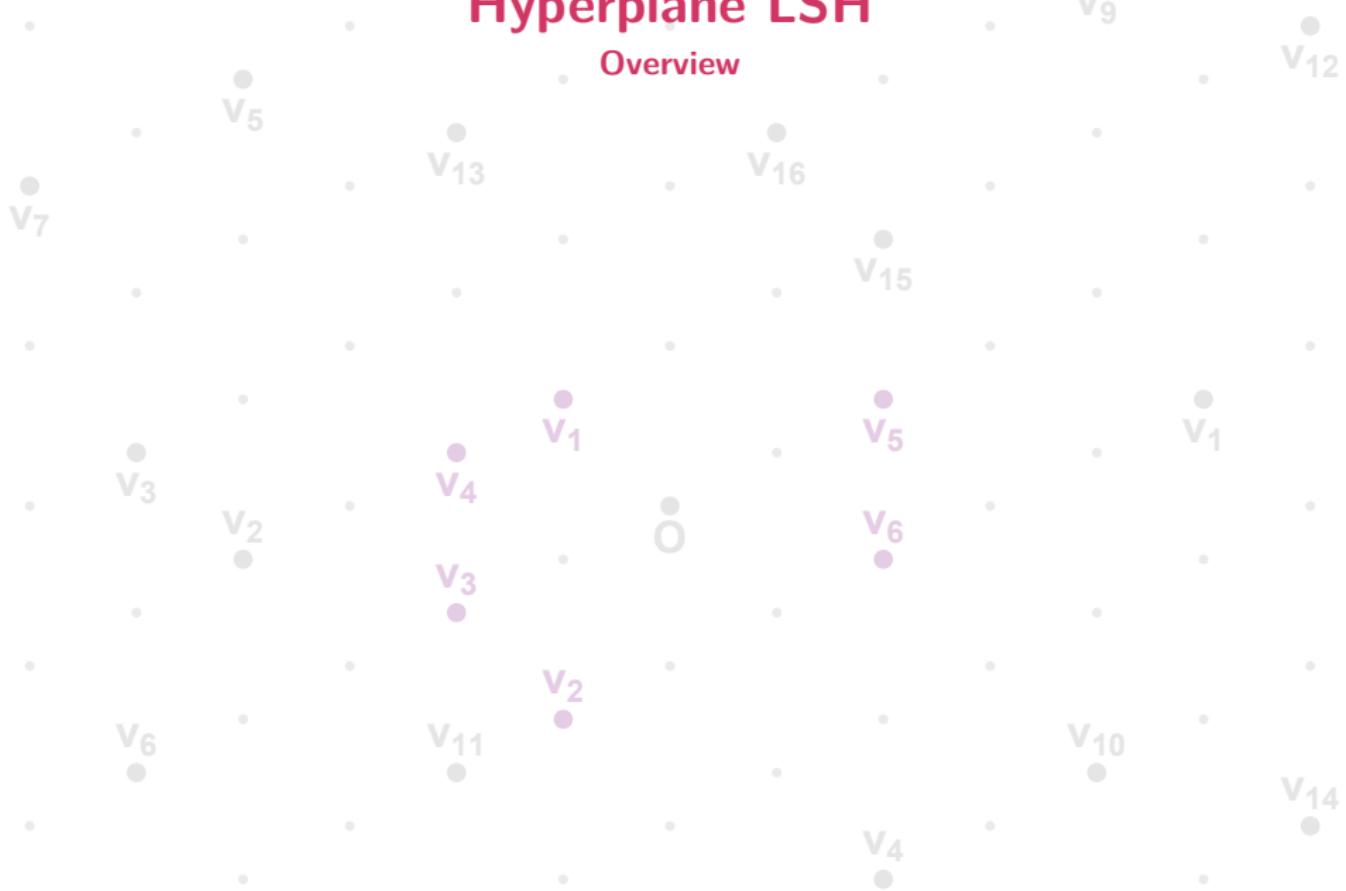
Hyperplane LSH

4. Repeat with $L \leftarrow R$ until we find a shortest vector



Hyperplane LSH

Overview



Hyperplane LSH

Overview

- Two parameters to tune
 - ▶ $k = O(n)$: Number of hyperplanes, leading to 2^k regions
 - ▶ $t = 2^{O(n)}$: Number of different, independent “hash tables”

Hyperplane LSH

Overview

- Two parameters to tune
 - ▶ $k = O(n)$: Number of hyperplanes, leading to 2^k regions
 - ▶ $t = 2^{O(n)}$: Number of different, independent “hash tables”
- Space complexity: $2^{0.337n+o(n)}$
 - ▶ Number of vectors: $2^{0.208n+o(n)}$
 - ▶ Number of hash tables: $2^{0.129n+o(n)}$
 - ▶ Each hash table contains all vectors

Hyperplane LSH

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 - ▶ Candidate nearest vectors: $2^{0.129n+o(n)}$
 - ▶ Repeat this for each list vector: $2^{0.208n+o(n)}$

Hyperplane LSH

Overview

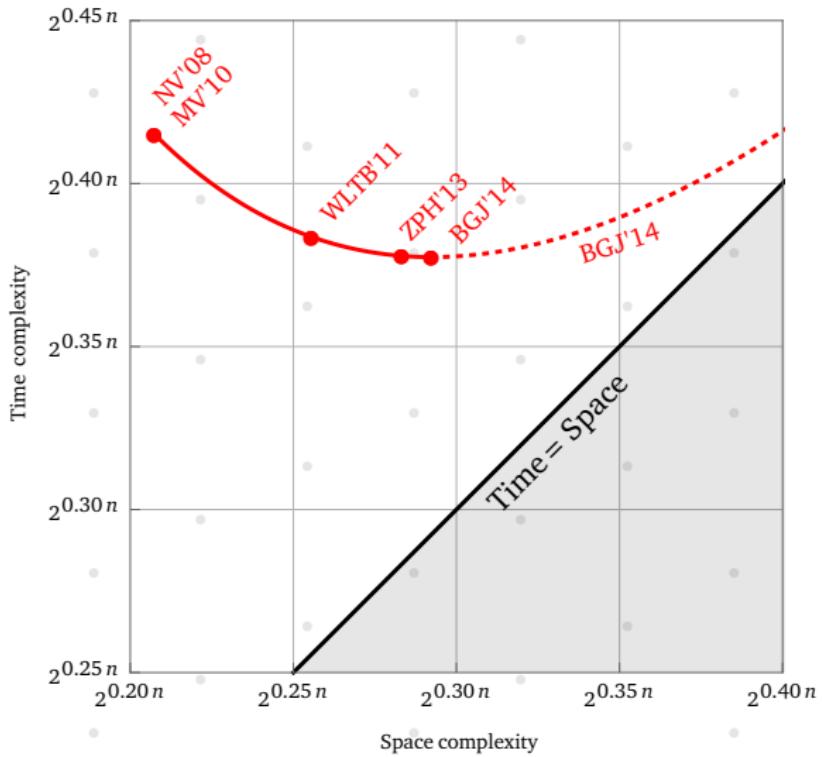
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Theorem

Sieving with hyperplane LSH heuristically solves SVP in time and space $2^{0.337n+o(n)}$.

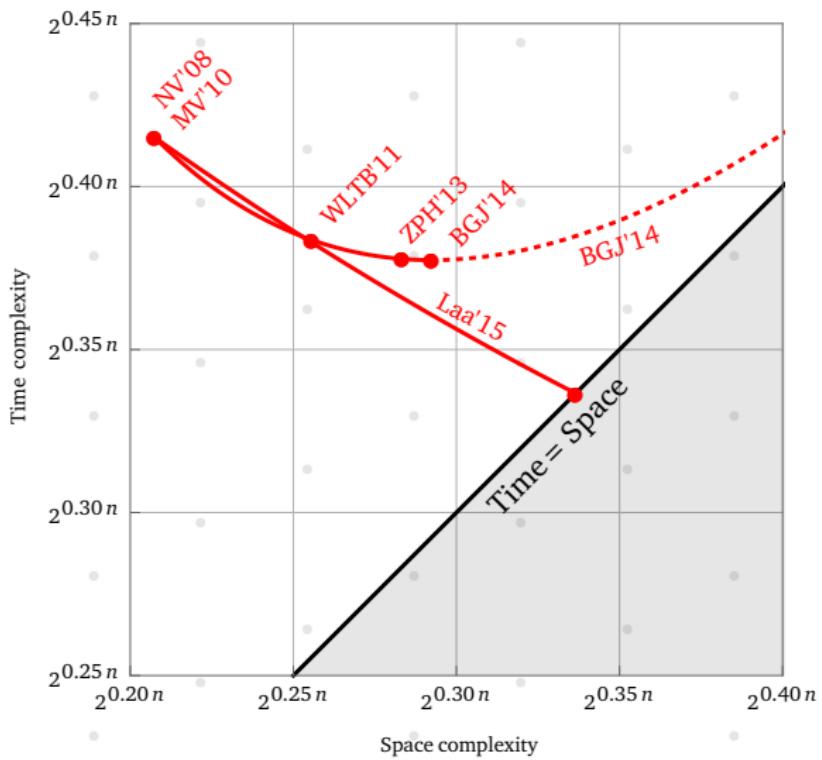
Hyperplane LSH

Space/time trade-off



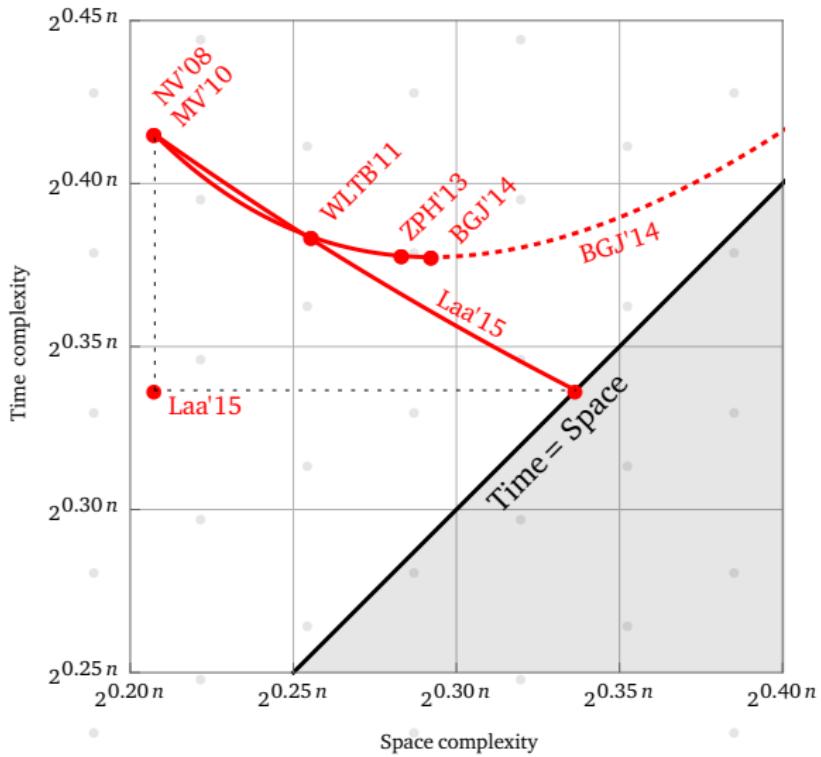
Hyperplane LSH

Space/time trade-off



Hyperplane LSH

Space/time trade-off



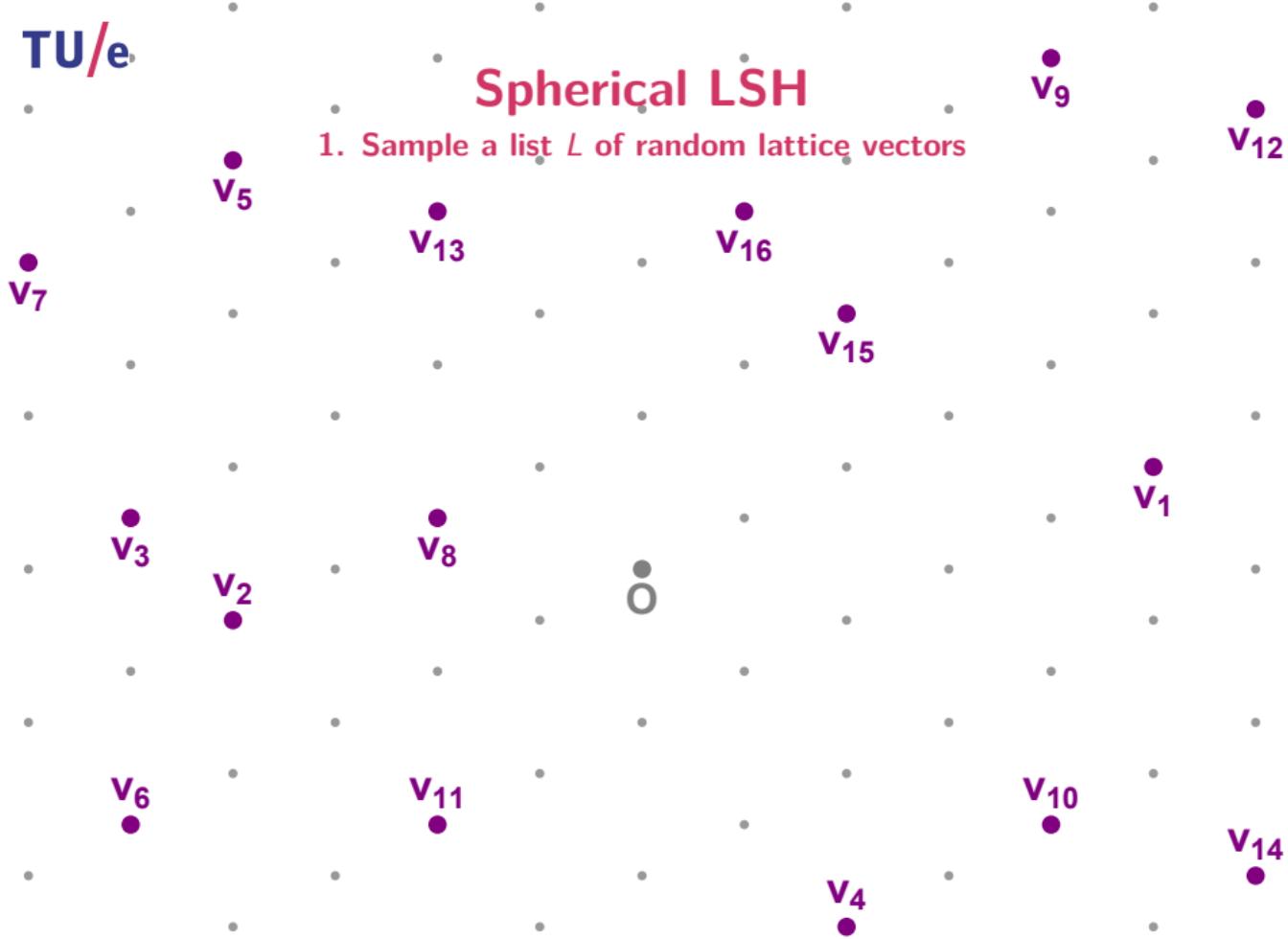
Spherical LSH

1. Sample a list L of random lattice vectors



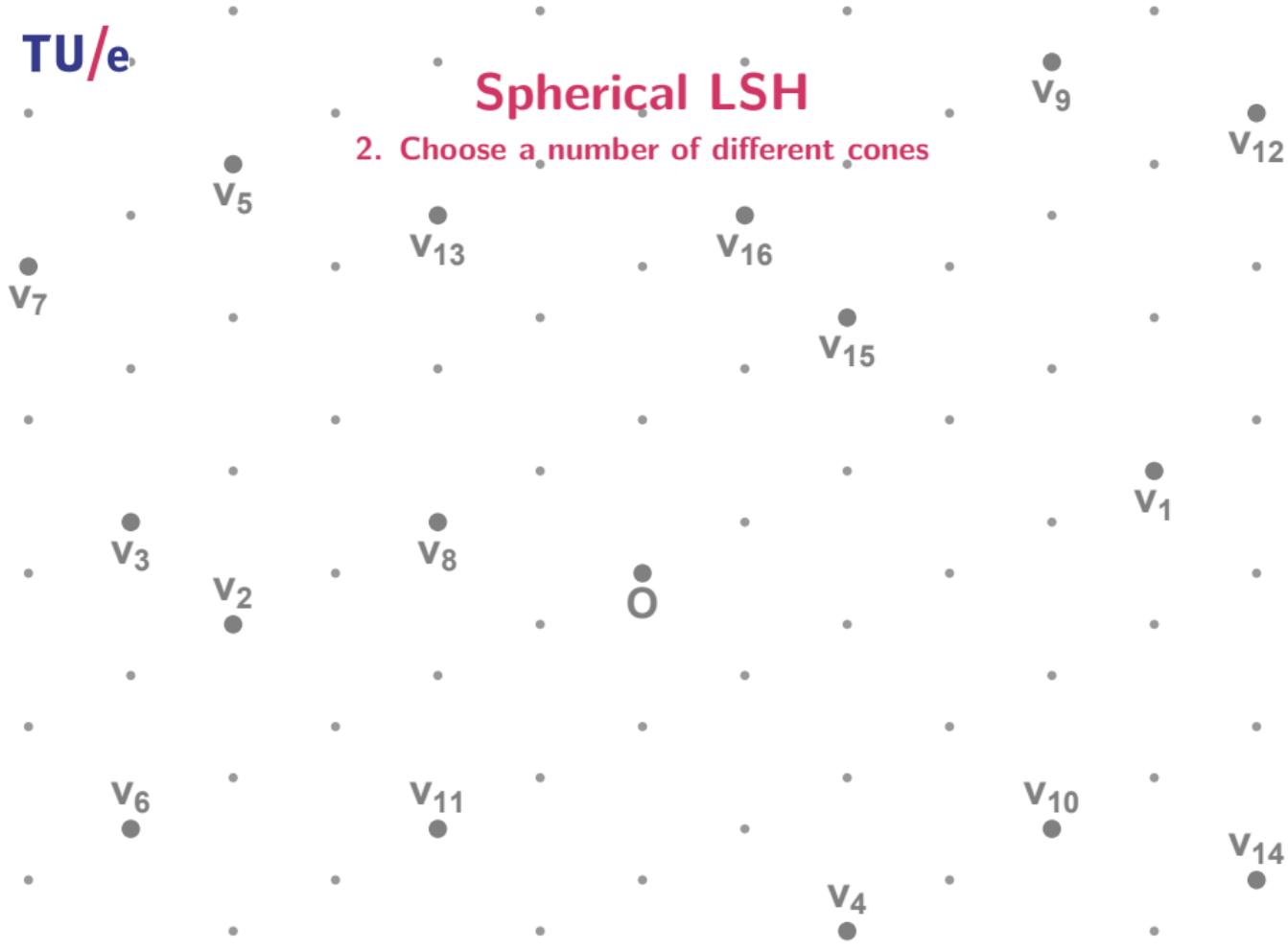
Spherical LSH

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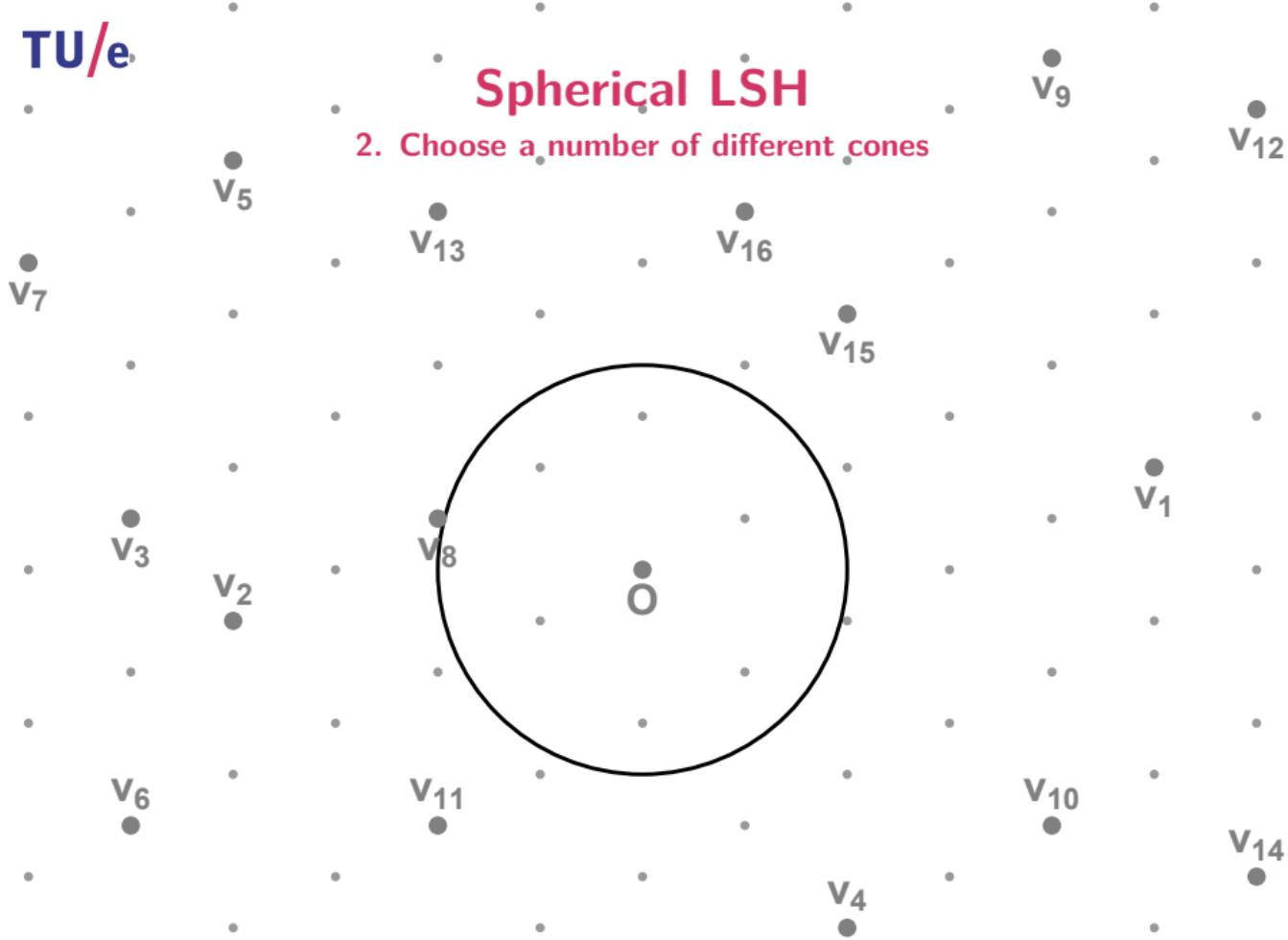
Spherical LSH

2. Choose a number of different cones



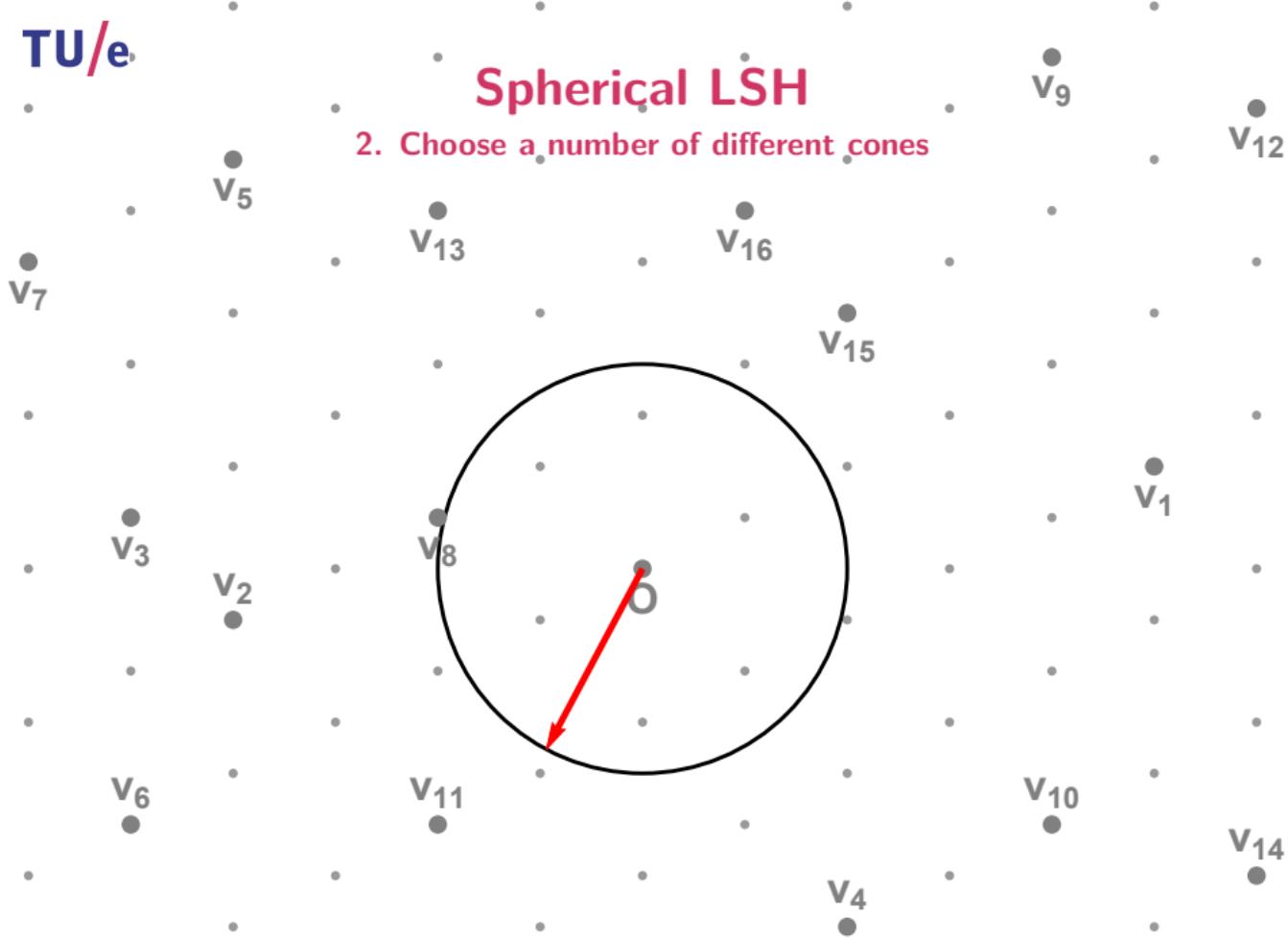
Spherical LSH

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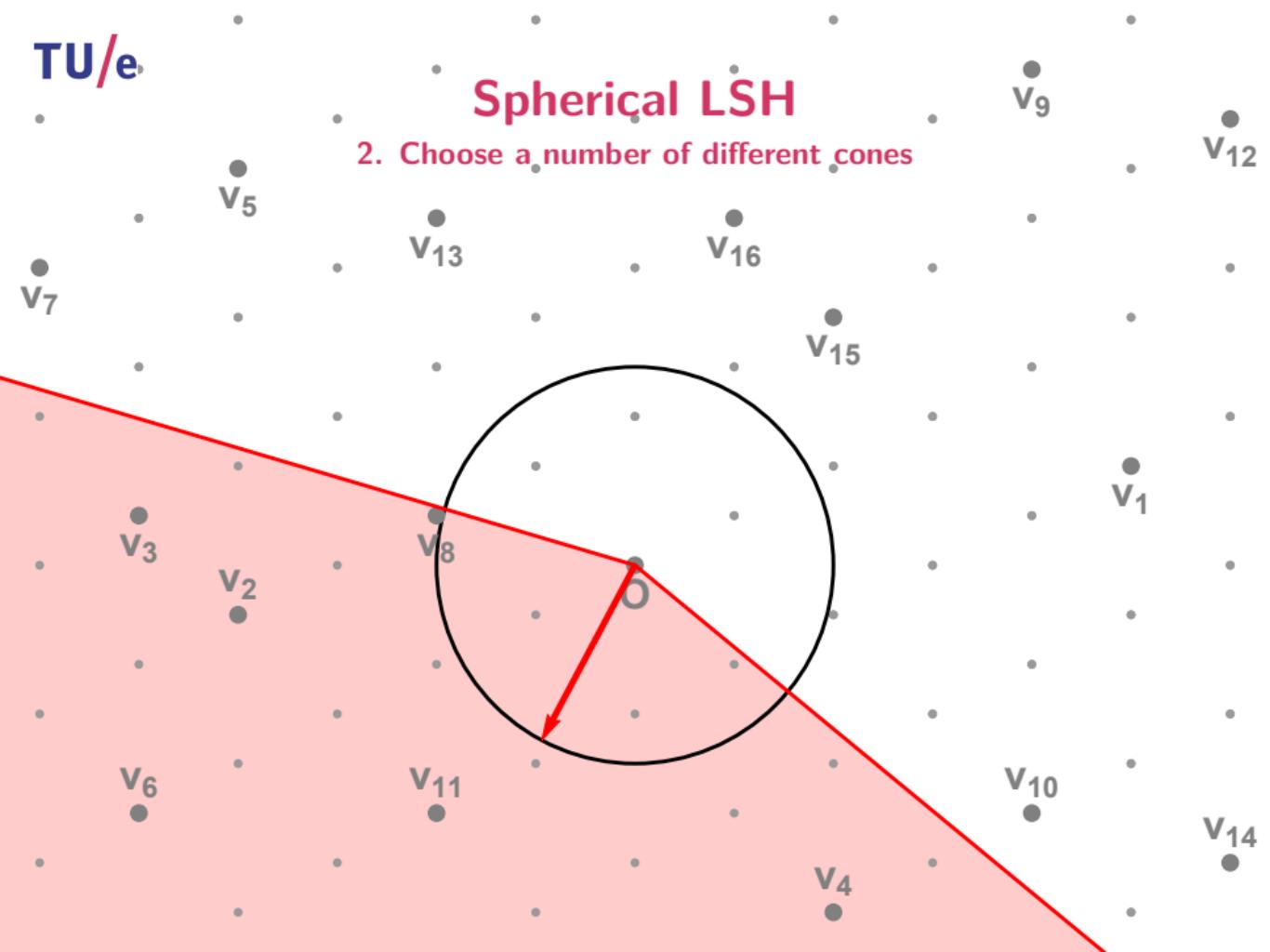
Spherical LSH

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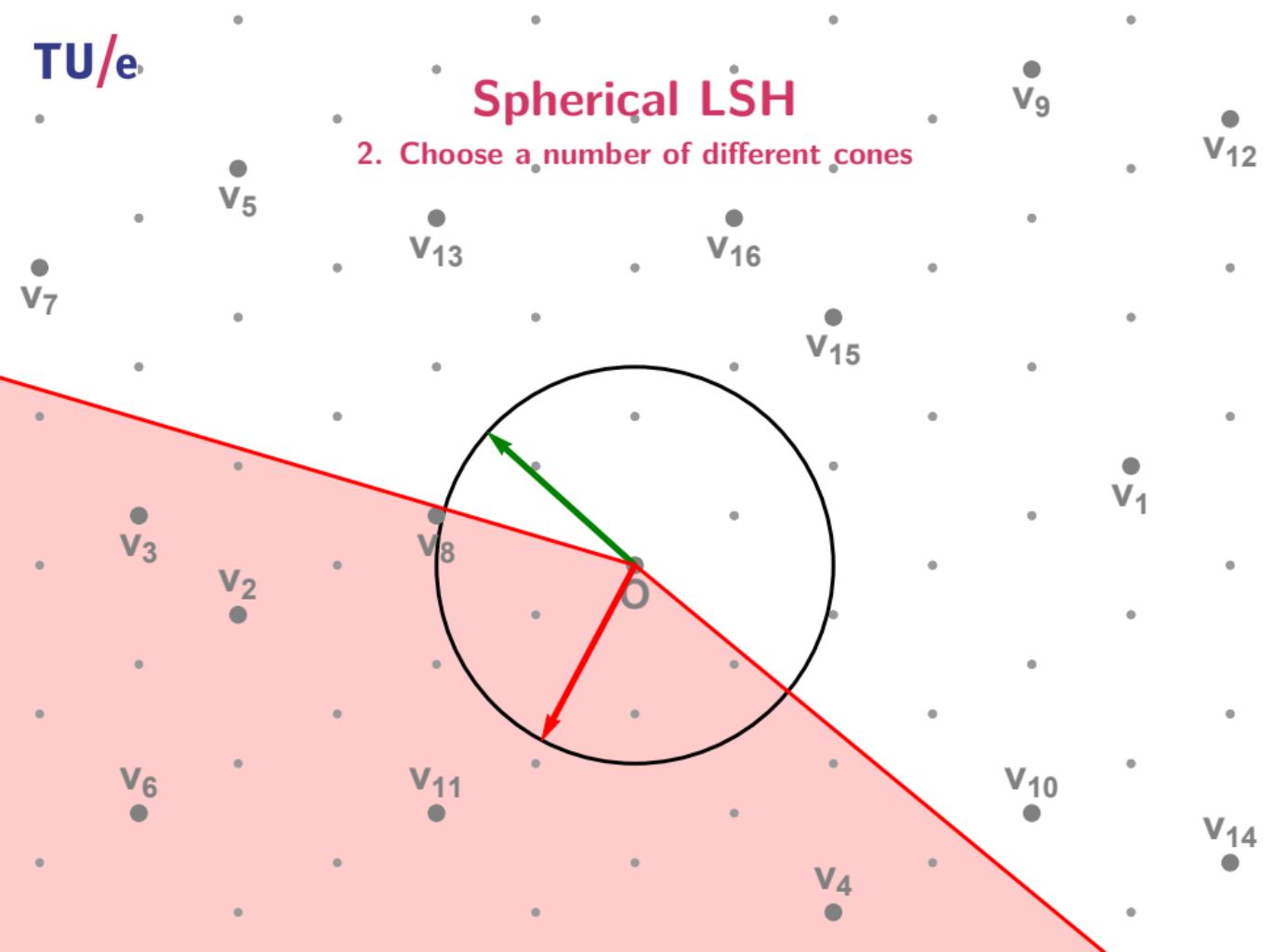
Spherical LSH

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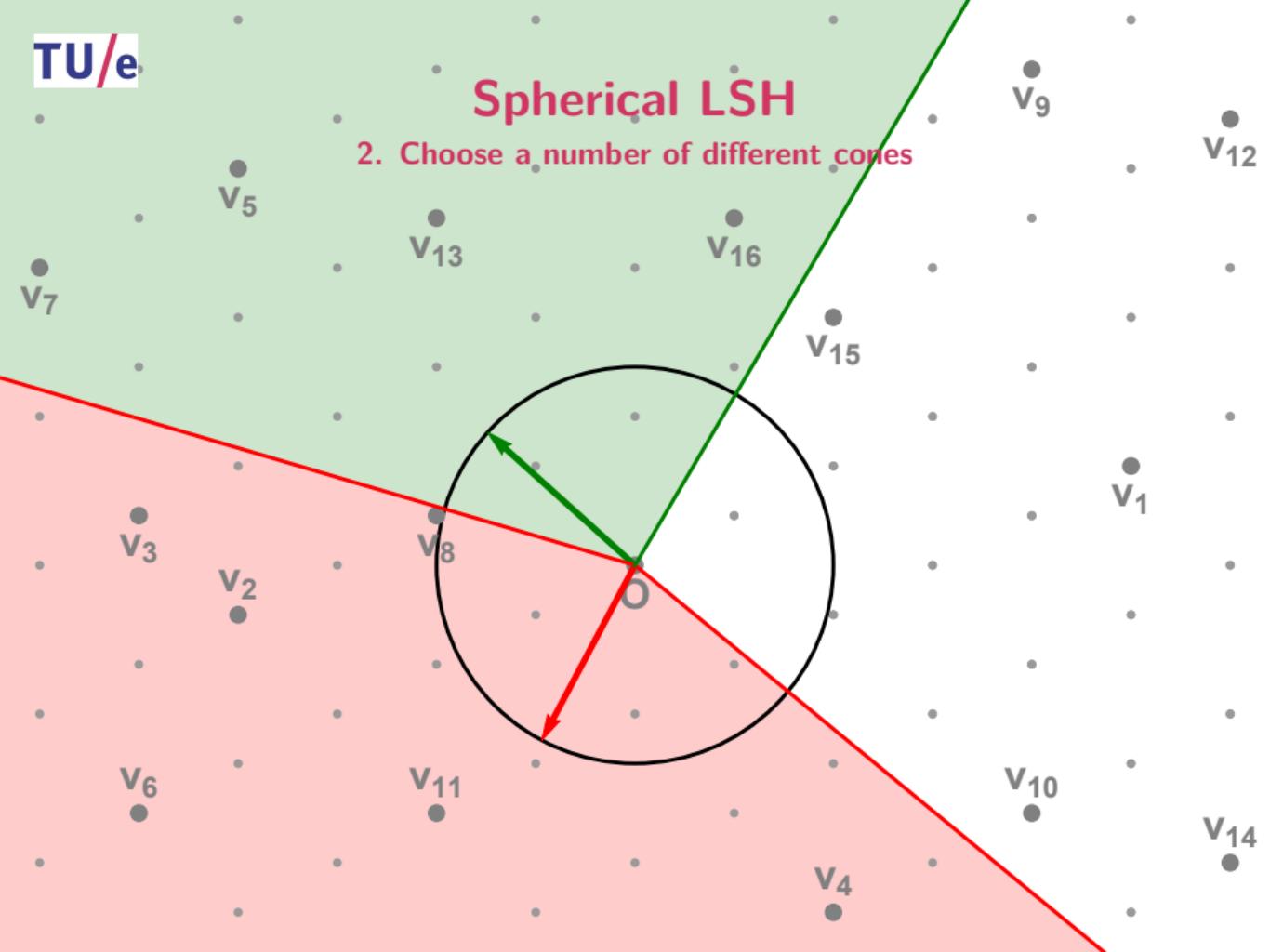
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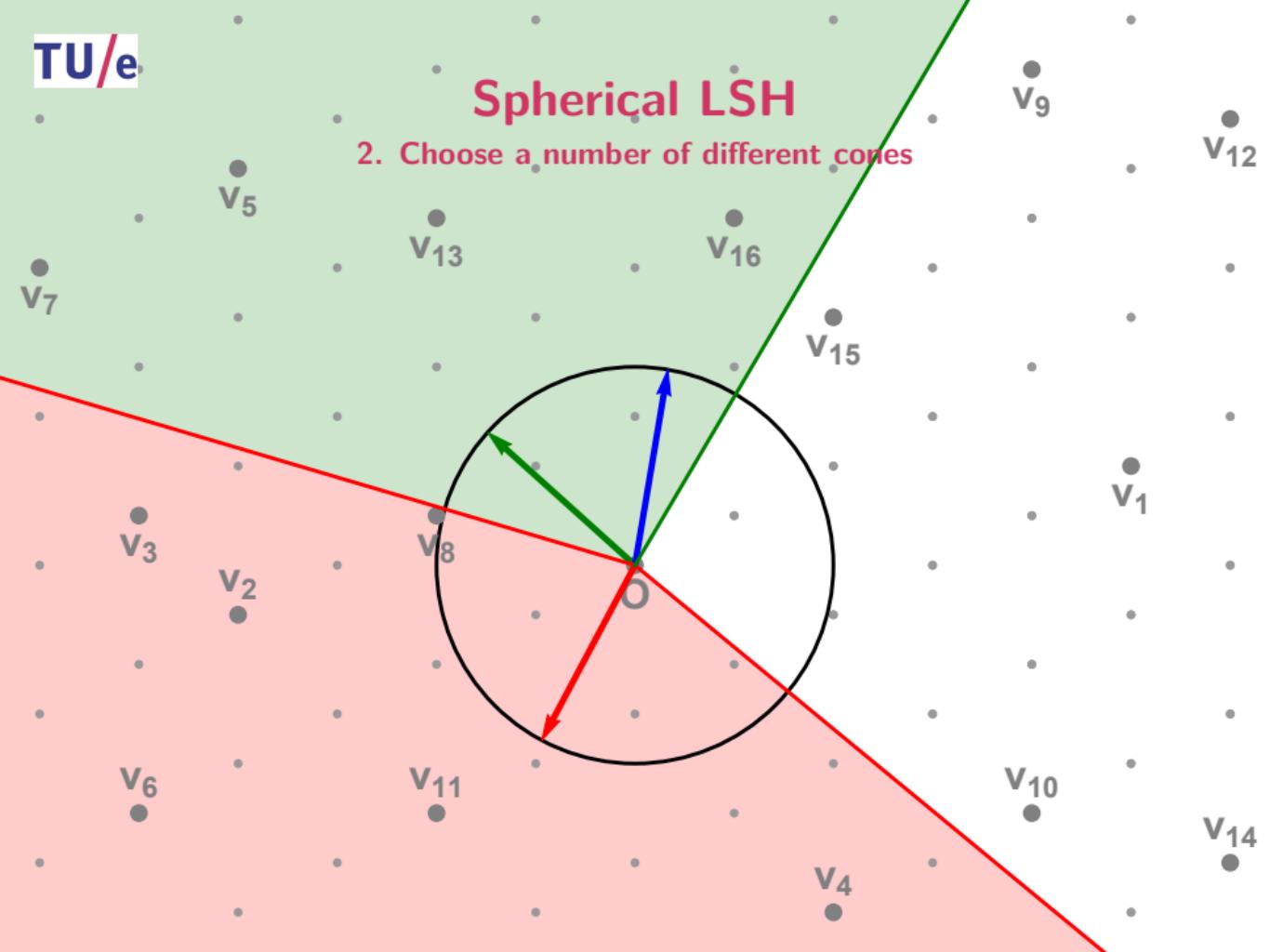
Spherical LSH

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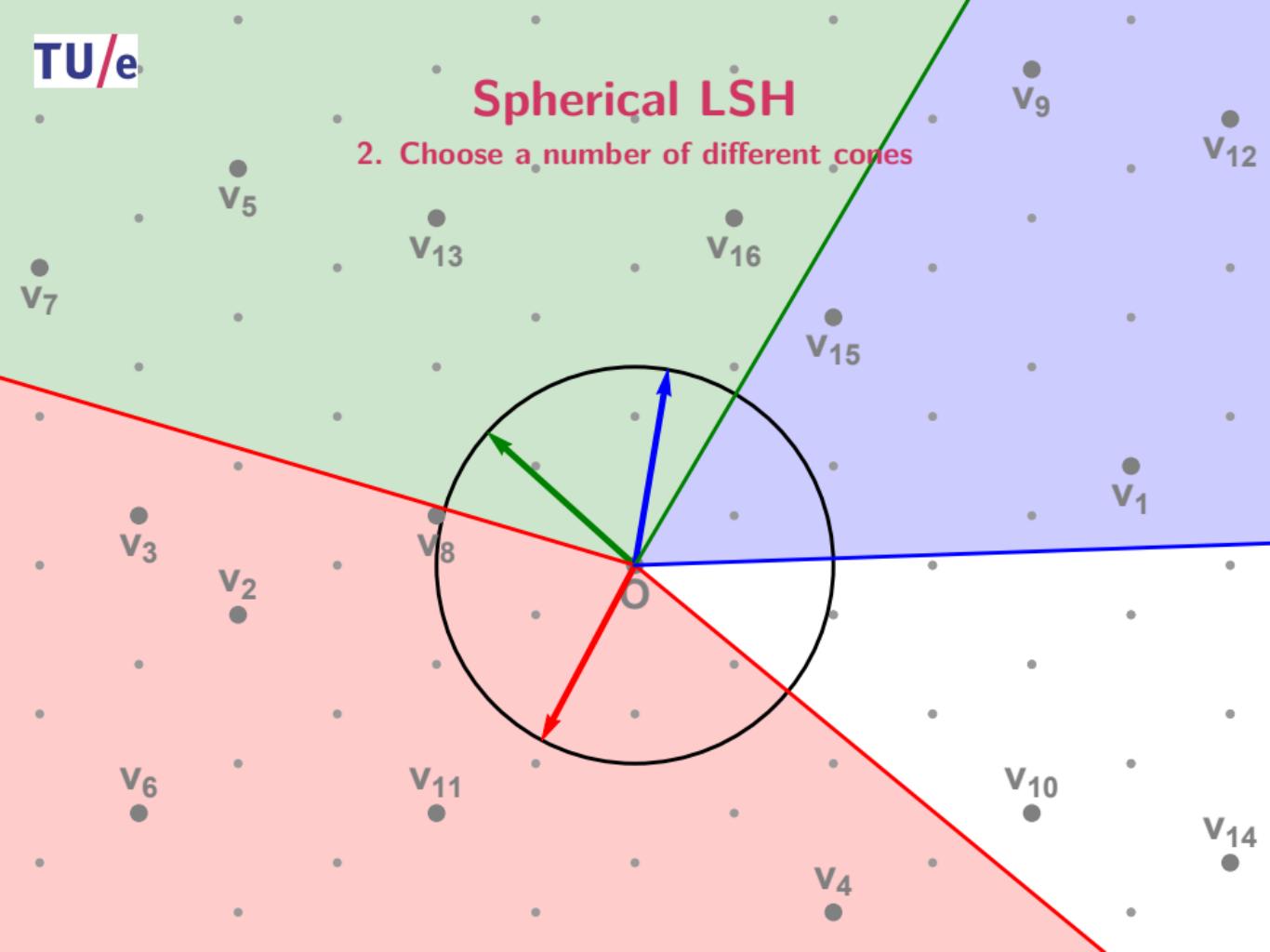
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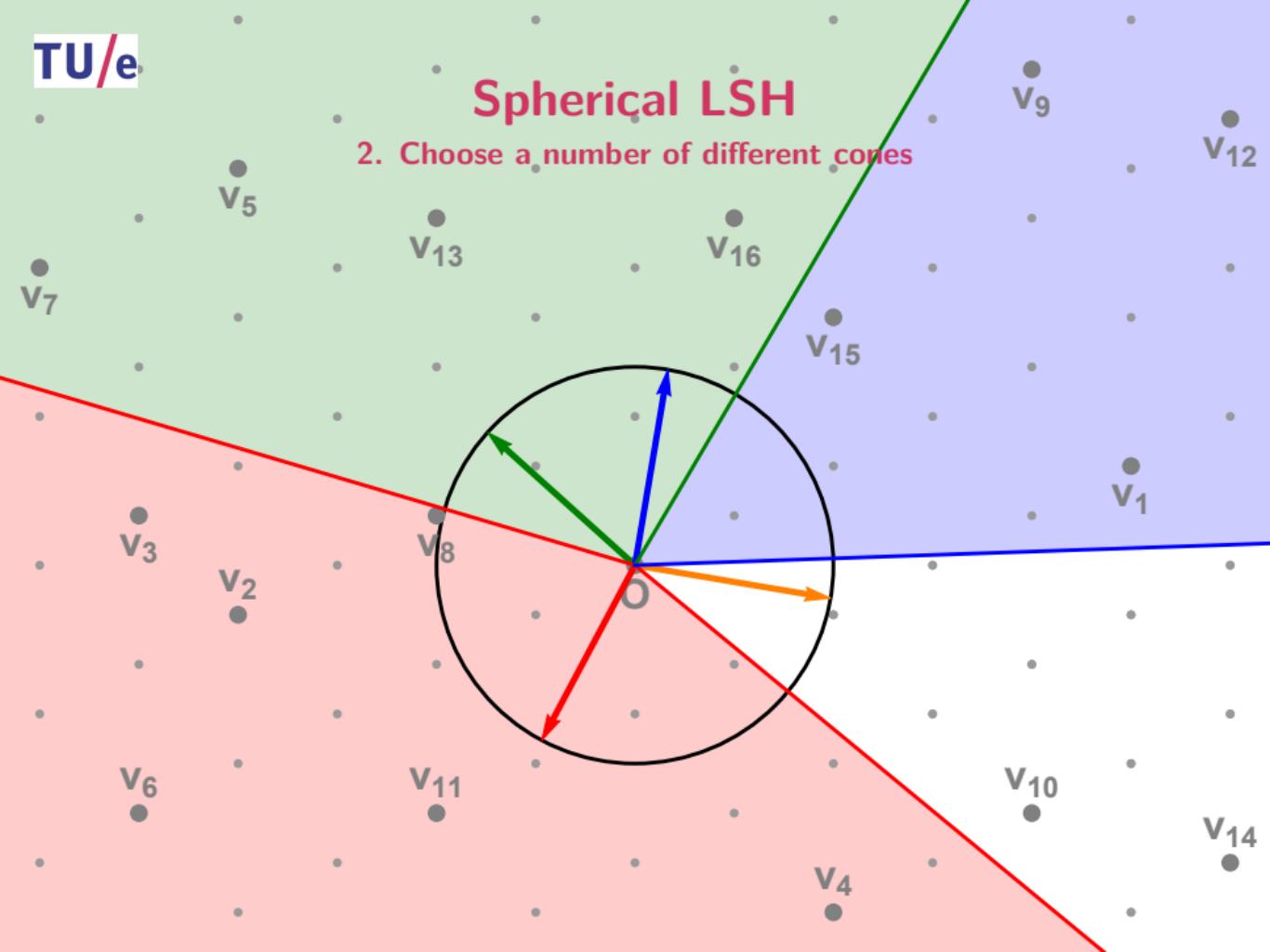
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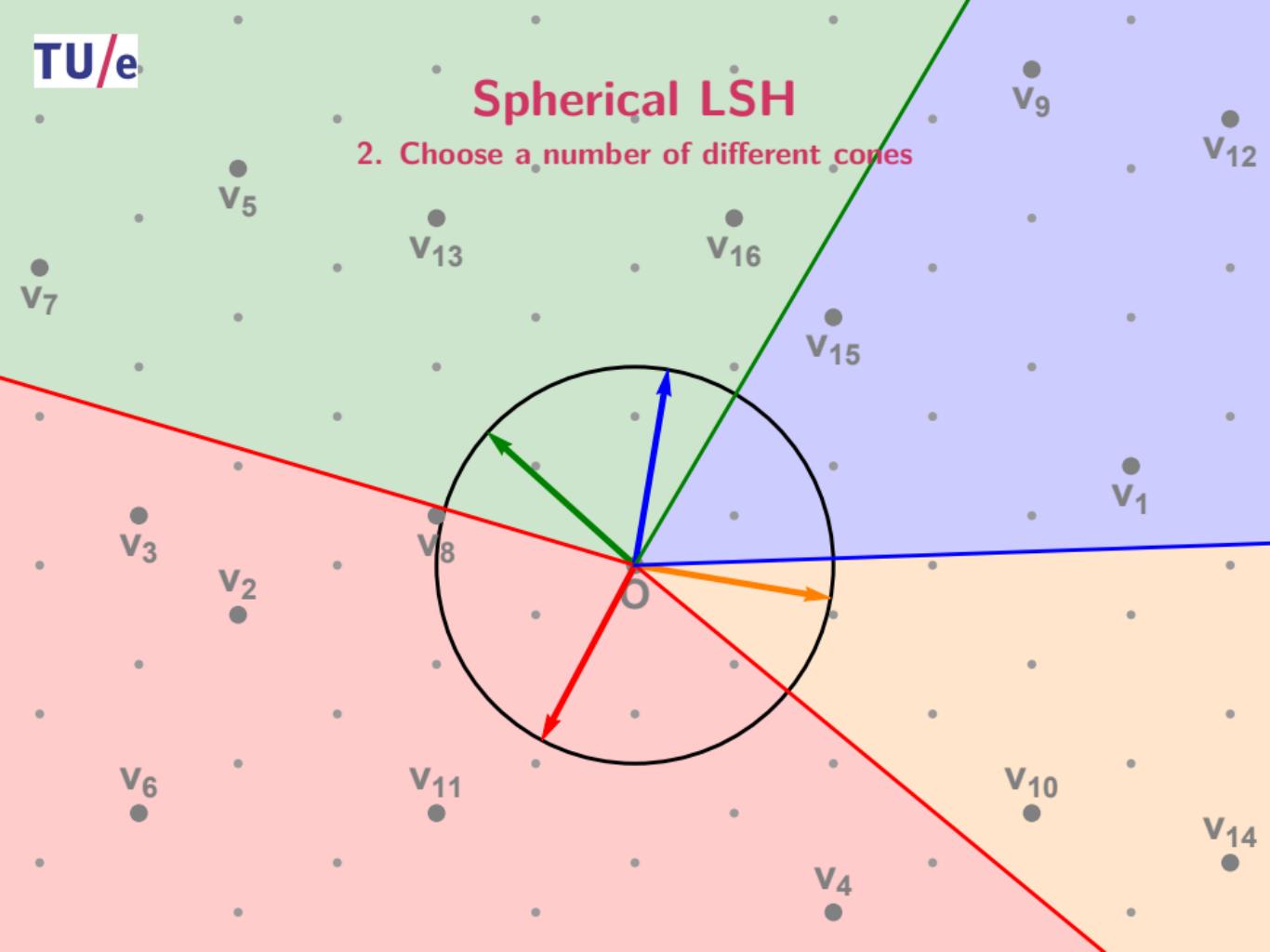
Spherical LSH

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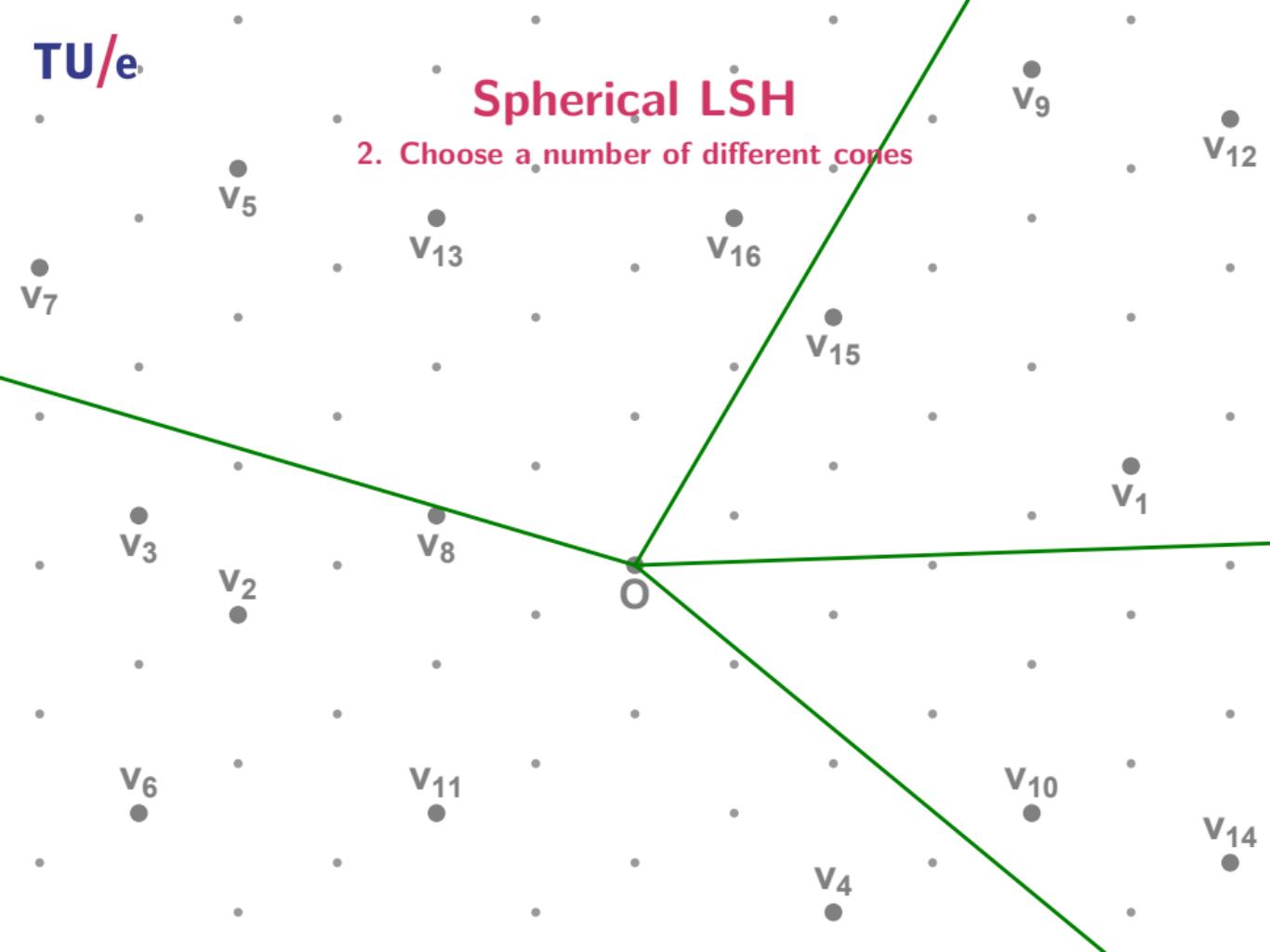
Spherical LSH

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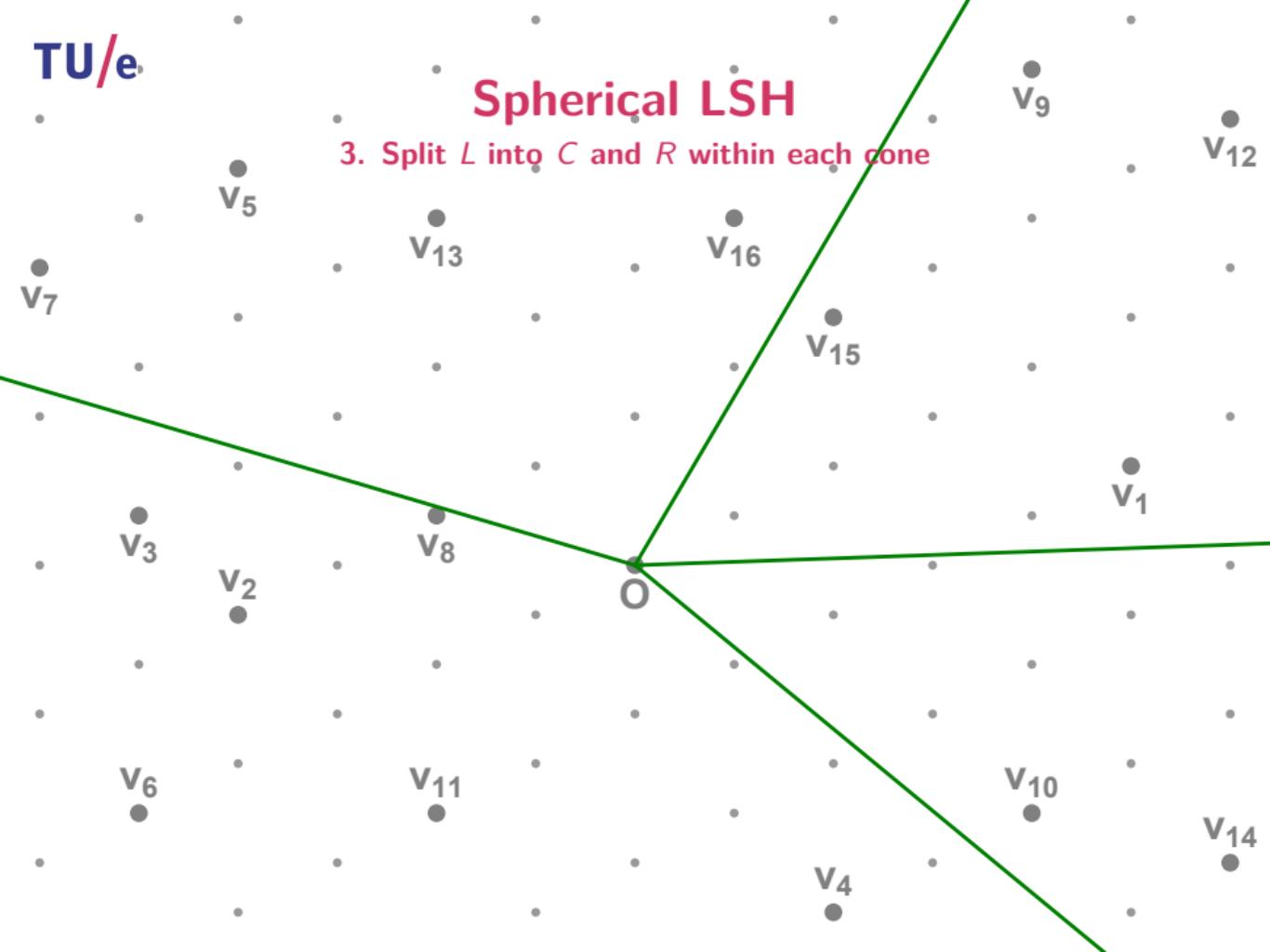
Spherical LSH

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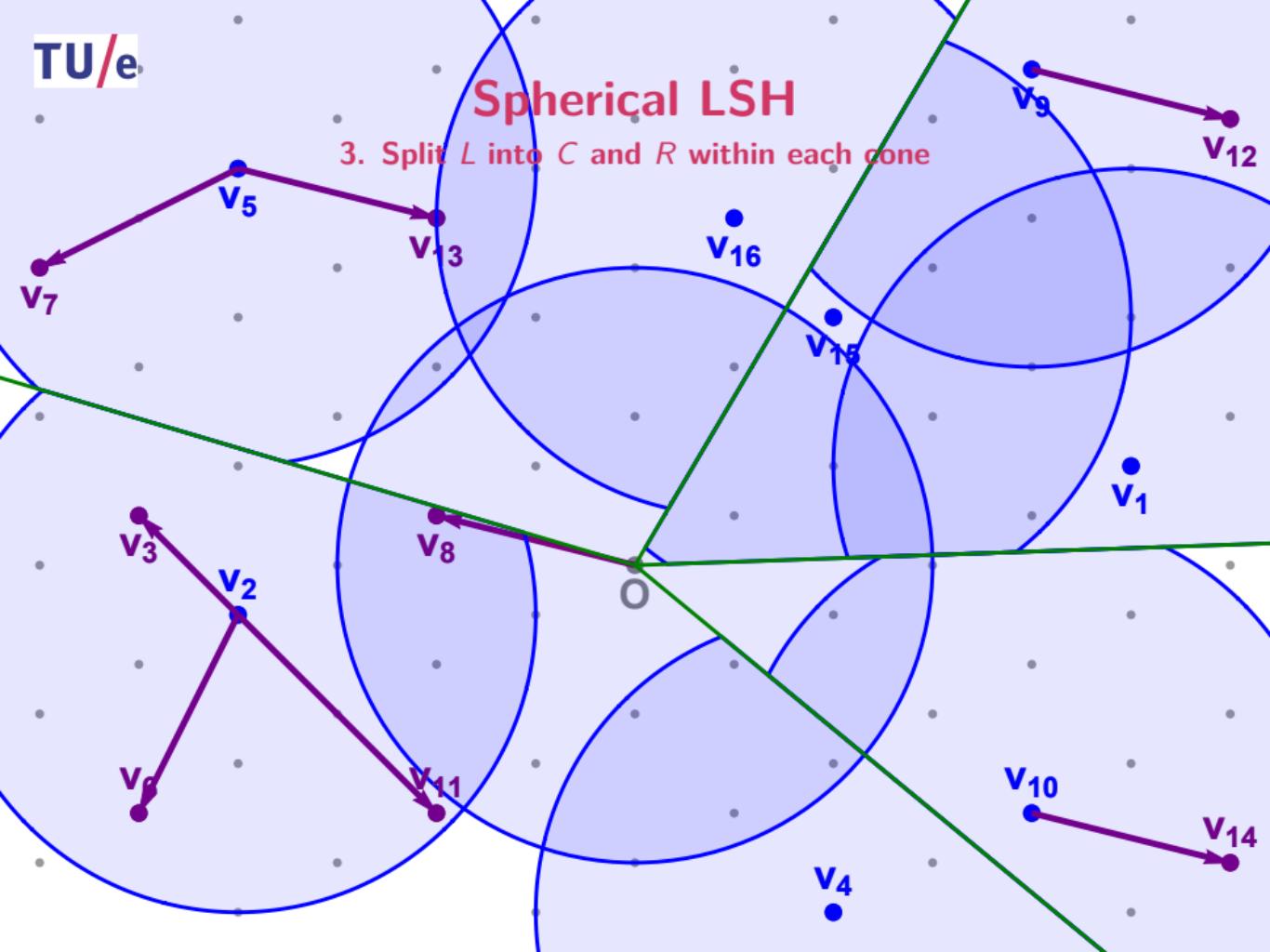
Spherical LSH

3. Split L into C and R within each cone



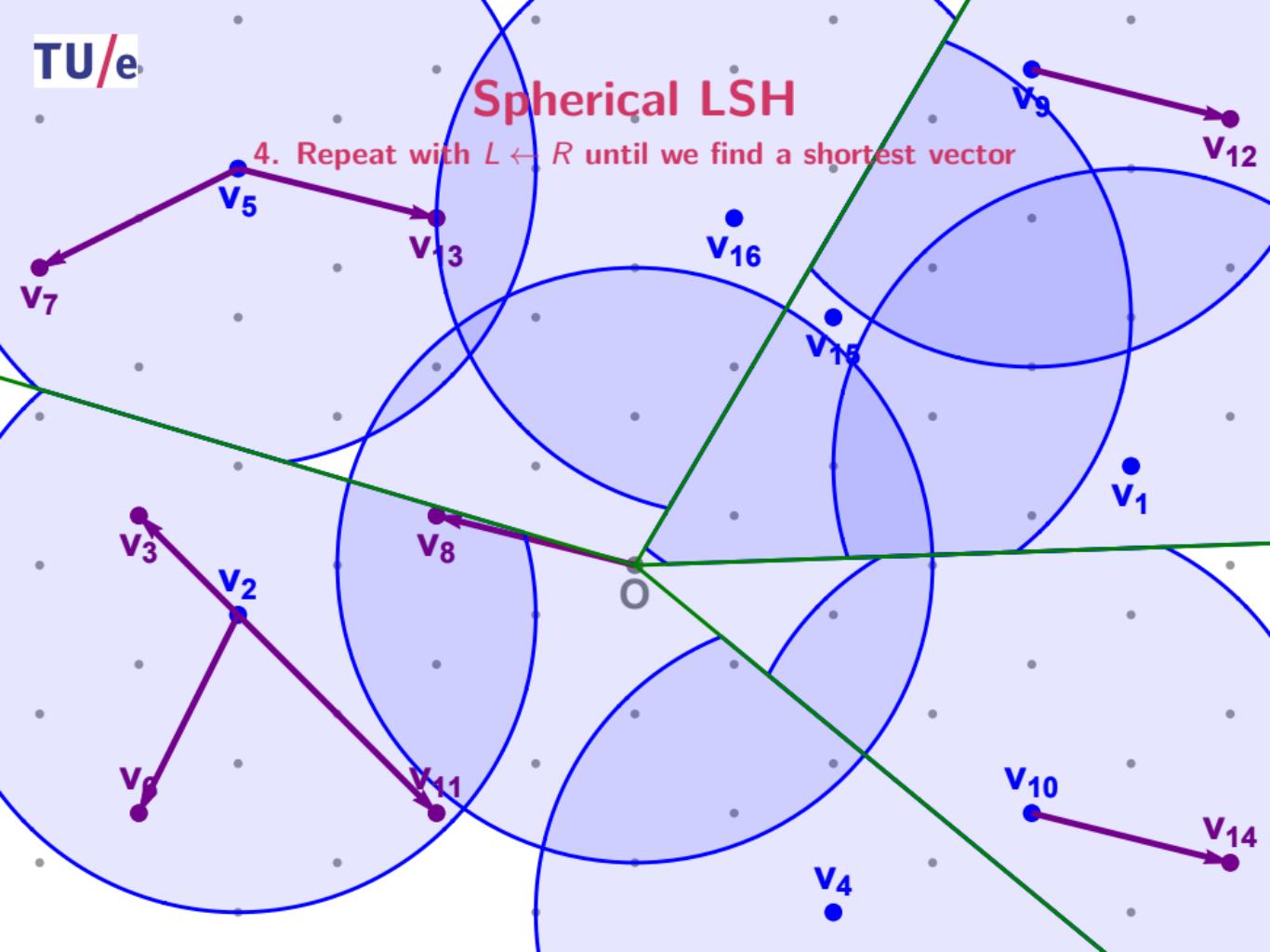
Spherical LSH

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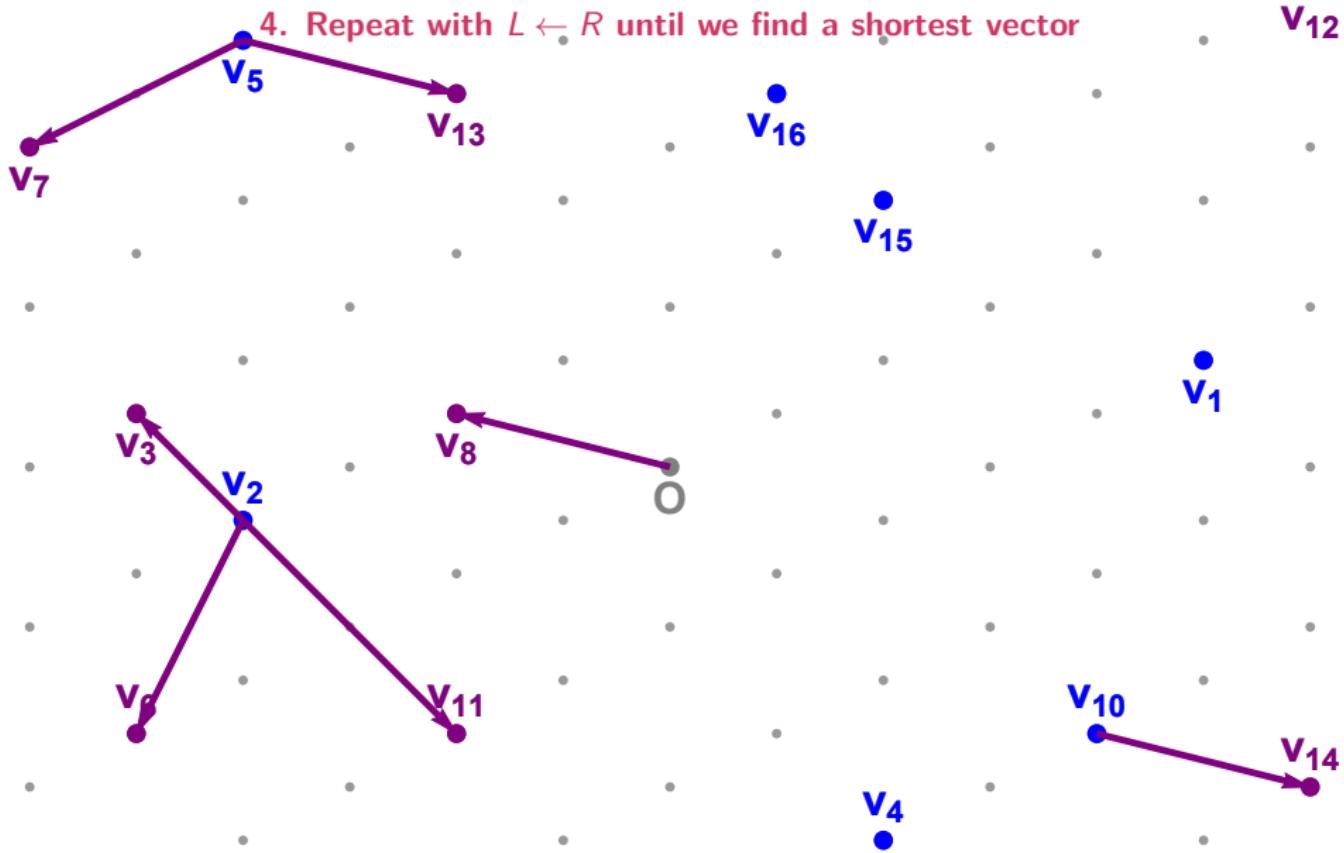
Spherical LSH

4. Repeat with $L \leftarrow R$ until we find a shortest vector



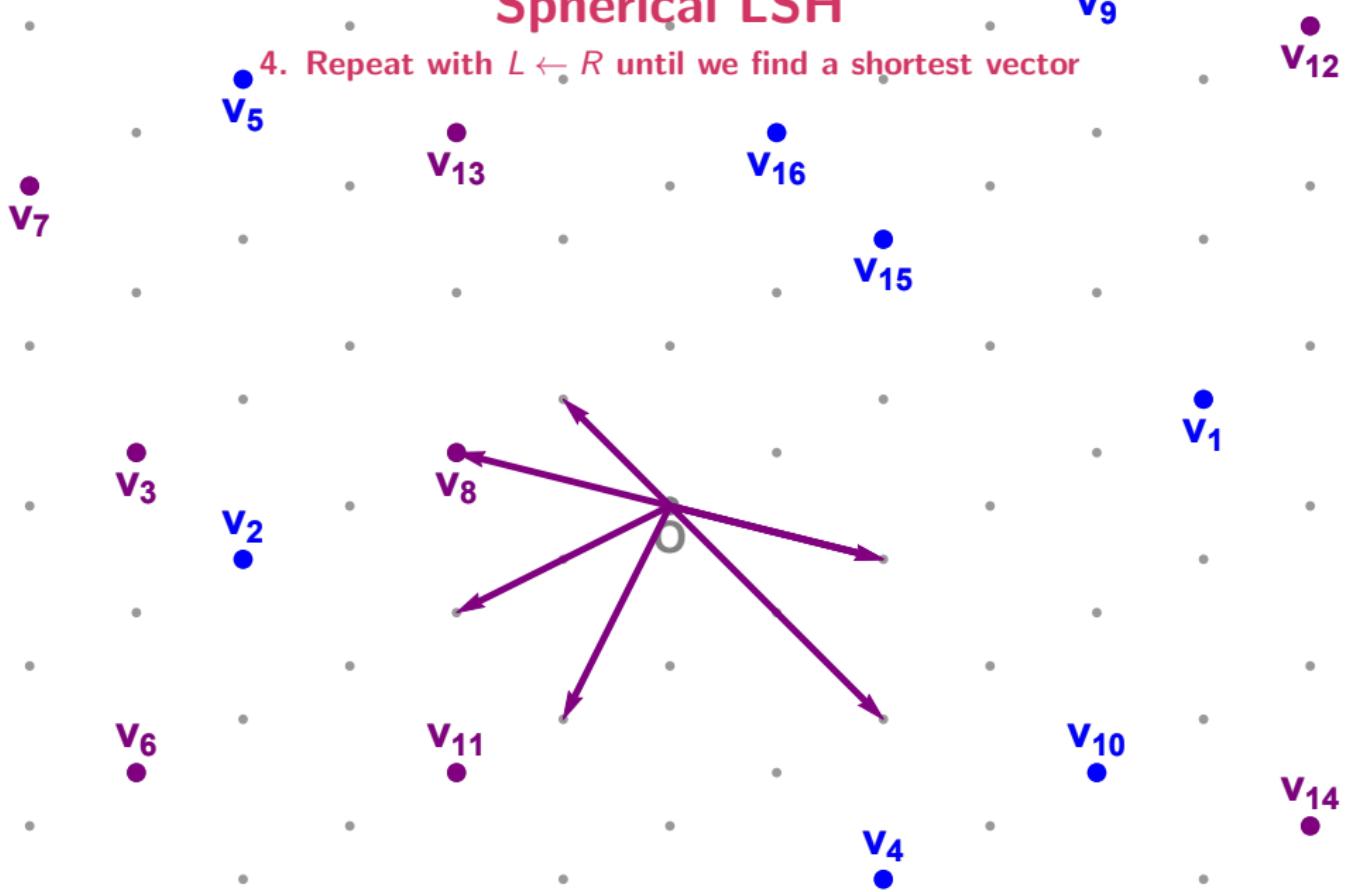
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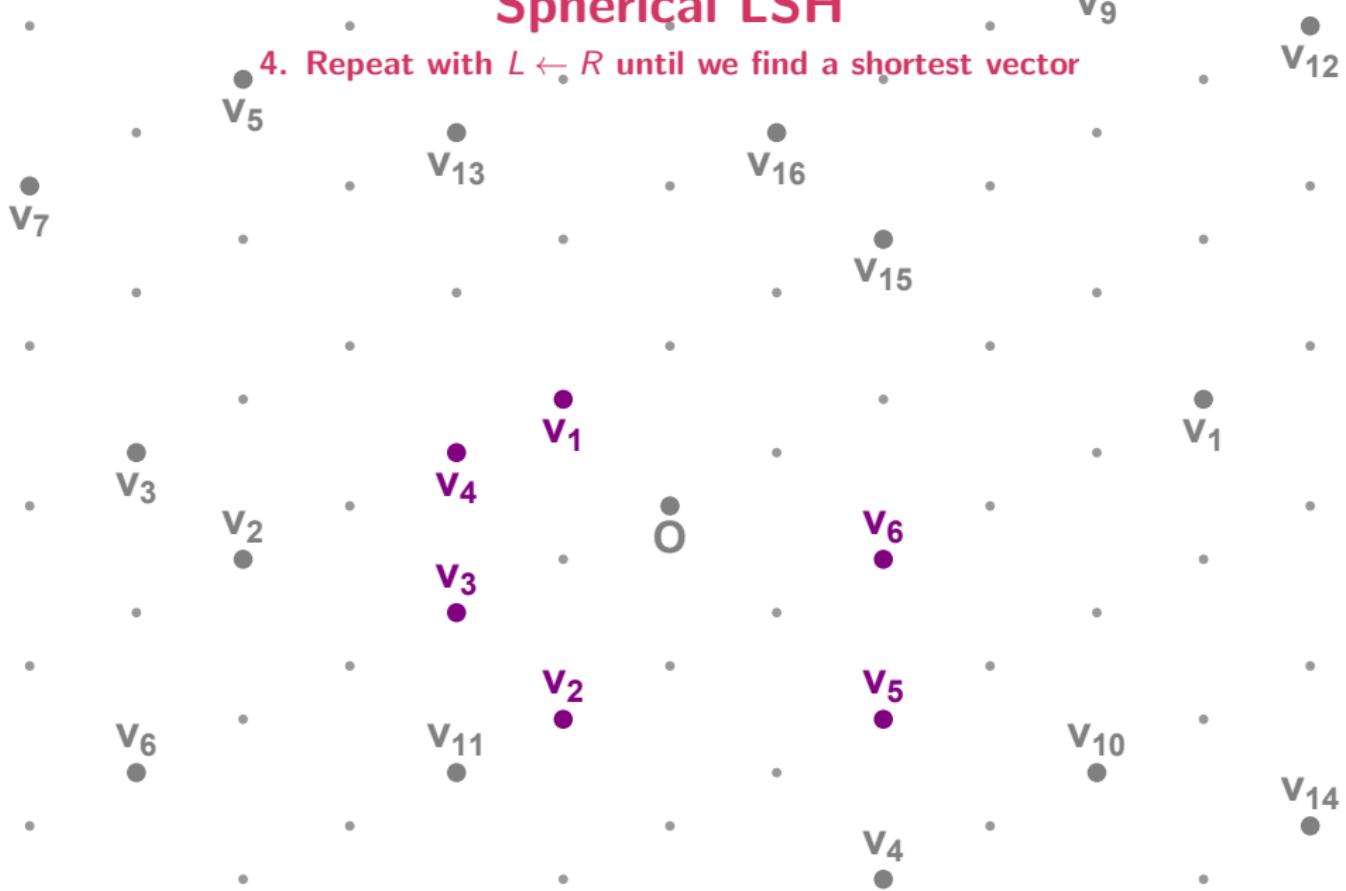
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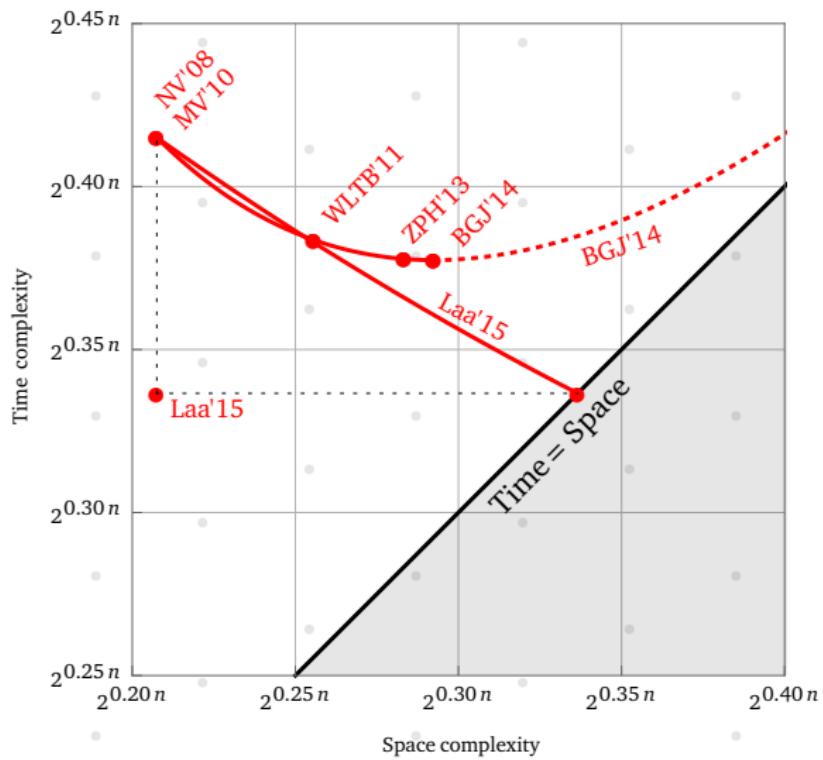
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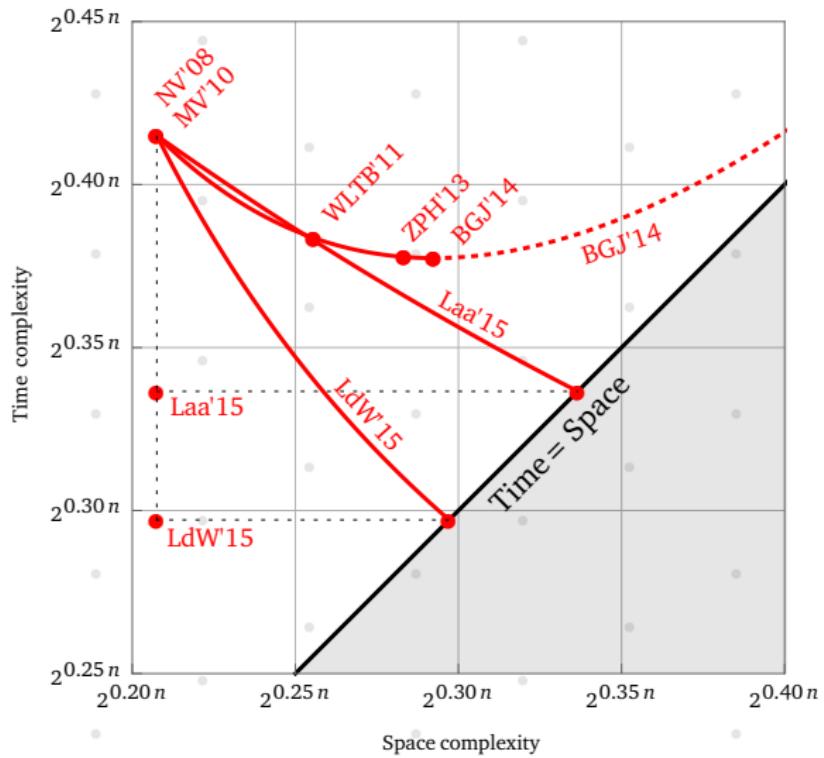
Spherical LSH

Space/time trade-off



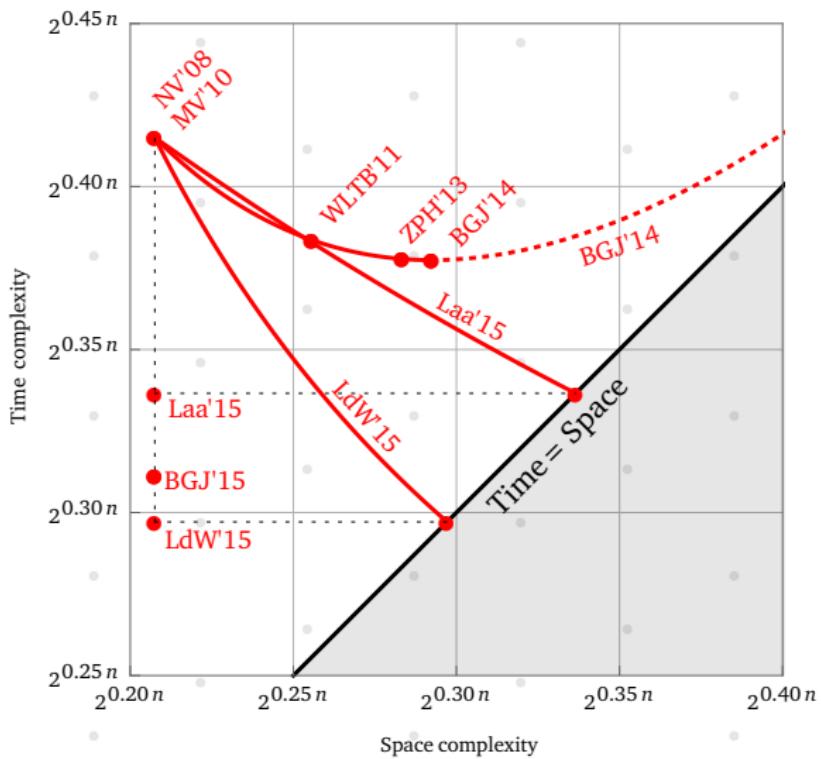
Spherical LSH

Space/time trade-off



May and Ozerov's NNS method

Space/time trade-off



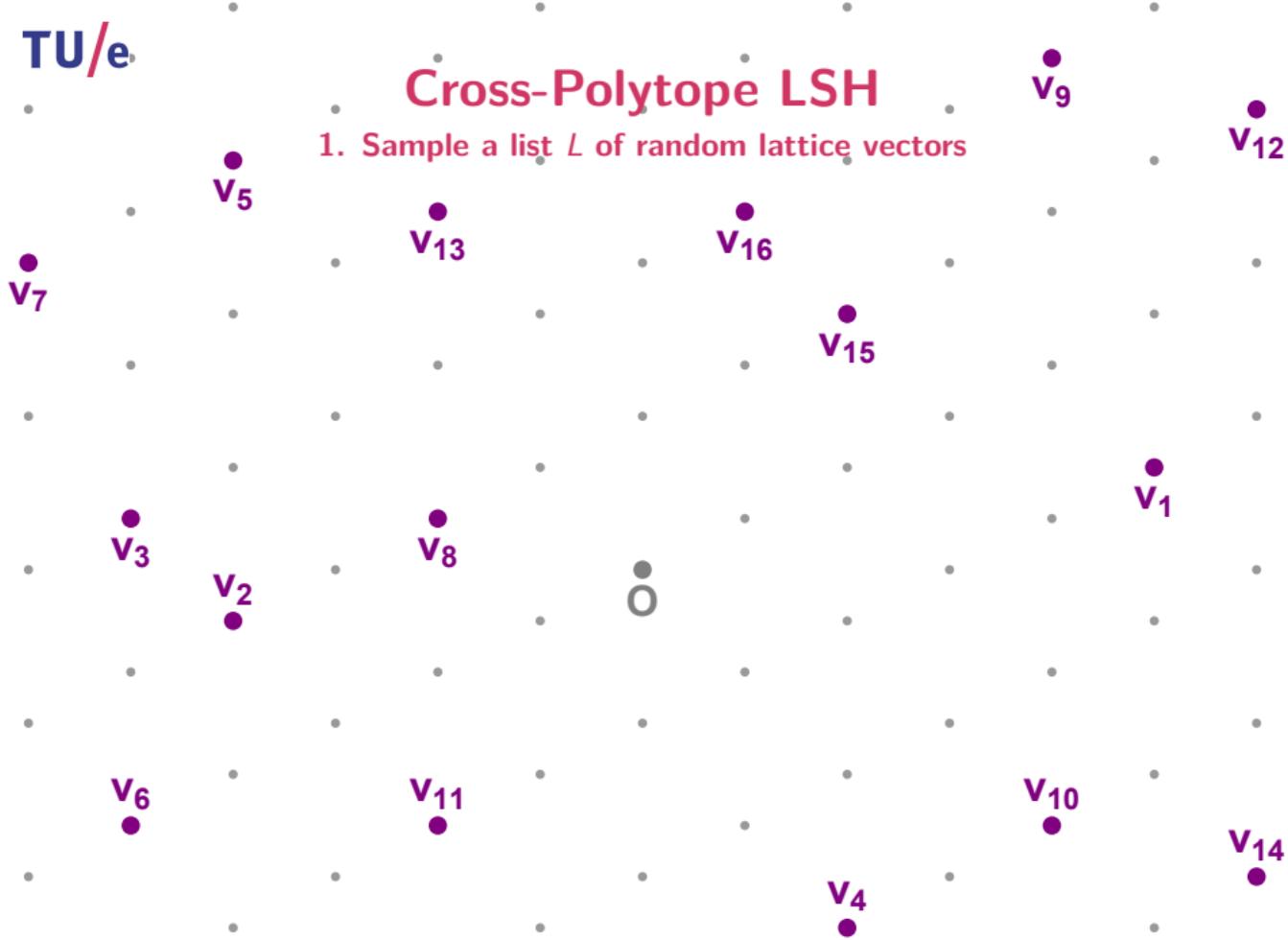
Cross-Polytope LSH

1. Sample a list L of random lattice vectors



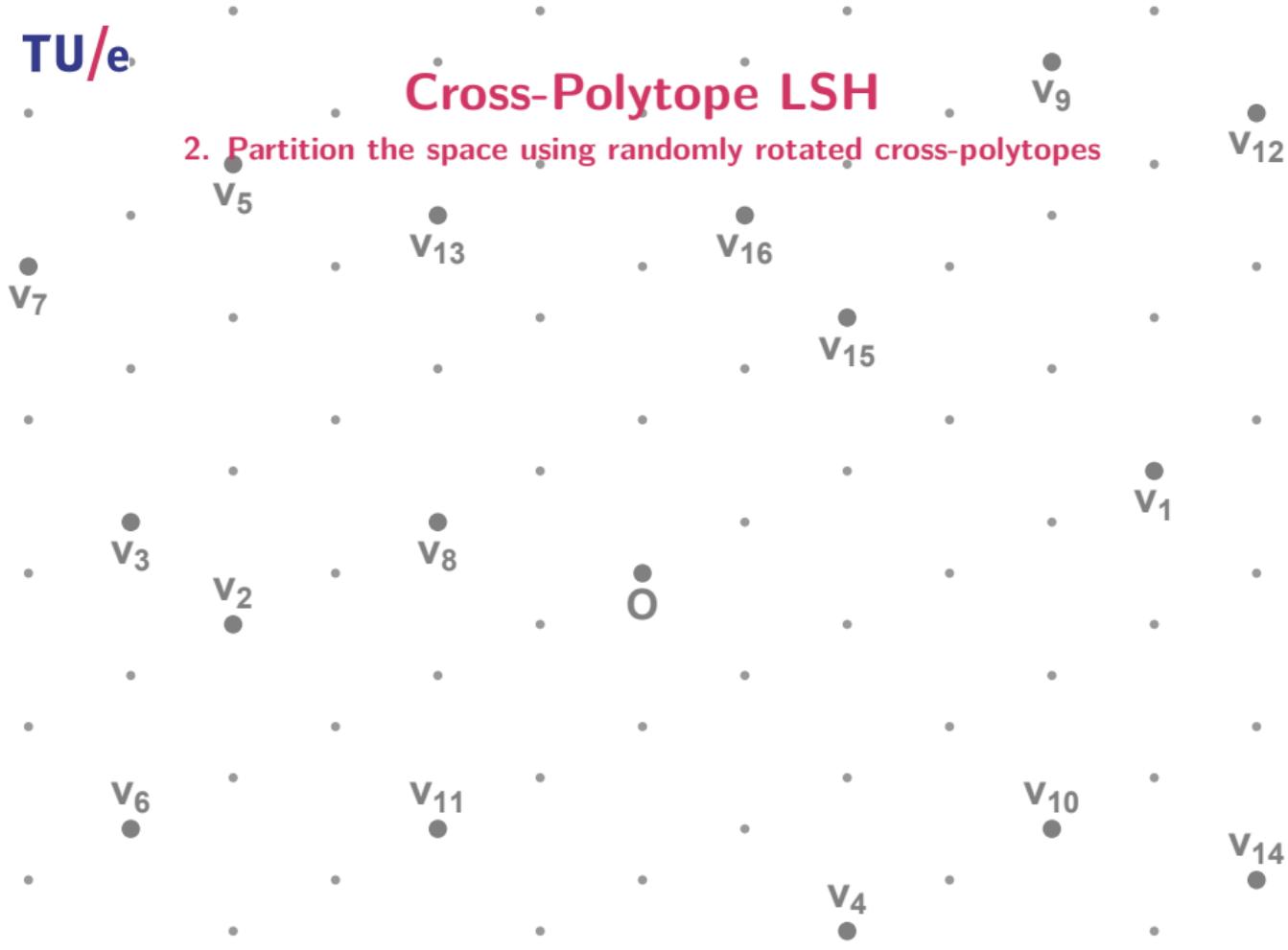
Cross-Polytope LSH

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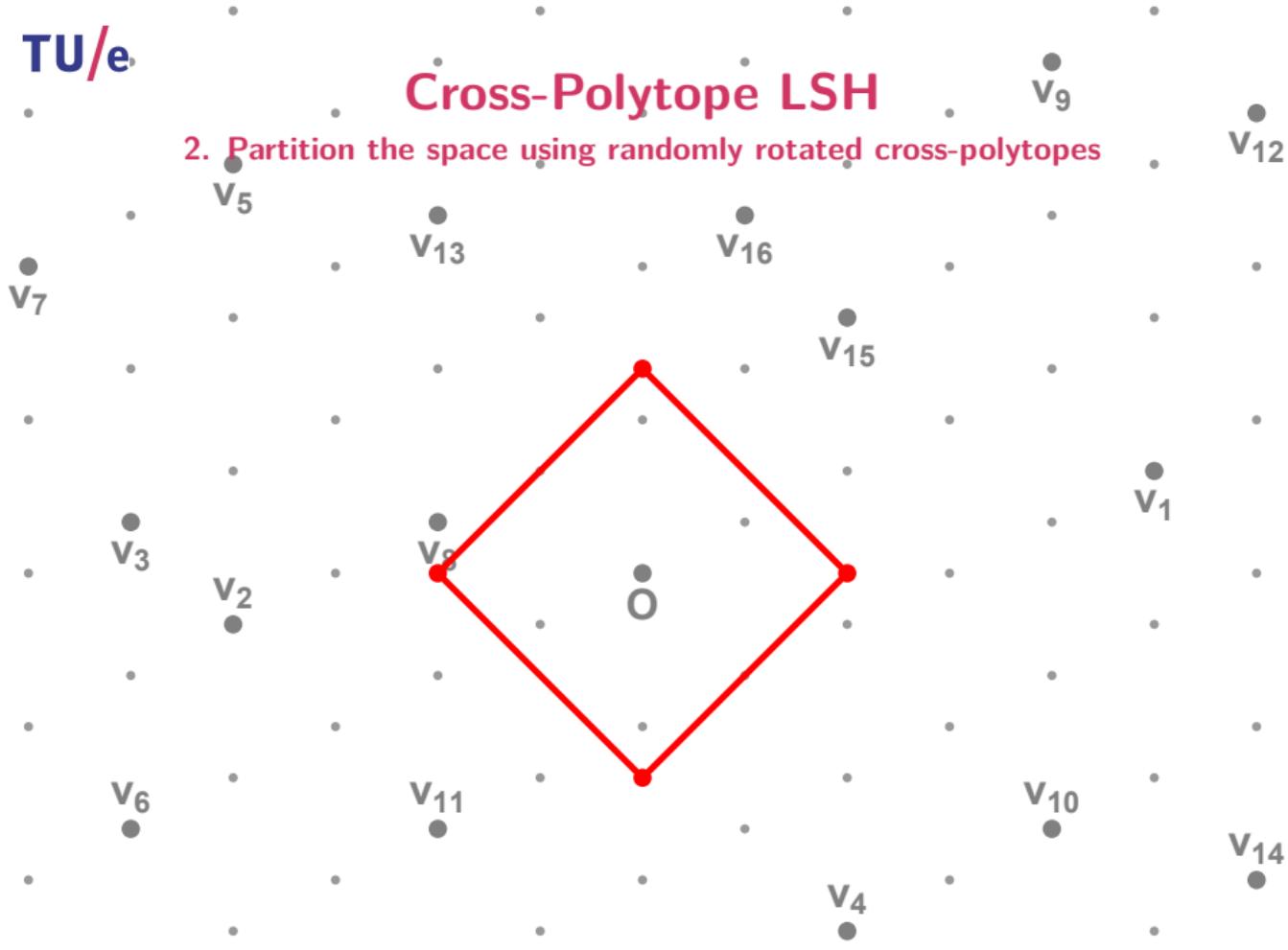
Cross-Polytope LSH

2. Partition the space using randomly rotated cross-polytopes



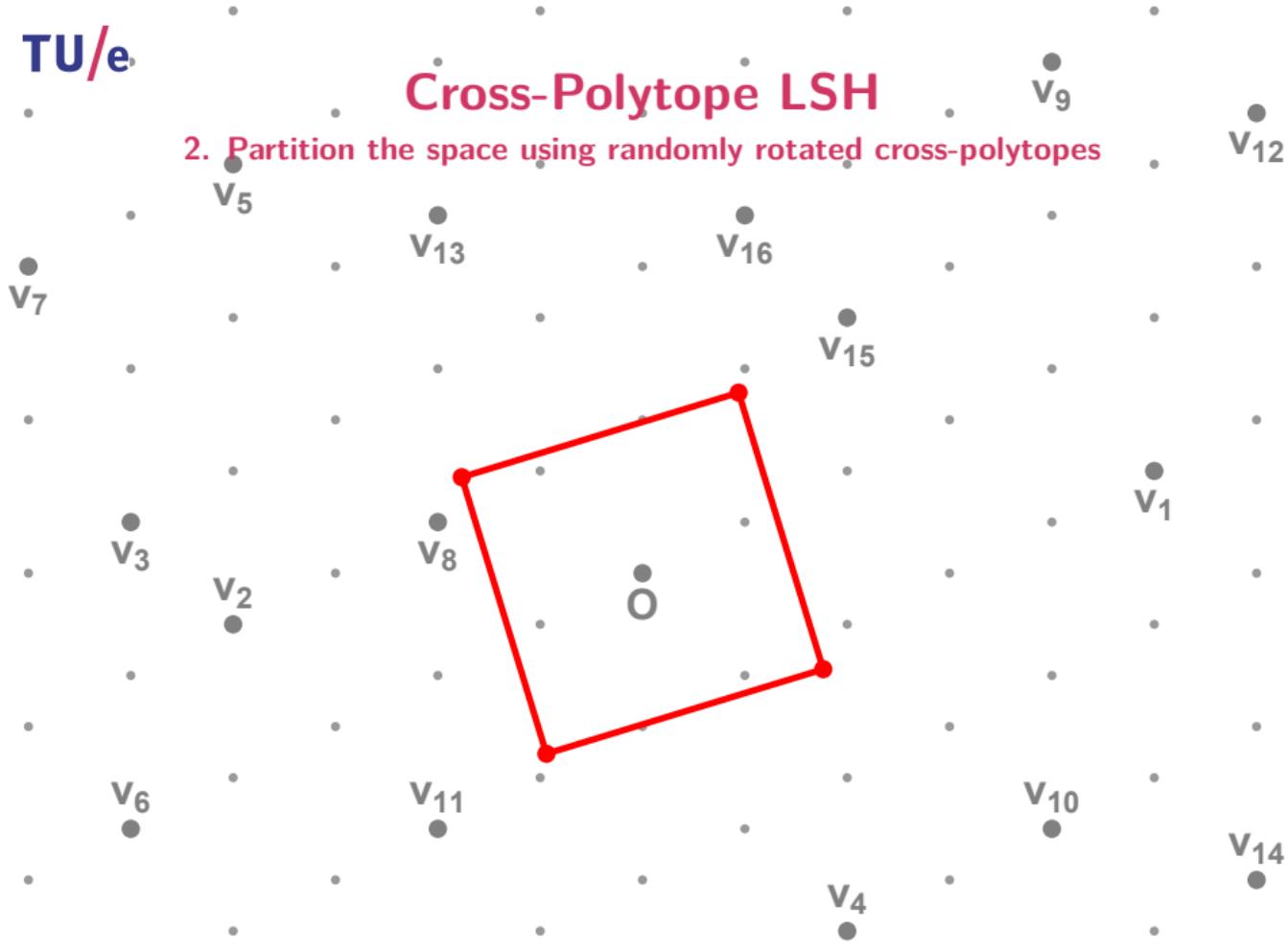
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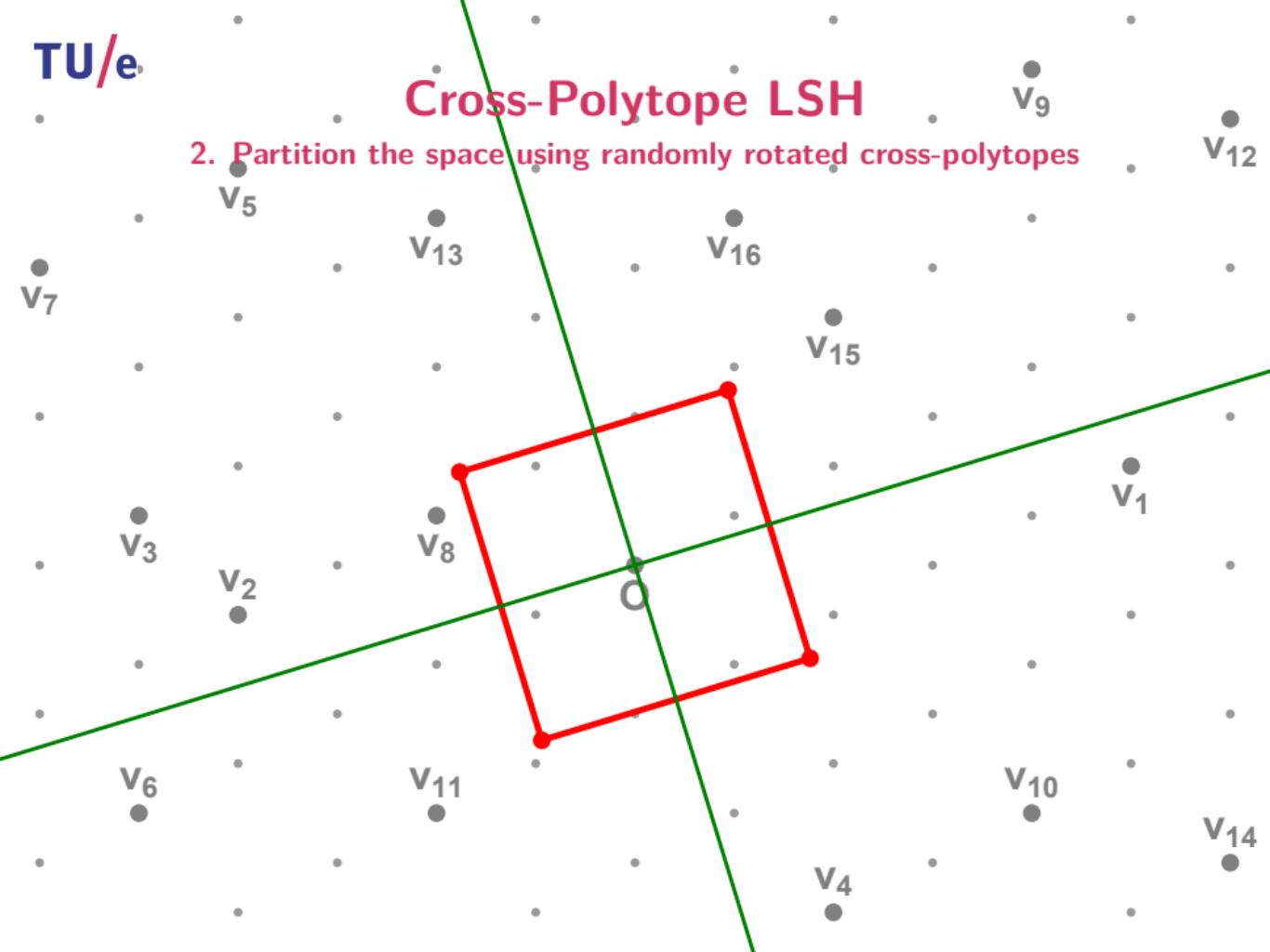
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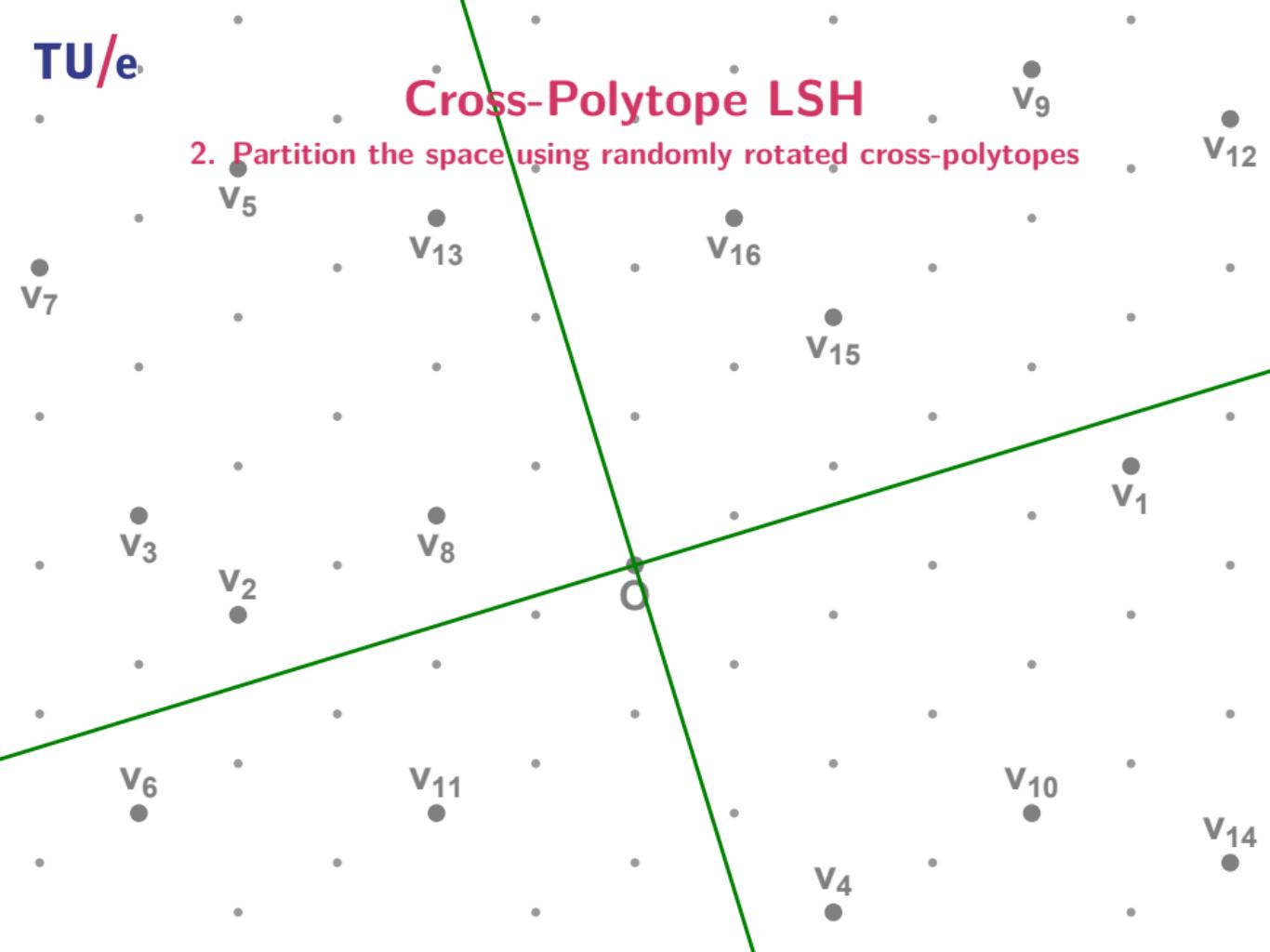
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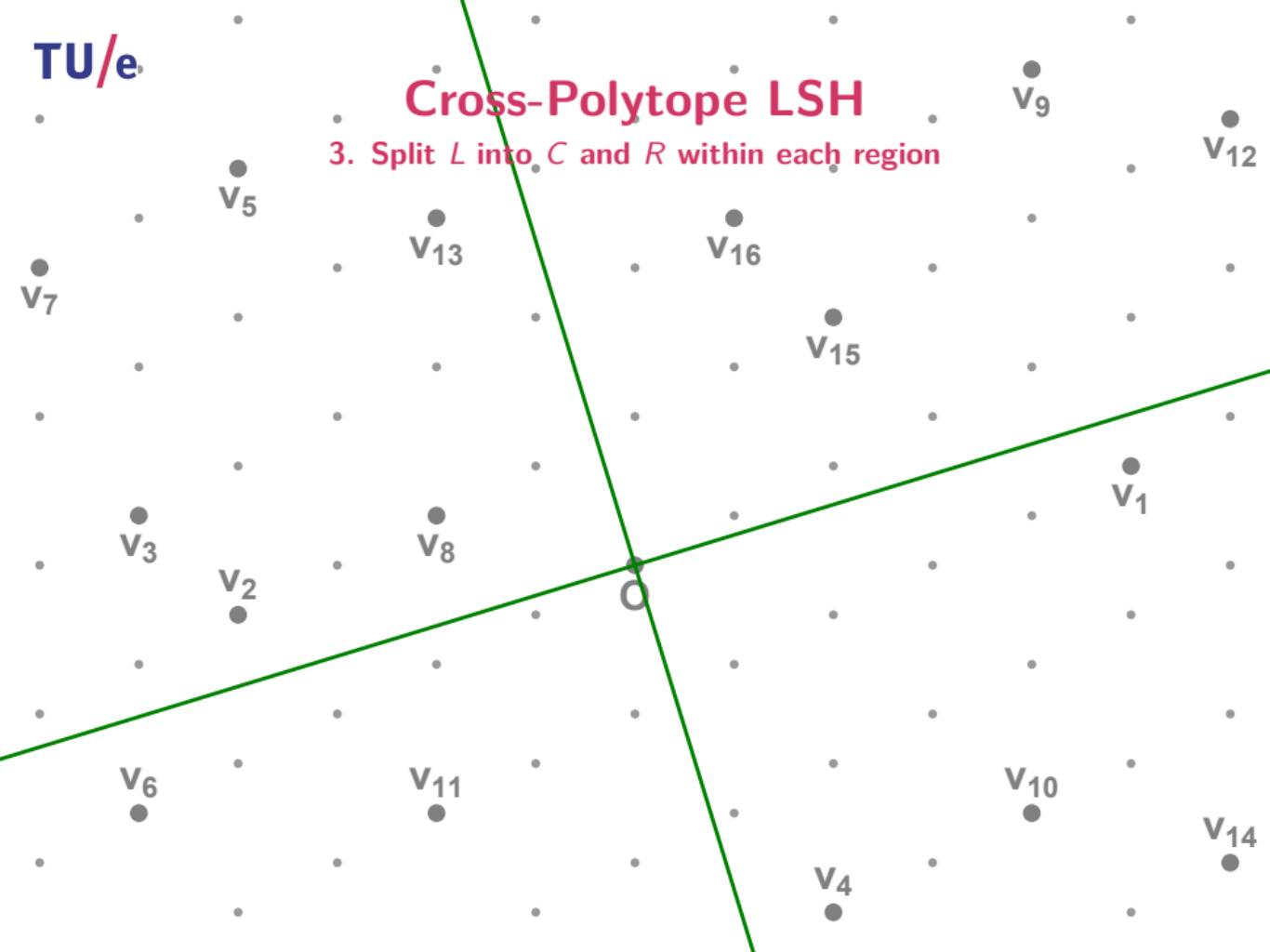
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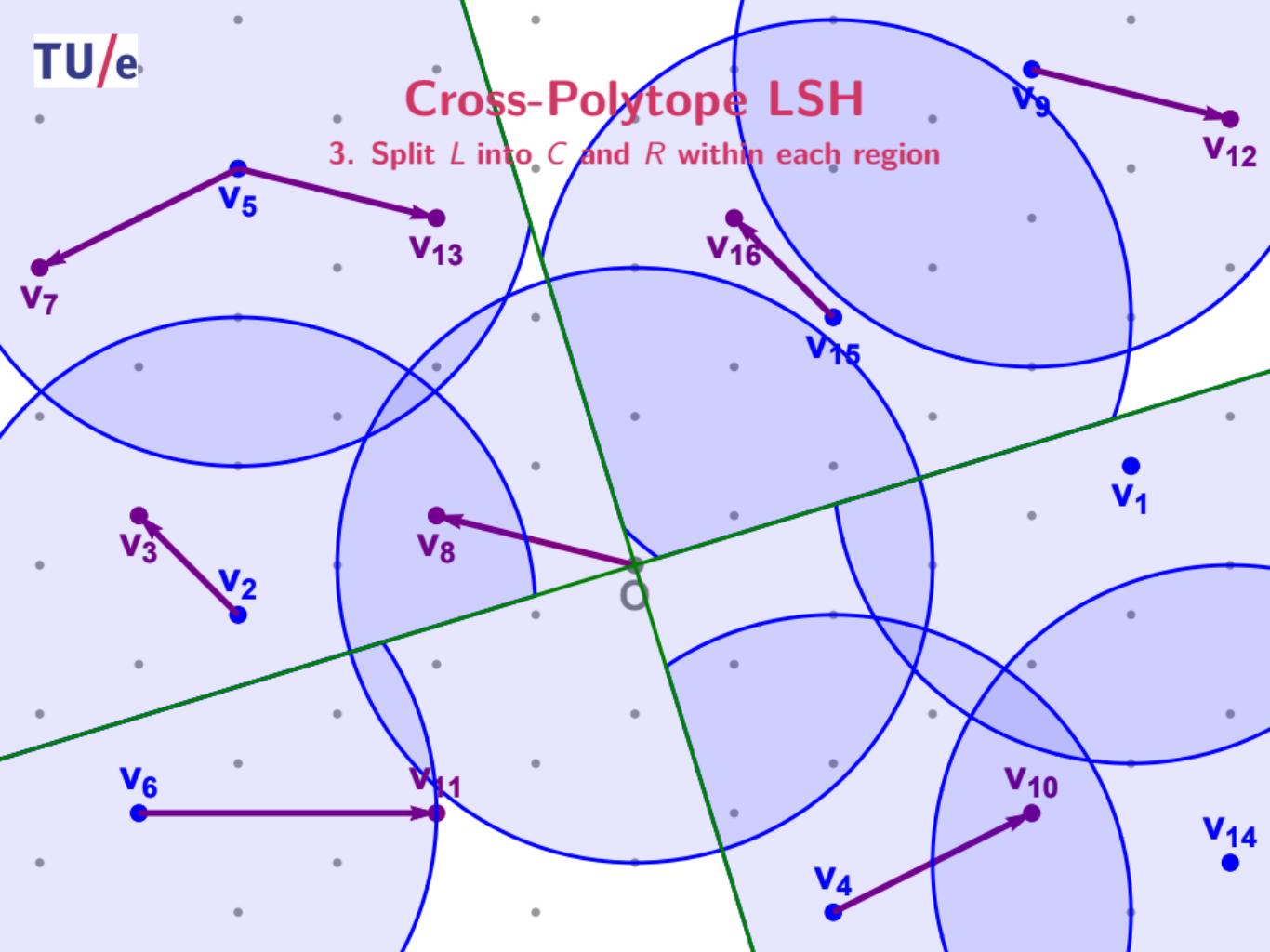
Cross-Polytope LSH

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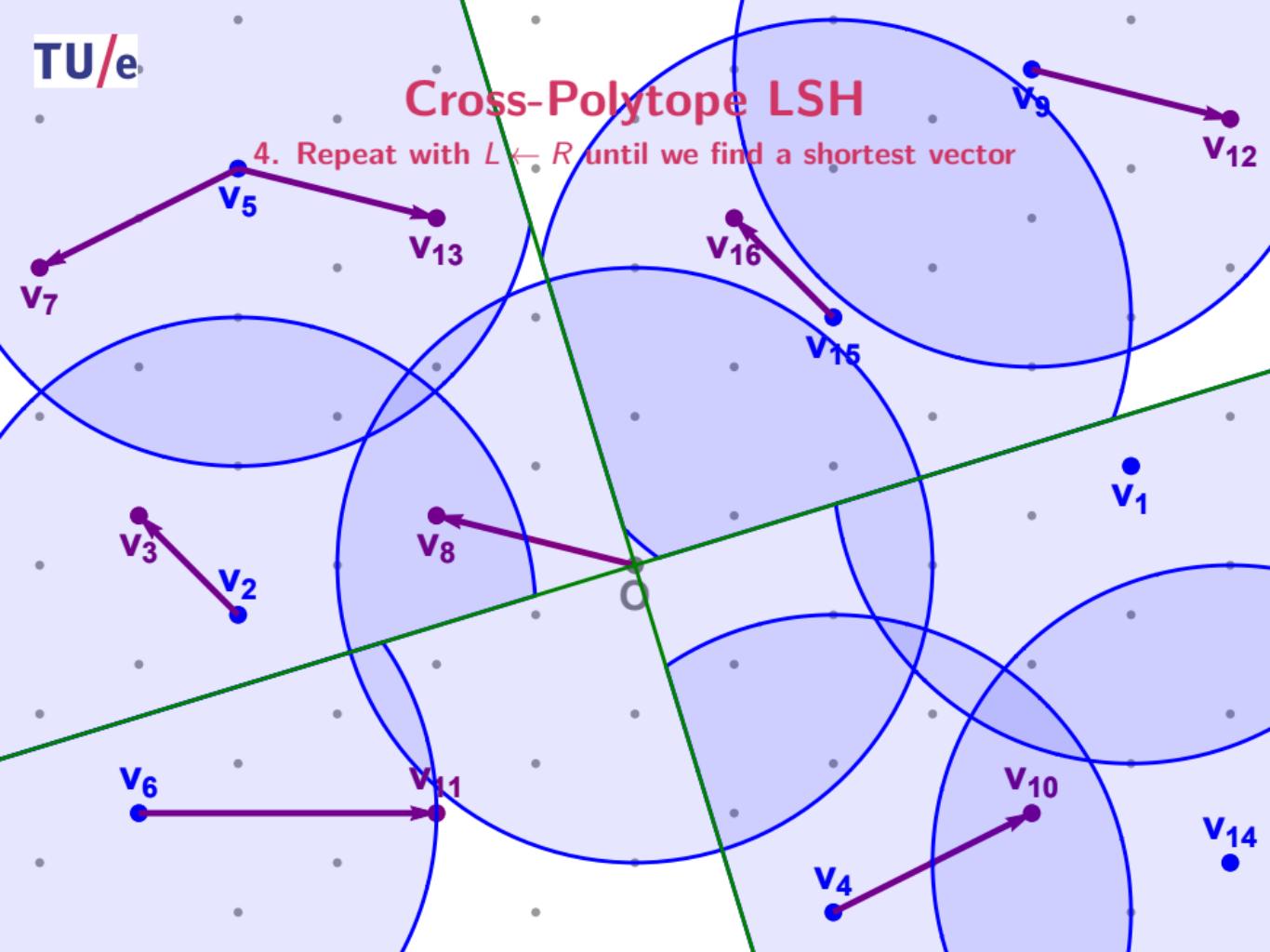
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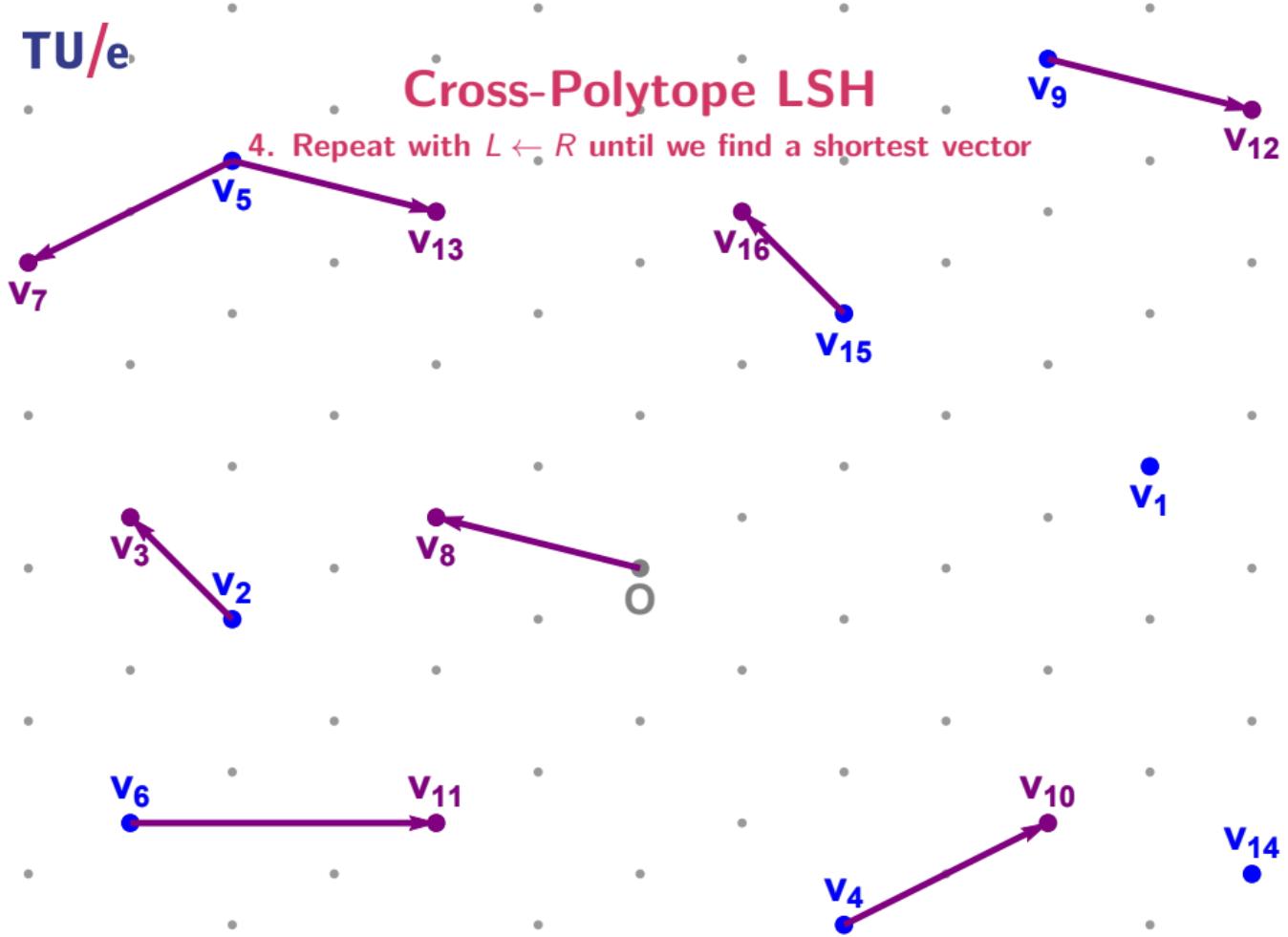
Cross-Polytope LSH

4. Repeat with $L \leftarrow R$ until we find a shortest vector



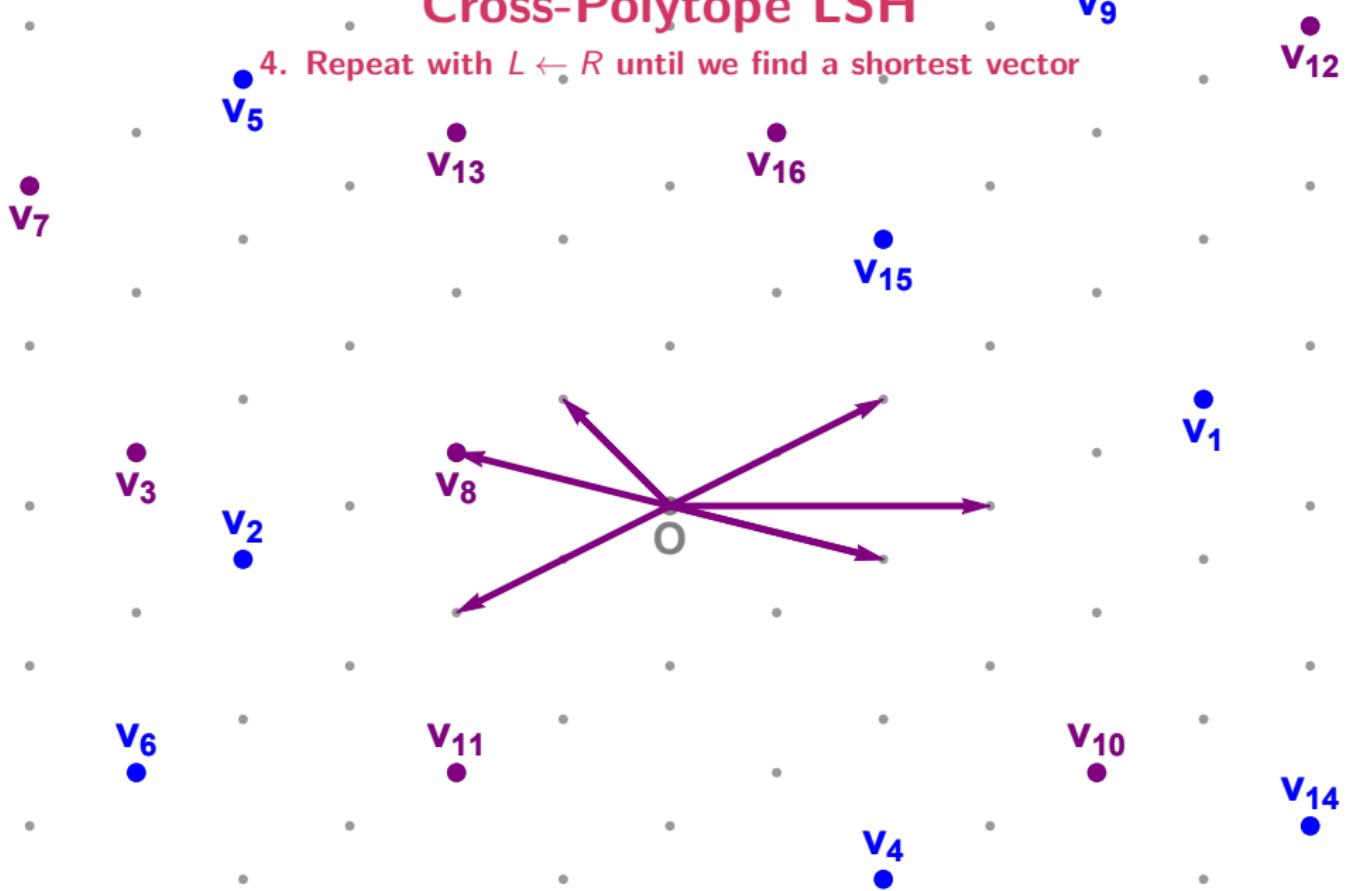
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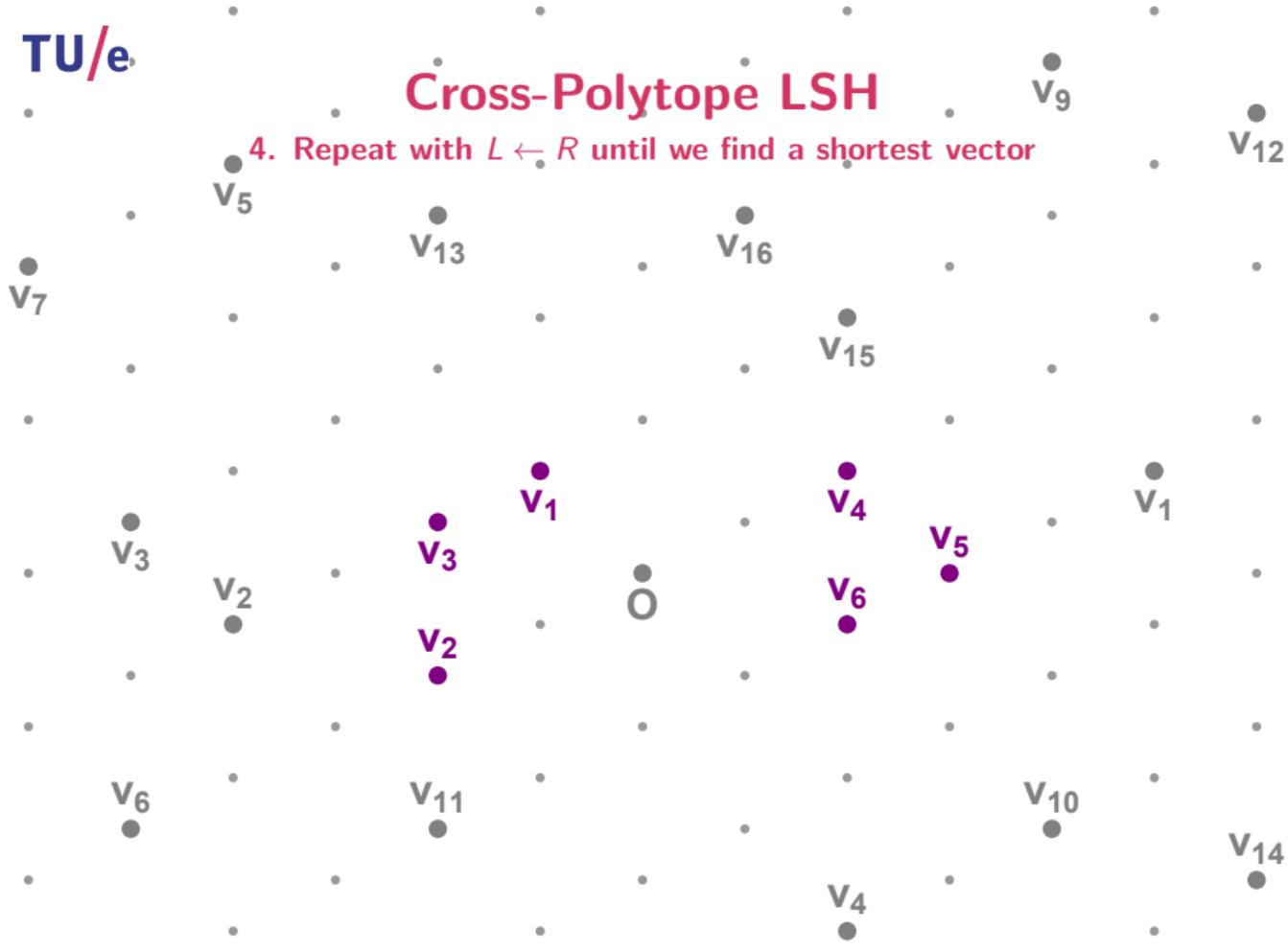
Cross-Polytope LSH

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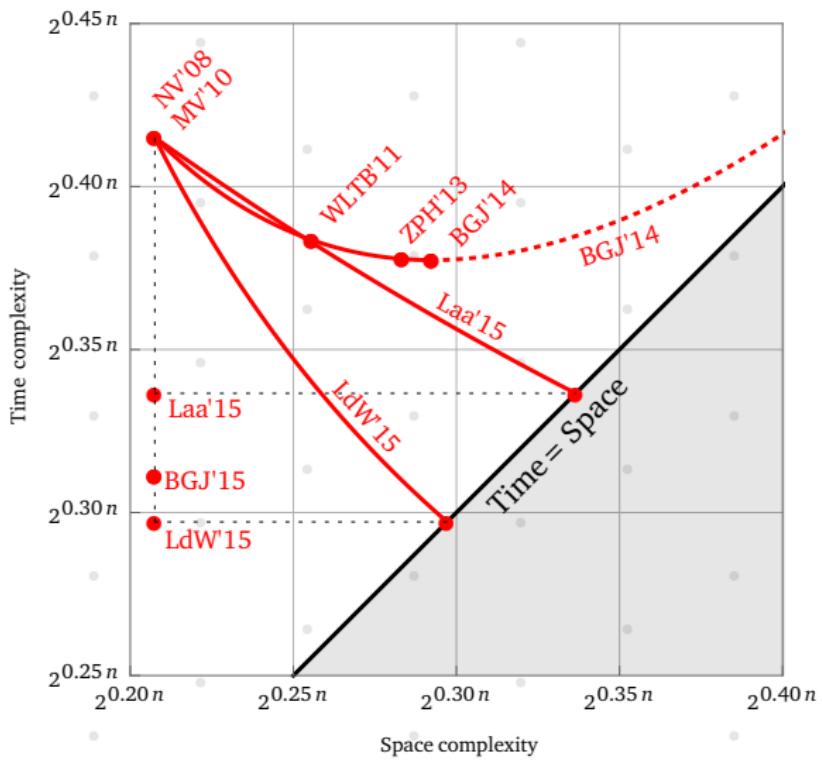
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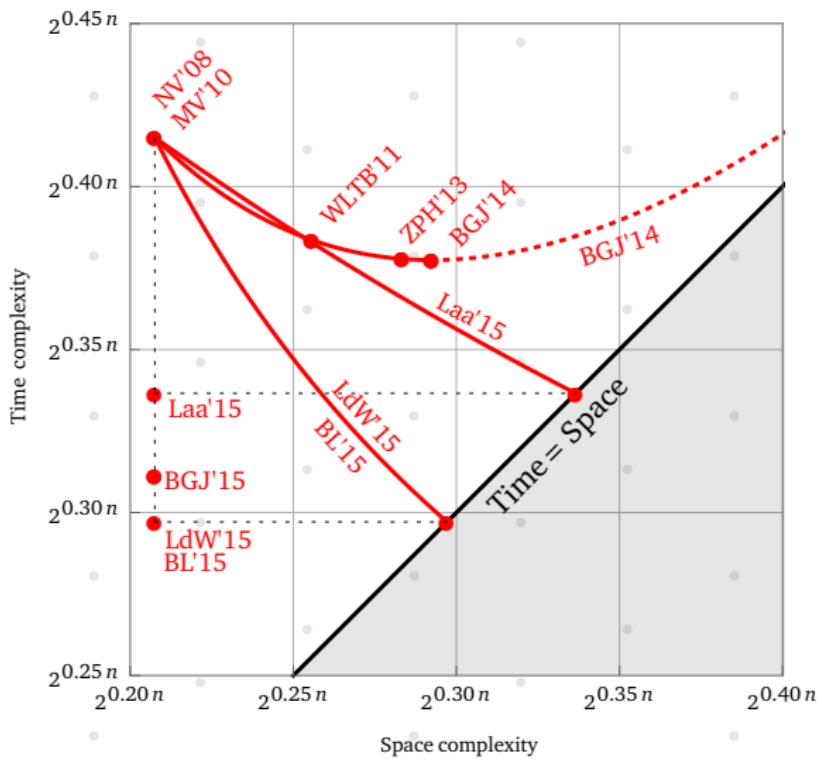
Cross-Polytope LSH

Space/time trade-off



Cross-Polytope LSH

Space/time trade-off



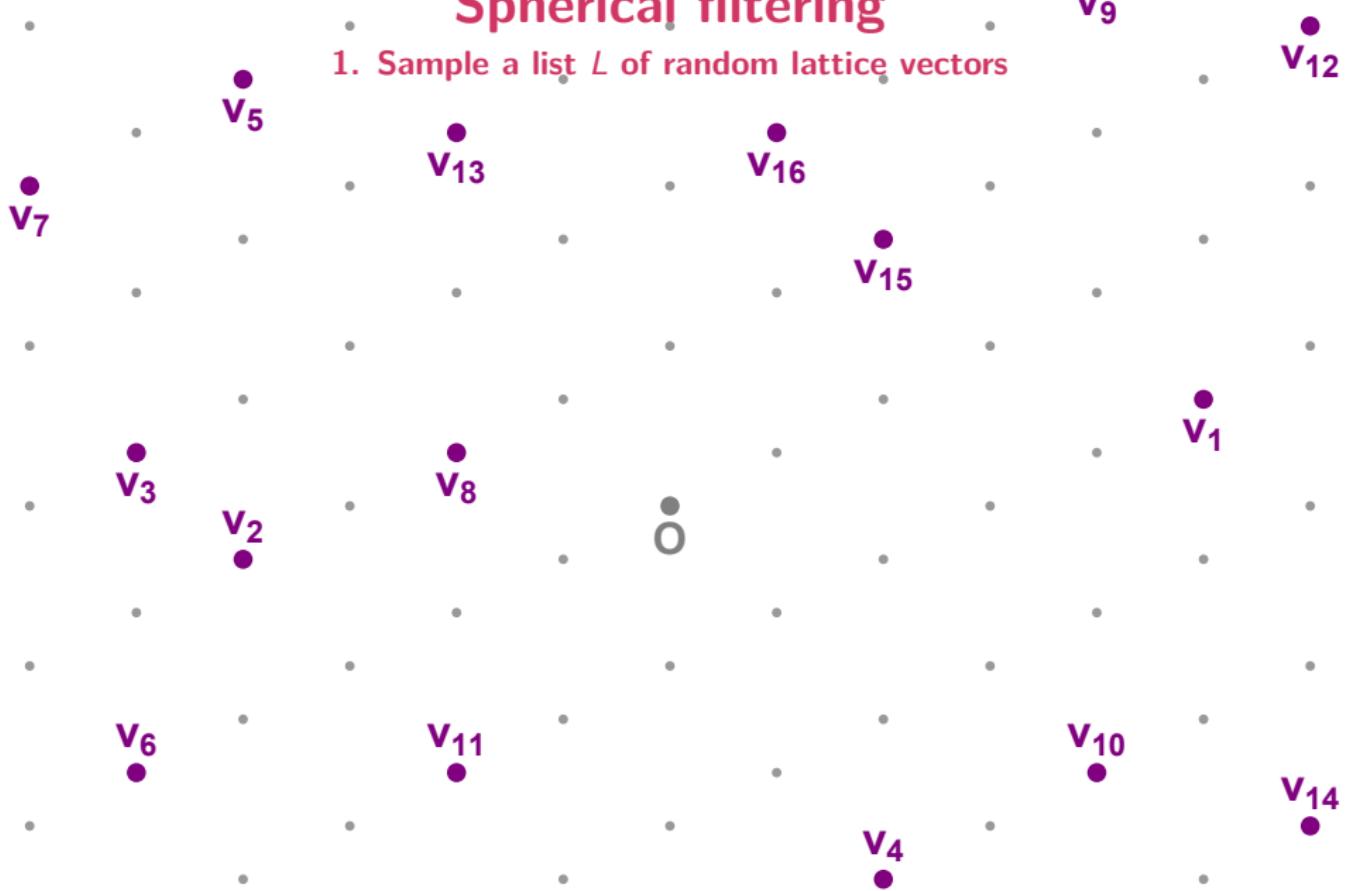
Spherical filtering

1. Sample a list L of random lattice vectors



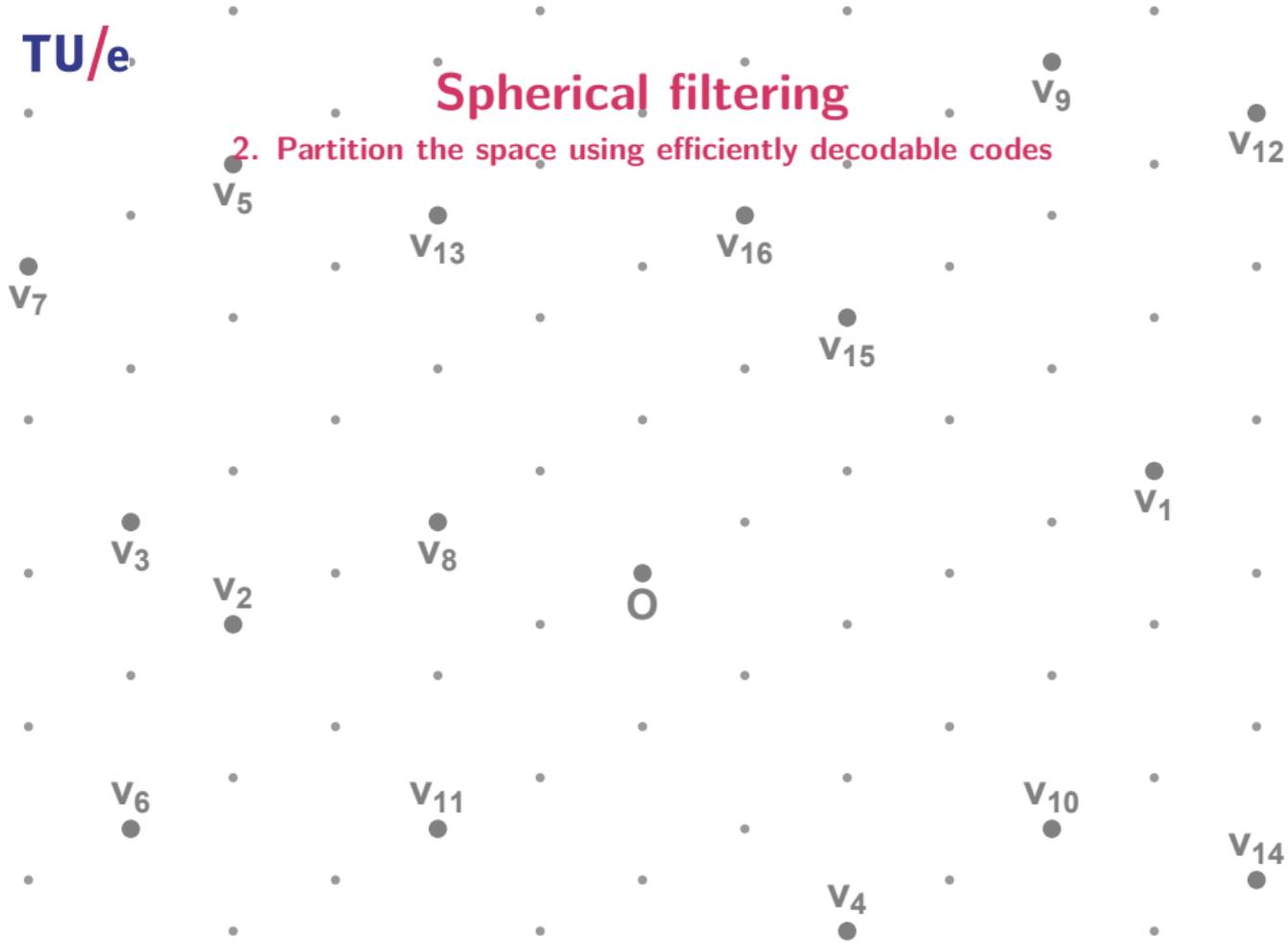
Spherical filtering

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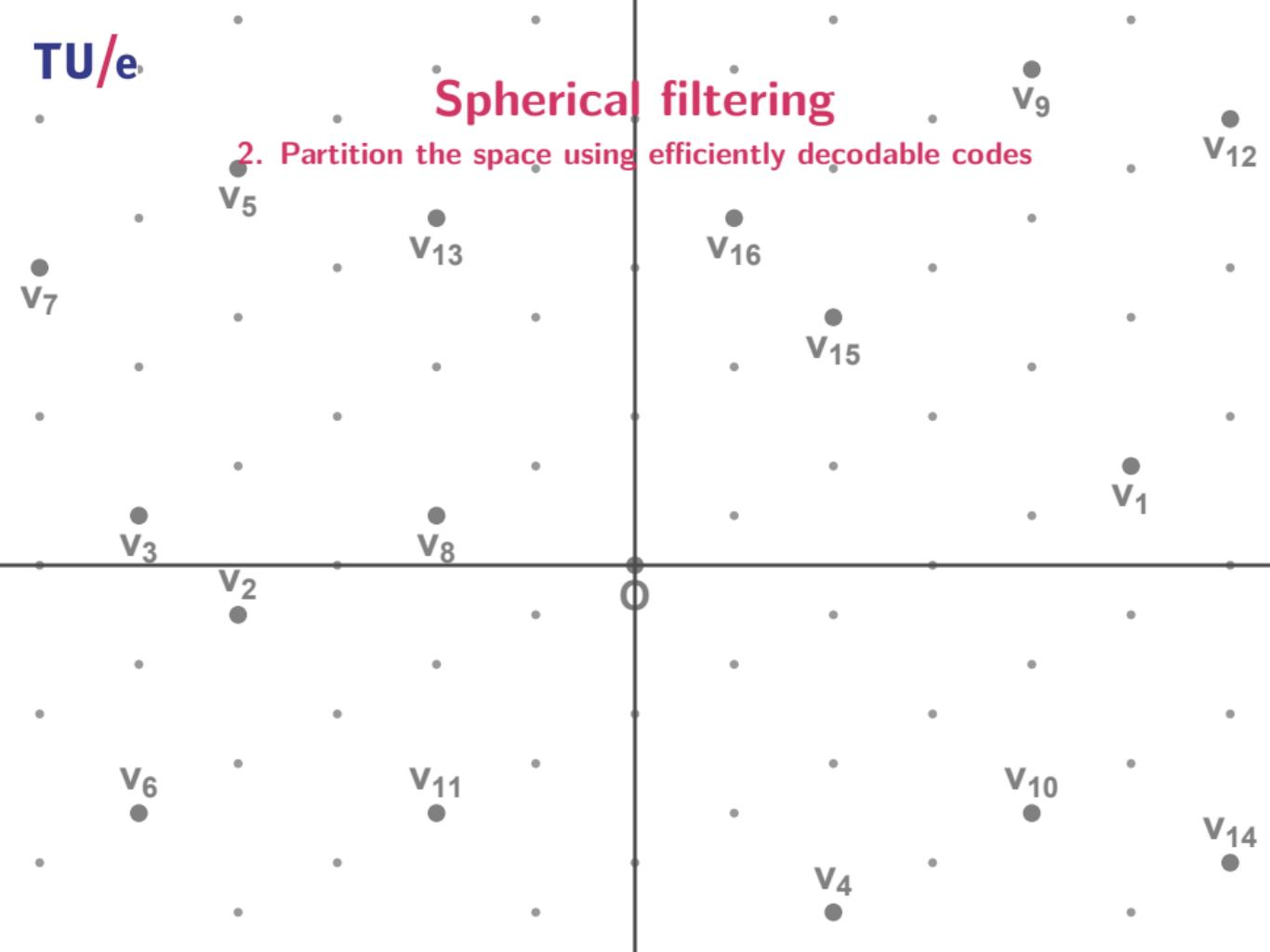
Spherical filtering

2. Partition the space using efficiently decodable codes



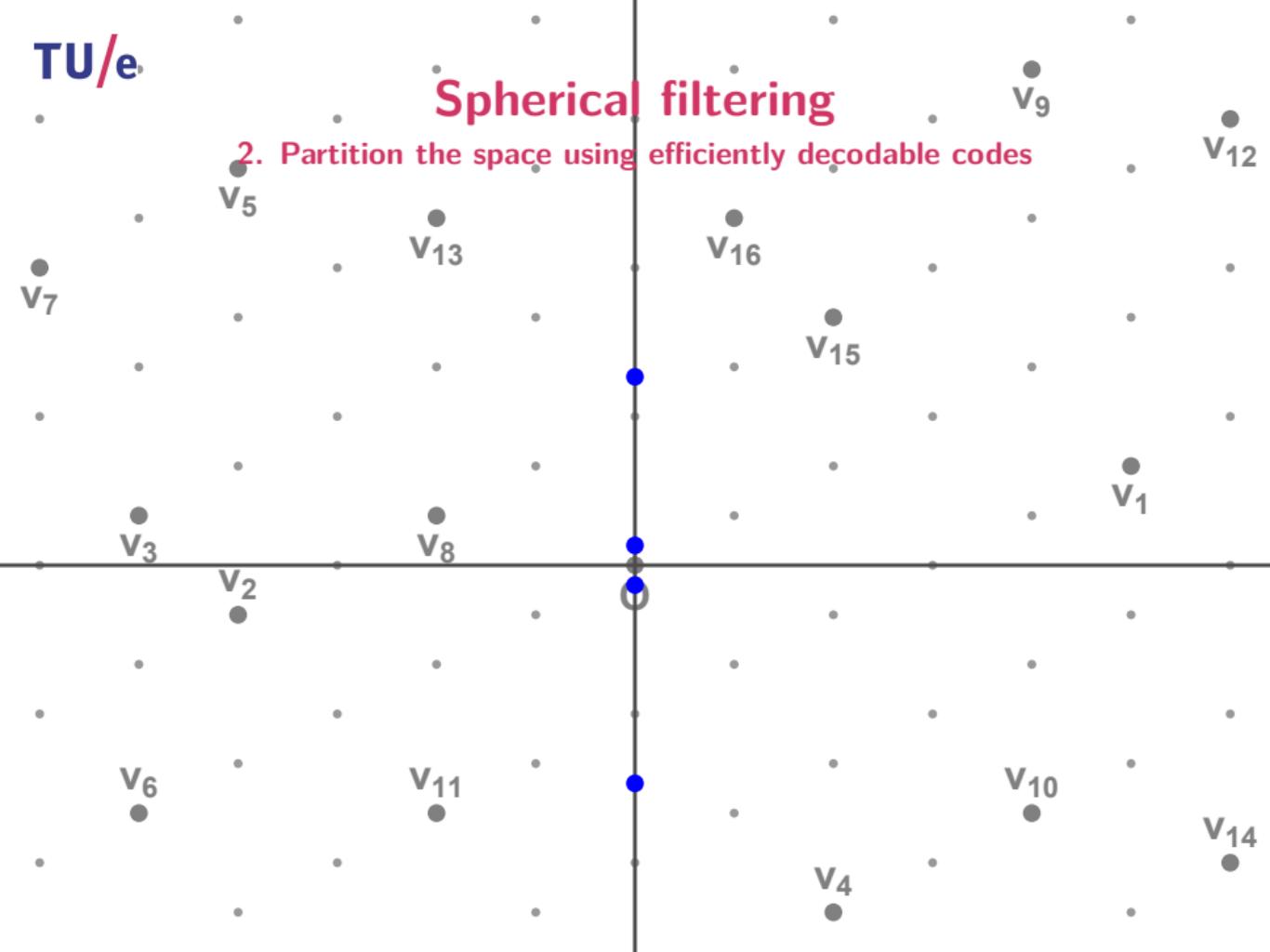
Spherical filtering

2. Partition the space using efficiently decodable codes



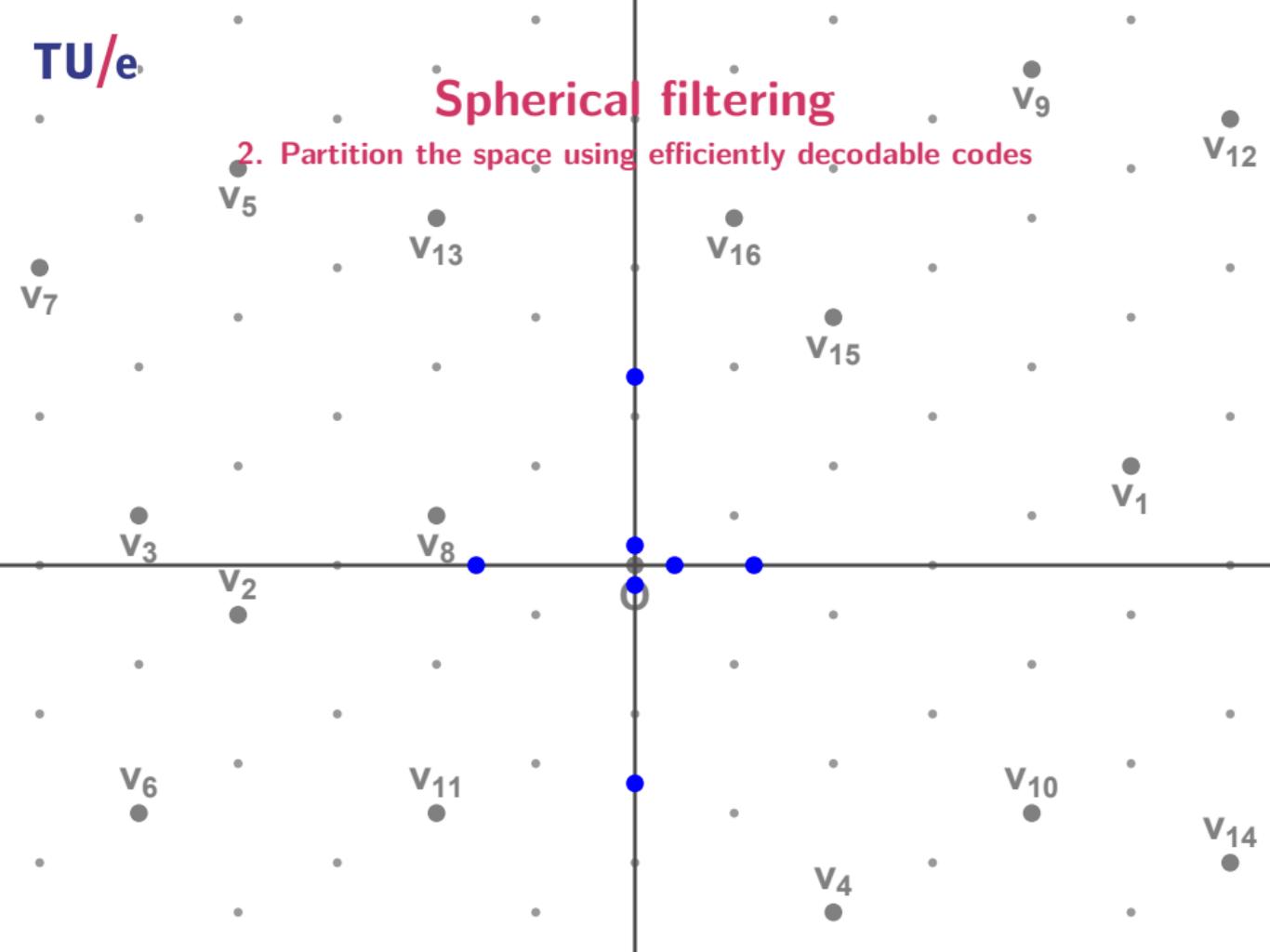
Spherical filtering

2. Partition the space using efficiently decodable codes



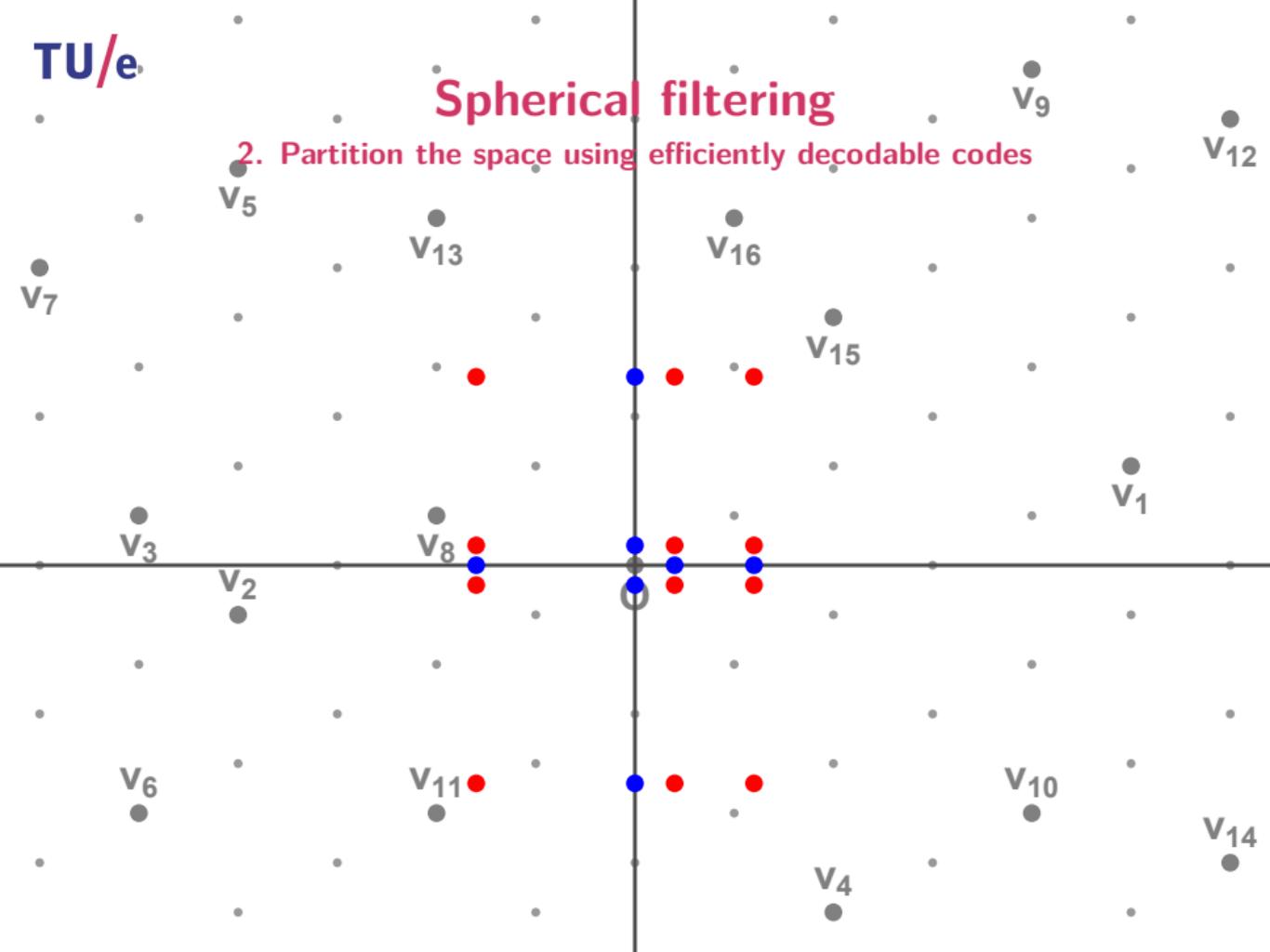
Spherical filtering

2. Partition the space using efficiently decodable codes



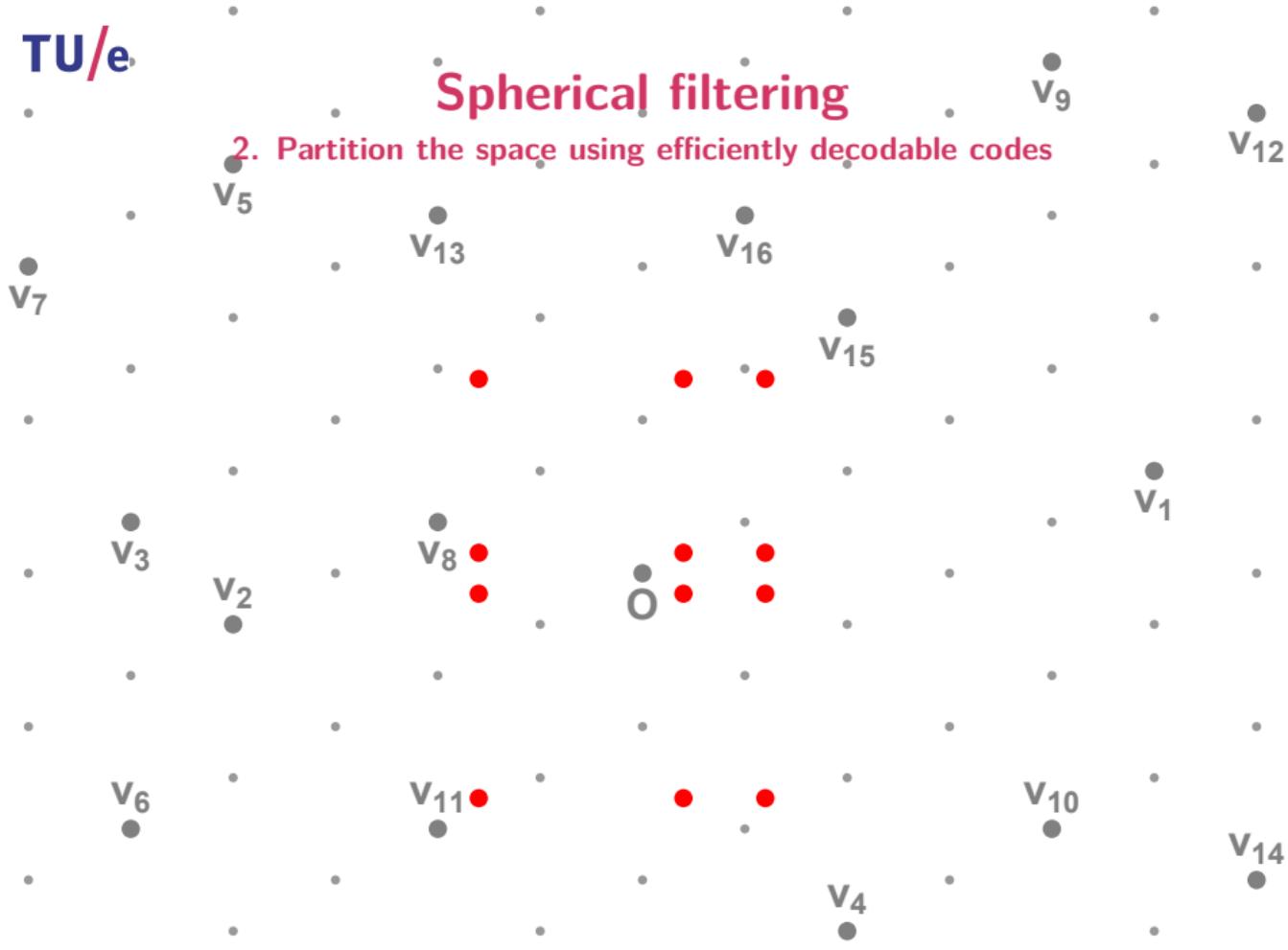
Spherical filtering

2. Partition the space using efficiently decodable codes



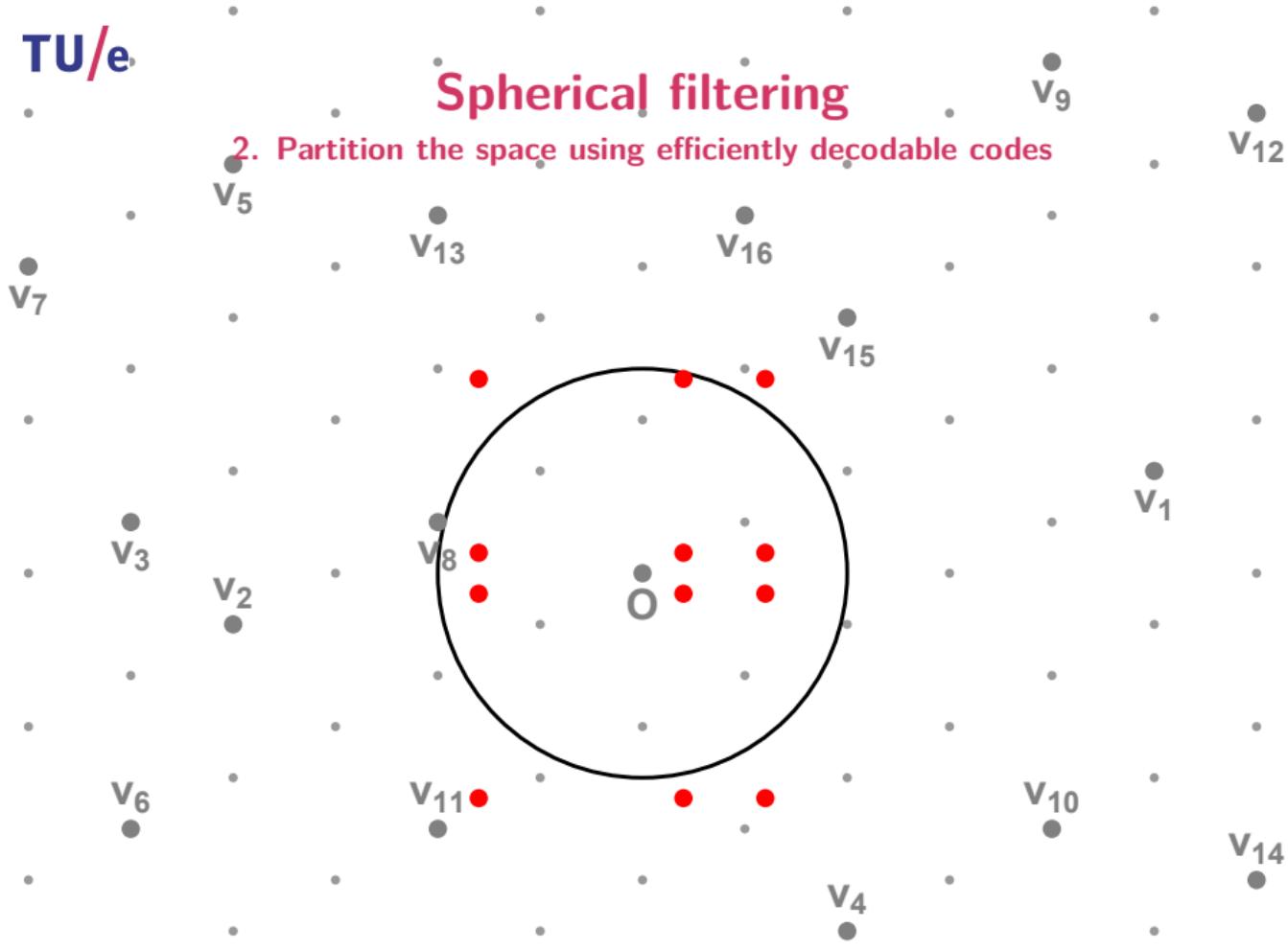
Spherical filtering

2. Partition the space using efficiently decodable codes



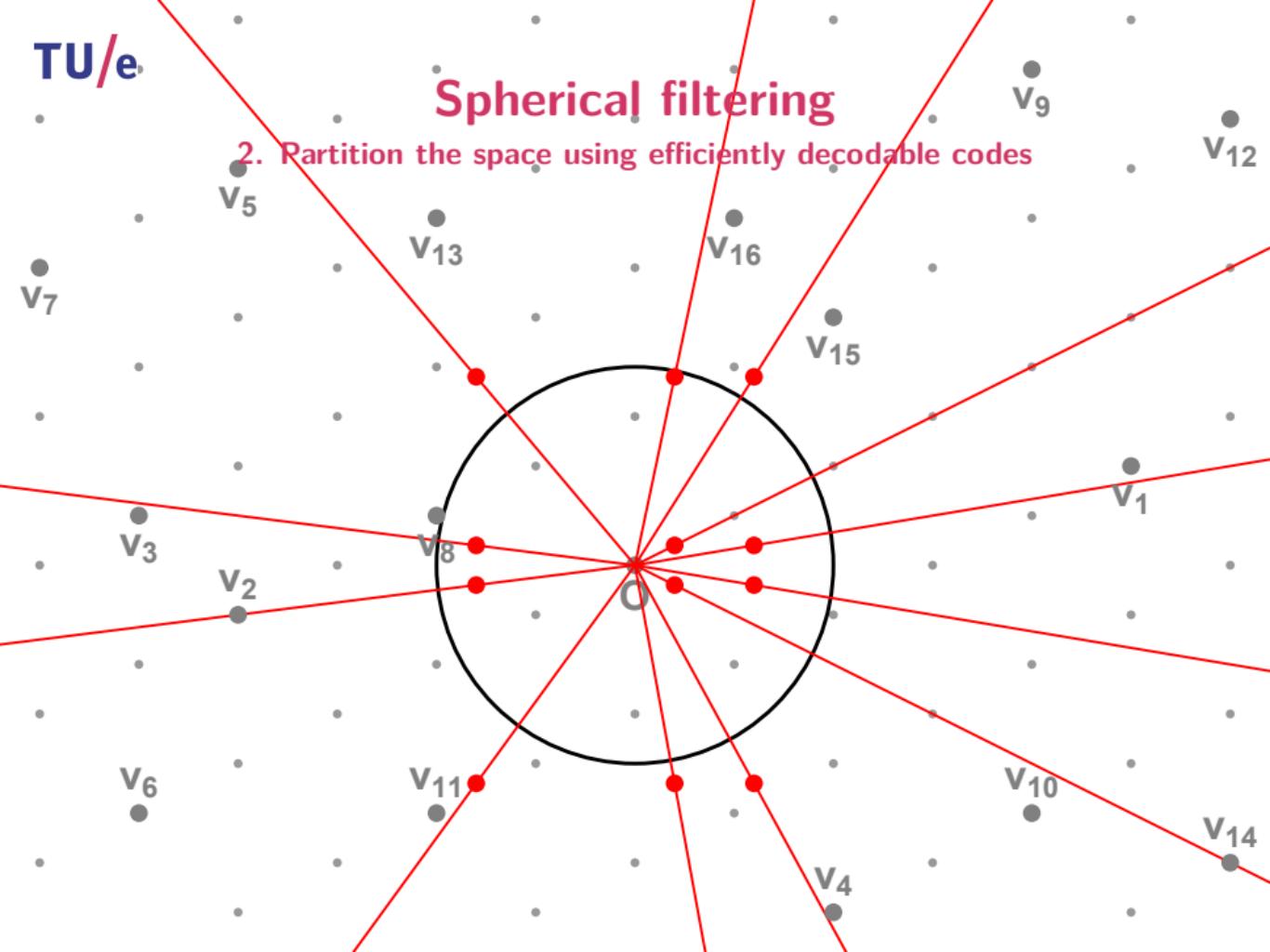
Spherical filtering

2. Partition the space using efficiently decodable codes



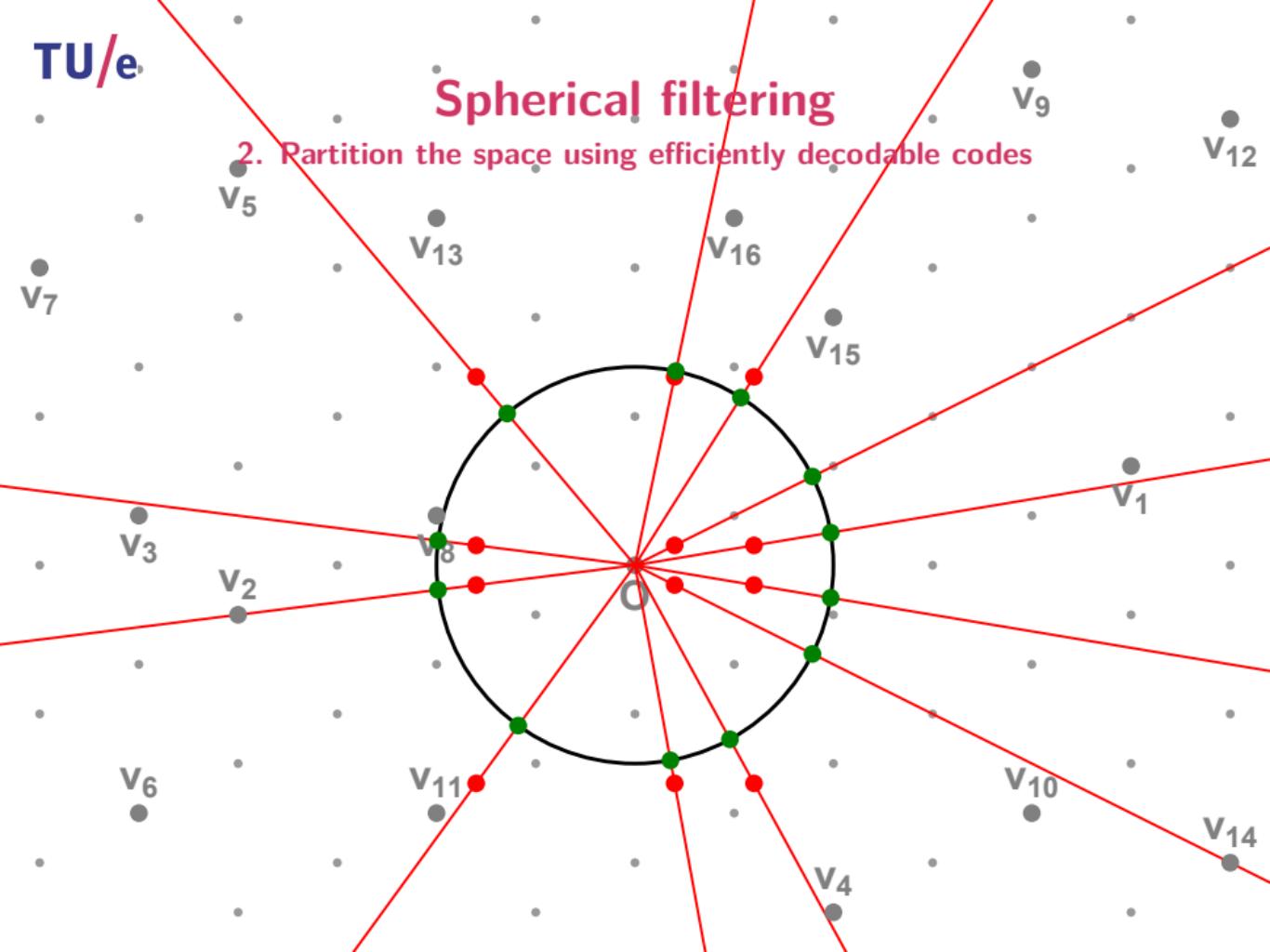
Spherical filtering

2. Partition the space using efficiently decodable codes



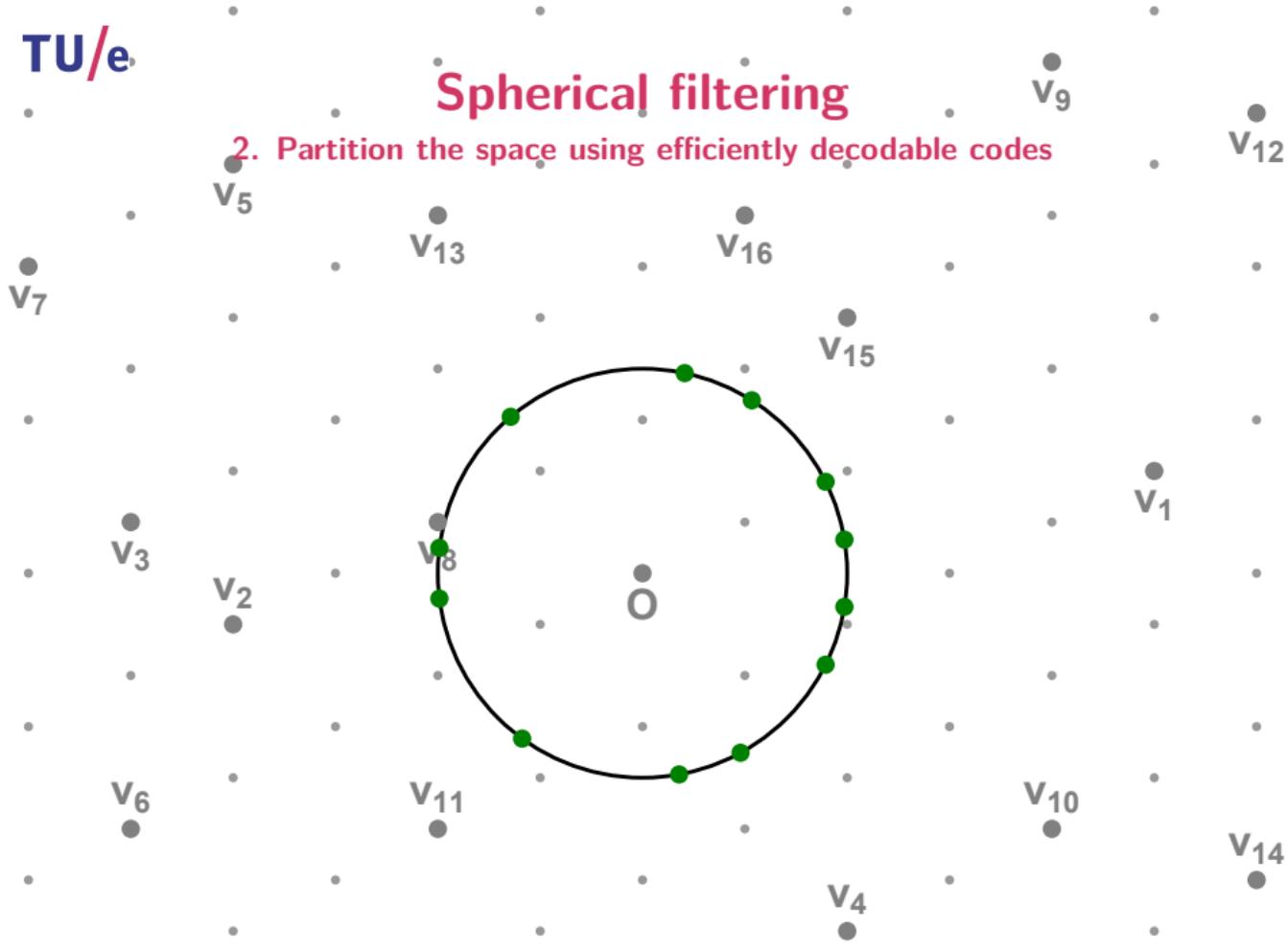
Spherical filtering

2. Partition the space using efficiently decodable codes



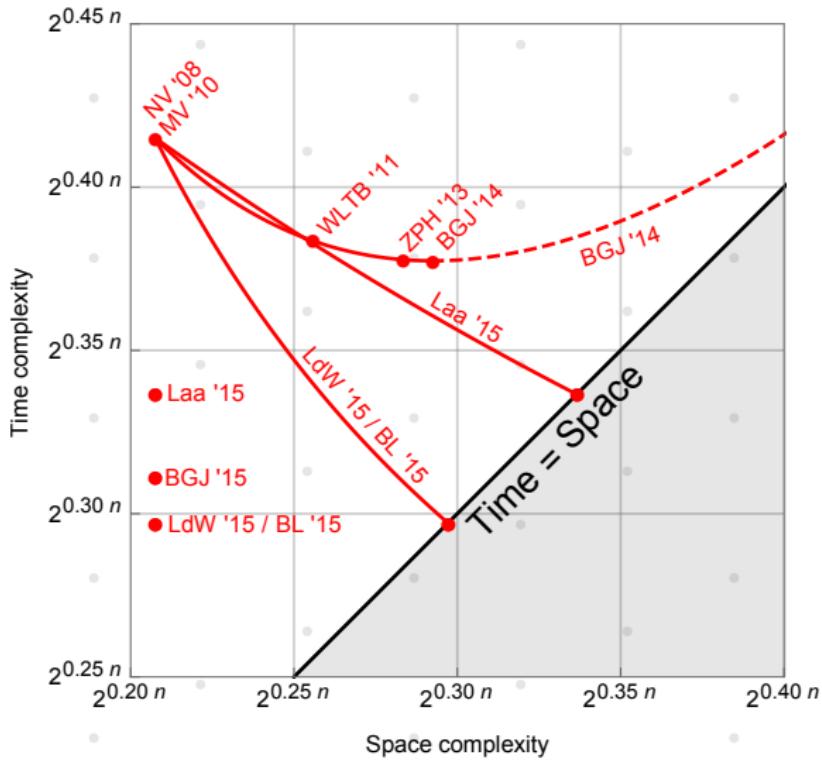
Spherical filtering

2. Partition the space using efficiently decodable codes



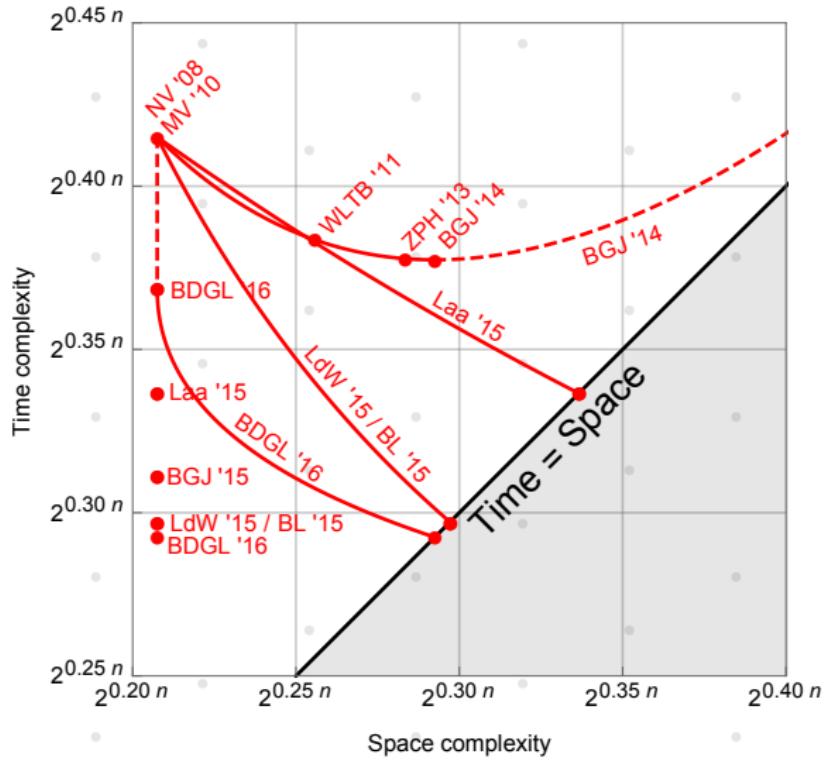
Spherical filtering

Space/time trade-off



Spherical filtering

Space/time trade-off



Questions

[vdP'12]

