

# Interpolation and Its Applications (Exercises)

Vinh Dinh Nguyen  
PhD in Computer Science

WHAT  
HAVE YOU  
LEARNED



## ❖ Image Processing

Grayscale Image  
(Height, Width)



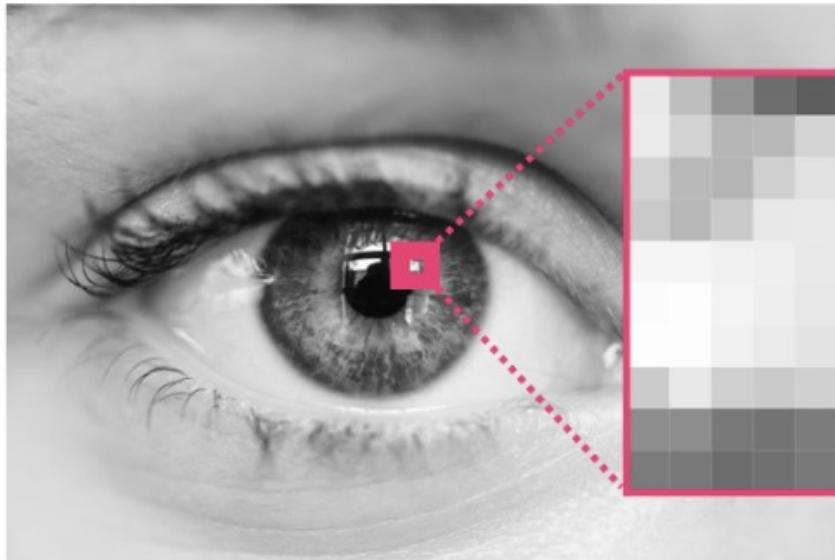
Color Image  
(Height, Width, Channel)



# Image Data

## ❖ Grayscale images

Height



Width

Pixel  $p$  = scalar

$0 \leq p \leq 255$

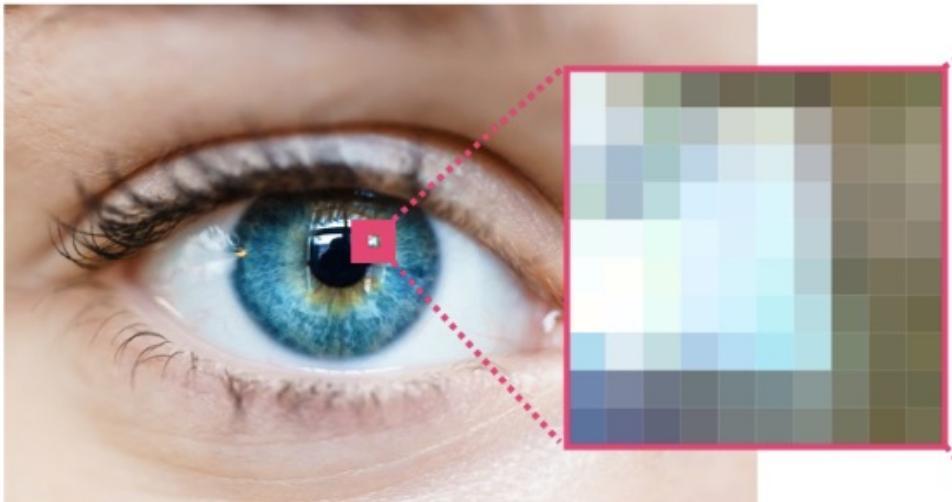
230	194	147	108	90	98	84	96	91	101
237	206	188	195	207	213	163	123	116	128
210	183	180	205	224	234	188	122	134	147
198	189	201	227	229	232	200	125	127	135
249	241	237	244	232	226	202	116	125	126
251	254	241	239	230	217	196	102	103	99
243	255	240	231	227	214	203	116	95	91
204	231	208	200	207	201	200	121	95	95
144	140	120	115	125	127	143	118	92	91
121	121	108	109	122	121	134	106	86	97

Resolution: #pixels

Resolution = Height  $\times$  Width

# Image Data

## ❖ Color images



(Height, Width, channel)

RGB color image

$$\text{Pixel } p = \begin{bmatrix} r \\ g \\ b \end{bmatrix}$$

$$0 \leq r,g,b \leq 255$$

233	188	137	96	90	95	63	73	73	82	
237	202	159	120	105	110	88	107	112	121	109
226	191	147	110	101	112	98	123	110	119	142
221	191	176	182	203	214	169	144	133	145	131
185	160	161	184	205	223	186	137	147	161	122
181	174	189	207	206	215	194	136	142	151	115
246	237	237	231	208	206	192	122	143	144	87
254	254	241	224	199	192	181	99	122	117	133
239	248	232	207	187	182	184	110	114	110	74
193	215	193	167	158	164	181	114	112	111	74
113	119	110	111	113	123	135	120	108	106	113
93	97	91	103	107	111	122	112	104	114	82

Resolution: #pixels  
Resolution = Height x Width

# Read Image by OpenCV

## ❖ Load an image

```
1 import cv2
2
3 # read a grayscale image
4 img = cv2.imread('nature.jpg', 0)
5
6 # show the image
7 cv2.imshow("img", img)
8
9 # copy them when showing an image
10 cv2.waitKey()
11 cv2.destroyAllWindows()

1 import cv2
2
3 # read a grayscale image
4 img = cv2.imread('nature.jpg', 0)
5
6 # save the image
7 cv2.imwrite('processed_image.jpg', img)
```



```
1 # get image info
2
3 import numpy as np
4 import cv2
5
6 # read a grayscale image
7 img = cv2.imread('nature.jpg', 0)
8
9 shape = img.shape
10 print(shape)
```

(500, 1200)

(Height, Width)

# Read Image by OpenCV

## ❖ Load an image

```
1 import cv2
2
3 # read a color image
4 img = cv2.imread('nature.jpg', 1)
5
6 # show the image
7 cv2.imshow("img", img)
8
9 # copy them when showing an image
10 cv2.waitKey()
11 cv2.destroyAllWindows()

1 import cv2
2
3 # read a color image
4 img = cv2.imread('nature.jpg', 1)
5
6 # save the image
7 cv2.imwrite('processed_image.jpg', img)
```



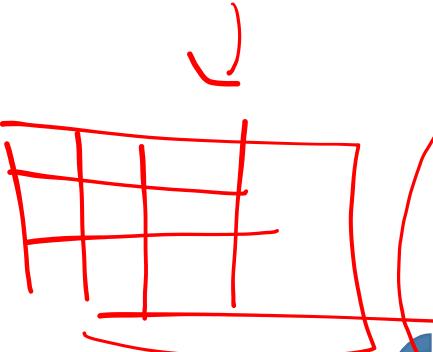
```
1 # get image info
2
3 import numpy as np
4 import cv2
5
6 # read a grayscale image
7 img = cv2.imread('nature.jpg', 1)
8
9 shape = img.shape
10 print(shape)
```

(500, 1200, 3)

(Height, Width, Channel)

# Image Resize

❖ How ???



Copy and  
interpolation

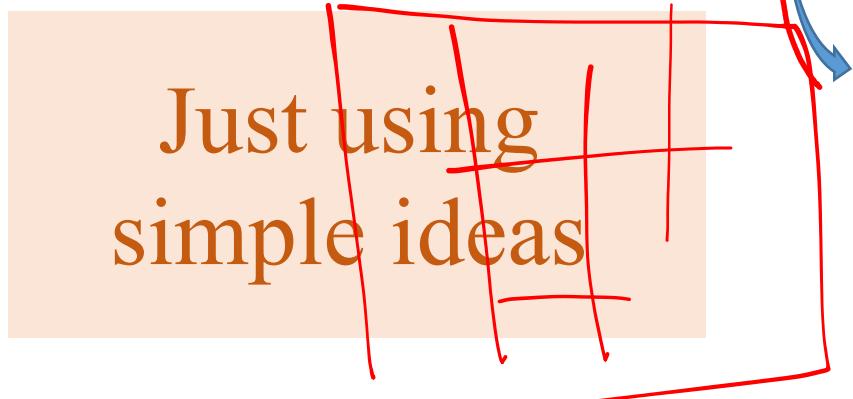
Mapping (scale is an integer)



(1000, 2400)



Just using  
simple ideas



# Image Resize



❖ How ???

Mapping (scale is an integer)

Old resolution: (5, 5)

Copy and  
interpolation

0	2	4	6	8
1	3	5	7	9
0	2	4	6	8
1	3	5	7	9
0	2	4	6	8



new resolution: (10, 10)

✓									
.	✓								
✓	.	✓							
			✓						
			.	✓					
				.	✓				
					✓				
					.	✓			
						✓			
						.	✓		

Just using  
simple ideas

# Image Resize



❖ How ???

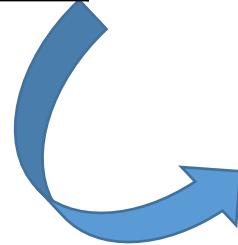
Mapping (scale is an integer)

Old resolution: (5, 5)

0	2	4	6	8
1	3	5	7	9
0	2	4	6	8
1	3	5	7	9
0	2	4	6	8

Copy and  
interpolation

Just using  
simple ideas



new resolution: (10, 10)

0									
1									
0									
1									
0									

How to fill in value in:



0		

# Missing Data

Formatted Date	Temperature (C)
2006-01-01 00:00:00.000 +0100	0.577778
2006-01-01 01:00:00.000 +0100	1.161111
2006-01-01 02:00:00.000 +0100	1.666667
2006-01-01 03:00:00.000 +0100	1.711111
2006-01-01 04:00:00.000 +0100	1.183333
2006-01-01 05:00:00.000 +0100	1.205556
2006-01-01 06:00:00.000 +0100	2.222222
2006-01-01 07:00:00.000 +0100	2.072222
2006-01-01 08:00:00.000 +0100	2.2
2006-01-01 09:00:00.000 +0100	NaN
2006-01-01 10:00:00.000 +0100	2.788889
2006-01-01 11:00:00.000 +0100	NaN
2006-01-01 12:00:00.000 +0100	4.911111
2006-01-01 13:00:00.000 +0100	6.205556
2006-01-01 14:00:00.000 +0100	NaN

Fill data

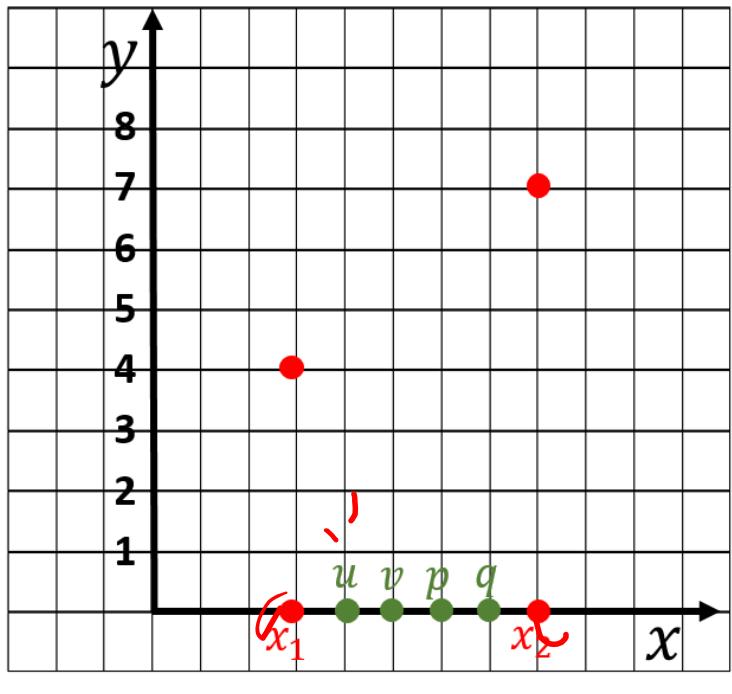
Formatted Date	Temperature (C)
2006-01-01 00:00:00.000 +0100	0.577778
2006-01-01 01:00:00.000 +0100	1.161111
2006-01-01 02:00:00.000 +0100	1.666667
2006-01-01 03:00:00.000 +0100	1.711111
2006-01-01 04:00:00.000 +0100	1.183333
2006-01-01 05:00:00.000 +0100	1.205556
2006-01-01 06:00:00.000 +0100	2.222222
2006-01-01 07:00:00.000 +0100	2.072222
2006-01-01 08:00:00.000 +0100	2.2
2006-01-01 09:00:00.000 +0100	???
2006-01-01 10:00:00.000 +0100	2.788889
2006-01-01 11:00:00.000 +0100	???
2006-01-01 12:00:00.000 +0100	4.911111
2006-01-01 13:00:00.000 +0100	6.205556
2006-01-01 14:00:00.000 +0100	???

How to fill in value in NaN

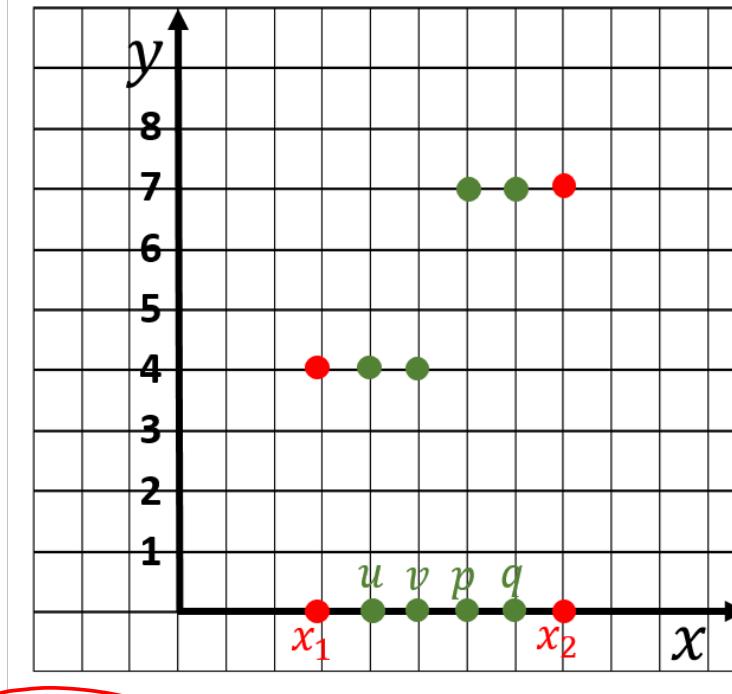
2.2
NaN
2.788889

# Interpolation

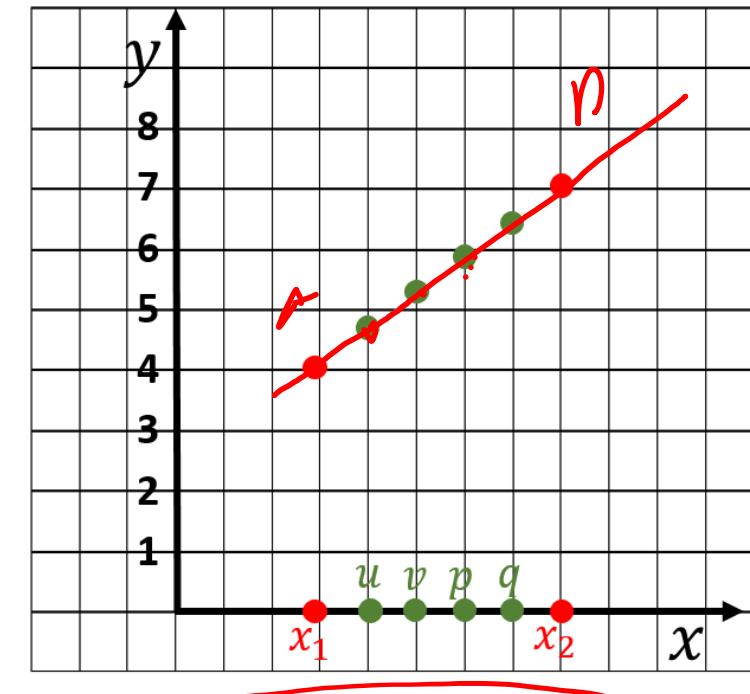
## ❖ Two simplest techniques



Tìm giá trị cho các vị trí  $u$ ,  $v$ ,  $p$  và  $q$

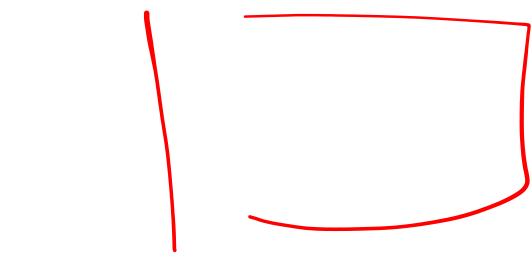


**Nearest neighbor:** Tính khoảng cách  
đến  $x_1$  và  $x_2$ , và lấy giá trị của  $x$  gần hơn

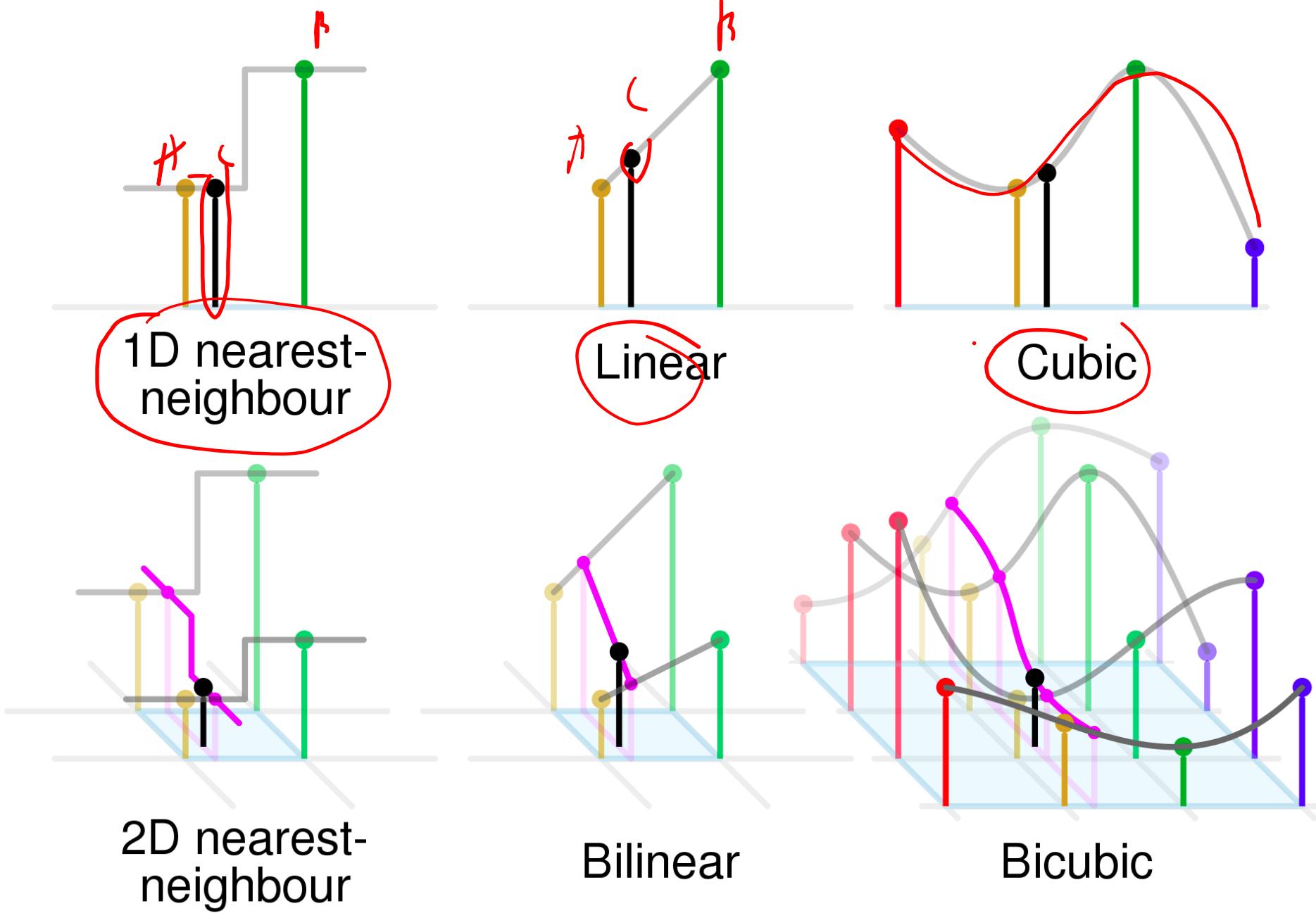


**Nội suy theo hàm tuyến tính**

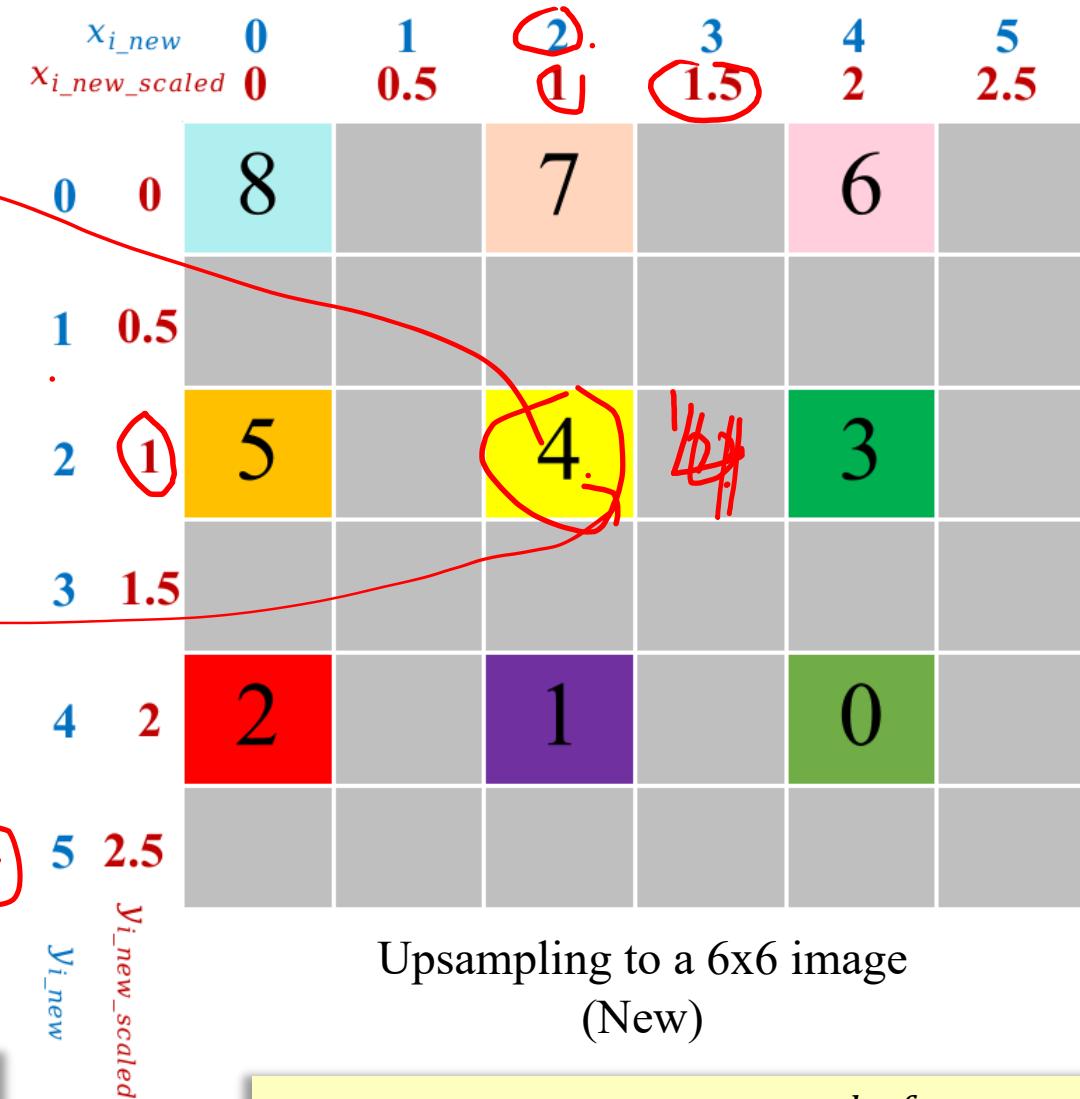
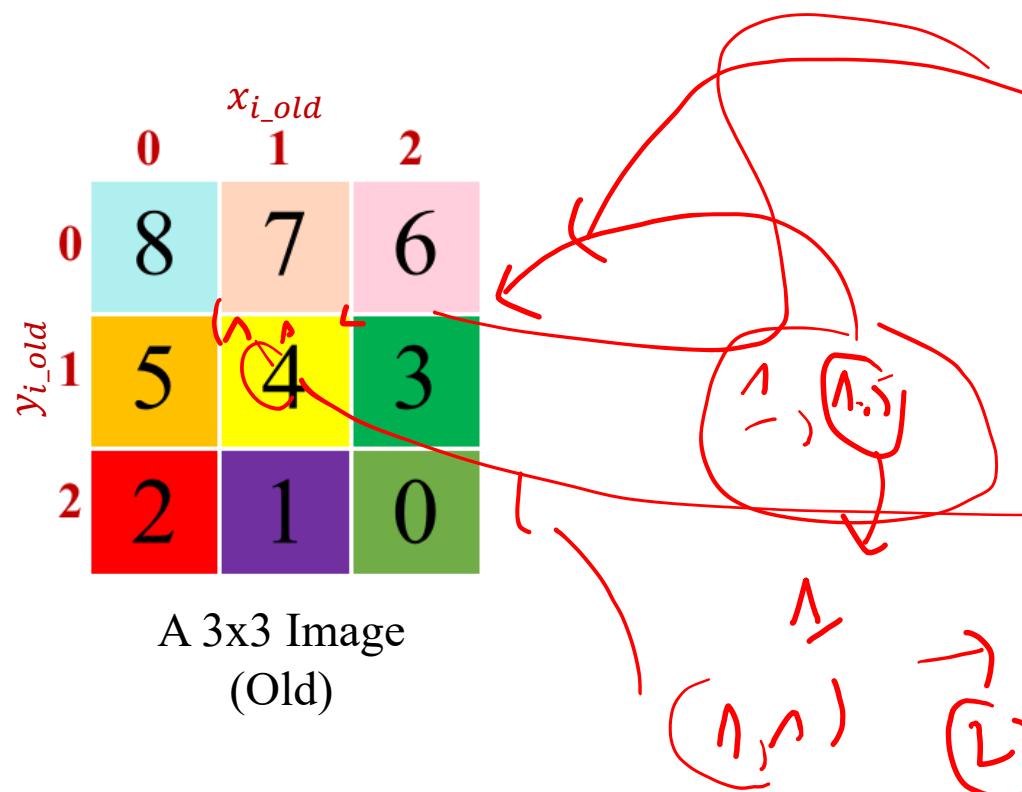
# Interpolation



[https://en.wikipedia.org/wiki/Nearest-neighbor\\_interpolation](https://en.wikipedia.org/wiki/Nearest-neighbor_interpolation)



# Nearest-neighbor interpolation



*old\_w: (width of Old) = 3*

*old\_h: (height of Old) = 3*

*new\_w: (width of New) = 6*

*new\_h: (height of New) = 6*

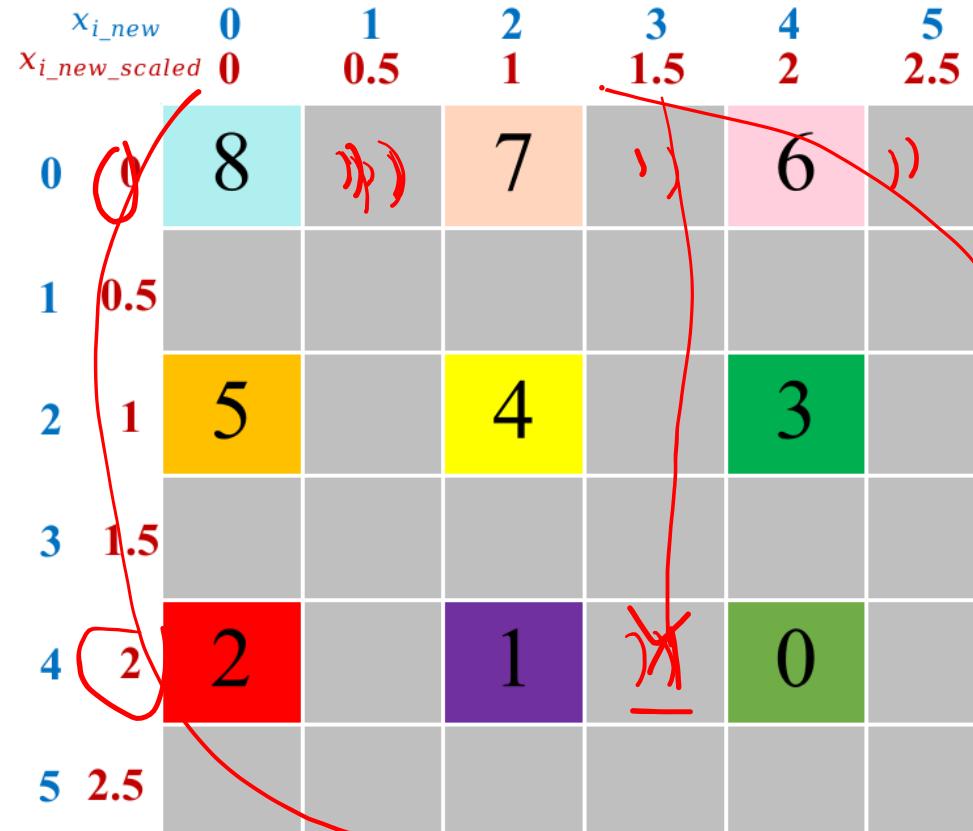
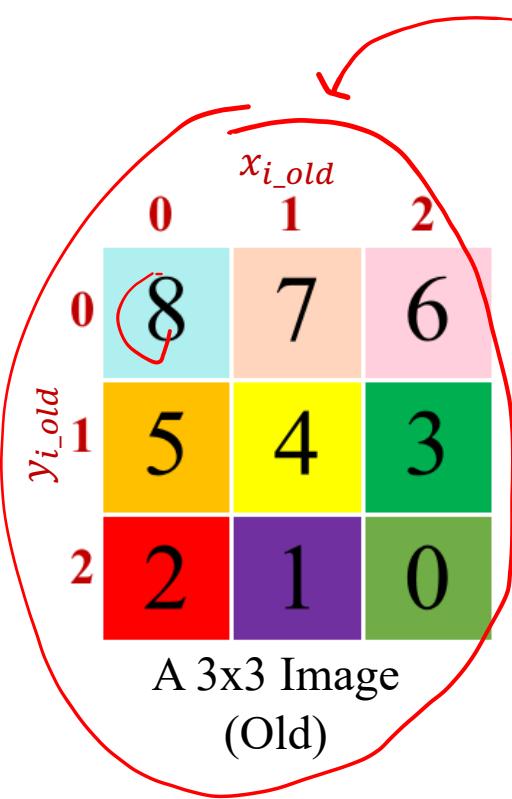
$$w\_scale\_factor = \frac{old\_w}{new\_w}$$

$$h\_scale\_factor = \frac{old\_h}{new\_h}$$

$$x_{i\_new\_scaled} = x_{i\_new} * w\_scale\_factor$$

$$y_{i\_new\_scaled} = y_{i\_new} * h\_scale\_factor$$

# Nearest-neighbor interpolation



*old\_w: (width of Old) = 3*

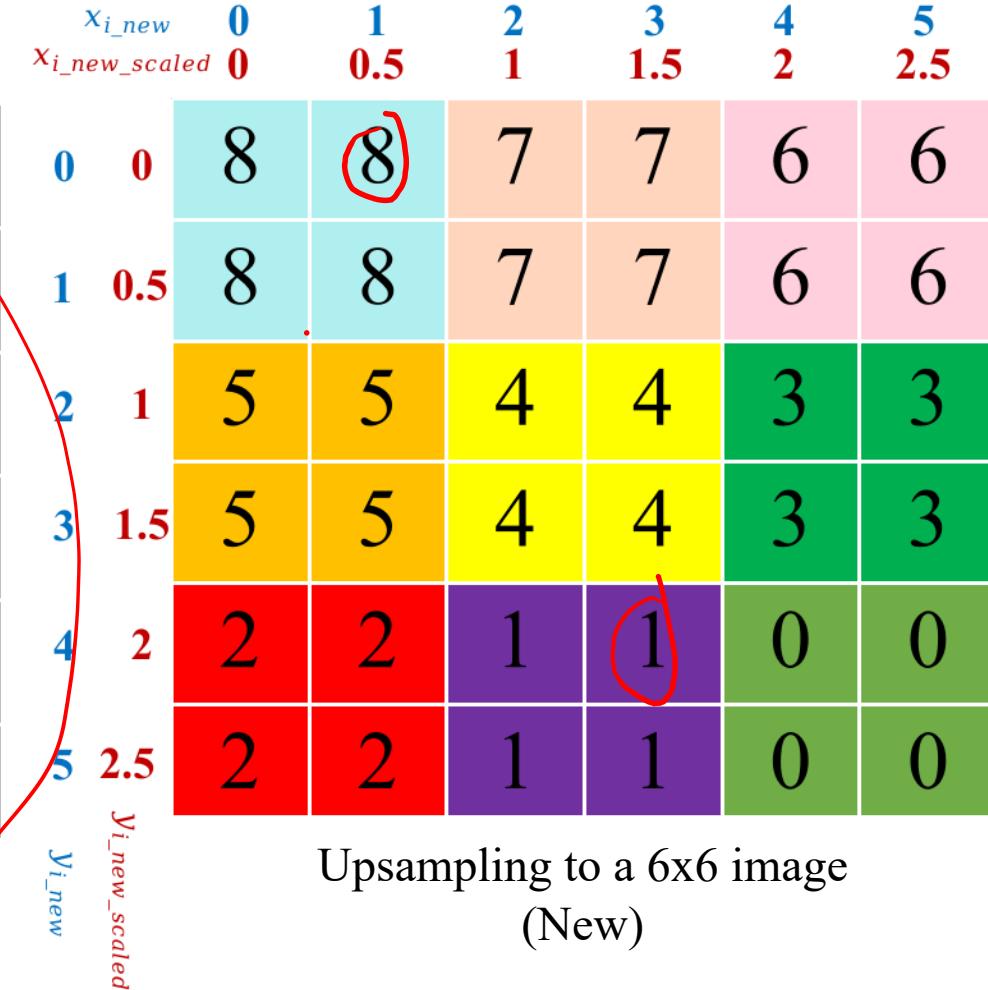
*old\_h: (height of Old) = 3*

*new\_w: (width of New) = 6*

*new\_h: (height of New) = 6*

$$w\_scale\_factor = \frac{old\_w}{new\_w}$$

$$h\_scale\_factor = \frac{old\_h}{new\_h}$$



*x<sub>i</sub>\_new\_scaled = x<sub>i</sub>\_new \* w\_scale\_factor*

*y<sub>i</sub>\_new\_scaled = y<sub>i</sub>\_new \* h\_scale\_factor*

# Nearest-neighbor interpolation

## Old resolution: (5, 5)

# Copy and interpolation

# Just using simple ideas

# Mapping (scale is an integer)

0	2	4	6	8
1	3	5	7	9
0	2	4	6	8
1	3	5	7	9
0	2	4	6	8



$\bar{b}_j v$ )

New resolution: (10, 10)

# Nearest-neighbor interpolation

Mapping (scale is an integer)

Old resolution: (5, 5)

Copy and  
interpolation

Just using  
simple ideas

0	2	4	6	8
1	3	5	7	9
0	2	4	6	8
1	3	5	7	9
0	2	4	6	8



New resolution: (10, 10)

0	0	2	2	4	4	6	6	8	8
0	0	2	2	4	4	6	6	8	8
1	1	3	3	5	5	7	7	9	9
1	1	3	3	5	5	7	7	9	9
0	0	2	2	4	4	6	6	8	8
0	0	2	2	4	4	6	6	8	8
1	1	3	3	5	5	7	7	9	9
1	1	3	3	5	5	7	7	9	9
0	0	2	2	4	4	6	6	8	8
0	0	2	2	4	4	6	6	8	8

# Nearest-neighbor interpolation

# Mapping (scale is an integer)

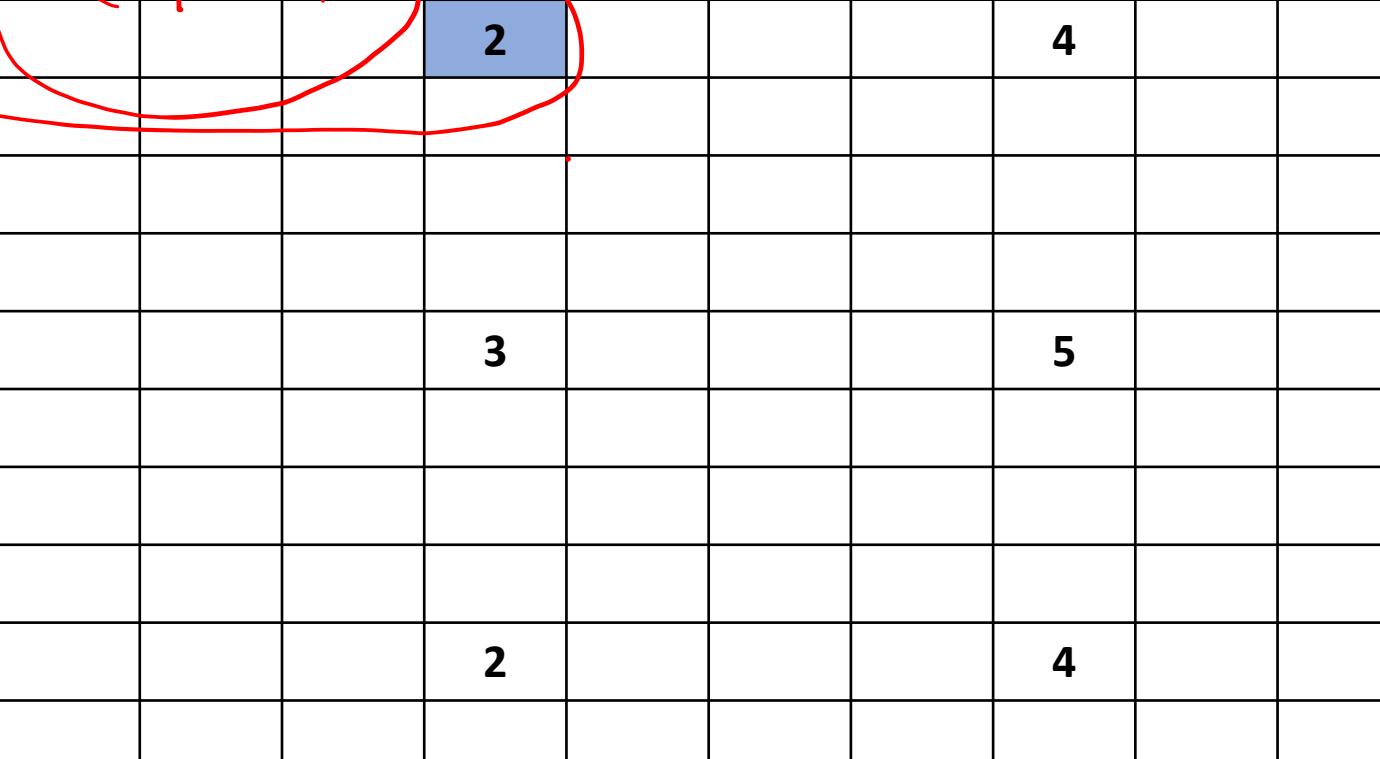
(3, 3)

0	2	4
1	3	5
0	2	4

# Copy and interpolation



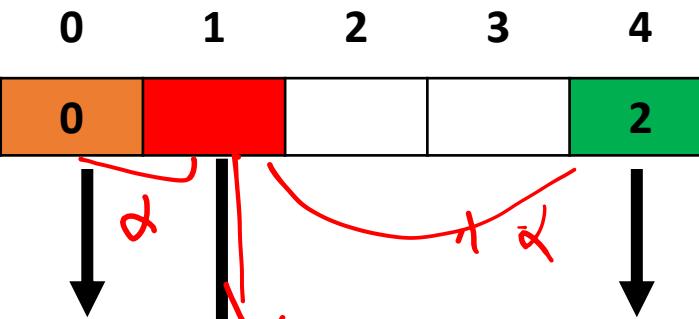
teger)



0	1	2	3	4	5	6	7	8	9	10	11
1		2	3	4	5						
0		2	4	6	8	10	12	14	16	18	20
1		3	5	7	9	11	13	15	17	19	21
2		4	6	8	10	12	14	16	18	20	22
3		5	7	9	11	13	15	17	19	21	23
4		6	8	10	12	14	16	18	20	22	24
5		7	9	11	13	15	17	19	21	23	25
6		8	10	12	14	16	18	20	22	24	26
7		9	11	13	15	17	19	21	23	25	27
8		10	12	14	16	18	20	22	24	26	28
9		11	13	15	17	19	21	23	25	27	29
10		12	14	16	18	20	22	24	26	28	30
11		13	15	17	19	21	23	25	27	29	31

X= index

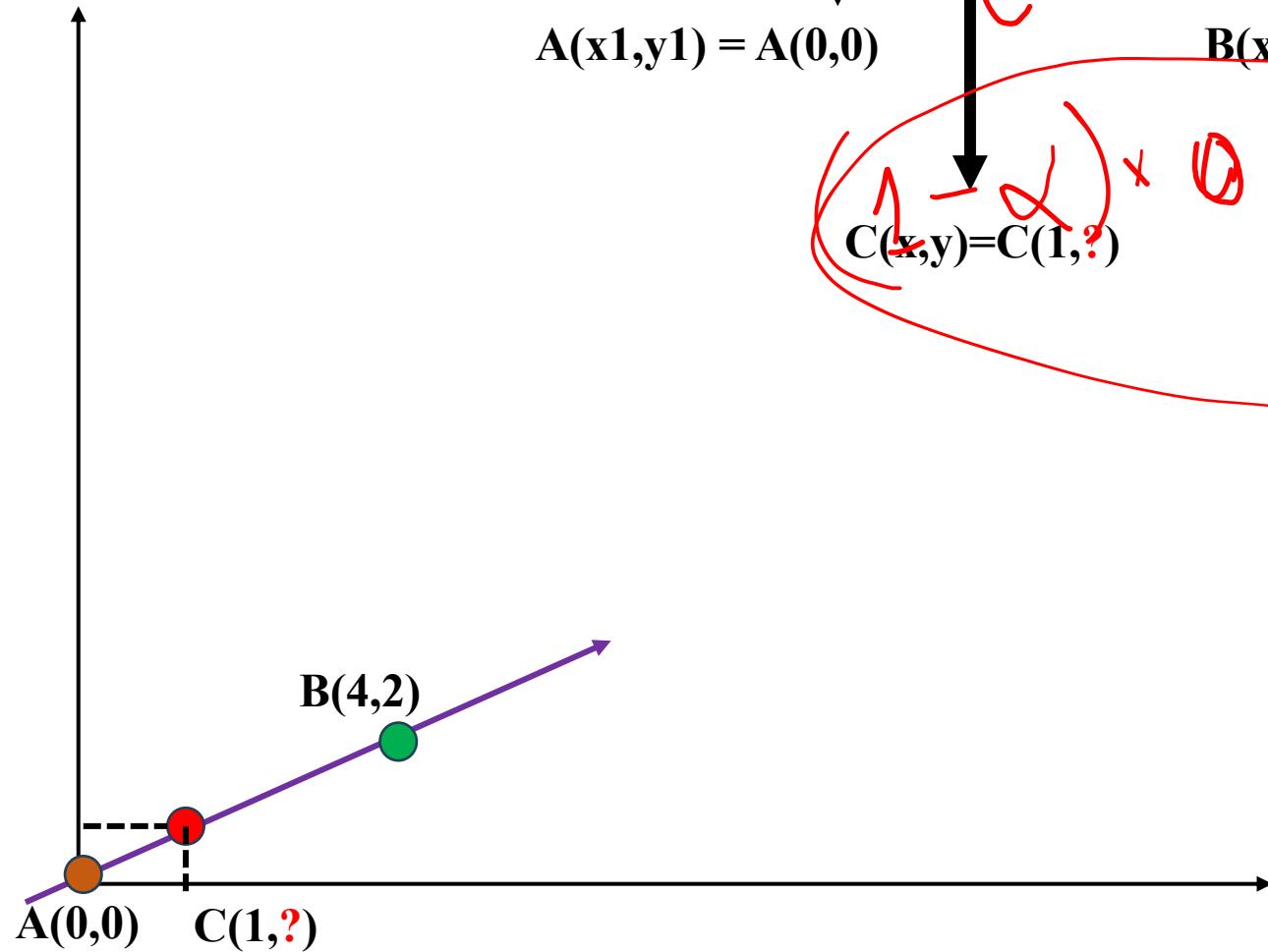
Y = value



$$A(x_1, y_1) = A(0, 0)$$

$$B(x_2, y_2) = B(4, 2)$$

$$C(x, y) = C(1, ?)$$
$$(1-x) \times 0 + x \times 2$$



#### Linear Interpolation Formula

$$\text{Linear Interpolation Formula } (y) = y_1 + \frac{(x - x_1)(y_2 - y_1)}{x_2 - x_1}$$

where,

- $x_1$  and  $y_1$  are the first coordinates
- $x_2$  and  $y_2$  are the second coordinates
- $x$  is the point to perform the interpolation
- $y$  is the interpolated value

Given A( $x_1, y_1$ ) and B( $x_2, y_2$ )

Line equation:  $y = ax + b$  (1)

From A and B, we have  $a = \frac{y_2 - y_1}{x_2 - x_1}$  (2)

$$(4) \Rightarrow y = \frac{(y_2 - y_1)x}{x_2 - x_1} + \frac{y_1(x_2 - x_1) - x_1(y_2 - y_1)}{x_2 - x_1}$$

$$= \frac{y_2x - y_1x + y_1x_2 - y_1x_1 - y_2x_1 + y_1x_1}{x_2 - x_1}$$

$$= \frac{x_2 - x}{x_2 - x_1} y_1 + \frac{x - x_1}{x_2 - x_1} y_2$$

$\alpha$        $1 - \alpha$

Replace A into (1), we have  $b = y_1 - \frac{y_2 - y_1}{x_2 - x_1} x_1$  (3)

From (1), (2), and (3)  $\Rightarrow y = \frac{y_2 - y_1}{x_2 - x_1} x + \left( y_1 - \frac{y_2 - y_1}{x_2 - x_1} x_1 \right)$  (4)

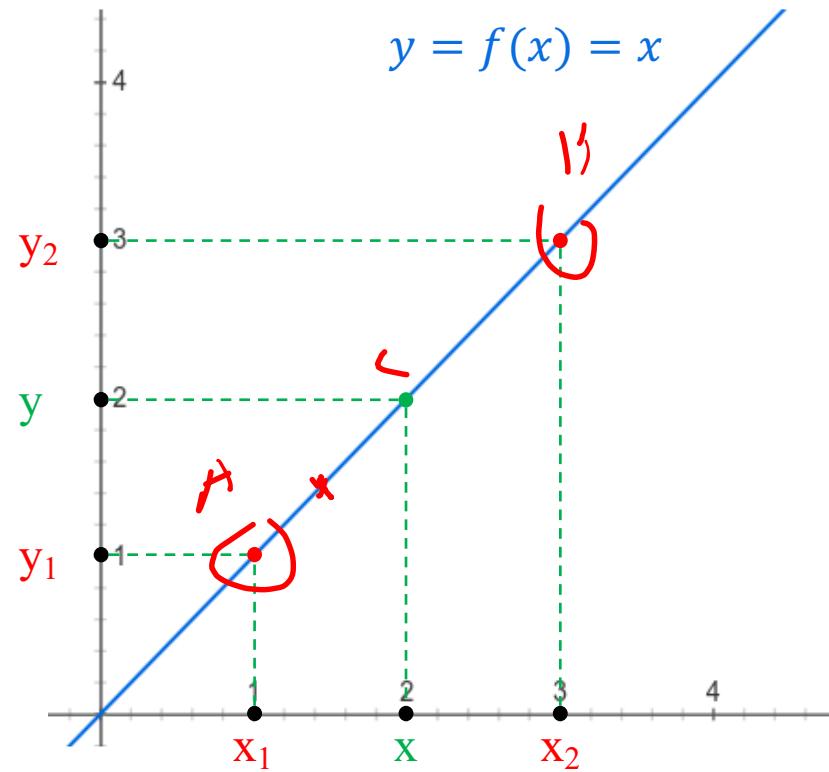
$$\frac{x_2 - x}{x_2 - x_1} + \frac{x - x_1}{x_2 - x_1} = 1$$



Line segment equation

$$v = \alpha v_1 + (1 - \alpha) v_2$$

# Linear interpolation



Giả sử không biết đường thẳng  $y = x$ , chỉ biết hai điểm  $(x_1=1, y_1=1)$  và  $(x_2=3, y_2=3)$ . Tìm điểm màu xanh lá có  $y=?$  khi biết  $x=2$ .

$$y \approx \frac{x_2 - x}{x_2 - x_1} y_1 + \frac{x - x_1}{x_2 - x_1} y_2 = \frac{3 - 2}{3 - 1} * 1 + \frac{2 - 1}{3 - 1} * 3 = 2$$

Tỉ lệ  $y$  phụ  
thuộc vào  $y_1$

Tỉ lệ  $y$  phụ  
thuộc vào  $y_2$

$(x=1.1, y=?)$

$$y \approx \frac{3 - 1.1}{3 - 1} * 1 + \frac{1.1 - 1}{3 - 1} * 3 \approx 0.95 * 1 + 0.05 * 3 \approx 1.1$$

Càng gần  $x_1$ ,  $y$  phụ thuộc nhiều vào  $y_1$  và ít phụ thuộc  $y_2$

$(x=2.9, y=?)$

$$y \approx \frac{3 - 2.9}{3 - 1} * 1 + \frac{2.9 - 1}{3 - 1} * 3 \approx 0.05 * 1 + 0.95 * 3 \approx 2.9$$

Càng gần  $x_2$ ,  $y$  phụ thuộc nhiều vào  $y_2$  và ít phụ thuộc  $y_1$

# Bilinear interpolation

# Mapping (scale is an integer)

(3, 3)

0	2	4
1	3	5
0	2	4

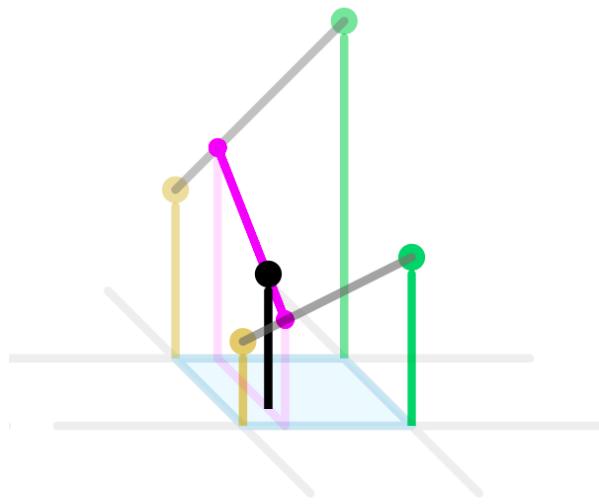
# Copy and interpolation



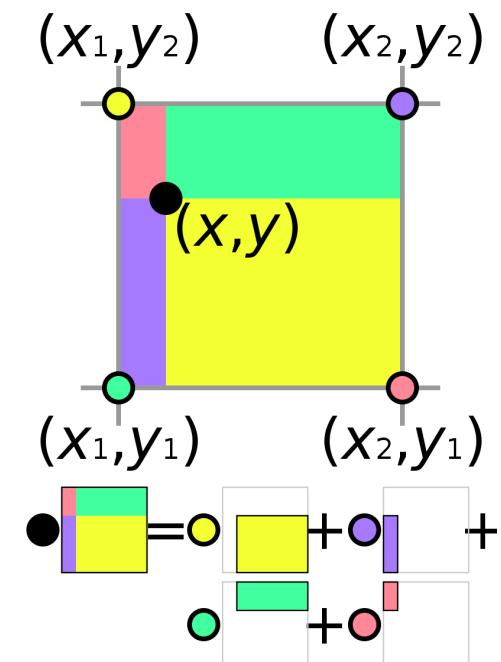
(12, 12)

## Scaled index

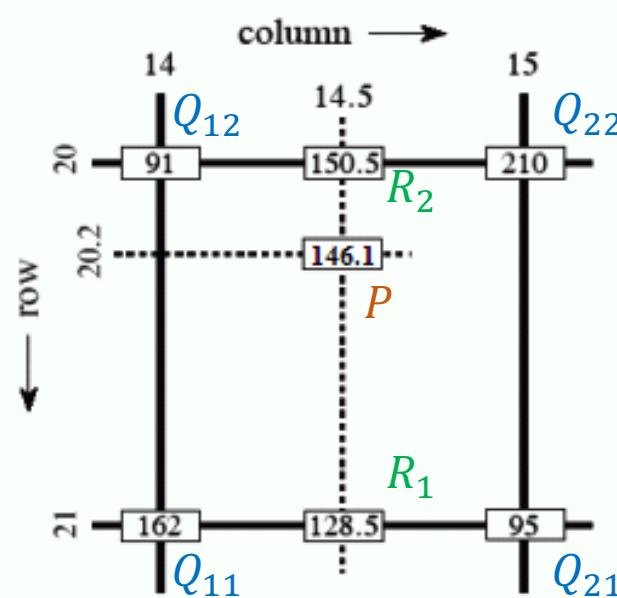
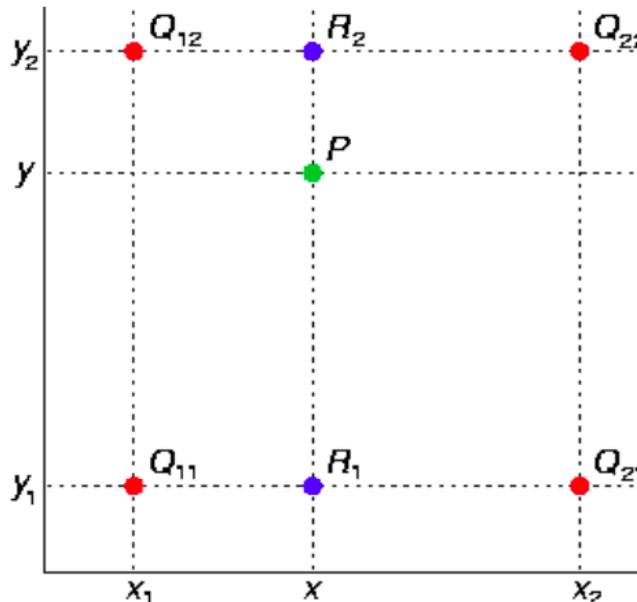
# Bilinear Interpolation



[https://en.wikipedia.org/wiki/Nearest\\_n-neighbor\\_interpolation](https://en.wikipedia.org/wiki/Nearest_n-neighbor_interpolation)



[https://en.wikipedia.org/wiki/Bilinear\\_interpolation](https://en.wikipedia.org/wiki/Bilinear_interpolation)



$$f(R_1) \approx \frac{x_2 - x}{x_2 - x_1} f(Q_{11}) + \frac{x - x_1}{x_2 - x_1} f(Q_{21})$$

$$f(R_2) \approx \frac{x_2 - x}{x_2 - x_1} f(Q_{12}) + \frac{x - x_1}{x_2 - x_1} f(Q_{22})$$

$$f(P) \approx \frac{y_2 - y}{y_2 - y_1} f(R_1) + \frac{y - y_1}{y_2 - y_1} f(R_2)$$

$$x_1 = 14, x_2 = 15, y_1 = 21, y_2 = 20$$

$$Q_{11} = (14, 21), Q_{21} = (15, 21)$$

$$Q_{12} = (14, 20), Q_{22} = (15, 20)$$

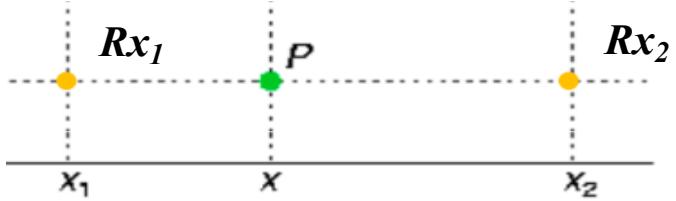
$$R_1 = (14.5, 21), R_2 = (14.5, 20), P = (14.5, 20.2)$$

$$f(R_1) \approx \frac{15 - 14.5}{15 - 14} 162 + \frac{14.5 - 14}{15 - 14} 95 = 128.5$$

$$f(R_2) \approx \frac{15 - 14.5}{15 - 14} 91 + \frac{14.5 - 14}{15 - 14} 210 = 150.5$$

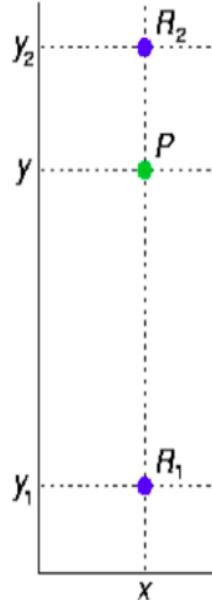
$$f(P) \approx \frac{20 - 20.2}{20 - 21} 128.5 + \frac{20.2 - 21}{20 - 21} 150.5 = 146.1$$

### Horizontal Linear Interpolation

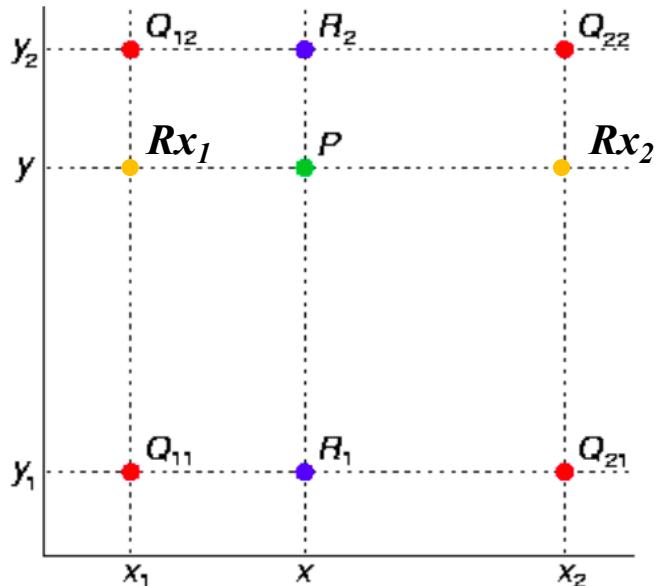


$$f(P) \approx \frac{x_2 - x}{x_2 - x_1} f(Rx_1) + \frac{x - x_1}{x_2 - x_1} f(Rx_2)$$

### Vertical Linear Interpolation



$$f(P) \approx \frac{y_2 - y}{y_2 - y_1} f(R_1) + \frac{y - y_1}{y_2 - y_1} f(R_2)$$



### Bilinear Interpolation

$$f(R_1) \approx \frac{x_2 - x}{x_2 - x_1} f(Q_{11}) + \frac{x - x_1}{x_2 - x_1} f(Q_{21})$$

$$f(R_2) \approx \frac{x_2 - x}{x_2 - x_1} f(Q_{12}) + \frac{x - x_1}{x_2 - x_1} f(Q_{22})$$

$$f(P) \approx \frac{y_2 - y}{y_2 - y_1} f(R_1) + \frac{y - y_1}{y_2 - y_1} f(R_2)$$



**TIME  
FOR  
ACTION**

# Problem 1

**Câu hỏi 1** (Kỹ thuật thao tác trên dữ liệu danh sách 2 chiều): Cho trước một danh sách các danh sách số tự nhiên được lưu trữ trong danh sách (List) 2 chiều gồm có m dòng và n cột. Phát triển chương trình tính tổng cho mỗi dòng của danh sách. Bên dưới là ví dụ minh họa các test case đầu vào và đầu ra của chương trình.

Examples

Input :

```
data = [[1, 2, 3],  
        [4, 5, 6],  
        [7, 8, 9]]
```

Output :

```
result = [6, 15, 24]
```

-----.

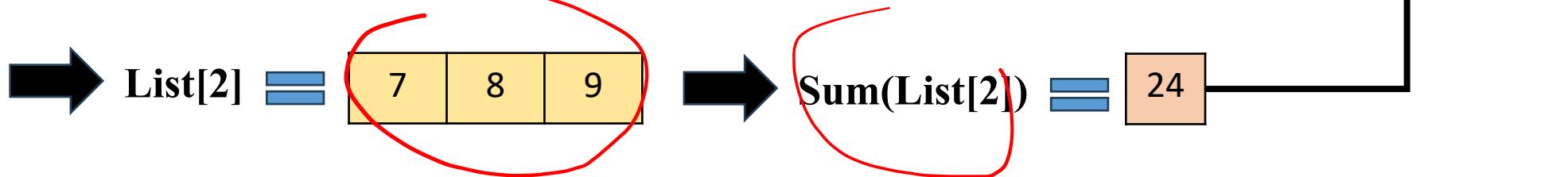
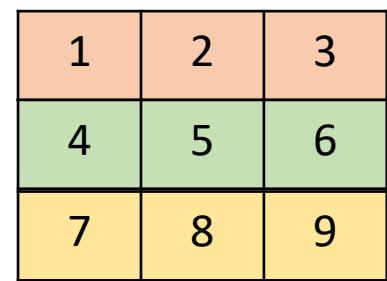
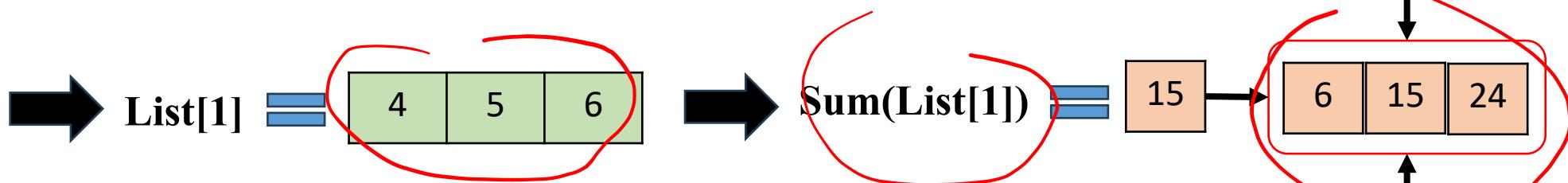
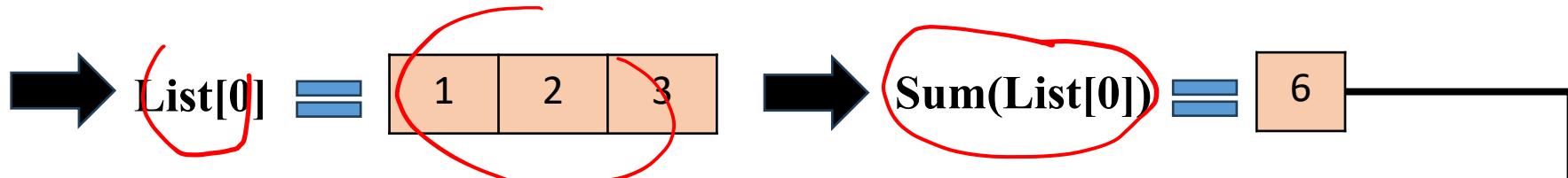
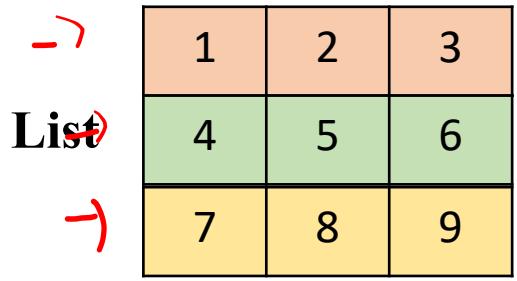
Input :

```
data = [[1, 2],  
        [3, 4],  
        [5, 6]]
```

Output :

```
[3, 7, 11]
```

# Problem Analysis



# Solution

```
# implementation
def compute_sum(data):
    result = []

    # your code here

    return result
.

# test case
data = [[1, 2, 3],
         [4, 5, 6],
         [7, 8, 9]]
assert compute_sum(data) == [6, 15, 24]

data = [[1, 2],
         [3, 4],
         [5, 6]]
assert compute_sum(data) == [3, 7, 11]

# multiple choice
data = [[6, 8],
         [3, 1],
         [7, 3]]
print(compute_sum(data))
```

# Problem 2

**Câu hỏi 2** (Kỹ thuật thao tác trên dữ liệu danh sách 2 chiều):

Cho trước một danh sách các danh sách số tự nhiên được lưu trữ trong danh sách (List) 2 chiều gồm có  $n$  dòng và  $n$  cột (ma trận vuông). Phát triển chương trình hoán vị dữ liệu (thay đổi dòng thành cột và ngược lại). Bên dưới là ví dụ minh họa các test case đầu vào và đầu ra của chương trình.

Examples

Input :

```
data = [[1, 2, 3],  
        [4, 5, 6],  
        [7, 8, 9]]
```

Output :

```
result = [[1, 4, 7],  
          [2, 5, 8],  
          [3, 6, 9]]
```

-----

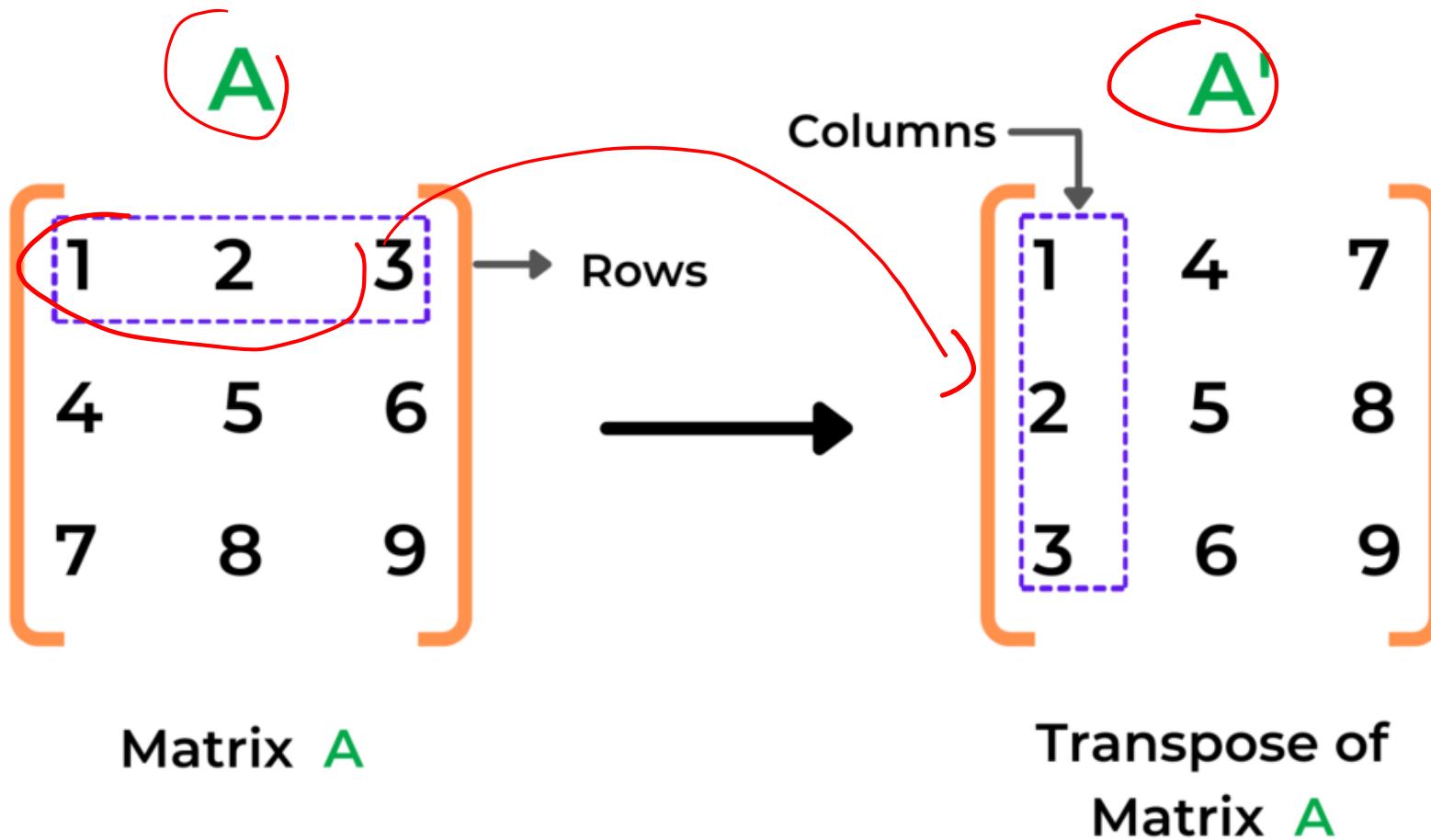
Input :

```
data = [[1, 2],  
        [4, 5]]
```

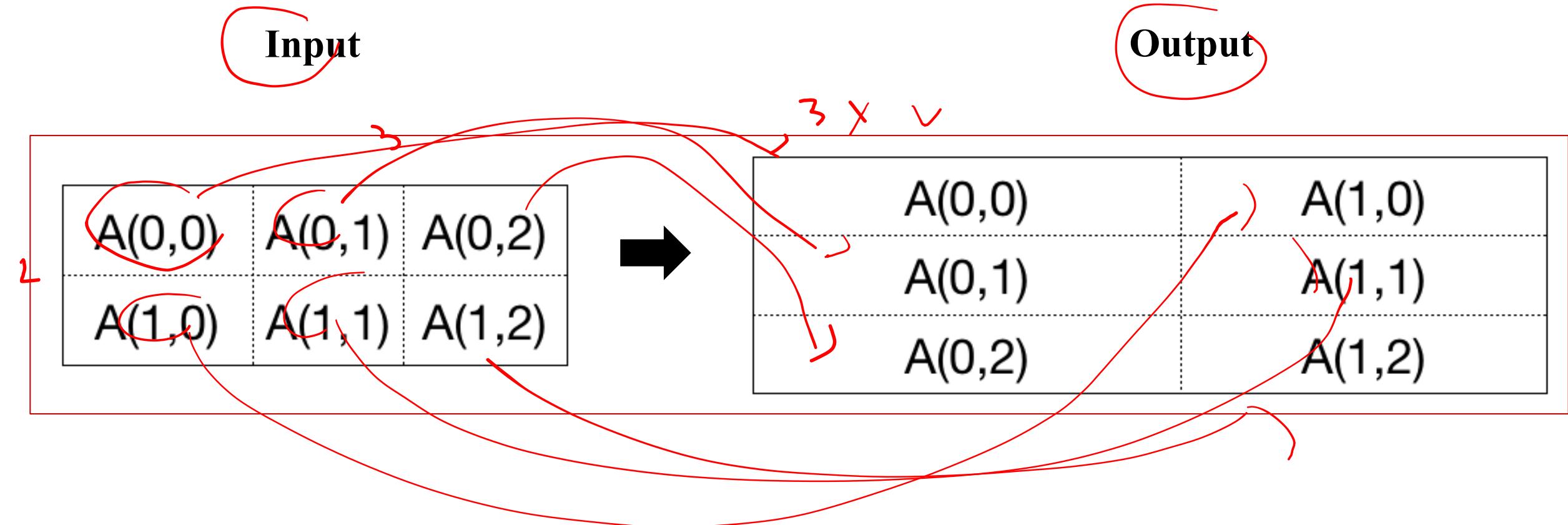
Output :

```
result = [[1, 4],  
          [2, 5]]
```

# Problem Analysis



# Problem Analysis



# Solution

```
# implementation
def transpose(data):
    # your code here

    return result

# test case
data = [[1, 2, 3],
        [4, 5, 6],
        [7, 8, 9]]
assert transpose(data) == [[1, 4, 7],
                          [2, 5, 8],
                          [3, 6, 9]]

data = [[1, 2],
        [4, 5]]
assert transpose(data) == [[1, 4],
                          [2, 5]]

# multiple choice
data = [[7, 2],
        [3, 5]]
print(transpose(data))
```

# Problem 3

**Câu hỏi 3** (kỹ thuật nội suy liền kề cho dữ liệu 1 chiều (1D)): Cho trước một danh sách các số tự nhiên có chiều dài là n và một tham số k. Phát triển chương trình nội suy danh sách đầu vào thành danh sách mới có độ dài là  $n*k$  sử dụng kỹ thuật nội suy liền kề (nearest neighbor interpolation). Bên dưới là ví dụ minh họa các test case đầu vào và đầu ra của chương trình.

Examples

Input :

```
data = [1, 2, 3]
k = 2
```

Output :

```
result = [1, 1, 2, 2, 3, 3]
-----
```

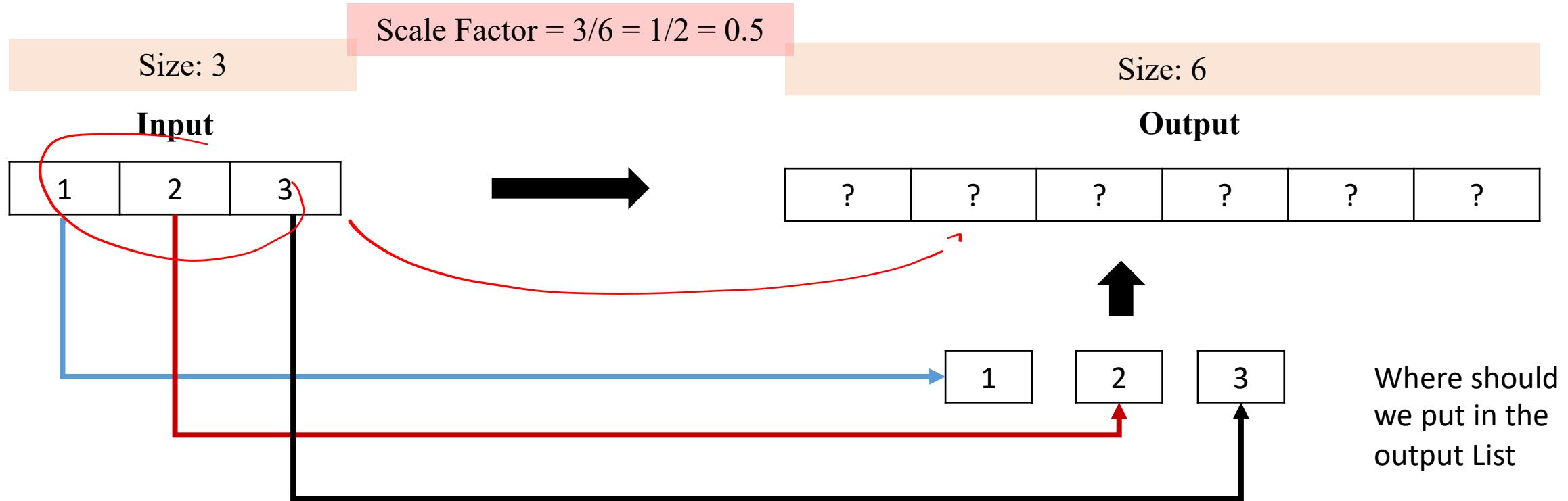
Input :

```
data = [1, 2, 3]
k = 3
```

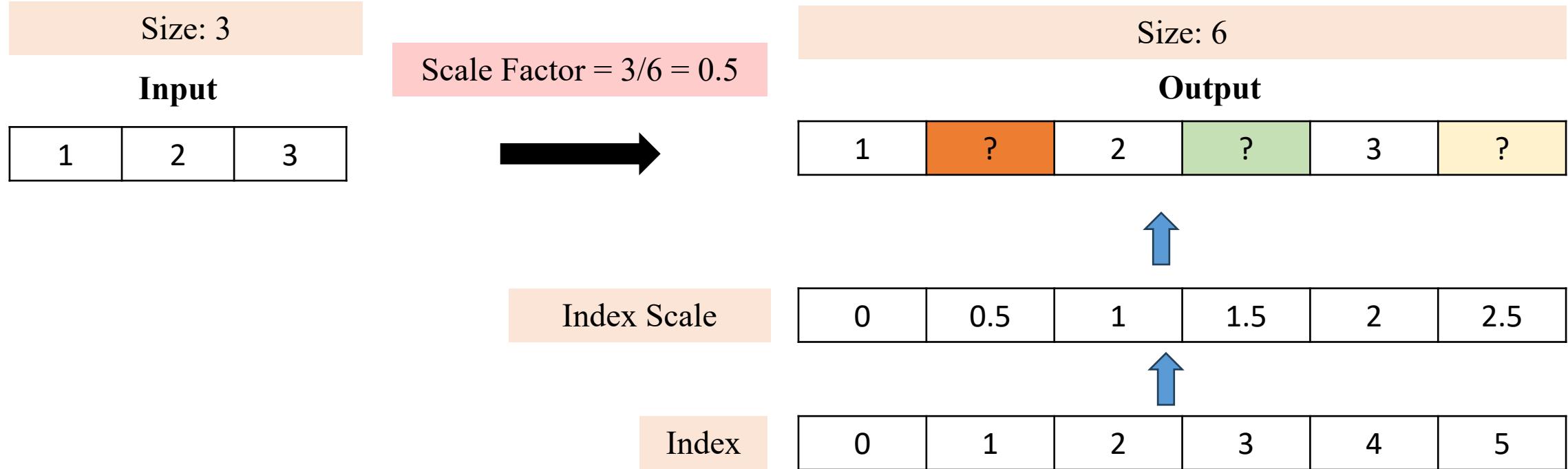
Output :

```
result = [1, 1, 1, 2, 2, 2, 3, 3, 3]
```

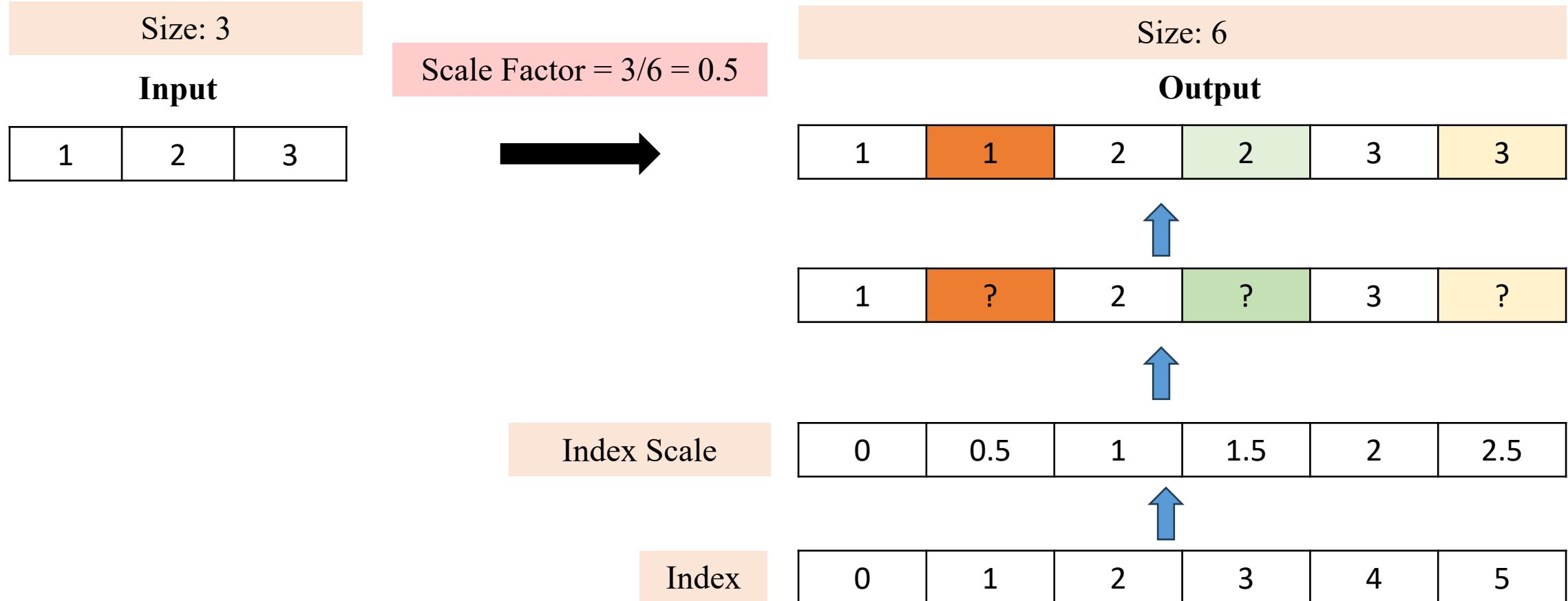
# Problem Analysis



# Problem Analysis



# Problem Analysis



**Input size: 3**

**Step 1:**

7	8	9
---	---	---

**Output size: 6**

--	--	--	--	--	--

**Step 2:**

$$\text{Scaled ratio: } 3/6 = 1/2 = 0.5$$

**Step 3:**

Calculate the index for the Output

Output Index

0	1	2	3	4	5
---	---	---	---	---	---



Scalled Output Index

0	0.5	1	1.5	2	2.5
---	-----	---	-----	---	-----



**Step 4:**

Floor the output index

0	0	1	1	2	2
---	---	---	---	---	---



**Step 4:**

Interpolation

7	7	8	8	9	9
---	---	---	---	---	---



**Input**

7	8	9
---	---	---

0 1 2

Operator	Name	Example
+	Addition	$x + y$
-	Subtraction	$x - y$
*	Multiplication	$x * y$
/	Division	$x / y$
%	Modulus	$x \% y$
**	Exponentiation	$x ** y$
//	Floor division	$x // y$

Floor division is a mathematical operation in Python. This operator will divide the first argument by the second and round the result down to the nearest whole number, making it equivalent to the `math.floor()` function.

# Solution

```
1 # implementation
2 def nni_1d(data, k):
3     # your code here
4
5     return result
6
7 # test case
8 data = [1, 2, 3]
9 k = 2
10 assert nni_1d(data, k) == [1, 1, 2, 2, 3, 3]
11
12 data = [1, 2, 3]
13 k = 3
14 assert nni_1d(data, k) == [1, 1, 1, 2, 2, 2, 3, 3, 3]
15
16 # multiple choice
17 data = [5, 7]
18 k = 3
19 print(nni_1d(data, k))
```

# Problem 4

**Câu hỏi 4** (kỹ thuật nội suy liền kề cho dữ liệu 2 chiều (2D)): Cho trước một danh sách các số tự nhiên (m dòng và n cột) và tham số k. Phát triển chương trình nội suy danh sách đầu vào thành danh sách mới (m dòng và  $n*k$  cột) sử dụng kỹ thuật nội suy liền kề (nearest neighbor interpolation). Bên dưới là ví dụ minh họa các test case đầu vào và đầu ra của chương trình.

Examples

Input:

```
data = [[1, 2, 3],  
        [4, 5, 6],  
        [7, 8, 9]]
```

k = 2

3x3  $\Rightarrow$  3x6

Output:

```
result = [[1, 1, 2, 2, 3, 3],  
          [4, 4, 5, 5, 6, 6],  
          [7, 7, 8, 8, 9, 9]]
```

Input:

```
data = [[1, 2],  
        [4, 5],  
        [7, 8]]
```

k = 2

Output:

```
result = [[1, 1, 2, 2],  
          [4, 4, 5, 5],  
          [7, 7, 8, 8]]
```

# Problem Analysis

Size: 3x3

Input

1	2	3
4	5	6
7	8	9

Scale Factor =  $3/6 = 0.5$

Index Scale



Size: 3x6

Output

0	0.5	1	1.5	2	2.5
---	-----	---	-----	---	-----


# Problem Analysis

Size: 3x3

Input

1	2	3
4	5	6
7	8	9

Scale Factor =  $3/6 = 0.5$

Index Scale

Column Index

Column Index

Size: 3x6

Output

0	0.5	1	1.5	2	2.5
0	0.5	1	1.5	2	2.5
0	0.5	1	1.5	2	2.5
0	0.5	1	1.5	2	2.5

0	0.5	1	1.5	2	2.5
0	0.5	1	1.5	2	2.5
0	0.5	1	1.5	2	2.5

Floor(Index Matrix)

0	0	1	1	2	2
0	0	1	1	2	2
0	0	1	1	2	2

# Problem Analysis

Size: 3x3

Input

1	2	3
4	5	6
7	8	9

Scale Factor =  $3/6 = 0.5$

Index Scale

Column  
Index

Value

1	1	2	2	3	3
4	4	5	5	6	6
7	7	8	8	9	9

Size: 3x6

Output

0	0.5	1	1.5	2	2.5
0	0.5	1	1.5	2	2.5
0	0.5	1	1.5	2	2.5

0	0.5	1	1.5	2	2.5
0	0.5	1	1.5	2	2.5
0	0.5	1	1.5	2	2.5

Floor(Index Matrix)

0	0	1	1	2	2
0	0	1	1	2	2
0	0	1	1	2	2

Final result

# Simpler Approach

# Problem Analysis

Size: 3x3

Input

1	2	3
4	5	6
7	8	9

Scale Factor =  $3/6 = 0.5$

Index Scale



0  
1  
2

0      0.5      1      1.5      2      2.5

(0,0)	(0,0.5)	(0,1)	(0,1.5)	(0,2)	(0,2.5)
(1,0)	(1,0.5)	(1,1)	(1, 1.5)	(1, 2)	(1, 2.5)
(2,0)	(2,0.5)	(2,1)	(2,1.5)	(2,2)	(2,2.5)

Size: 3x6

Output

(0,0)	(0,0)	(0,1)	(0,1)	(0,2)	(0,2)
(1,0)	(1,0)	(1,1)	(1, 1)	(1, 2)	(1, 2)
(2,0)	(2,0)	(2,1)	(2,1)	(2,2)	(2,2)

1	1	2	2	3	3
4	4	5	5	6	6
7	7	8	8	9	9

# Solution

```
1 # implementation
2 def h_nni_2d(data, k):
3     # your code here
4     return result
5
6 # test case
7 data = [[1, 2, 3],
8         [4, 5, 6],
9         [7, 8, 9]]
10 k = 2
11 assert h_nni_2d(data, k) == [[1, 1, 2, 2, 3, 3],
12                               [4, 4, 5, 5, 6, 6],
13                               [7, 7, 8, 8, 9, 9]]
14
15 data = [[1, 2],
16          [4, 5],
17          [7, 8]]
18 k = 2
19 assert h_nni_2d(data, k) == [[1, 1, 2, 2],
20                               [4, 4, 5, 5],
21                               [7, 7, 8, 8]]
22
23 # multiple choice
24 data = [[3],
25          [4]]
26 k = 3
27 print(h_nni_2d(data, k))
```

# Problem 5

**Câu hỏi 5** (kỹ thuật nội suy liền kề cho dữ liệu 2 chiều (2D)): Cho trước một danh sách các số tự nhiên (m dòng và n cột) và tham số k. Phát triển chương trình nội suy dữ liệu cột của danh sách đầu vào thành danh sách mới (m\*k dòng và n cột) sử dụng kỹ thuật nội suy liền kề (nearest neighbor interpolation). Bên dưới là ví dụ minh họa các test case đầu vào và đầu ra của chương trình.

## Examples

### Input :

```
data = [[1, 2, 3],  
        [4, 5, 6],  
        [7, 8, 9]]  
  
k = 2
```

### Output :

```
result = [[1, 2, 3],  
          [1, 2, 3],  
          [4, 5, 6]  
          [4, 5, 6],  
          [7, 8, 9],  
          [7, 8, 9]]  
-----  
T ----- .
```

# Problem Analysis

Size: 3x3

Input

1	2	3
4	5	6
7	8	9

Scale Factor =  $3/6 = 0.5$

Index Scale



Size: 3x6

Output


# Problem Analysis

Size: 3x3

**Input**

1	2	3
4	5	6
7	8	9

Scale Factor =  $3/6 = 0.5$

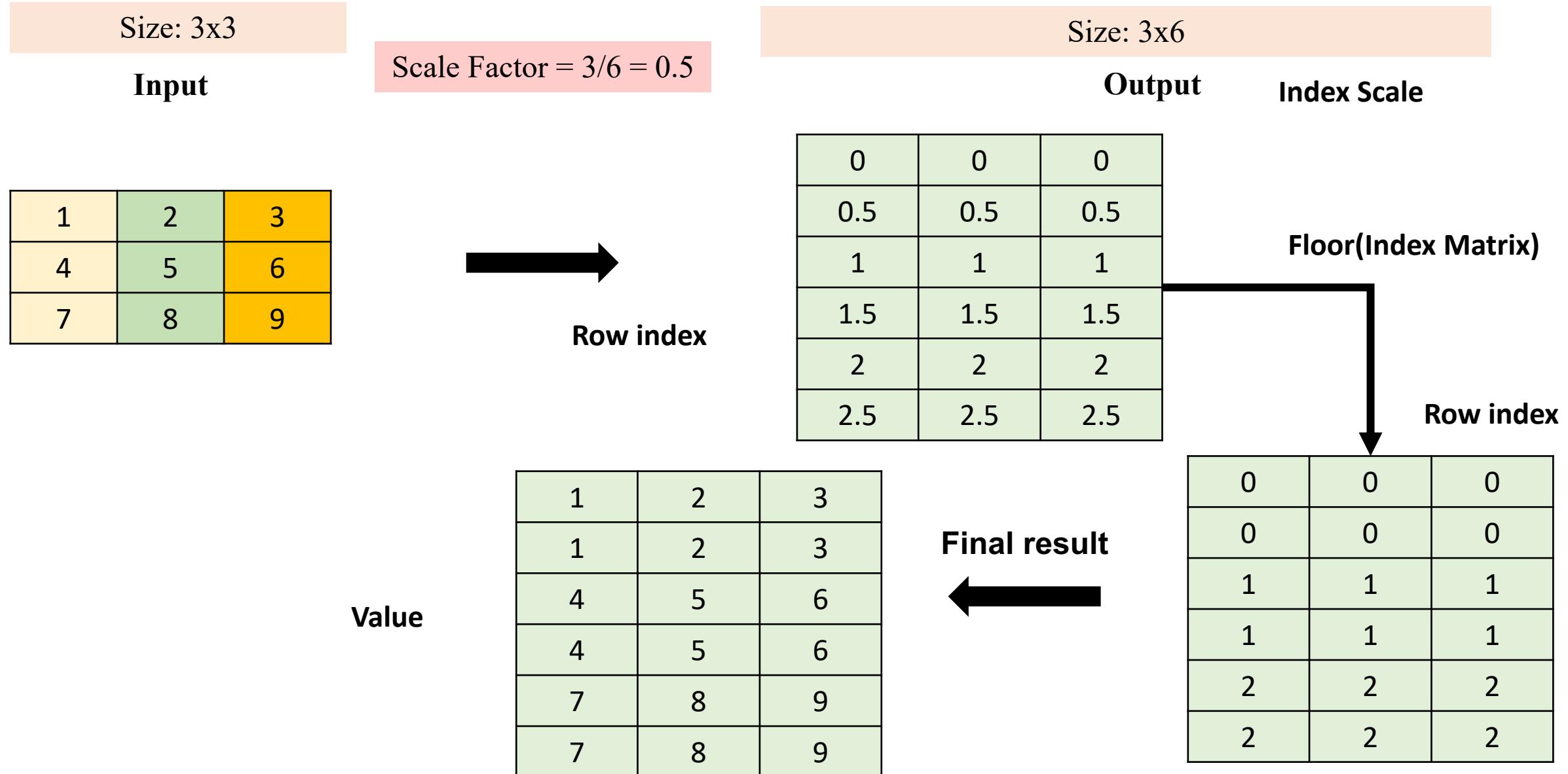
**Row index  
Matrix**

Size: 3x6

**Output**


0
0.5
1
1.5
2
2.5

# Problem Analysis



# Simpler Approach

# Problem Analysis

Size: 3x3

Input

1	2	3
4	5	6
7	8	9

1	2	3
1	2	3
4	5	6
4	5	6
7	8	9
7	8	9

Scale Factor =  $3/6 = 0.5$



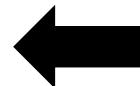
Output

Size: 3x6

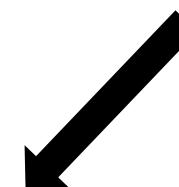
0      1      2      Index

(0,0)	(0,1)	(0,2)
(0.5,0)	(0.5,1)	(0.5,2)
(1,0)	(1,1)	(1,2)
(1.5,0)	(1.5,1)	(1.5,2)
(2,0)	(2,1)	(2,2)
(2.5,0)	(2.5,1)	(2.5,2)

0      0.5      1      1.5      2      2.5



(0,0)	(0,1)	(0,2)
(0,0)	(0,1)	(0,2)
(1,0)	(1,1)	(1,2)
(1,0)	(1,1)	(1,2)
(2,0)	(2,1)	(2,2)
(2,0)	(2,1)	(2,2)



# Solution

```
# implementation
def v_nni_2d(data, k):
    # your code here

    return result

# test case
data = [[1, 2, 3],
         [4, 5, 6],
         [7, 8, 9]]
k = 2
assert v_nni_2d(data, k) == [[1, 2, 3],
                               [1, 2, 3],
                               [4, 5, 6],
                               [4, 5, 6],
                               [7, 8, 9],
                               [7, 8, 9]]
```

# Problem 6

**Câu hỏi 6** (kỹ thuật nội suy liền kề cho dữ liệu 2 chiều (2D)): Cho trước một danh sách các số tự nhiên (m dòng và n cột) và tham số k. Phát triển chương trình nội suy dữ liệu danh sách đầu vào thành danh sách mới ( $m^*k$  dòng và  $n^*k$  cột) sử dụng kỹ thuật nội suy liền kề (nearest neighbor interpolation). Bên dưới là ví dụ minh họa các test case đầu vào và đầu ra của chương trình.

Examples

Input:

```
data = [[1, 2, 3],  
        [4, 5, 6],  
        [7, 8, 9]]
```

k = 2

Output:

```
result = [[1, 1, 2, 2, 3, 3],  
          [1, 1, 2, 2, 3, 3],  
          [4, 4, 5, 5, 6, 6],  
          [4, 4, 5, 5, 6, 6],  
          [7, 7, 8, 8, 9, 9],  
          [7, 7, 8, 8, 9, 9]]
```

-----

Input:

```
data = [[1, 2],  
        [4, 5],  
        [7, 8]]
```

k = 2

(3x)

6x

# Problem Analysis

Scale Factor =  $3/6 = 0.5$

Size: 3x3

Input

1	2	3
4	5	6
7	8	9

Size: 6x6

Output

Column index

column interpolation


# Problem Analysis

Scale Factor =  $3/6 = 0.5$

Size: 3x3

Input

$\begin{smallmatrix} 5 \\ 1 \end{smallmatrix}$

0	1	2
1	2	3
4	5	6
7	8	9

Column index

column interpolation

Size: 6x6

Output

0	0.5	1	1.5	2	2.5
0	0.5	1	1.5	2	2.5
0	0.5	1	1.5	2	2.5

Nearest index

0	0	1	1	2	2
0	0	1	1	2	2
0	0	1	1	2	2

Result

# Problem Analysis

Scale Factor =  $3/6 = 0.5$

Size: 3x3

Input

1	2	3
4	5	6
7	8	9

column interpolation



Size: 6x6

Output

1	1	2	2	3	3
4	4	5	5	6	6
7	7	8	8	9	9

row interpolation

Index Scale

0  
0.5  
1  
1.5  
2  
2.5


# Problem Analysis

Scale Factor =  $3/6 = 0.5$

Size: 3x3

Input

1	2	3
4	5	6
7	8	9

column  
interpolation

0  
1  
2

1	1	2	2	3	3
4	4	5	5	6	6
7	7	8	8	9	9

Size: 6x6

Output

0	0	0	0	0	0
0.5	0.5	0.5	0.5	0.5	0.5
1	1	1	1	1	1
1.5	1.5	1.5	1.5	1.5	1.5
2	2	2	2	2	2
2.5	2.5	2.5	2.5	2.5	2.5

Row  
Interpolation  
By Index

1	1	2	2	3	3
1	1	2	2	3	3
4	4	5	5	6	6
4	4	5	5	6	6
7	7	8	8	9	9
7	7	8	8	9	9

0	0	0	0	0	0
0	0	0	0	0	0
1	1	1	1	1	1
1	1	1	1	1	1
2	2	2	2	2	2
2	2	2	2	2	2

Nearest index

# Simpler Approach

# Problem Analysis

Scale Factor =  $3/6 = 0.5$

Size: 6x6

Size: 3x3

Input

1	2	3
4	5	6
7	8	9

Column index

(0,0)	(0,0)	(0,1)	(0,1)	(0,2)	(0,2)
(0,0)	(0,0)	(0,1)	(0,1)	(0,2)	(0,2)
(1,0)	(1,0)	(1,1)	(1,1)	(1,2)	(1,2)
(1,0)	(1,0)	(1,1)	(1,1)	(1,2)	(1,2)
(2,0)	(2,0)	(2,1)	(2,1)	(2,2)	(2,2)
(2,0)	(2,0)	(2,1)	(2,1)	(2,2)	(2,2)

Output

0	0.5	1	1.5	2	2.5
(0,0)	(0,0.5)	(0,1)	(0,1.5)	(0,2)	(0,2.5)
(0.5,0)	(0.5,0.5)	(0.5,1)	(0.5,1.5)	(0.5,2)	(0.5,2.5)
(1,0)	(1,0.5)	(1,1)	(1,1.5)	(1,2)	(1,2.5)
(1.5,0)	(1.5,0.5)	(1.5,1)	(1.5,1.5)	(1.5,2)	(1.5,2.5)
(2,0)	(2,0.5)	(2,1)	(2,1.5)	(2,2)	(2,2.5)
(2.5,0)	(2.5,0.5)	(2.5,1)	(2.5,1.5)	(2.5,2)	(2.5,2.5)

0  
0.5  
1  
1.5  
2  
2.5

Row index

Nearest index

# Problem Analysis

Scale Factor =  $3/6 = 0.5$

Size: 6x6

Size: 3x3

Input

1	2	3
4	5	6
7	8	9

Column index

1	1	2	2	3	3
1	1	2	2	3	3
4	4	5	5	6	6
4	4	5	5	6	6
7	7	8	8	9	9
7	7	8	8	9	9

(0,0)	(0,0.5)	(0,1)	(0,1.5)	(0,2)	(0,2.5)
(0.5,0)	(0.5,0.5)	(0.5,1)	(0.5,1.5)	(0.5,2)	(0.5,2.5)
(1,0)	(1,0.5)	(1,1)	(1,1.5)	(1,2)	(1,2.5)
(1.5,0)	(1.5,0.5)	(1.5,1)	(1.5,1.5)	(1.5,2)	(1.5,2.5)
(2,0)	(2,0.5)	(2,1)	(2,1.5)	(2,2)	(2,2.5)
(2.5,0)	(2.5,0.5)	(2.5,1)	(2.5,1.5)	(2.5,2)	(2.5,2.5)

0

0.5

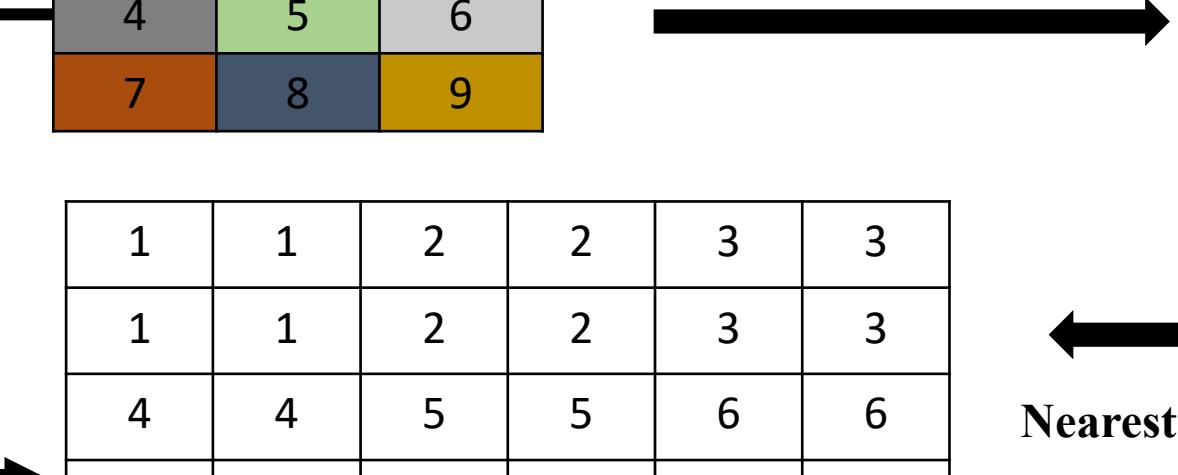
1

1.5

2

2.5

Row index



# Solution

```
1 # implementation
2 def nni_2d(data, k):
3     # your code here
4
5     return result
6
7 # test case
8 data = [[1, 2, 3],
9         [4, 5, 6],
10        [7, 8, 9]]
11 k = 2
12 assert nni_2d(data, k) == [[1, 1, 2, 2, 3, 3],
13                             [1, 1, 2, 2, 3, 3],
14                             [4, 4, 5, 5, 6, 6],
15                             [4, 4, 5, 5, 6, 6],
16                             [7, 7, 8, 8, 9, 9],
17                             [7, 7, 8, 8, 9, 9]]
18
19 data = [[1, 2],
20         [4, 5],
21         [7, 8]]
22 k = 2
23 assert nni_2d(data, k) == [[1, 1, 2, 2],
24                             [1, 1, 2, 2],
25                             [4, 4, 5, 5],
26                             [4, 4, 5, 5],
27                             [7, 7, 8, 8],
28                             [7, 7, 8, 8]]
```

# Problem 7

**Câu hỏi 7** (Kỹ thuật nội suy linear cho dữ liệu 1 chiều (1D)): Cho trước danh sách số thực có chiều dài n và tham số k. Phát triển chương trình nội suy danh sách đầu vào thành danh sách mới có độ dài là  $n*k$  sử dụng kỹ thuật nội suy linear. Lưu ý rằng, để đơn giản, bạn có thể bỏ qua phần tử cuối cùng ( $k-1$ ). Bên dưới là ví dụ minh họa các test case đầu vào và đầu ra của chương trình.

## Examples

**Input :**

```
data = [1, 2, 3]  
k = 2
```

**Output :**

```
result = [1.0, 1.5, 2.0, 2.5, 3.0, x]
```

---

**Input :**

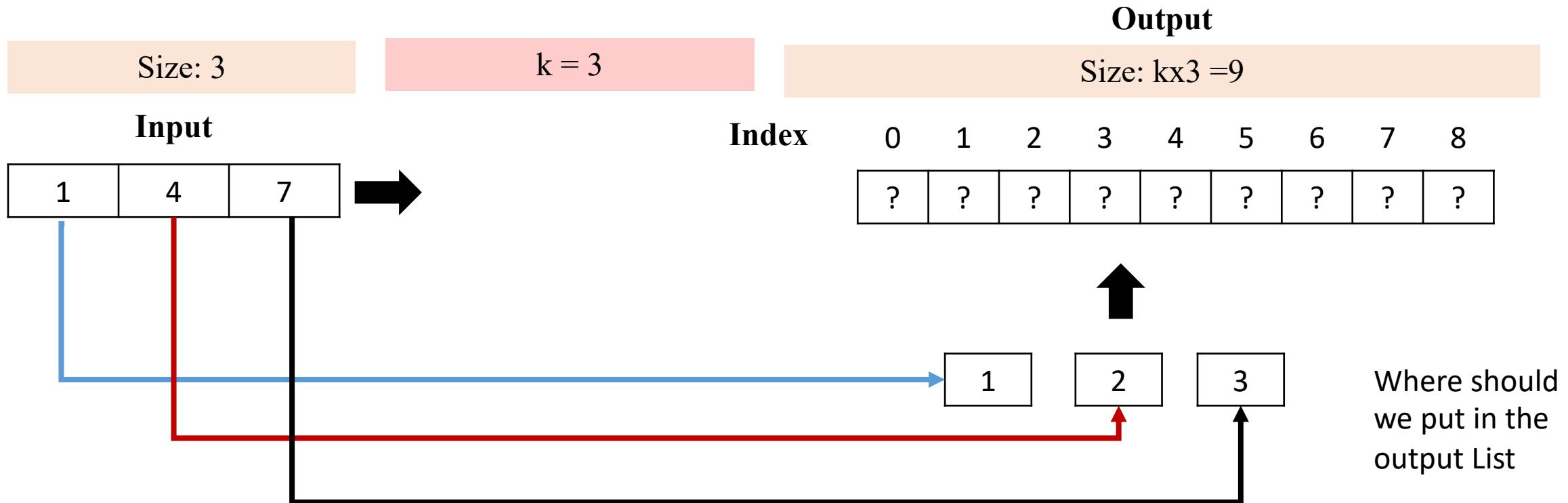
```
data = [1, 4, 7]  
k = 3
```

**Output :**

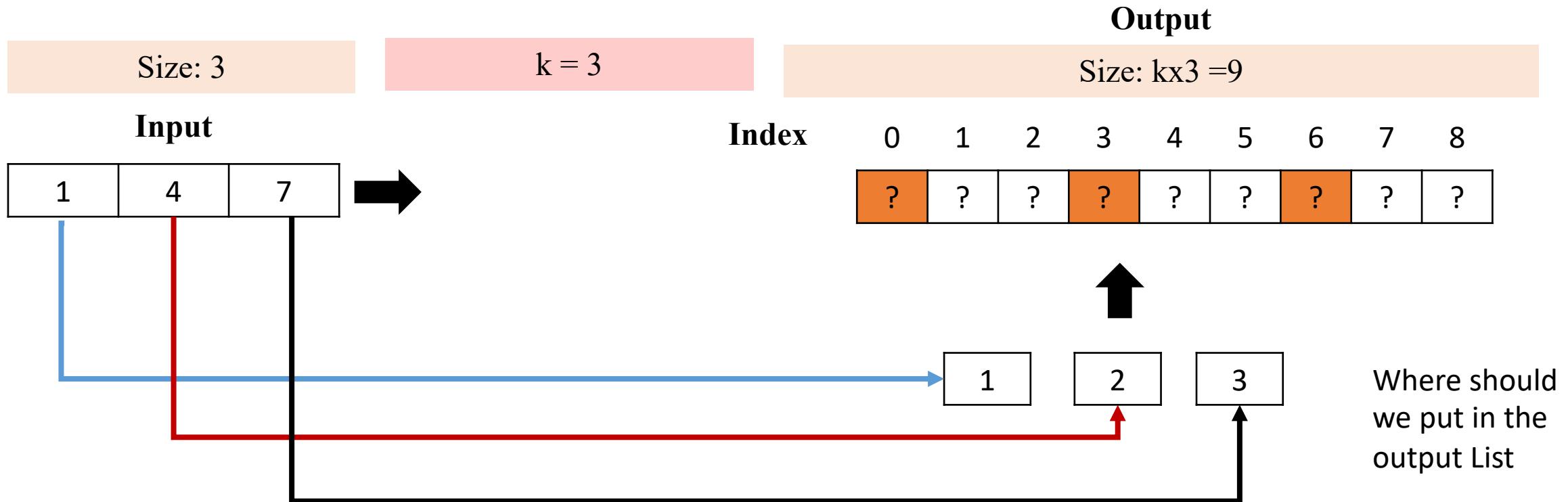
```
result = [1, 2.0, 3.0, 4, 5.0, 6.0, 7, x, x]
```

Here , x is any number .

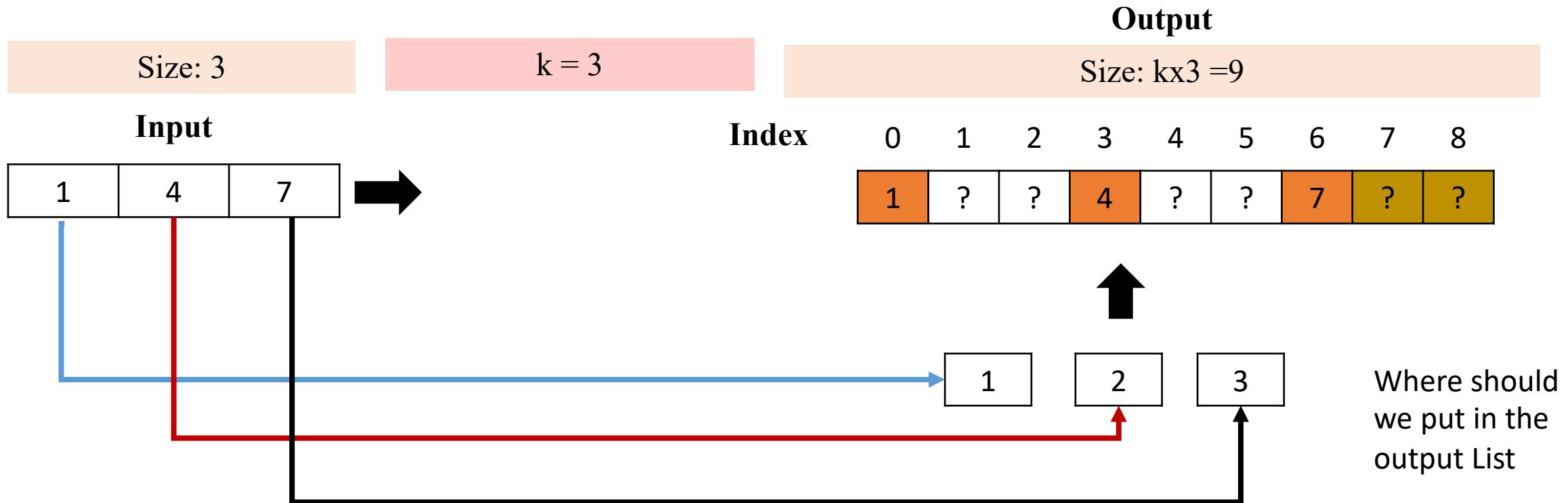
# Analyze Problem



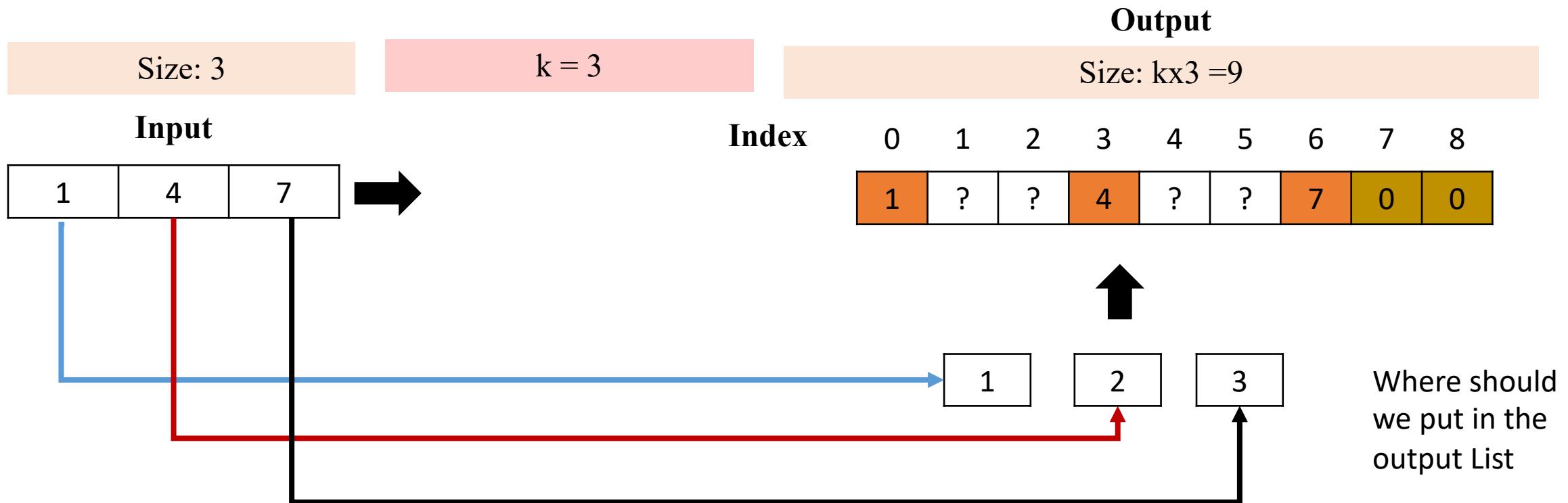
# Analyze Problem



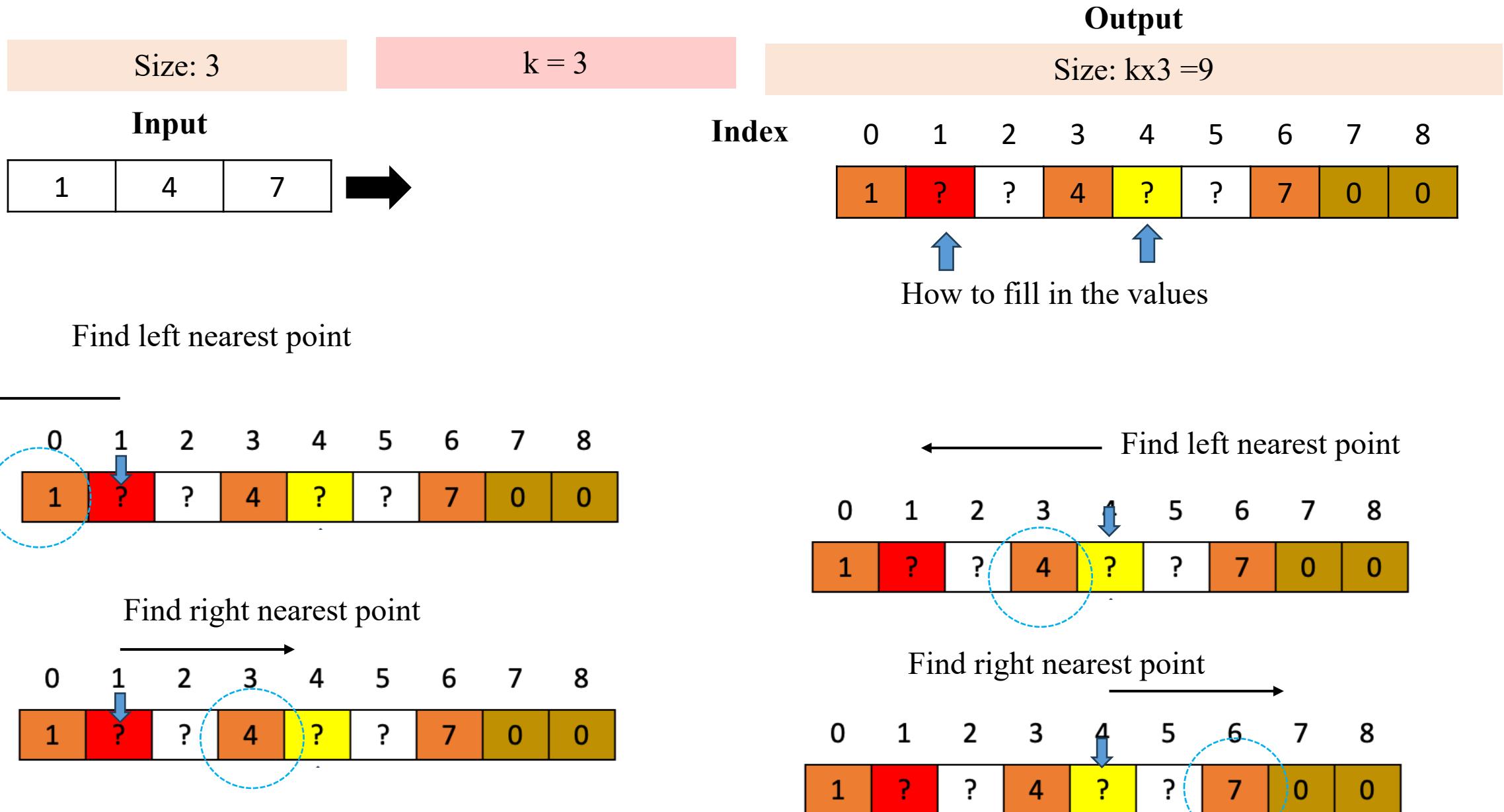
# Analyze Problem



# Analyze Problem

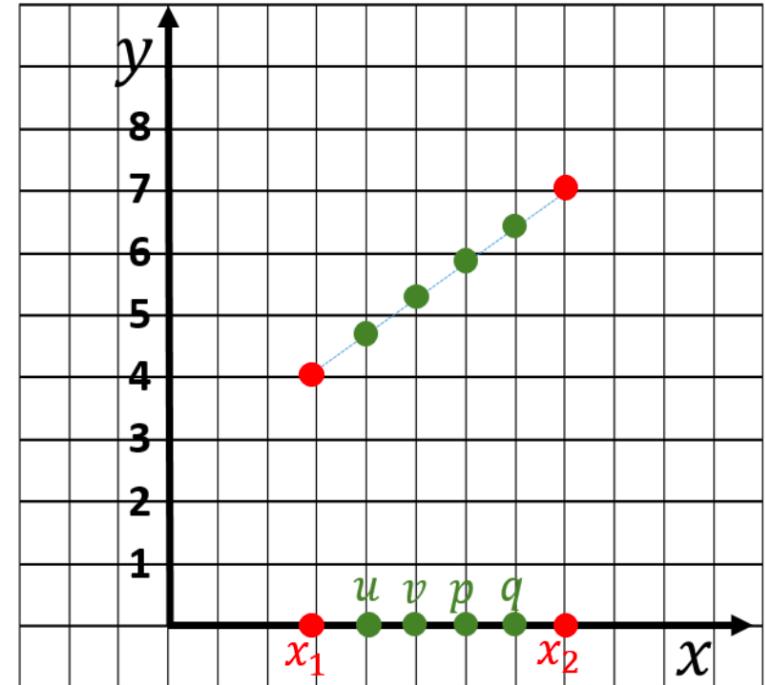


# Analyze Problem



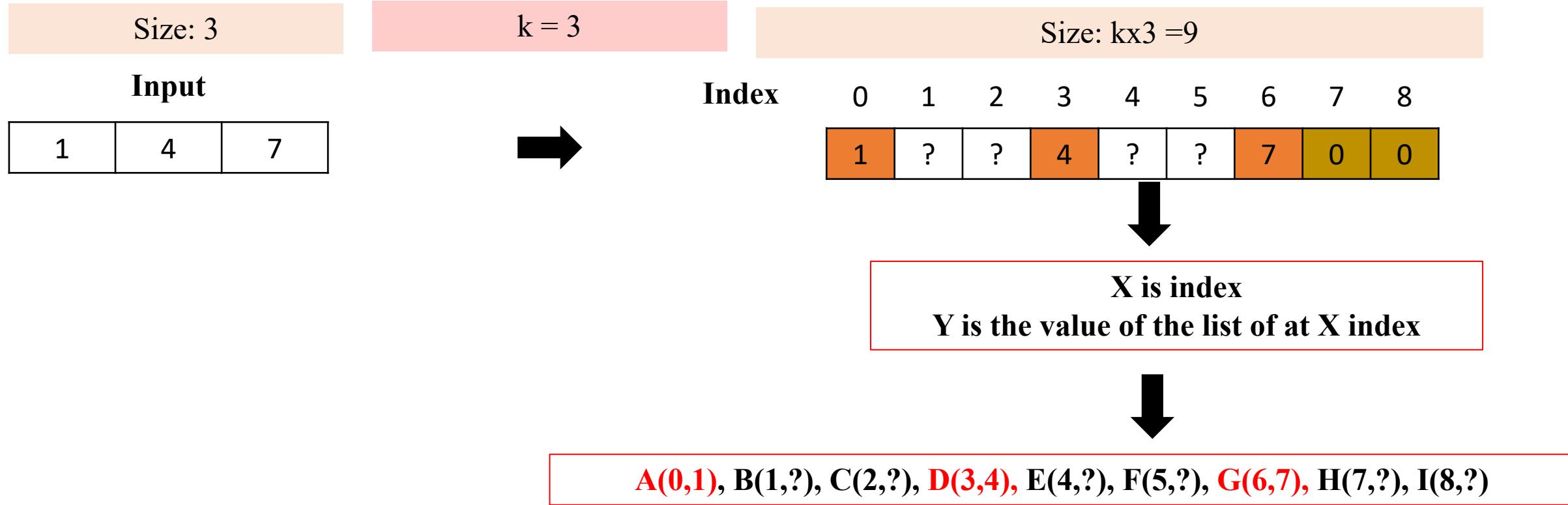
# Analyze Problem

```
def linear_interpolation(p1, p2, x):
    x1, y1 = p1
    x2, y2 = p2
    if x1 != x2:
        y = ((x2-x)/(x2-x1))*y1 + ((x-x1)/(x2-x1))*y2
    else:
        y = y1
    return y
```

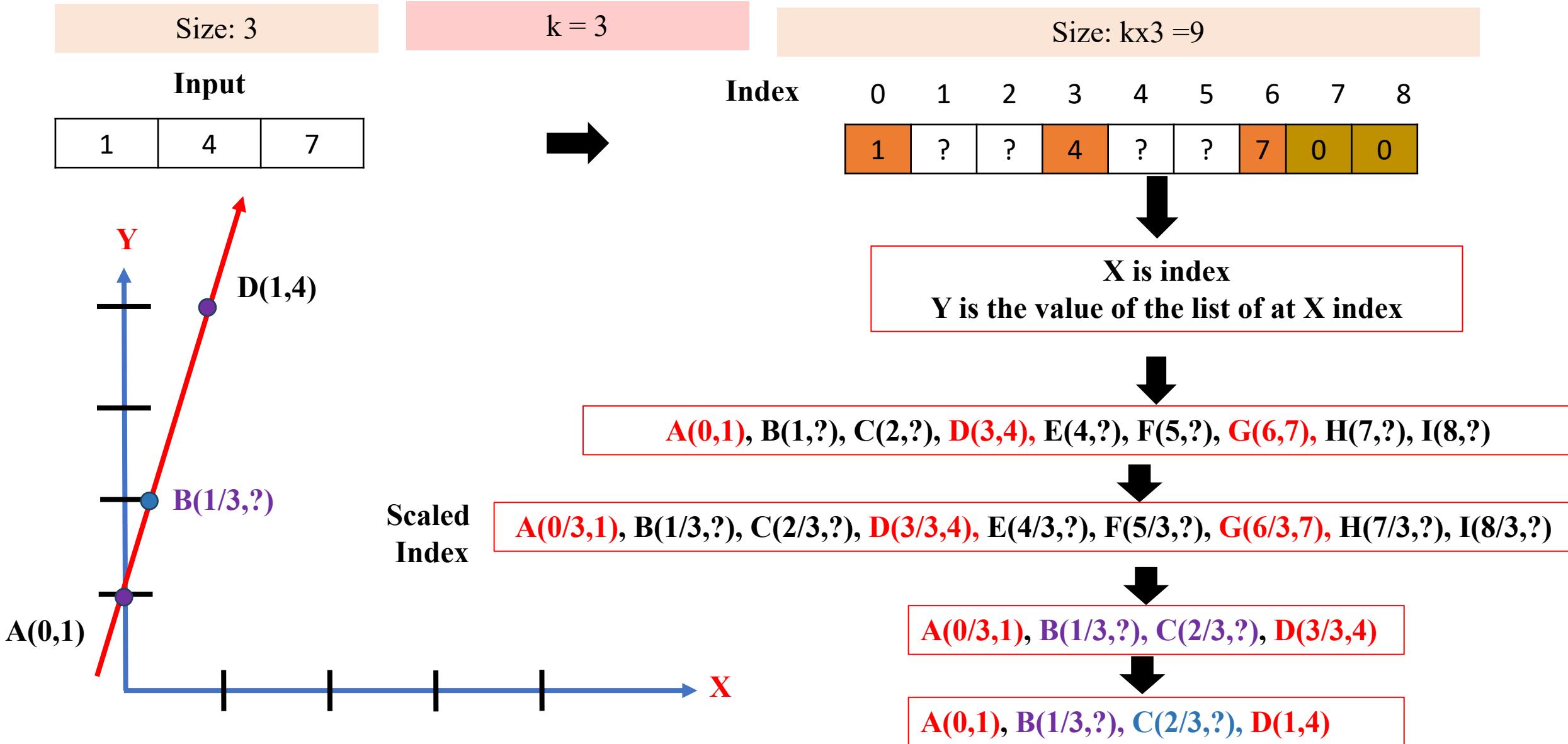


Linear interpolation

# Analyze Problem



# Analyze Problem



# Analyze Problem

Size: 3

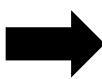
$k = 3$

Output

Size:  $k \times 3 = 9$

Input

1	4	7
---	---	---



Index

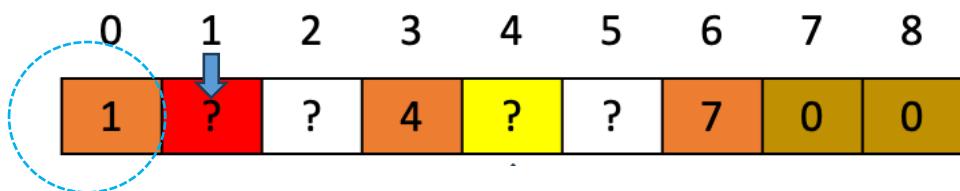
0 1 2 3 4 5 6 7 8

1	?	?	4	?	?	7	0	0
---	---	---	---	---	---	---	---	---

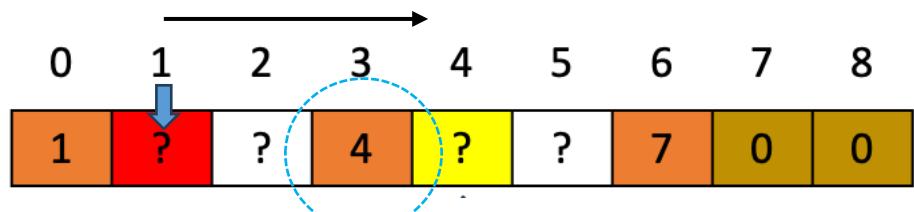


How to fill in the values

Find left nearest point



Find right nearest point



```
def li_1d(data, k):
    n = len(data)
    result_len = n*k
    result = [0]*result_len

    for i in range(result_len-k+1):

        if i%k == 0:
            result[i] = data[i//k]
        else:
            begin = math.floor(i/k)
            end = math.ceil(i/k)
            p1 = (begin, data[begin])
            p2 = (end, data[end])
            y = linear_interpolation(p1=p1, p2=p2, x=i//k)
            result[i] = y

    return result
```

# Solution

```
1 # implementation
2 import math
3
4 def linear_interpolation(p1, p2, x):
5     x1, y1 = p1
6     x2, y2 = p2
7     if x1 != x2:
8         y = ((x2-x)/(x2-x1))*y1 + ((x-x1)/(x2-x1))*y2
9     else:
10        y = y1
11    return y
12
13 def li_1d(data, k):
14     # your code here
15
16     return result
17
18 # test case
19 data = [1, 2, 3]
20 k = 2
21 assert li_1d(data, k)[-1] == [1, 1.5, 2, 2.5, 3, 0][-1]
```

# Problem 8

**Câu hỏi 8** (optional): Cho trước ảnh đầu vào ([link1](#) hoặc [link2](#)) có kích thước dài 400 và rộng 400 như bên dưới:



Bạn có thể sử dụng thư viện opencv để đọc ảnh đầu vào và chuyển dữ liệu ảnh đầu vào thành danh sách 2 chiều (400 dòng và 400 cột).

```
1 import cv2  
2  
3 image = cv2.imread('tree.jpg', 0).tolist()  
4  
5 height = len(image)  
6 width = len(image[0])
```

Ví dụ, *image* có kiểu dữ liệu là 2 chiều (a list of lists). Phát triển chương trình để tăng kích thước ảnh đầu (400x400) và thành ảnh mới (1200 x1200) sử dụng kỹ thuật nội suy nearest-neigbor hoặc linear.

**400x400**



Nearest-  
neighbor  
Interpolation



**1200 x 1200**

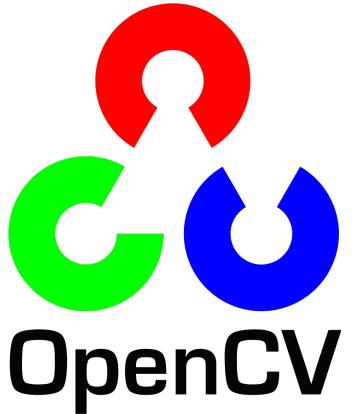


# Analyze Problem

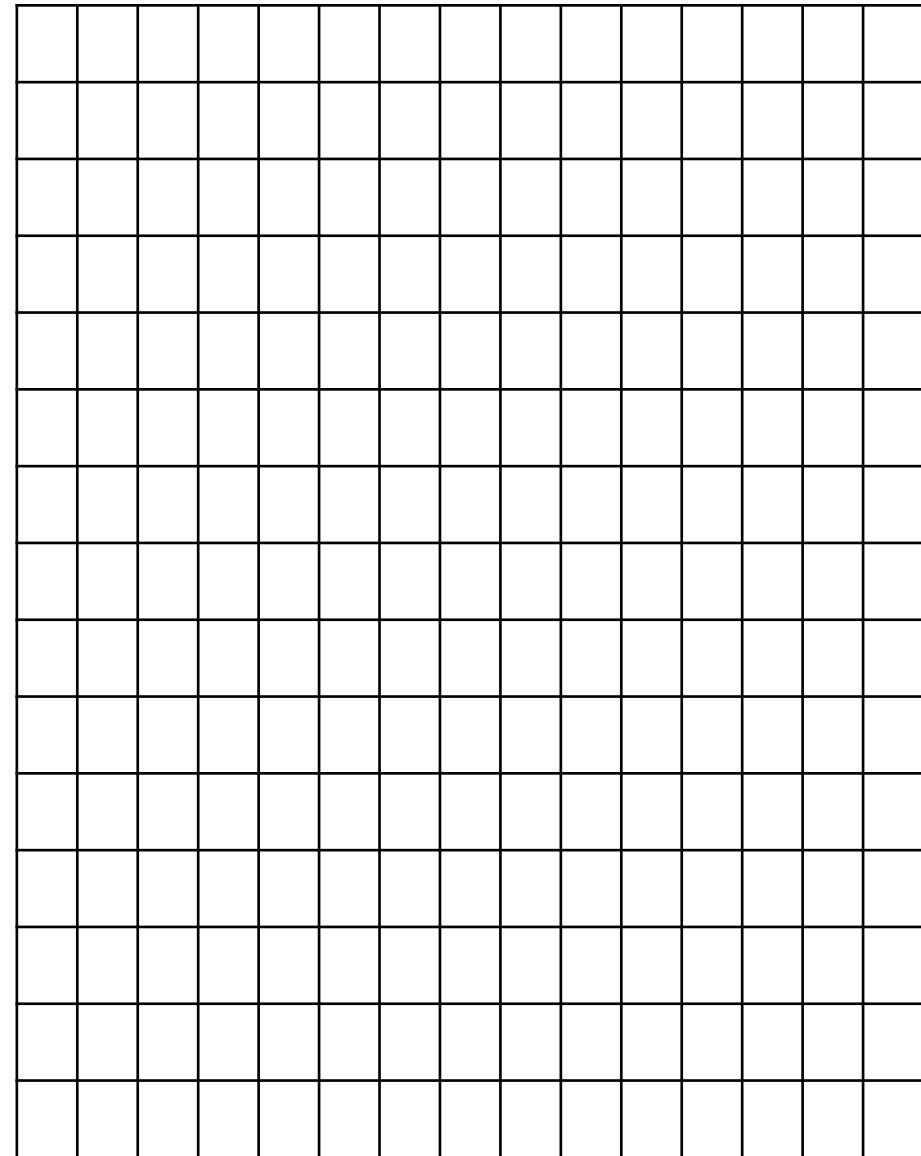
400x400



Read by



To List



# Solution

```
import cv2
import numpy as np
import math

def resize_nni(source, source_h, source_w, target_h, target_w):
    new_data = [[0]*target_w for _ in range(target_h)]

    #Calculate horizontal and vertical scaling factor
    w_scale_factor = source_w/target_w
    h_scale_factor = source_h/target_h

    #your code here

    return new_data

image = cv2.imread('tree.jpg', 0).tolist()

height = len(image)
width = len(image[0])

new_image = resize_nni(image, height, width, height*3, width*3)
cv2.imwrite('tree_2x.jpg', np.array(new_image))
```

# Bilinear interpolation

## Mapping (scale is an integer)

(3, 3)

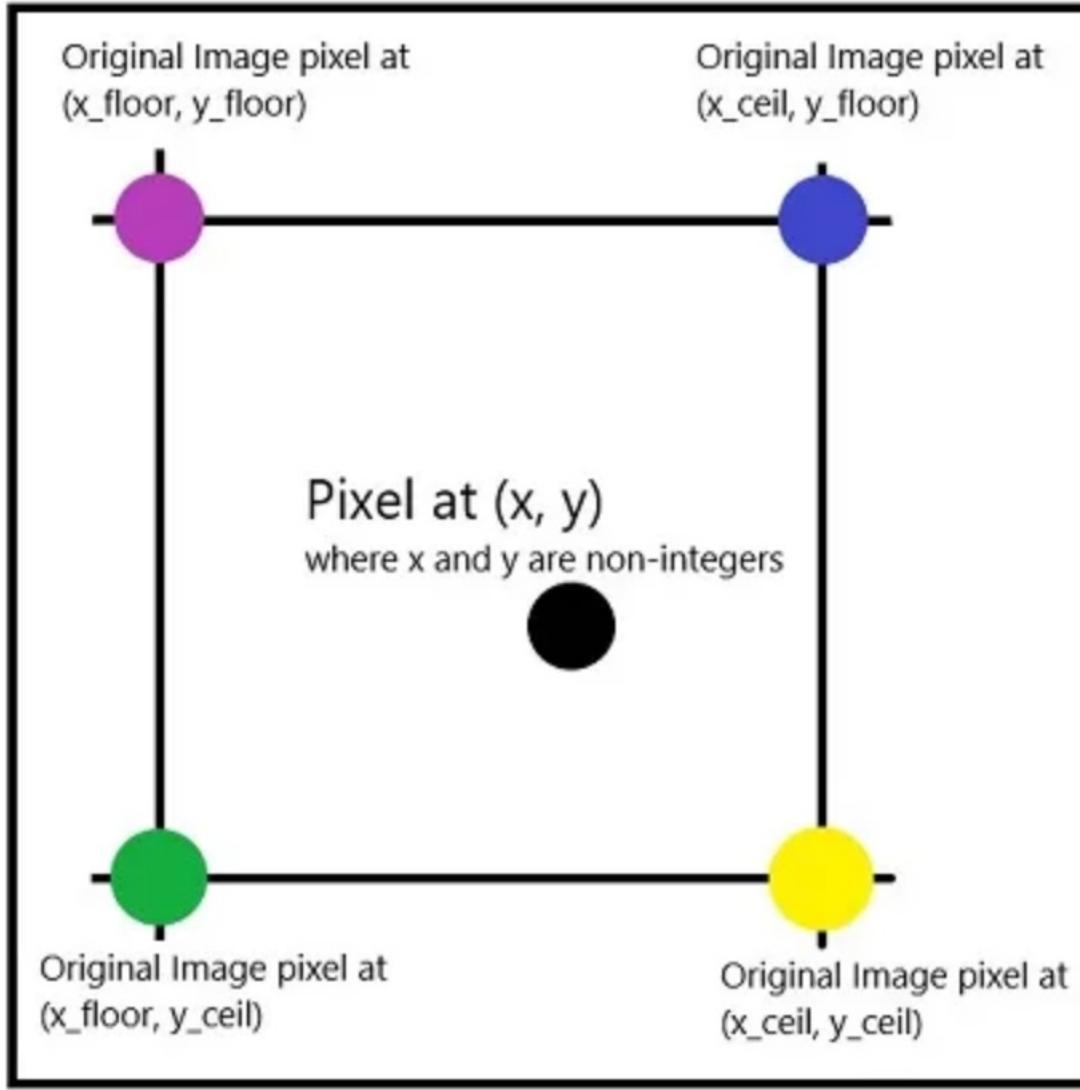
0	2	4
1	3	5
0	2	4

Copy and  
interpolation

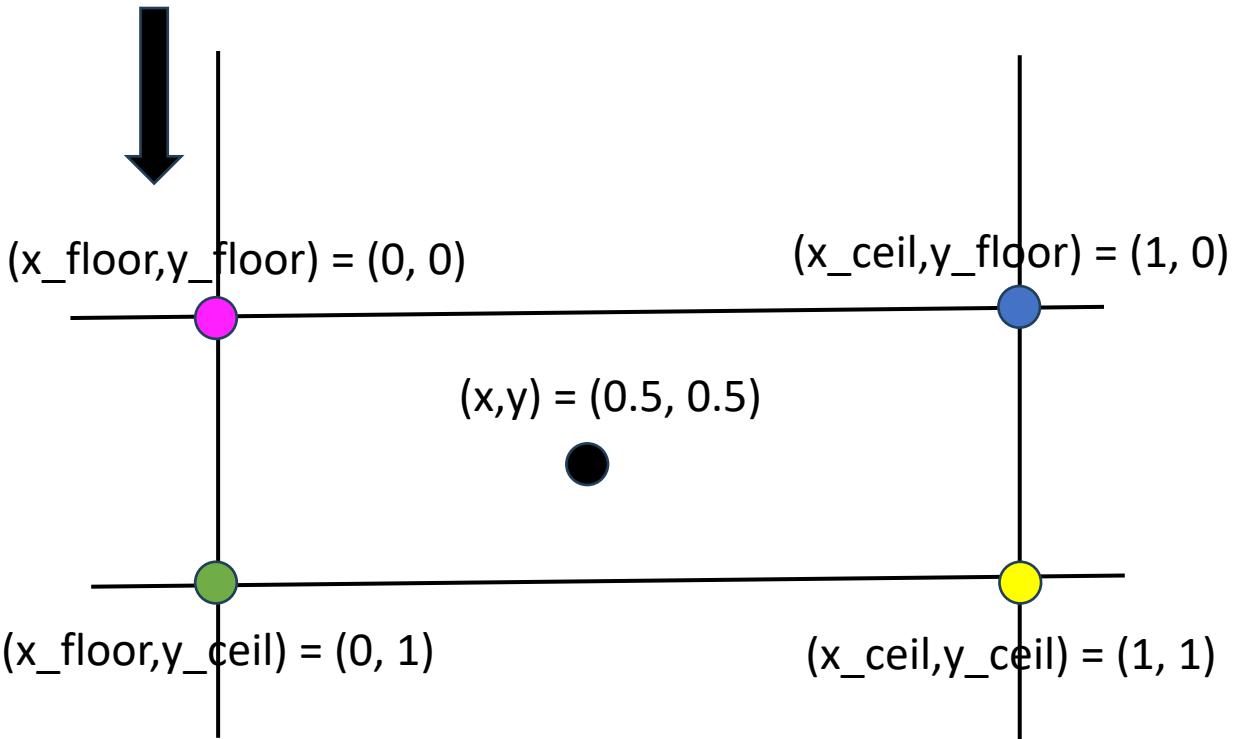
Scaled index	0	1/4	2/4	3/4	4/4	5/4	6/4	7/4	8/4	9/4	10/4	11/4
0	0				2				4			
1/4												
2/4												
3/4												
4/4	1				3				5			
5/4												
6/4												
7/4												
8/4	0				2				4			
9/4												
10/4												
11/4												



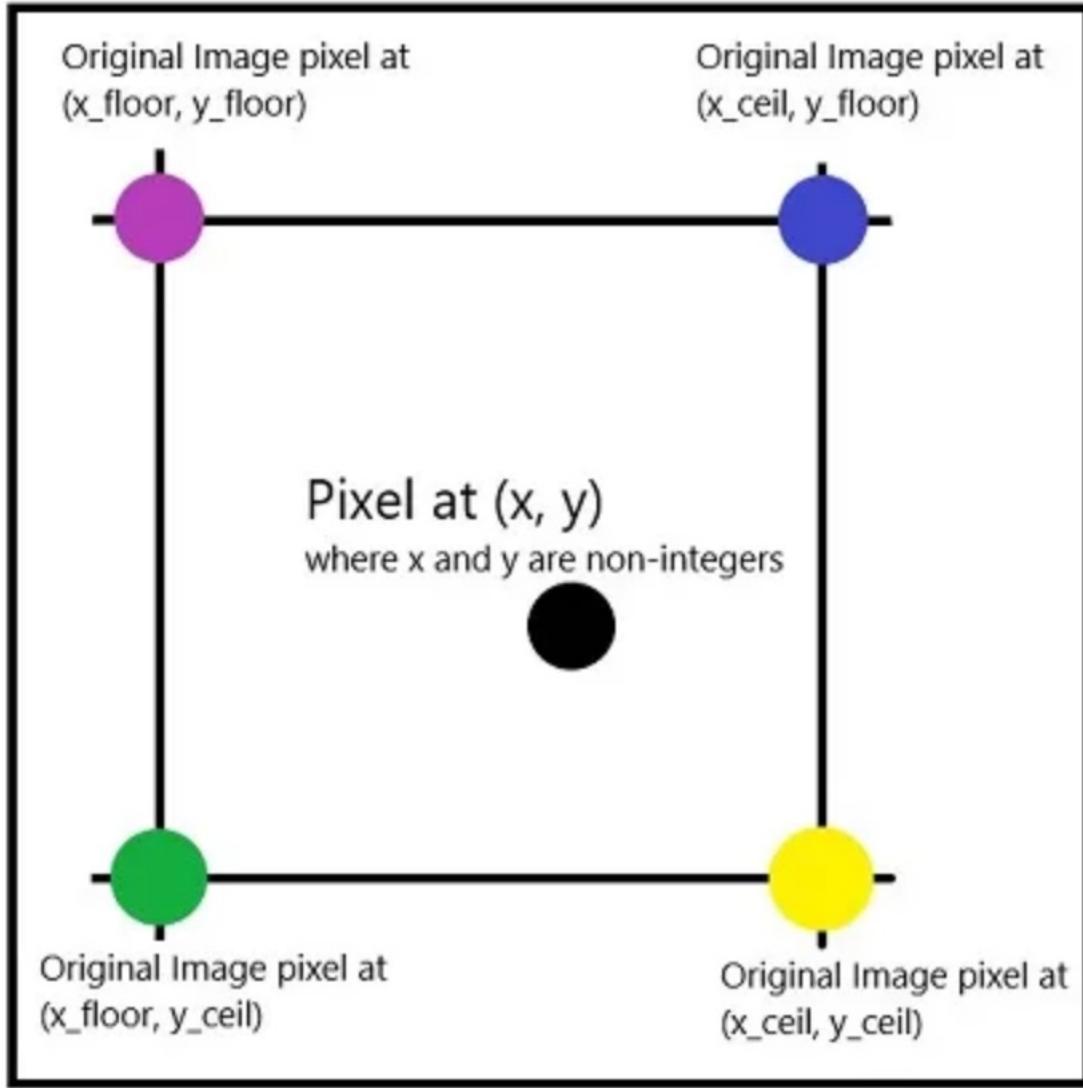
# Bi-linear Interpolation for Image Resize



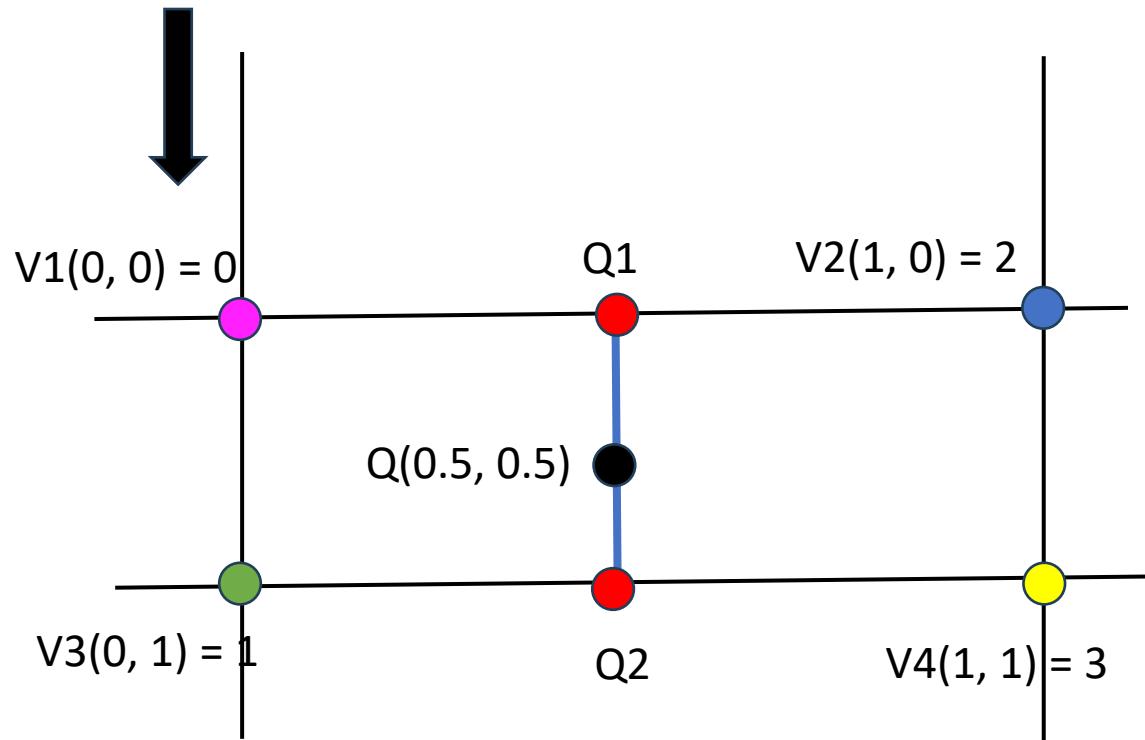
Scaled index	0	$1/4$	$2/4$	$3/4$	$4/4$
0	0				2
$1/4$					
$2/4$					
$3/4$					
$4/4$	1				3



# Bi-linear Interpolation for Image Resize



Scaled index	0	1/4	2/4	3/4	4/4
0	0				2
1/4					
2/4					
3/4					
4/4	1				3



# Solution

```
def bl_resize_normalcase(original_img, old_h, old_w, new_h, new_w):
    resized = np.zeros((new_h, new_w))
    #Calculate horizontal and vertical scaling factor
    w_scale_factor = (old_w ) / (new_w ) if new_h != 0 else 0
    h_scale_factor = (old_h ) / (new_h ) if new_w != 0 else 0
    for i in range(new_h):
        for j in range(new_w):
            #map the coordinates back to the original image
            x = i * h_scale_factor
            y = j * w_scale_factor
            #calculate the coordinate values for 4 surrounding pixels.
            x_floor = math.floor(x)
            x_ceil = min( old_h - 1, math.ceil(x))
            y_floor = math.floor(y)
            y_ceil = min(old_w - 1, math.ceil(y))

            v1 = original_img[x_floor][y_floor]
            v2 = original_img[x_ceil][y_floor]
            v3 = original_img[x_floor][y_ceil]
            v4 = original_img[x_ceil] [y_ceil]

            q1 = v1 * (x_ceil - x) + v2 * (x - x_floor)
            q2 = v3 * (x_ceil - x) + v4 * (x - x_floor)
            q = q1 * (y_ceil - y) + q2 * (y - y_floor)

            resized[i,j] = q
    return resized.astype(np.uint8)
```



# Solution

```
def bl_resize_normalcase(original_img, old_h, old_w, new_h, new_w):
    resized = np.zeros((new_h, new_w))
    #Calculate horizontal and vertical scaling factor
    w_scale_factor = (old_w ) / (new_w ) if new_h != 0 else 0
    h_scale_factor = (old_h ) / (new_h ) if new_w != 0 else 0
    for i in range(new_h):
        for j in range(new_w):
            #map the coordinates back to the original image
            x = i * h_scale_factor
            y = j * w_scale_factor
            #calculate the coordinate values for 4 surrounding pixels.
            x_floor = math.floor(x)
            x_ceil = min( old_h - 1, math.ceil(x))
            y_floor = math.floor(y)
            y_ceil = min(old_w - 1, math.ceil(y))

            v1 = original_img[x_floor][y_floor]
            v2 = original_img[x_ceil][y_floor]
            v3 = original_img[x_floor][y_ceil]
            v4 = original_img[x_ceil] [y_ceil]

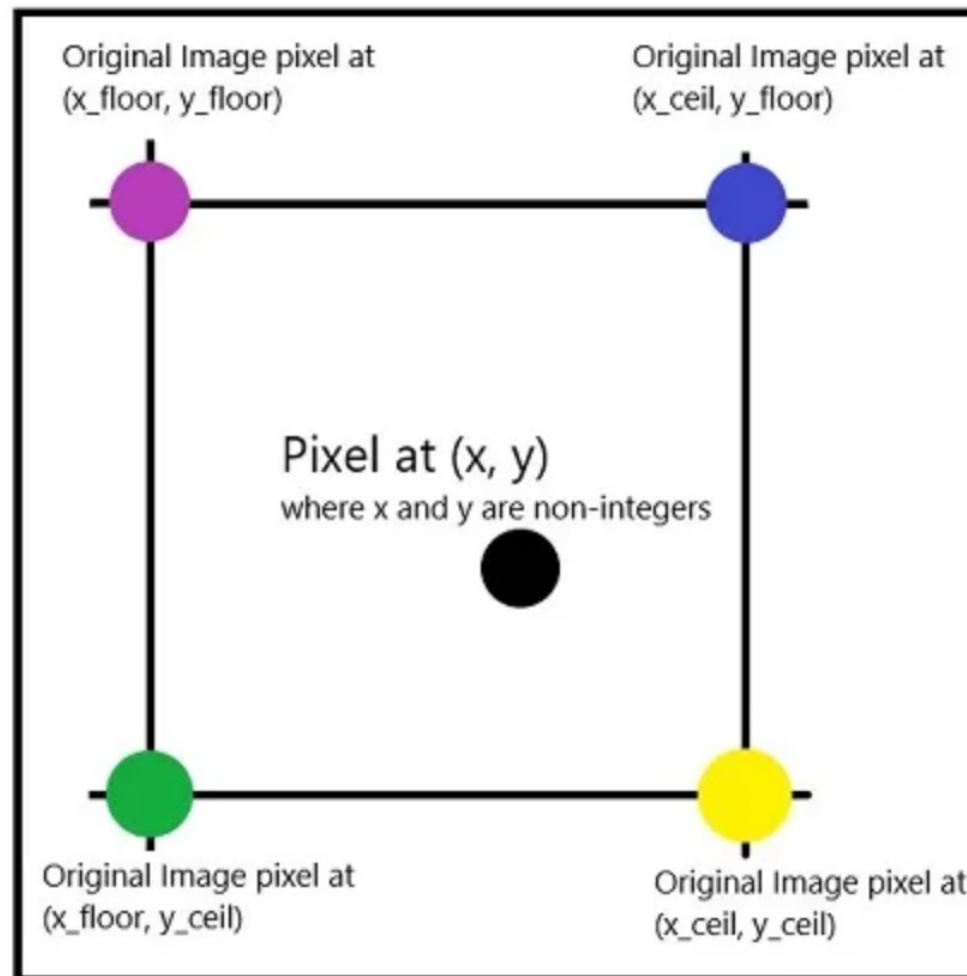
            q1 = v1 * (x_ceil - x) + v2 * (x - x_floor)
            q2 = v3 * (x_ceil - x) + v4 * (x - x_floor)
            q = q1 * (y_ceil - y) + q2 * (y - y_floor)

            resized[i,j] = q
    return resized.astype(np.uint8)
```

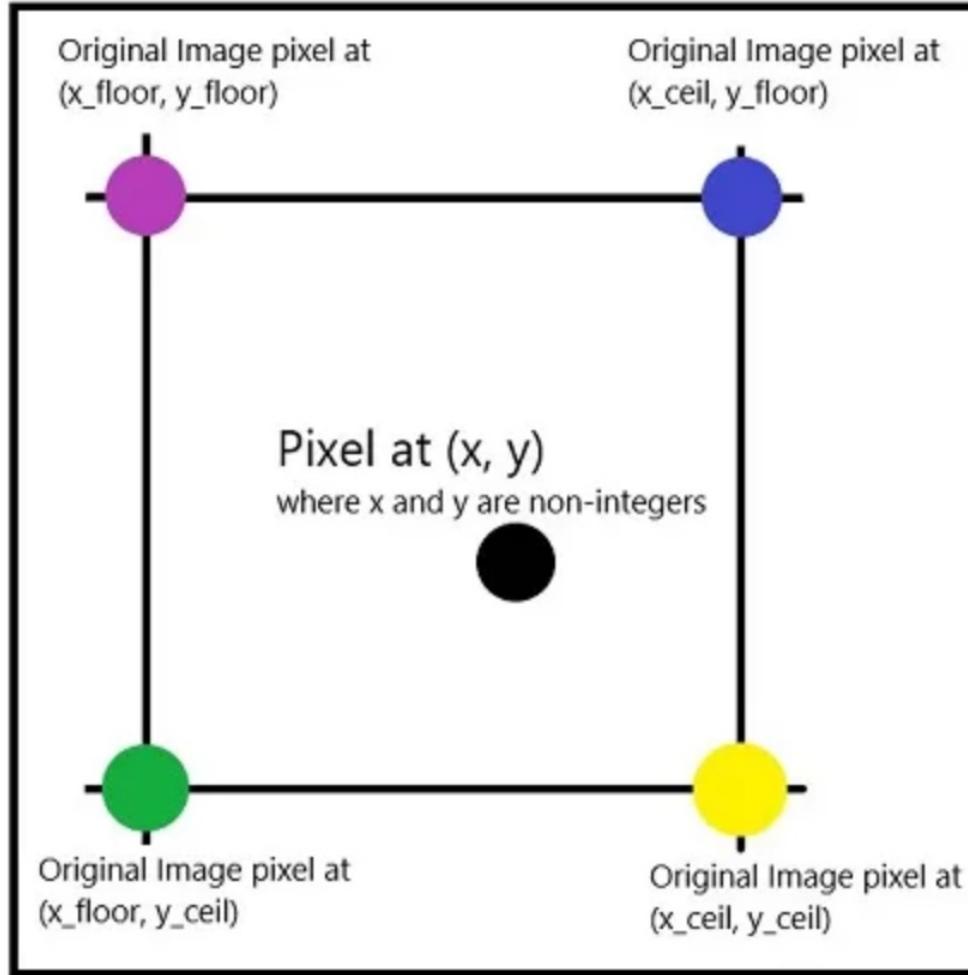


**What's wrong?**

This happens when either  $x$  or  $y$  is an integer resulting in  $q=0$

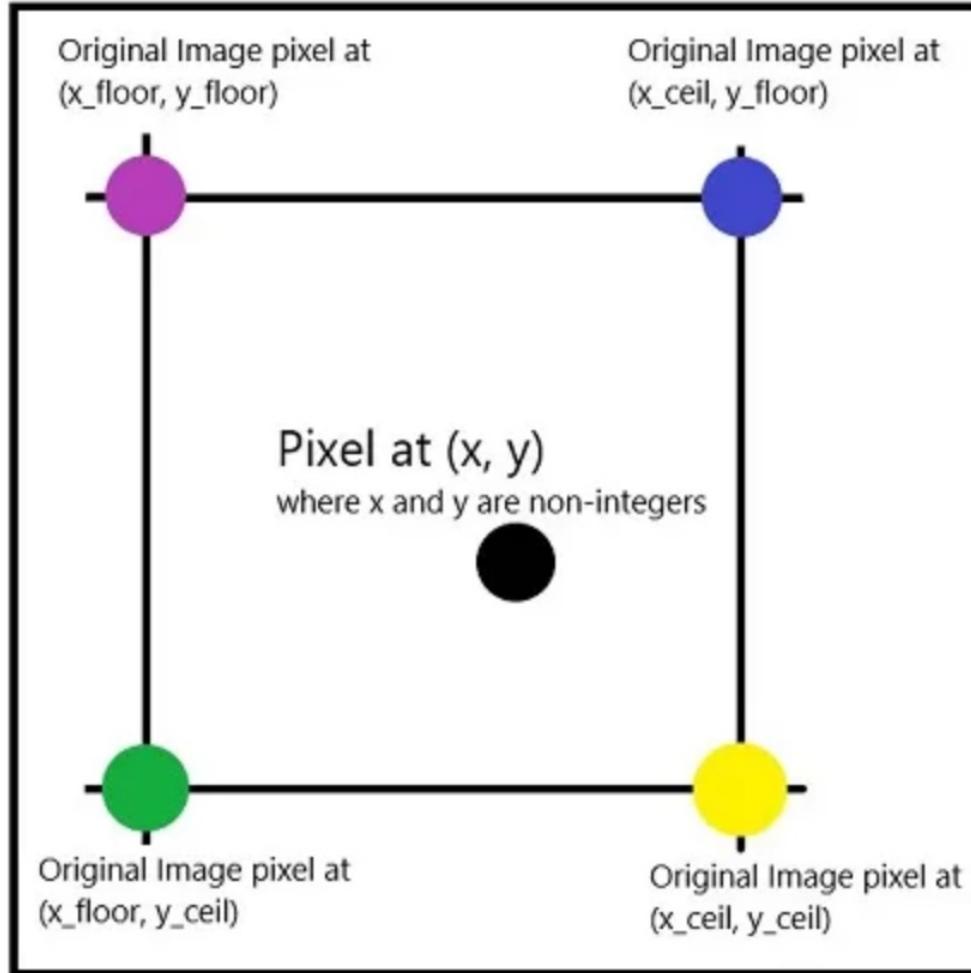


This happens when either  $x$  or  $y$  is an integer resulting in  $q=0$



When  $x$  is an integer,  $x_{floor}$  and  $x_{ceil}$  will have the same value as  $x$ . As a result, both  $q1$  and  $q2$  will be 0 and finally,  $q$  will be 0 as well

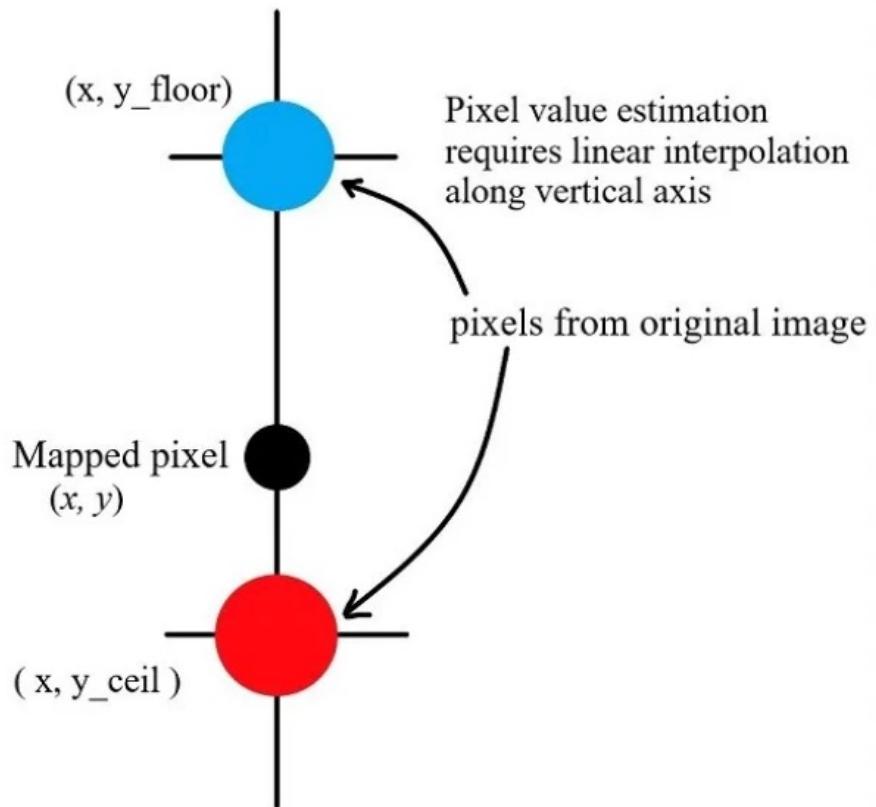
This happens when either  $x$  or  $y$  is an integer resulting in  $q=0$



if  $y$  is an integer,  $y_{floor}$  and  $y_{ceil}$  will have the same value as  $y$ . Consequently, we will get non-zero values for  $q_1$  and  $q_2$  but  $q$  will be 0 since  $y = y_{floor} = y_{ceil}$ .

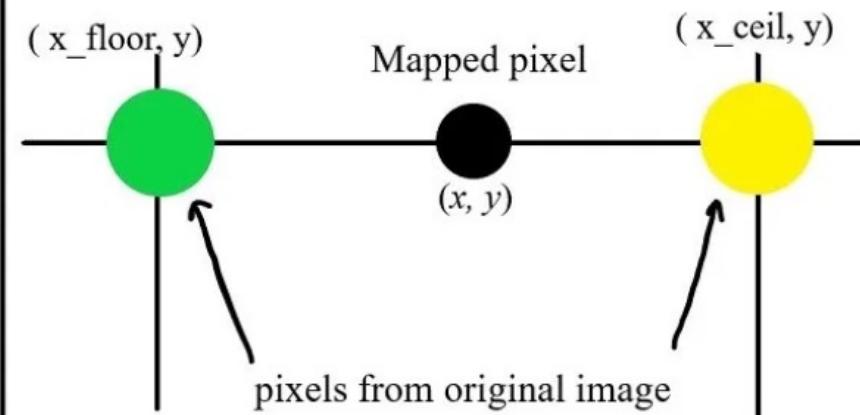
How to solve the problem

When  $x$  is an integer :



When  $y$  is an integer :

Pixel value estimation requires linear interpolation along horizontal axis.



# Solution

```
if (x_ceil == x_floor) and (y_ceil == y_floor):
    q = original_img[int(x)][int(y)]
elif (x_ceil == x_floor):
    q1 = original_img[int(x)][int(y_floor)]
    q2 = original_img[int(x)][int(y_ceil)]
    q = q1 * (y_ceil - y) + q2 * (y - y_floor)
elif (y_ceil == y_floor):
    q1 = original_img[int(x_floor)][int(y)]
    q2 = original_img[int(x_ceil)][int(y)]
    q = (q1 * (x_ceil - x)) + (q2 * (x - x_floor))
else:
    v1 = original_img[x_floor][y_floor]
    v2 = original_img[x_ceil][y_floor]
    v3 = original_img[x_floor][y_ceil]
    v4 = original_img[x_ceil][y_ceil]

    q1 = v1 * (x_ceil - x) + v2 * (x - x_floor)
    q2 = v3 * (x_ceil - x) + v4 * (x - x_floor)
    q = q1 * (y_ceil - y) + q2 * (y - y_floor)
```



