

Skip List and HNSW

Big Data Management and Governance

Prof. Giovanni Simonini, Giovanni Malaguti

University of Modena and Reggio Emilia

November 13, 2025

Details of the Lab (1)

- We will implement Skip List and HNSW data structures
- Simple implementations, no optimization and we won't cover all the details
- Goal: to obtain two new data structures that support creation, insert and some kind of search operations
- At home, attempt to implement other operations (e.g., key-value semantics, different distance functions, update, delete) and identify necessary code changes

Details of the Lab (2)

- Clone the repository
<https://github.com/Stravanni/bdm.git>
- Inside the cloned repository, navigate to
"bdm/lab/skiplist-hnsw"
- There are three files: skip_list.py, hnsw.py, and utils.py
- We will implement algorithms and data structures inside the first two files from scratch. In utils.py there are helper functions for visualization and experiments

Details of the Lab (3)

Open a shell and move to the lab folder. There create the virtual environment:

```
(skiplist-hnsw)$ python -m venv .venv
```

Activate the environment:

```
(skiplist-hnsw)$ source .venv/bin/activate
```

Install the required packages:

```
(skiplist-hnsw)$ pip install -r requirements.txt
```

Skip Lists

SkipList: Definition

- Linked List + Binary-search tree/sorted arrays
- Each node has 1+ layers, and each layer might be connected to different nodes
- Layers are assigned following a geometric distribution
- We'll use a simple implementation of Skip Lists, but you can extend it to include key-value pairs inside nodes

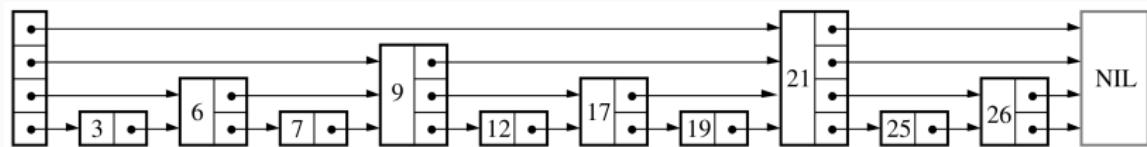


Figure 1: Example of Skip List (from [2])

Search

Algorithm 1 Search

Require: Skip list S with header $head$, value to search q

Ensure: Node with q if present, NIL otherwise

```
1: current  $\leftarrow head$ 
2: level  $\leftarrow$  max level of the skip list
3: for  $i = level$  to 0 do
4:   while current.forward[ $i$ ]  $\neq NIL$ 
      and current.forward[ $i$ ].item  $< q$  do
5:     current  $\leftarrow current.forward[i]$ 
6:   end while
7: end for
8: current  $\leftarrow current.forward[0]$ 
9: if current  $\neq NIL$  and current.item  $= q$  then
10:   return current
11: else
12:   return NIL
13: end if
```

Random Level Generation

Algorithm 2 Random Level Generation

Require: Probability p , max level maxLevel

Ensure: Random level for the new node

- 1: $\text{level} \leftarrow 0$
- 2: **while** $\text{random()} < p$ **and** $\text{level} < \text{maxLevel}$ **do**
- 3: $\text{level} \leftarrow \text{level} + 1$
- 4: **end while**
- 5: **return** level

Insertion

Algorithm 3 Insert

Require: Skip list S with header $head$, value q

Ensure: Skip list updated with the new node

```
1: update  $\leftarrow$  array of pointers with size  $maxLevel + 1$ 
2: current  $\leftarrow head$ 
3: level  $\leftarrow$  current max level
4: for  $i = level$  to 0 do
5:   while current.forward[ $i$ ]  $\neq NIL$  and current.forward[ $i$ ].item  $<$   $q$  do
6:     current  $\leftarrow$  current.forward[ $i$ ]
7:   end while
8:   update[ $i$ ]  $\leftarrow$  current
9: end for
10: current  $\leftarrow$  current.forward[0]
11: if current  $\neq NIL$  and current.item  $= q$  then
12:   return
13: end if
14: newLevel  $\leftarrow$  RandomLevel()
15: newNode  $\leftarrow$  createNode( $q$ , newLevel)
16: if newLevel  $>$  currentLevel then
17:   for  $i = currentLevel + 1$  to newLevel + 1 do
18:     update[ $i$ ]  $\leftarrow head$ 
19:   end for
20:   currentLevel  $\leftarrow$  newLevel
21: end if
22: for  $i = 0$  to newLevel + 1 do
23:   newNode.forward[ $i$ ]  $\leftarrow update[ $i$ ].forward[ $i$ ]
24:   update[ $i$ ].forward[ $i$ ]  $\leftarrow$  newNode
25: end for$ 
```

HNSW

From Skip Lists to HNSW

- HNSW[1] is a generalization of Skip List data structure
- Each node can have ≥ 1 neighbors at each level, not just one
- The neighbors are a proximity graph of the node

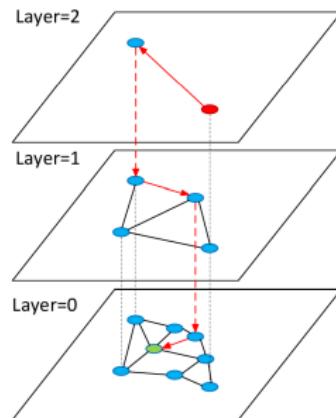


Figure 2: An HNSW graph

Search

Algorithm 4 Search

Require: multilayer graph $hnws$, query element q , number of nearest neighbors to return K , size of the dynamic candidate list $efSearch$

Ensure: K nearest neighbors to q

- 1: $W \leftarrow \emptyset$
 - 2: $ep \leftarrow$ enter point for $hnsw$
 - 3: $L \leftarrow$ level of ep
 - 4: **for** $I_c \leftarrow L \dots 0$ **do**
 - 5: $W \leftarrow \text{SearchLayer}(q, ep, ef = 1, I_c)$
 - 6: $ep \leftarrow$ get nearest element from W to q
 - 7: **end for**
 - 8: $W \leftarrow \text{SearchLayer}(q, ep, ef = efSearch, I_c = 0)$
 - 9: **return** $\text{SelectNeighborsSimple}(q, W, K)$
-

Search: how to search candidates on each layer?

Algorithm 5 SearchLayer

Require: query element q , enter points ep , number of nearest neighbors to q to return ef , layer number l_c

Ensure: ef nearest neighbors to q

```
1:  $v \leftarrow ep$  // visited nodes. Note that the passed  $ep$  is a set of points, not a single node!
2:  $W \leftarrow ep$  // set of final results (updated during search)
3:  $C \leftarrow ep$  // set of candidate nodes for search steps
4: while  $|C| > 0$  do
5:    $c \leftarrow$  extract nearest element from  $C$  to  $q$ 
6:    $f \leftarrow$  get furthest element from  $W$  to  $q$ 
7:   if  $distance(q, c) > distance(q, f)$  then
8:     break
9:   end if
10:  for each  $e \in neighbourhood(c)$  at layer  $l_c$  do
11:    if  $e \notin v$  then
12:       $v \leftarrow v \cup e$ 
13:       $f \leftarrow$  get furthest element from  $W$  to  $q$ 
14:      if  $distance(q, e) < distance(q, f)$  or  $|W| < ef$  then
15:         $C \leftarrow C \cup e$ 
16:         $W \leftarrow W \cup e$ 
17:        if  $|W| > ef$  then
18:          remove furthest element from  $W$  to  $q$ 
19:        end if
20:      end if
21:    end if
22:  end for
23: end while
24: return  $W$ 
```

Search: how to select neighbors (simple version)?

- This is the simplest version of a selection method for neighbors
- It may lead disconnected graphs, where few nodes are not properly linked to the others
- It's simpler and faster than other heuristics (perfect for us)

Algorithm 6 Select Neighbors Simple

Require: base element q , candidate elements C , number of neighbors to return K

Ensure: K nearest elements to q

1: **return** K nearest elements from C to q

Insert

Algorithm 7 Insert

Require: multilayer graph $hnsw$, new element q , number of connections per node M_{max} , size of dynamic candidate list $efConstruction$, normalization factor m_L

Ensure: update $hnsw$ inserting new element q

```
1:  $W \leftarrow \emptyset$  // list for the currently found nearest neighbors
2:  $ep \leftarrow$  get enter point for  $hnsw$ 
3:  $L \leftarrow$  level of  $ep$  // current top-layer for  $hnsw$ 
4:  $I \leftarrow \text{RandomLevel}(m_L)$ 
5: for  $I_c \leftarrow L \dots I + 1$  do
6:    $W \leftarrow \text{SearchLayer}(q, ep, ef = 1, I_c)$ 
7:    $ep \leftarrow$  get the nearest element from  $W$  to  $q$ 
8: end for
9: for  $I_c \leftarrow \min(L, I) \dots 0$  do
10:   $W \leftarrow \text{SearchLayer}(q, ep, efConstruction, I_c)$ 
11:   $\text{neighbors} \leftarrow \text{SelectNeighborsSimple}(q, W, M)$ 
12:  add bidirectional connections from  $\text{neighbors}$  to  $q$  at layer  $I_c$ 
13:  for each  $e \in \text{neighbors}$  do
14:     $eConn \leftarrow \text{neighbourhood}(e)$  at layer  $I_c$ 
15:    if  $|eConn| < M_{max}$  then
16:       $eNewConn \leftarrow \text{SelectNeighborsSimple}(e, eConn, M_{max})$ 
17:      set  $\text{neighbourhood}(e)$  at layer  $I_c$  to  $eNewConn$ 
18:    end if
19:  end for
20:   $ep \leftarrow W$ 
21: end for
22: if  $I > L$  then
23:   set enter point for  $hnsw$  to  $q$ 
24: end if
```

References

-  MALKOV, Y. A., AND YASHUNIN, D. A.
Efficient and robust approximate nearest neighbor search using
hierarchical navigable small world graphs, 2018.
-  PUGH, W.
Skip lists: a probabilistic alternative to balanced trees.
Commun. ACM 33, 6 (June 1990), 668–676.