3_1_Dynamic_Optimization__Tracking_Problem

January 1, 2018

1 Dynamic Programming to find Optimal Path

Here we have an square area with hills and a grided road network. A person has to drive from point A to B on the network. The person comes from an esoteric cult where it is not allowed to drive your car towards west or south on sundays. Today is a sunday. So this person can only drive towards north or east.

At every moment, this person has to make a decision to either go north or east such that he reaches B and minimizes fuel consumption. Climbing a slope consumes fuel proportional to the height gained while driving down the slope does not consume any fuel.

This would be a classic Dynammic programming problem and I will walk you through the code to help this person find the optimal path.

Once you have mastered this, then we will write code to find the optimal path for non sundays when the person is allowed to go in either of the 4 directions!

1.1 Part 1: Sundays

Follow the comments to understand the process. Don't get bogged down by the code. Understand the idea.

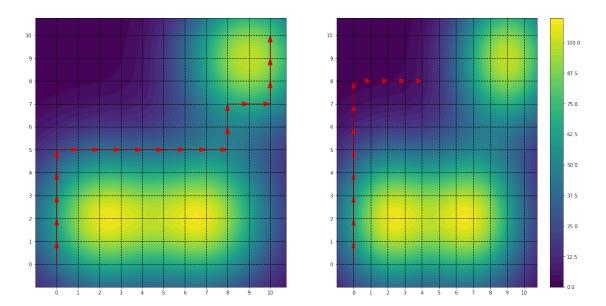
```
In [2]: from math import exp
   import numpy as np

# First we will write a function that takes
# in a x,y coordinate on a map and returns the height of that point
# This is a function that draws 3 small hills
# in a square region going from 0,0 to 10,10
@np.vectorize
def height(x, y):
   return 100 * exp(
        -(((x - 9) / 3) ** 2 + ((y - 9) / 3) ** 2)) + 100 * exp(
        -(((x - 2) / 3) ** 2 + ((y - 2) / 3) ** 2))
# Then we write a function that takes current x,y
# and previous x,y and calculates the height gained
```

```
# If the height reduces, it returns 0
# This is a good cost function since, more the height
# increase, more the fuel consumption
# but going down the slope takes 0 fuel
def cost(x, y, x prev, y prev):
    c = height(x, y) - height(x_prev, y_prev)
    return c if c > 0 else 0
# This is the main function that calculates the
# minimum cost as well as the optimal path to reach given x,y
def minimum_cost_path(x, y):
    # If x=0 and y=0, reaching this point is trivial
    # and takes 0 fuel and optimal path is [[0,0]]
    # Just do not drive
    if (x == 0 \text{ and } y == 0):
        return 0, [[0, 0]]
    else:
        # But if both not O...we have following cases
        if x > 0 and y > 0:
            # If both greater than 0, what are the points,
            # from which a person can reach x, y?
            # Ans - since person can move only up or right,
            # only points that can reach x,y are
            # (x-1,y) and (x,y-1)
            # so we first calculate cost of approaching
            # left point (x-1,y)
            cost_from_left = minimum_cost_path(x - 1, y)
            # Then we calculate cost of approaching
            # bottom point (x,y-1)
            cost_from_below = minimum_cost_path(x, y - 1)
            \# cost1 = cost \ of \ approaching \ left \ point \ (x-1,y)
            # + cost of going from (x-1,y) to (x,y)
            \# cost2 = cost \ of \ approaching \ bottom \ point \ (x,y-1)
            # + cost of going from (x,y-1) to (x,y)
            # return the minimum cost1 and cost2
            # as cost to approach (x,y)
            # cost of approaching left point
            # (x-1,y) is calculated by
            # recursively calling minimum_cost_path(x-1,y)
            if cost_from_below[0] + \
                    cost(x, y, x, y - 1) < \setminus
                             cost_from_left[0] + cost(x, y, x - 1, y):
                return cost_from_below[0] + cost(x, y, x, y - 1), \
                       cost_from_below[1] + [[x, y]]
            else:
```

```
return cost_from_left[0] + cost(x, y, x - 1, y), \
                                     cost_from_left[1] + [[x, y]]
                   elif x > 0 and y == 0:
                        # If y=0, only way to approach (x,y) is from left
                       mc = minimum cost path(x - 1, y)
                       return mc[0] + cost(x, y, x - 1, y), mc[1] + [[x, y]]
                   else:
                        # If x=0, only way to approach (x,y) is from bottom
                       mc = minimum_cost_path(x, y - 1)
                       return mc[0] + cost(x, y, x, y - 1), mc[1] + [[x, y]]
         # Call the function to calculate optimal path to point 10,10
         cost_tenten, path_tenten = minimum_cost_path(10, 10)
         # Call the function to calculate optimal path to point 4,8
         cost_foureight, path_foureight = minimum_cost_path(4, 8)
         # Print the optimal costs and paths. Ignore the mumbo jumbo
         path string = ""
         for i,p in enumerate(path_tenten):
              if i\%5 == 0:
                   path string+="\n"
              path\_string += "(" + str(p[0]) + ", " + str(p[1]) + ")-->"
         path_string = path_string[:-3]
         print("Minimum cost to reach 10,10 = {0}\n Optimal Path - {1}"
                 .format(cost_tenten, path_string))
         path_string = ""
         print("\n")
         for i,p in enumerate(path_foureight):
              if i%5==0:
                   path_string+="\n"
              path_string += "(" + str(p[0]) + ", " + str(p[1]) + ")-->"
         path_string = path_string[:-3]
         print("Minimum cost to reach 4,8 = {0}\n Optimal Path - {1}"
                 .format(cost_foureight, path_string))
Minimum cost to reach 10,10 = 100.35381234337648
Optimal Path -
(0, 0) \longrightarrow (0, 1) \longrightarrow (0, 2) \longrightarrow (0, 3) \longrightarrow (0, 4) \longrightarrow
(0, 5) \longrightarrow (1, 5) \longrightarrow (2, 5) \longrightarrow (3, 5) \longrightarrow (4, 5) \longrightarrow
(5, 5) \longrightarrow (6, 5) \longrightarrow (7, 5) \longrightarrow (8, 5) \longrightarrow (8, 6) \longrightarrow
(8, 7) \longrightarrow (9, 7) \longrightarrow (10, 7) \longrightarrow (10, 8) \longrightarrow (10, 9) \longrightarrow
(10, 10)
Minimum cost to reach 4,8 = 29.38051878887093
 Optimal Path -
```

```
(0, 0) \longrightarrow (0, 1) \longrightarrow (0, 2) \longrightarrow (0, 3) \longrightarrow (0, 4) \longrightarrow
(0, 5) \longrightarrow (0, 6) \longrightarrow (0, 7) \longrightarrow (0, 8) \longrightarrow (1, 8) \longrightarrow
(2, 8) \longrightarrow (3, 8) \longrightarrow (4, 8)
In [206]: # Lets visualize it via plots
           # You can ignore the code and focus on
           # plot and get an intuitive feel
           fig, ax = plt.subplots(1, 2, figsize=(20, 10))
           ax[0].grid(color='k', linestyle='--', linewidth=1)
           ax[0].set xticks(np.arange(0, 11, 1))
           ax[0].set_yticks(np.arange(0, 11, 1))
           for i in range(1,len(path_tenten)):
               ax[0].arrow(path_tenten[i-1][0], path_tenten[i-1][1],
                             path_tenten[i][0]-path_tenten[i-1][0],
                             path_tenten[i][1]-path_tenten[i-1][1],
                             head_width=0.2, head_length=0.3, fc='r',
                             ec='r', length_includes_head=True)
           ax[0].scatter(x, y)
           ax[1].grid(color='k', linestyle='--', linewidth=1)
           ax[1].set_xticks(np.arange(0, 11, 1))
           ax[1].set_yticks(np.arange(0, 11, 1))
           for i in range(1,len(path foureight)):
               ax[1].arrow(path_foureight[i-1][0], path_foureight[i-1][1],
                             path_foureight[i][0]-path_foureight[i-1][0],
                             path_foureight[i][1]-path_foureight[i-1][1],
                             head_width=0.2, head_length=0.3, fc='r',
                             ec='r', length_includes_head=True)
           ax[1].scatter(x, y)
           ax[1].scatter(10,10,marker="o",cmap="r", s=300)
           ax[0].scatter(10,10,marker="o",cmap="r", s=300)
           # Our 2-dimensional distribution will be over variables X and Y
           delta = 0.25
           xg = np.arange(-1, 11, delta)
           yg = np.arange(-1, 11, delta)
           Xg, Yg = np.meshgrid(xg, yg)
           Z = height(Xg, Yg)
           CS = ax[0].contourf(Xg, Yg, Z, 50)
           CS = ax[1].contourf(Xg, Yg, Z, 50)
           cbar = plt.colorbar(CS)
           plt.show()
```



1.2 Why is it so slow?

1.3 How do we speed it up?

We will use a technique called Memoization, where we store the values we have previously calculated

I have marked in comments the change from previous code

```
In [3]: # This is the main function that calculates the
        # minimum cost as well as the optimal path to reach given x,y
       ## Create a global storage ##
       ## and initialize all ######
       ## values to -1 ##########
       ##############################
       storage_for_computed_answers = [[-1 for _ in range(11)]
                                      for _ in range(11)]
       def minimum_cost_path(x, y):
           # If x=0 and y=0, reaching this point is trivial
           # and takes 0 fuel and optimal path is [[0,0]]
           # Just do not drive
           if (x == 0 \text{ and } y == 0):
               return 0, [[0, 0]]
           else:
               # But if both not O...we have following cases
               if x > 0 and y > 0:
```

```
# If both greater than O, what are the points, from
# which a person can reach x, y?
# Ans - since person can move only up or right,
# only points that can reach x,y are
# (x-1,y) and (x,y-1)
# so we first calculate cost of approaching
# left point (x-1,y)
##############################
## Check if answer ########
## already stored in #######
## global storage. If not ###
## Calculate and store ######
##################################
if storage_for_computed_answers[x - 1][y] == -1:
    storage_for_computed_answers[x - 1][y] = \
        minimum_cost_path(x - 1, y)
cost_from_left = storage_for_computed_answers[x - 1][y]
# Then we calculate cost of approaching
# bottom point (x,y-1)
##############################
## Check if answer ########
## already stored in #######
## global storage. If not ###
## Calculate and store #####
if storage_for_computed_answers[x][y - 1] == -1:
    storage_for_computed_answers[x][y - 1] = \
        minimum_cost_path(x, y - 1)
cost_from_below = storage_for_computed_answers[x][y - 1]
\# cost1 = cost \ of \ approaching \ left \ point \ (x-1,y) +
# cost of going from (x-1,y) to (x,y)
# cost2 = cost of approaching bottom point (x,y-1) +
# cost of going from (x,y-1) to (x,y)
# return the minimum cost1 and cost2 as cost
# to approach (x,y)
# cost of approaching left point (x-1,y) is
# calculated by
# recursively calling minimum_cost_path(x-1,y)
if cost\_from\_below[0] + cost(x, y, x, y - 1) < \setminus
                cost_from_left[0] + cost(x, y, x - 1, y):
   return cost_from_below[0] + cost(x, y, x, y - 1), \
           cost_from_below[1] + [[x, y]]
else:
   return cost_from_left[0] + cost(x, y, x - 1, y), \
```

```
cost_from_left[1] + [[x, y]]
        elif x > 0 and y == 0:
            ## Check if answer ########
            ## already stored in #######
            ## global storage. If not ###
            ## Calculate and store ######
            ###############################
            if storage for computed answers [x - 1][y] == -1:
                storage_for_computed_answers[x - 1][y] = \
                    minimum_cost_path(x - 1, y)
            cost_from_left = storage_for_computed_answers[x - 1][y]
            return cost_from_left[0] + cost(x, y, x - 1, y), \
                   cost_from_left[1] + [[x, y]]
        else:
            ##################################
            ## Check if answer ########
            ## already stored in #######
            ## global storage. If not ###
            ## Calculate and store ######
            ##############################
            if storage for computed answers [x][y-1] == -1:
                storage_for_computed_answers[x][y - 1] = \
                    minimum cost path(x, y - 1)
            cost_from_below = storage_for_computed_answers[x][y - 1]
            return cost_from_below[0] + cost(x, y, x, y - 1), \
                   cost_from_below[1] + [[x, y]]
# Print the optimal costs and paths. Ignore the mumbo jumbo
path_string = ""
for i,p in enumerate(path_tenten):
    if i%5==0:
        path_string+="\n"
    path string += "(" + str(p[0]) + ", " + str(p[1]) + ")-->"
path_string = path_string[:-3]
print("Minimum cost to reach 10,10 = {0}\n Optimal Path - {1}"
      .format(cost_tenten, path_string))
path_string = ""
print("\n")
for i,p in enumerate(path_foureight):
    if i%5==0:
        path_string+="\n"
   path\_string += "(" + str(p[0]) + ", " + str(p[1]) + ")-->"
path_string = path_string[:-3]
print("Minimum cost to reach 4,8 = {0}\n Optimal Path - {1}"
      .format(cost_foureight, path_string))
```

```
Minimum cost to reach 10,10 = 100.35381234337648 Optimal Path - (0,0)->(0,1)->(0,2)->(0,3)->(0,4)-> (0,5)->(1,5)->(2,5)->(3,5)->(4,5)-> (5,5)->(6,5)->(7,5)->(8,5)->(8,6)-> (8,7)->(9,7)->(10,7)->(10,8)->(10,9)-> (10,10)

Minimum cost to reach 4,8 = 29.38051878887093 Optimal Path - (0,0)->(0,1)->(0,2)->(0,3)->(0,4)-> (0,5)->(0,6)->(0,7)->(0,8)->(1,8)->(1,8)-> (2,8)->(3,8)->(4,8)
```

Lets again plot and visualize. Notice that this time answers were calculated MUCH MUCH FASTER than before

```
In [209]: # Lets visualize it via plots
          # You can ignore the code and focus on plot and
          # get an intuitive feel
          fig, ax = plt.subplots(1, 2, figsize=(20, 10))
          ax[0].grid(color='k', linestyle='--', linewidth=1)
          ax[0].set_xticks(np.arange(0, 11, 1))
          ax[0].set_yticks(np.arange(0, 11, 1))
          for i in range(1,len(path_tenten)):
              ax[0].arrow(path_tenten[i-1][0], path_tenten[i-1][1],
                          path_tenten[i][0]-path_tenten[i-1][0],
                          path_tenten[i][1]-path_tenten[i-1][1],
                          head_width=0.2, head_length=0.3, fc='r',
                          ec='r', length_includes_head=True)
          ax[0].scatter(x, y)
          ax[1].grid(color='k', linestyle='--', linewidth=1)
          ax[1].set_xticks(np.arange(0, 11, 1))
          ax[1].set_yticks(np.arange(0, 11, 1))
          for i in range(1,len(path_foureight)):
              ax[1].arrow(path_foureight[i-1][0], path_foureight[i-1][1],
                          path_foureight[i][0]-path_foureight[i-1][0],
                          path_foureight[i][1]-path_foureight[i-1][1],
                          head width=0.2, head length=0.3, fc='r',
                          ec='r', length_includes_head=True)
          ax[1].scatter(x, y)
          ax[1].scatter(10,10,marker="o",cmap="r", s=300)
          ax[0].scatter(10,10,marker="o",cmap="r", s=300)
```

```
# Our 2-dimensional distribution will be over variables X and Y
delta = 0.25
xg = np.arange(-1, 11, delta)
yg = np.arange(-1, 11, delta)
Xg, Yg = np.meshgrid(xg, yg)
Z = height(Xg,Yg)
CS = ax[0].contourf(Xg, Yg, Z, 50)
CS = ax[1].contourf(Xg, Yg, Z, 50)
cbar = plt.colorbar(CS)

plt.show()
```

1.4 Part 2: Weekdays

Try to guess what change will you make to calculate optimal path on weekdays? You know that the person can now travel north, south, east and west.

Hint: Now there are 4 ways to approach every point

I have marked places where you need to add corresponding code with bold comments. Try doing it!

```
storage for computed answers = [[-1 for _ in range(11)] for _ in range(11)]
def minimum_cost_path(x, y, top_corner):
    # If x=0 and y=0, reaching this point is trivial
    # and takes 0 fuel and optimal path is [[0,0]]
    # Just do not drive
    if (x == 0 \text{ and } y == 0):
        return 0, [[0, 0]]
    else:
        # But if both not O...we have following cases
        if x > 0 and y > 0:
            if x == top_corner[0] and y < top_corner[1]:</pre>
                # There are 3 ways to approach.
                # This one is already wirtten for you
                if storage_for_computed_answers[x - 1][y] == -1:
                    storage_for_computed_answers[x - 1][y] = \
                        minimum_cost_path(x - 1, y, top_corner)
                cost_from_left = storage_for_computed_answers[x - 1][y] + \
                                 cost(x, y, x - 1, y)
                if storage for computed answers [x][y + 1] == -1:
                    storage_for_computed_answers[x][y + 1] = \
                        minimum_cost_path(x, y + 1, top_corner)
                cost_from_top = storage_for_computed_answers[x][y + 1] + \
                                 cost(x, y, x, y + 1)
                if storage_for_computed_answers[x][y - 1] == -1:
                    storage_for_computed_answers[x][y - 1] = \
                        minimum_cost_path(x, y - 1, top_corner)
                cost_from_bottom = storage_for_computed_answers[x][y - 1] + \
                                   cost(x, y, x, y - 1)
                list_of_cost_and_paths = [cost_from_left, cost_from_top,
                                           cost from bottom]
                list_of_costs = [e[0] for e in list_of_cost_and_paths]
                idx = np.argmin(list of costs)
                return list_of_cost_and_paths[idx][0], \
                       list_of_cost_and_paths[idx][1] + [[x, y]]
            elif y == top_corner[1] and x < top_corner[0]:</pre>
                # There are 3 ways to approach
                if storage_for_computed_answers[x - 1][y] == -1:
                    storage_for_computed_answers[x - 1][y] = \
                        minimum_cost_path(x - 1, y, top_corner)
                cost_from_left = storage_for_computed_answers[x - 1][y] + \
                                 cost(x, y, x - 1, y)
```

```
if storage_for_computed_answers[x + 1][y] == -1:
        storage_for_computed_answers[x + 1][y] = \
           minimum_cost_path(x + 1, y, top_corner)
   cost from right = storage for computed answers [x + 1][y] + \
                      cost(x, y, x + 1, y)
   if storage_for_computed_answers[x][y - 1] == -1:
       storage for computed answers [x][y-1] = \
           minimum_cost_path(x, y - 1, top_corner)
   cost_from_bottom = storage_for_computed_answers[x][y - 1] + \
                       cost(x, y, x, y - 1)
   list_of_cost_and_paths = [cost_from_left, cost_from_right,
                              cost_from_bottom]
   list_of_costs = [e[0] for e in list_of_cost_and_paths]
   idx = np.argmin(list_of_costs)
   return list_of_cost_and_paths[idx][0], \
           list_of_cost_and_paths[idx][1] + [[x, y]]
elif x == top corner[0] and y == top corner[1]:
    # There are 2 ways to approach\
   if storage_for_computed_answers[x - 1][y] == -1:
        storage_for_computed_answers[x - 1][y] = \
           minimum_cost_path(x - 1, y, top_corner)
   cost_from_left = storage_for_computed_answers[x - 1][y] + \
                     cost(x, y, x - 1, y)
   if storage_for_computed_answers[x][y - 1] == -1:
       storage_for_computed_answers[x][y - 1] = \
           minimum_cost_path(x, y - 1, top_corner)
   cost_from_bottom = storage_for_computed_answers[x][y - 1] + \
                       cost(x, y, x, y - 1)
   list_of_cost_and_paths = [cost_from_left, cost_from_bottom]
   list_of_costs = [e[0] for e in list_of_cost_and_paths]
   idx = np.argmin(list_of_costs)
   return list_of_cost_and_paths[idx][0], \
           list_of_cost_and_paths[idx][1] + [[x, y]]
else: # Both <10 and >0
   # There are 4 ways to approach
   if storage_for_computed_answers[x - 1][y] == -1:
        storage_for_computed_answers[x - 1][y] = \
           minimum_cost_path(x - 1, y, top_corner)
   cost_from_left = storage_for_computed_answers[x - 1][y] + \
```

```
if storage_for_computed_answers[x + 1][y] == -1:
            storage_for_computed_answers[x + 1][y] = \
                minimum cost path(x + 1, y, top corner)
        cost_from_right = storage_for_computed_answers[x + 1][y] + \
                          cost(x, y, x + 1, y)
        if storage_for_computed_answers[x][y + 1] == -1:
            storage_for_computed_answers[x][y + 1] = \
                minimum_cost_path(x, y + 1, top_corner)
        cost_from_top = storage_for_computed_answers[x][y + 1] + \
                        cost(x, y, x, y + 1)
        if storage_for_computed_answers[x][y - 1] == -1:
            storage_for_computed_answers[x][y - 1] = \
                minimum_cost_path(x, y - 1, top_corner)
        cost_from_bottom = storage_for_computed_answers[x][y - 1] + \
                           cost(x, y, x, y - 1)
        list_of_cost_and_paths = [cost_from_left, cost_from_right,
                                  cost from top, cost from bottom]
        list_of_costs = [e[0] for e in list_of_cost_and_paths]
        idx = np.argmin(list_of_costs)
        return list_of_cost_and_paths[idx][0], \
               list_of_cost_and_paths[idx][1] + [[x, y]]
elif x > 0 and y == 0:
    # There are 3 ways to approach
    if storage_for_computed_answers[x - 1][y] == -1:
        storage_for_computed_answers[x - 1][y] =
        minimum_cost_path(x - 1, y, top_corner)
    cost_from_left = storage_for_computed_answers[x - 1][y] +
    cost(x, y, x - 1, y)
    if storage for computed answers [x + 1][y] == -1:
        storage_for_computed_answers[x + 1][y] = \
            minimum_cost_path(x + 1, y, top_corner)
    cost_from_right = storage_for_computed_answers[x + 1][y] + \
                      cost(x, y, x + 1, y)
    if storage_for_computed_answers[x][y + 1] == -1:
        storage_for_computed_answers[x][y + 1] = \
            minimum_cost_path(x, y + 1, top_corner)
    cost_from_top = storage_for_computed_answers[x][y + 1] + \
                    cost(x, y, x, y + 1)
```

cost(x, y, x - 1, y)

```
cost_from_top]
                  list_of_costs = [e[0] for e in list_of_cost_and_paths]
                  idx = np.argmin(list_of_costs)
                  return list of cost and paths[idx][0], \
                         list of cost and paths[idx][1] + [[x, y]]
              else:
                  # There are 3 ways to approach
                  if storage_for_computed_answers[x + 1][y] == -1:
                      storage_for_computed_answers[x + 1][y] = \
                          minimum_cost_path(x + 1, y, top_corner)
                  cost_from_right = storage_for_computed_answers[x + 1][y] + \
                                    cost(x, y, x + 1, y)
                  if storage_for_computed_answers[x][y + 1] == -1:
                      storage_for_computed_answers[x][y + 1] = \
                          minimum_cost_path(x, y + 1, top_corner)
                  cost_from_top = storage_for_computed_answers[x][y + 1] + \
                                  cost(x, y, x, y + 1)
                  if storage_for_computed_answers[x][y - 1] == -1:
                      storage_for_computed_answers[x][y - 1] = \
                          minimum_cost_path(x, y - 1, top_corner)
                  cost_from_bottom = storage_for_computed_answers[x][y - 1] + \
                                     cost(x, y, x, y - 1)
                  list_of_cost_and_paths = [cost_from_bottom, cost_from_right,
                                            cost_from_top]
                  list_of_costs = [e[0] for e in list_of_cost_and_paths]
                  idx = np.argmin(list_of_costs)
                  return list_of_cost_and_paths[idx][0], \
                         list_of_cost_and_paths[idx][1] + [[x, y]]
      # Call the function to calculate optimal path to point 10,10
      cost_tenten, path_tenten = minimum_cost_path(10, 10, (10, 10))
    RecursionError
                                              Traceback (most recent call last)
    <ipython-input-218-e36a334c687f> in <module>()
    144 # Call the function to calculate optimal path to point 4,8
--> 145 cost_foureight, path_foureight = minimum_cost_path(1,1,(1,1))
```

list_of_cost_and_paths = [cost_from_left, cost_from_right,

```
146
    147 # Print the optimal costs and paths. Ignore the mumbo jumbo
    <ipython-input-218-e36a334c687f> in minimum_cost_path(x, y, top_corner)
                        # There are 2 ways to approach\
                        if storage_for_computed_answers[x-1][y]==-1:
     64
                            storage_for_computed_answers[x-1][y] = minimum_cost_path(x-1,y
---> 65
                        cost_from_left = storage_for_computed_answers[x-1][y] + cost(x,y,x)
     66
     67
    <ipython-input-218-e36a334c687f> in minimum_cost_path(x, y, top_corner)
                    # There are 3 ways to approach
    122
                    if storage_for_computed_answers[x+1][y]==-1:
    123
                        storage_for_computed_answers[x+1][y] = minimum_cost_path(x+1,y,top
--> 124
    125
                    cost_from_right = storage_for_computed_answers[x+1][y] + cost(x,y,x+1,
    126
    ... last 2 frames repeated, from the frame below ...
    <ipython-input-218-e36a334c687f> in minimum_cost_path(x, y, top_corner)
     63
                        # There are 2 ways to approach\
                        if storage_for_computed_answers[x-1][y]==-1:
     64
                            storage_for_computed_answers[x-1][y] = minimum_cost_path(x-1,y
---> 65
                        cost_from_left = storage_for_computed_answers[x-1][y] + cost(x,y,x)
     66
     67
```

 ${\tt RecursionError:\ maximum\ recursion\ depth\ exceeded\ in\ comparison}$

1.5 What happened?

The code throws an error that says maximum recursion depth exceeded! Why does this happen? Hint: Paths can have cycles that go indefinitely and therefore number of path to reach a point are infinite!

How is this problem solved? This is a traditional problem of shortest path and can be solved by graph based algorithms like breadth first search. Do read them if interested. See how those algorithms tackle this problem and distinguish it from our dynamic programming approach