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The Lecture Contains:

- Bitmap algorithm
- Example: representation
- Example: dominance check
- Nearest-neighbor (NN) algorithm
- NN in higher dimensions

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Bitmap algorithm

- Assume that the domain of each dimension is discretized into n values $\{0, \dots, n-1\}$
- Represent each value as a bitmap of size n bits
- Assume that the dominating function is $<$
- Represent a value v as a bitmap whose v MSBs are set to 0 and $n - v$ LSBs set to 1

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- For an object, extract bit slices for appropriate dimensions
- Bit slice for bit i contains the corresponding bits of *all* objects
- Bit slice A_i corresponds to the $(v_i + 1)$ -th MSB for dimension i
- Bitwise AND of A_i of all dimensions produces A
- A captures all objects that are better or equal in all dimensions

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- Bit slice B_i corresponds to the v_i -th MSB for dimension
- Bitwise OR of B_i of all dimensions produces B
- B captures all objects that are strictly better in at least one dimension

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- Bitwise OR of B_i of all dimensions produces B
- B captures all objects that are strictly better in *at least one* dimension
- Bitwise AND of A and B produces dominator list
- Produces approximate answers when domain is quantized

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Example: representation

Object	Value	Dimension 1	Dimension 2
O_1	(1,6)	0111 1111	0000 0011
O_2	(2,5)	0011 1111	0000 0111
O_3	(4,5)	0000 1111	0000 0111
O_4	(4,4)	0000 1111	0000 1111
O_5	(3,2)	0001 1111	0011 1111
O_6	(6,4)	0000 0011	0000 1111
O_7	(6,3)	0000 0011	0001 1111
O_8	(5,1)	0000 0111	0111 1111
O_9	(7,1)	0000 0001	0111 1111

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Example: dominance check

Object	Value	Dimension 1	Dimension 2
O_1	(1,6)	0111 1111	0000 0011
O_2	(2,5)	0011 1111	0000 0111
O_3	(4,5)	0000 1111	0000 0111
O_4	(4,4)	0000 1111	0000 1111
O_5	(3,2)	0001 1111	0011 1111
O_6	(6,4)	0000 0011	0000 1111
O_7	(6,3)	0000 0011	0001 1111
O_8	(5,1)	0000 0111	0111 1111
O_9	(7,1)	0000 0001	0111 1111

O_1 : Bit slice $A = 1\ 0000\ 0000 \wedge 1\ 1111\ 1111 = 1\ 0000\ 0000$

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O_5	(3,2)	0001 1111	0011 1111
O_6	(6,4)	0000 0011	0000 1111
O_7	(6,3)	0000 0011	0001 1111
O_8	(5,1)	0000 0111	0111 1111
O_9	(7,1)	0000 0001	0111 1111

O_1 : Bit slice $A = 1\ 0000\ 0000 \wedge 1\ 1111\ 1111 = 1\ 0000\ 0000$

O_1 : Bit slice $B = 0\ 0000\ 0000 \vee 0\ 1111\ 1111 = 0\ 1111\ 1111$

Example: dominance check

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O_5	(3,2)	0001 1111	0011 1111
O_6	(6,4)	0000 0011	0000 1111
O_7	(6,3)	0000 0011	0001 1111
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O_1 : Bit slice $A = 1\ 0000\ 0000 \wedge 1\ 1111\ 1111 = 1\ 0000\ 0000$

O_1 : Bit slice $B = 0\ 0000\ 0000 \vee 0\ 1111\ 1111 = 0\ 1111\ 1111$

O_1 : Bit slice $C = A \wedge B = 0\ 0000\ 0000$

Therefore, O_1 is a skyline object as it is not dominated by any other object

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Example: dominance check

Object	Value	Dimension 1	Dimension 2
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O_4	(4,4)	0000 1111	0000 1111
O_5	(3,2)	0001 1111	0011 1111
O_6	(6,4)	0000 0011	0000 1111
O_7	(6,3)	0000 0011	0001 1111
O_8	(5,1)	0000 0111	0111 1111
O_9	(7,1)	0000 0001	0111 1111

O_3 : Bit slice $A = 1\ 0000\ 0000 \wedge 1\ 1111\ 1111 = 1\ 0000\ 0000$

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Example: dominance check

Object	Value	Dimension 1	Dimension 2
O_1	(1,6)	0111 1111	0000 0011
O_2	(2,5)	0011 1111	0000 0111
O_3	(4,5)	0000 1111	0000 0111
O_4	(4,4)	0000 1111	0000 1111
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O_7	(6,3)	0000 0011	0001 1111
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O_8	(5,1)	0000 0111	0111 1111
O_9	(7,1)	0000 0001	0111 1111

O_3 : Bit slice $A = 1\ 0000\ 0000 \wedge 1\ 1111\ 1111 = 1\ 0000\ 0000$

O_3 : Bit slice $B = 0\ 0000\ 0000 \vee 0\ 1111\ 1111 = 0\ 1111\ 1111$

O_3 : Bit slice $C = A \wedge B = 0\ 0000\ 0000$

Therefore, O_3 is not a skyline object as it is dominated by O_2 , O_4 and O_5

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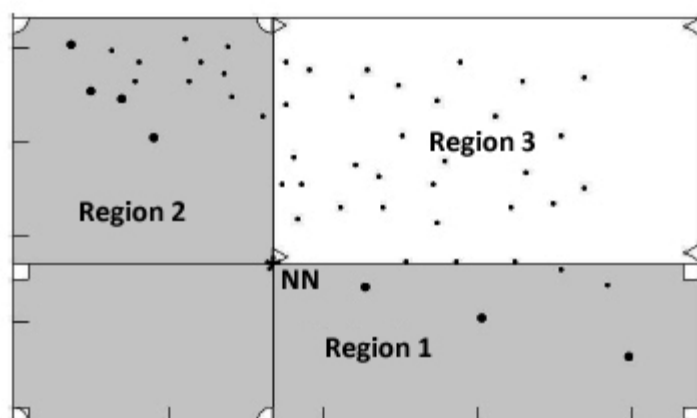
Nearest-neighbor (NN) algorithm

- Find nearest neighbor according to some monotone function
- This object is a skyline
- Exact form of monotone function does not matter

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- Find nearest neighbor according to some monotone function
- This object is a skyline
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- Based on NN, split into 3 regions (for 2D)
- Region 3 can be pruned
- Recurse in regions 1 and 2



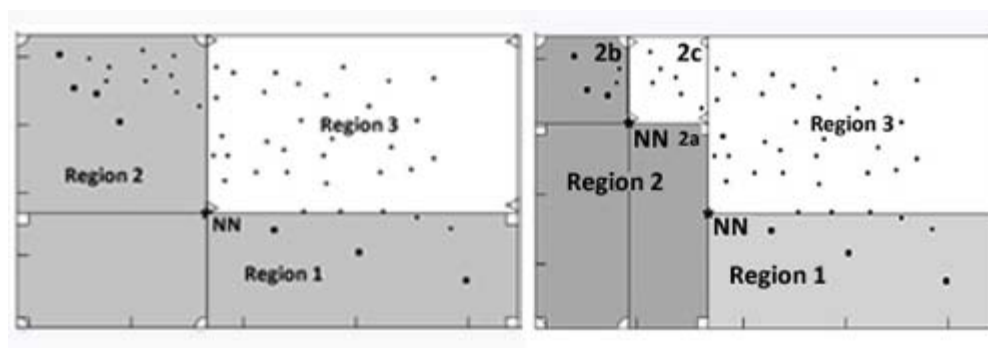
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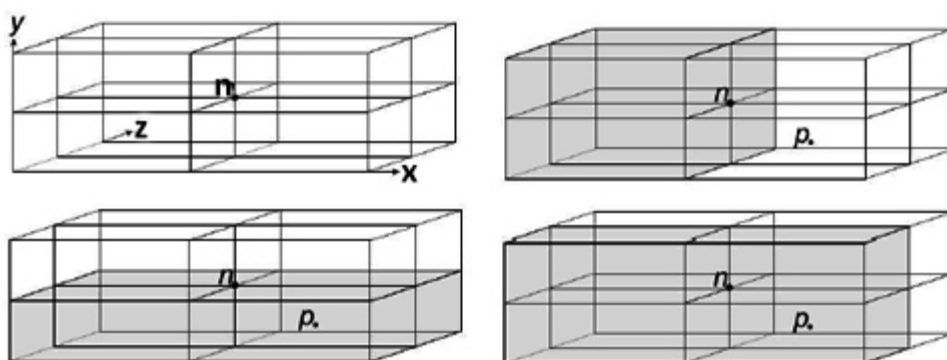
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NN in higher dimensions

- In d dimensions, requires visiting d regions
- These d regions overlap; therefore, an object may be visited multiple times (object p in figure)



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NN in higher dimensions

- In d dimensions, requires visiting d regions
- These d regions overlap; therefore, an object may be visited multiple times (object p in figure)
- Requires use of index structures to find NNs

