- Exercice 02 Local Beam Seach
 - Une version modifiée de l'algorithme Best First Search
 - La sélection des nœuds se fait à base d'une probabilité conditionnelle
 - Chaque itération implique plusieurs chemins
 - Chemins ordonnés et choisis selon la longueur du chemin
 - Enchainement plus court ou moins coûteux

- Exercice 02 Local Beam Seach
 - input du programme

Input: START and GOAL states

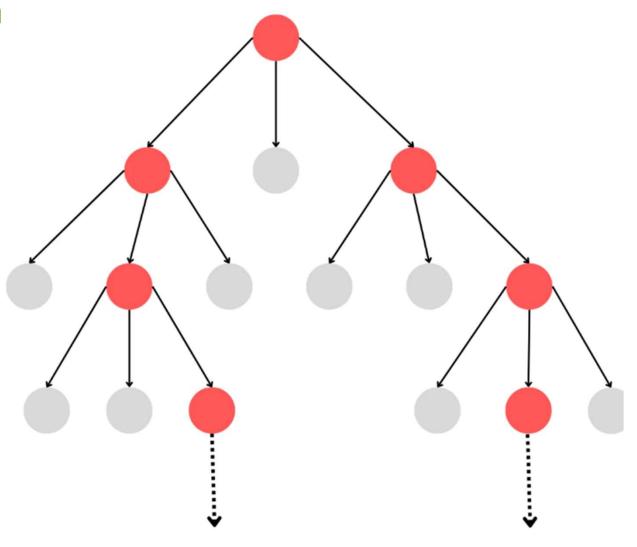
Local Variables: OPEN, NODE, SUCCs, W_OPEN, FOUND

Output: Yes or No

• Output du programme : YES / NO

- Exercice 02 Local Beam Seach
 - Méthode
- → NODE = Root_node: Found = false:
- if NODE is the goal node, then Found = true else find SUCCs of NODE. if any with its estimated cost and store in the OPEN list:
- → while (FOUND false and not able to proceed further) do
 - sort OPEN list:
 - → select the top W elements from the OPEN list and put them in the W_OPEN list and empty the OPEN list.
 - for each node in the W_OPEN list
 - { if NODE = goal state then FOUND = true find SUCCs of NODE. if any with its estimated cost and store in the OPEN list. }
 - ending while loop
- if FOUND = true then return Yes otherwise return No.
- Stop.

- Exercice 02 Local Beam Seach
 - Exemple d'exécution
 - W = 2; B = 3



- Exercice 02 Local Beam Seach
 - Ecrire un programme python pour implémenter l'algorithme LBS
 - Utilisez une fonction principale beam_search() qui itère plusieurs fois afin de trouver le plus court chemin
 - Les paramètres de cette fonction doivent être la distance
 entre deux nouds « distance » + la largeur du beam « berta »

- Exercice 02 Local Beam Seach: Solution
 - Définition de la fonction beam_search

- Exercice 02 Local Beam Seach: Solution
 - Traverser à travers les nœuds voisins ligne par ligne

```
LBS.py > ...
          #traverse through the neighbouring vertices row by row.
11
          for idx, tier in enumerate(distances):
12
              if idx > 0:
                  print(f'Paths kept after tier {idx-1}:')
                  print(*paths so far, sep='\n')
              paths at tier = list()
              for i in range(len(paths so far)):
                  path, distance = paths so far[i]
21
22
                  # Extending the paths
                  for j in range(len(tier)):
                      path_extended = [path + [j], distance + tier[j]]
                      paths_at tier.append(path_extended)
              paths_ordered = sorted(paths_at_tier, key=lambda element: element[1])
              # The best paths are saved
              paths so far = paths ordered[:beta]
30
              print(f'\nPaths reduced to after tier {idx}: ')
              print(*paths ordered[beta:], sep='\n')
          return paths so far
```

- Exercice 02 Local Beam Seach: Solution
 - Définir la matrice de distance et déterminer les meilleurs chemins

```
LBS.py >  beam_search
      #Distance matrix
      dists = [[1, 4, 6, 8],
37
               [5, 2, 3, 4]]
      dists = array(dists)
      # Calculating the best paths
41
      best paths = beam search(dists, 2)
42
      print('\nThe best paths:')
43
      for beta path in best paths:
44
45
          print(beta path)
```

- Exercice 02 Local Beam Seach: Solution
 - Résultat d'exécution

```
PS C:\Users\KHALED\Strategies> & C:/Users/KHALED/AppData/Local/Programs/Python/Python39/python.exe c:/Users/KHALED/Strategies/LB
 S.py
 Paths reduced to after tier 0:
 [[2], 6]
 [[3], 8]
 Paths kept after tier 0:
 [[0], 1]
 [[1], 4]
 Paths reduced to after tier 1:
 [[0, 3], 5]
 [[0, 0], 6]
 [[1, 1], 6]
 [[1, 2], 7]
 [[1, 3], 8]
 [[1, 0], 9]
 The best paths:
 [[0, 1], 3]
 [[0, 2], 4]
PS C:\Users\KHALED\Strategies>
```