

SCSA SC2001 Lab

Example Class

Project 3 Team 7

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(1) 2 Recursive Definitions

- Naive Approach
- Optimised Approach
 - #1 Improved optimised approach - 2 column
 - #2 Further improved optimised approach - 1 column

Recursive Definition: Naive Approach

Remaining capacity for sub problem

$$P(C, j) = \max_{k=0}^{\left\lfloor \frac{C}{w_{j-1}} \right\rfloor} \{ \underbrace{P(C - kw_{j-1}, j-1)}_{\text{Sub-problem: If don't (j-1)-th item}} + \underbrace{kp_{j-1}}_{\text{Number of (j-1)-th item we take}} \}$$

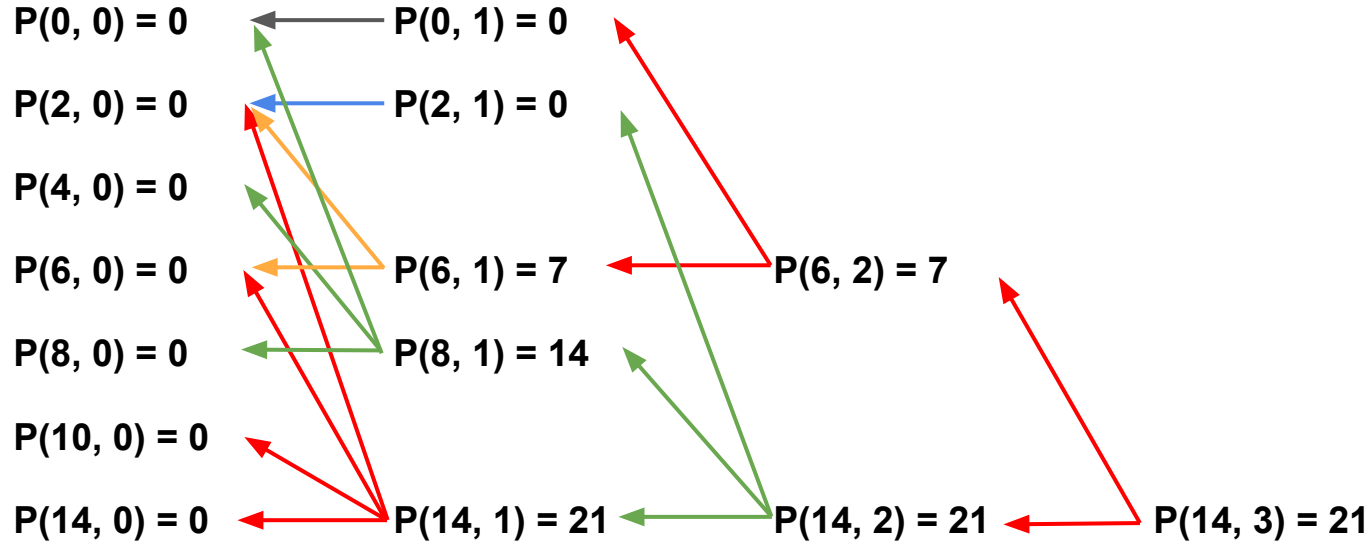
Capacity

Types of items

range of k


Base Case: $P(C, 0) = 0$

Subproblem Graph: Naive Approach



Example

	0	1	2	3 j
0	0	0		
1	0	0		
2	0	0		
3	0	0		
4	0	7		
5	0			
6	0			
7	0			
8	0			
9	0			
10	0			
11	0			
12	0			
13	0			
14	0			
Capacity				



	0	1	2
w_i	4	6	8
p_i	7	6	9

$$P(C, j) = \max_{k=0}^{\lfloor \frac{C}{w_{j-1}} \rfloor} \{P(C - kw_{j-1}, j-1) + kp_{j-1}\}$$

$$C = 4$$

$$w_{j-1} = 4$$

$$k = 4/4 = 1 \rightarrow k \text{ stops at } 1$$

$$k=0,$$

$$P(4, 0) = 0$$

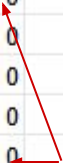
$$k=1,$$

$$P(0,0) + 1*7 = 7$$

Example

	0	1	2
w_i	4	6	8
p_i	7	6	9

	0	1	2	3 j
0	0	0		
1	0	0		
2	0	0		
3	0	0		
4	0	7		
5	0	7		
6	0			
7	0			
8	0			
9	0			
10	0			
11	0			
12	0			
13	0			
14	0			
Capacity				



$$P(C, j) = \max_{k=0}^{\lfloor \frac{C}{w_{j-1}} \rfloor} \{P(C - kw_{j-1}, j-1) + kp_{j-1}\}$$

$$C = 5$$

$$w_{j-1} = 4$$

$$k = 5/4 = 1 \rightarrow k \text{ stops at } 1$$

$$k=0,$$

$$P(5, 0) = 0$$

$$k=1,$$

$$P(1,0) + 1*7 = 7$$

Example

	0	1	2	3 j
0	0	0	0	
1	0	0	0	
2	0	0	0	
3	0	0	0	
4	0	7	7	
5	0	7	7	
6	0	7	7	
7	0	7	7	
8	0	14	14	
9	0	14	14	
10	0	14	14	
11	0	14	14	
12	0	21	21	
13	0	21	21	
14	0	21	21	

	0	1	2
w_i	4	6	8
p_i	7	6	9

$$P(C, j) = \max_{k=0}^{\lfloor \frac{C}{w_{j-1}} \rfloor} \{P(C - kw_{j-1}, j-1) + kp_{j-1}\}$$

$$C = 14$$

$$w_{j-1} = 6$$

$$k = \text{floor}(14/6) = 2 \rightarrow k \text{ stops at } 2$$

$$k=0, \\ P(14, 1) = 21$$

$$k=1, \\ P(8, 1) + 1*6 = 19$$

$$k=2, \\ P(2, 1) + 2*6 = 12$$

Complexity: Naive Approach

Average Time Complexity:

$$O\left(C^2 \cdot \sum_{j=0}^{n-1} \frac{1}{w_j}\right)$$

Worst time complexity: when $w_j = 1$

$$O(C^2 \cdot n)$$

$$P(C, j) = \max_{k=0}^{\lfloor \frac{C}{w_{j-1}} \rfloor} \{P(C - kw_{j-1}, j-1) + kp_{j-1}\}$$

```
for c in range(C): # O(C)
    for j in range(n): # O(n)
        for k in range(C // w[j-1]): # O(C/w_{j-1})
            # Calculate P(C, j)
```


Code for Naive Approach

```
int knapsack(const Item *items_begin, const Item *items_end, const int capacity) {
    size_t len = items_end - items_begin;
    int **f = calloc(len + 1, sizeof(int *));
    for (int i = 0; i <= len; ++i) {
        f[i] = calloc(capacity + 1, sizeof(int));
        memset(f[i], 0, (capacity + 1) * sizeof(int));
    }
    for (int i = 1; i <= len; ++i) {
        for (int j = 0; j <= capacity; ++j) {
            for (int k = 0; k * items_begin[i - 1].weight <= j; ++k) {
                f[i][j] = max(f[i][j],
                               f[i - 1][j - k * items_begin[i - 1].weight] +
                               k * items_begin[i - 1].profit);
            }
        }
    }
    int ans = f[len][capacity];
    free(f);
    return ans;
}
```

Recursive Definition: Optimised Approach

$$P(C, j) = \max \left\{ \overbrace{P(C, j-1)}^{\text{Don't choose } j\text{th element}}, \overbrace{P(C - w_{j-1}, j) + p_{j-1}}^{\text{Choose } j\text{th element}} \right\}$$

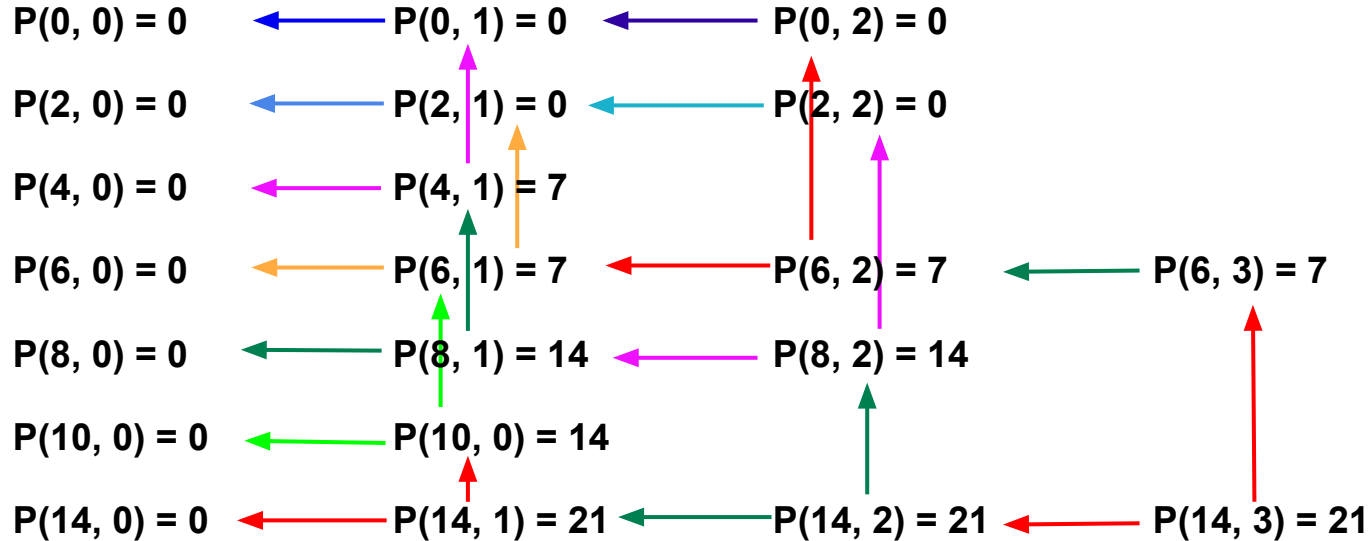
Reserve w_j capacity for the item

You can still choose item j due to unlimited supplies

Base Case: $P(C, 0) = 0$

Subproblem Graph : Optimised Approach

	0	1	2
w_i	4	6	8
p_i	7	6	9



Example

	0	1	2
w_i	4	6	8
p_i	7	6	9

	0	1	2	3	j
0	0	0	0	0	
1	0	0			
2	0	0			
3	0	0			
4	0	7			
5	0				
6	0				
7	0				
8	0				
9	0				
10	0				
11	0				
12	0				
13	0				
14	0				
Capacity					

$$P(C, j) = \max\{P(C, j - 1), P(C - w_{j-1}, j) + p_{j-1}\}$$

$$\begin{aligned}
 P(4, 1) &= \max((P(4, 0), P(4-4, 1) + 7)) \\
 &= \max((P(4, 0), P(0, 1) + 7)) \\
 &= \max(0, 7) \\
 &= 7
 \end{aligned}$$

Example

	0	1	2
w_i	4	6	8
p_i	7	6	9

	0	1	2	3	j
0	0	0	0	0	
1	0	0	0	0	
2	0	0	0	0	
3	0	0	0	0	
4	0	7	7	7	
5	0	7	7	7	
6	0	7	7	7	
7	0	7	7	7	
8	0	14			
9	0	14			
10	0	14			
11	0	14			
12	0	21			
13	0	21			
14	0	21			
Capacity					

$$P(C, j) = \max\{P(C, j - 1), P(C - w_{j-1}, j) + p_{j-1}\}$$

$$\begin{aligned}
 P(12, 1) &= \max((P(12, 0), P(12-4, 1) + 7)) \\
 &= \max((P(12, 0), P(8, 1) + 7)) \\
 &= \max(0, 21) \\
 &= \mathbf{21}
 \end{aligned}$$

$W_j < C$

Notice the pattern!

Example

	0	1	2
w_i	4	6	8
p_i	7	6	9

	0	1	2	3	j
0	0	0	0	0	
1	0	0	0	0	
2	0	0	0	0	
3	0	0	0	0	
4	0	7	7	7	
5	0	7	7	7	
6	0	7	7	7	
7	0	7	7	7	
8	0	14	14		
9	0	14	14		
10	0	14	14		
11	0	14	14		
12	0	21	21		
13	0	21	21		
14	0	21	21		
Capacity					

$$P(C, j) = \max\{P(C, j - 1), P(C - w_{j-1}, j) + p_{j-1}\}$$

$$\begin{aligned}
 P(11, 2) &= \max((P(11, 1), P(11-6, 2) + 6)) \\
 &= \max((P(11, 1), P(5, 2) + 6)) \\
 &= \max(14, 13) \\
 &= \mathbf{14}
 \end{aligned}$$

Example

	0	1	2	3	j
0	0	0	0	0	
1	0	0	0	0	
2	0	0	0	0	
3	0	0	0	0	
4	0	7	7	7	
5	0	7	7	7	
6	0	7	7	7	
7	0	7	7	7	
8	0	14	14	14	
9	0	14	14	14	
10	0	14	14	14	
11	0	14	14	14	
12	0	21	21	21	
13	0	21	21	21	
14	0	21	21	21	
Capacity					

	0	1	2
w_i	4	6	8
p_i	7	6	9

Object 0 costs lesser weight + profit-friendly
 Example : 2 times of Object 0 is >> 1 Object 2

$$P(C, j) = \max\{P(C, j - 1), P(C - w_{j-1}, j) + p_{j-1}\}$$

$$\begin{aligned} P(8, 3) &= \max((P(8, 2), P(8-8, 3) + 9)) \\ &= \max((P(8, 2), P(0, 3) + 9)) \\ &= \max(14, 9) \\ &= \mathbf{14} \end{aligned}$$

$$\begin{aligned} P(14, 3) &= \max((P(14, 2), P(14-8, 3) + 9)) \\ &= \max((P(14, 2), P(6, 3) + 9)) \\ &= \max(21, 15) \\ &= \mathbf{21} \end{aligned}$$

Complexity : Optimised Approach

```
for c in range(C + 1):      # O(C)
    for j in range(n):      # O(n)
        # O(1) calculate P(c, j)
```

$$\Theta(C \cdot n)$$

Running result for

	0	1	2
w_i	5	6	8
p_i	7	6	9

	0	1		
0	0	0		
1	0	0		
2	0	0	0	0
3	0	0	0	0
4	0	0	0	0
5	0	7	7	7
6	0	7	7	7
7	0	7	7	7
8	0	7	7	9
9	0	7	7	9
10	0	14	14	14
11	0	14	14	14
12	0	14	14	14
13	0	14	14	16
14	0	14	14	16
Capacity				

Maximum
Profit = 16

Code for Optimised Approach

```
int knapsack(const Item *items_begin, const Item *items_end, const int capacity) {
    int **f = calloc((items_end - items_begin + 1), sizeof(int *));
    for (int i = 0; i <= items_end - items_begin; ++i) {
        f[i] = calloc(capacity + 1, sizeof(int));
    }
    memset(f[0], 0, (capacity + 1) * sizeof(int));
    for (const Item *item = items_begin; item != items_end; ++item) {
        for (int i = 0; i <= capacity; ++i) {
            if (i >= item->weight) {
                f[item - items_begin + 1][i] = max(f[item - items_begin][i],
                f[item - items_begin + 1][i - item->weight] + item->profit);
            } else {
                f[item - items_begin + 1][i] = f[item - items_begin][i];
            }
        }
    }
    int ans = f[items_end - items_begin][capacity];
    for (int i = 0; i <= items_end - items_begin; ++i) {
        free(f[i]);
    }
    free(f);
    return ans;
}
```

Analysis of both approaches

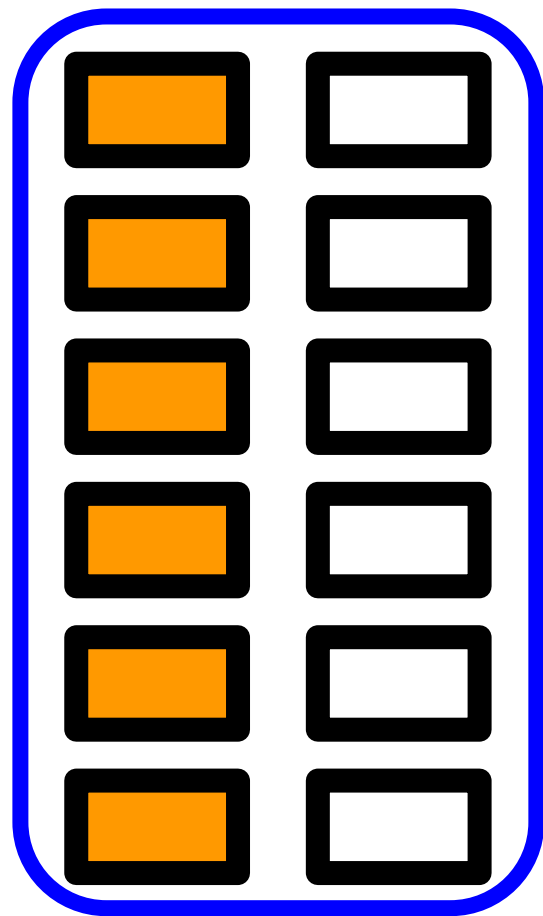
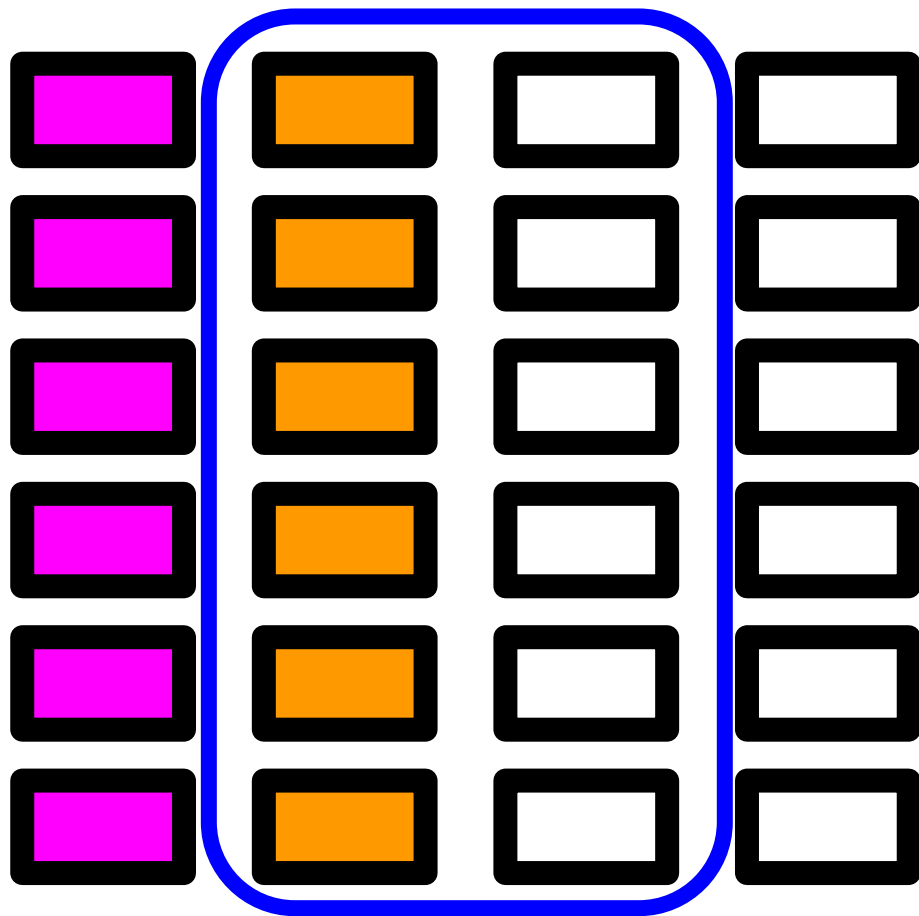
- Time Complexity for Optimised Approach is better than Naive
 - $O(Cn) \ll O(C^2n)$

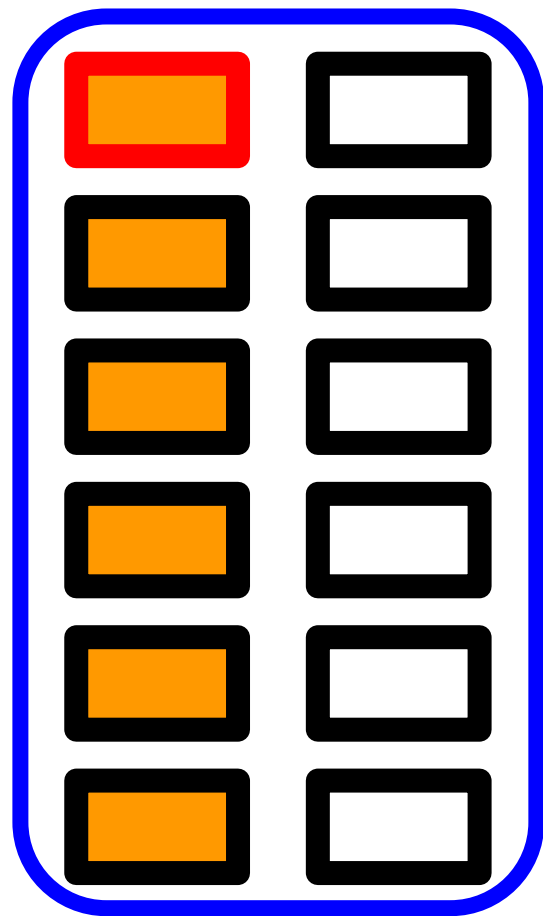
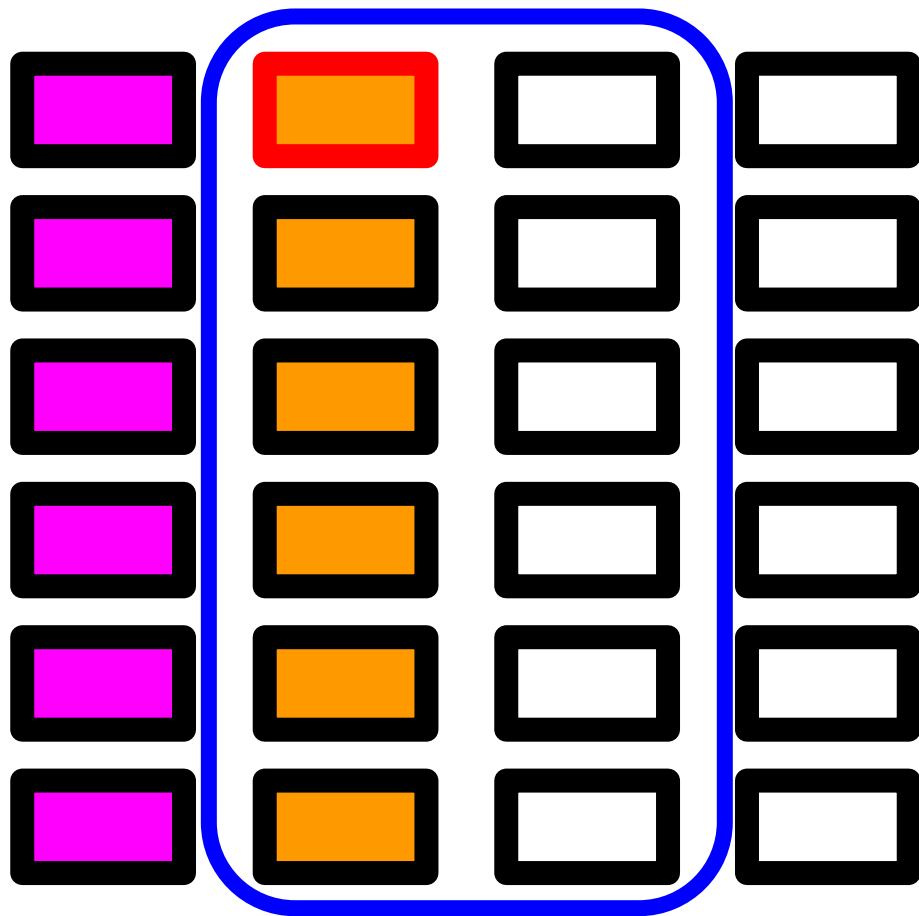
Improved optimised approach : 2 columns + 1 column

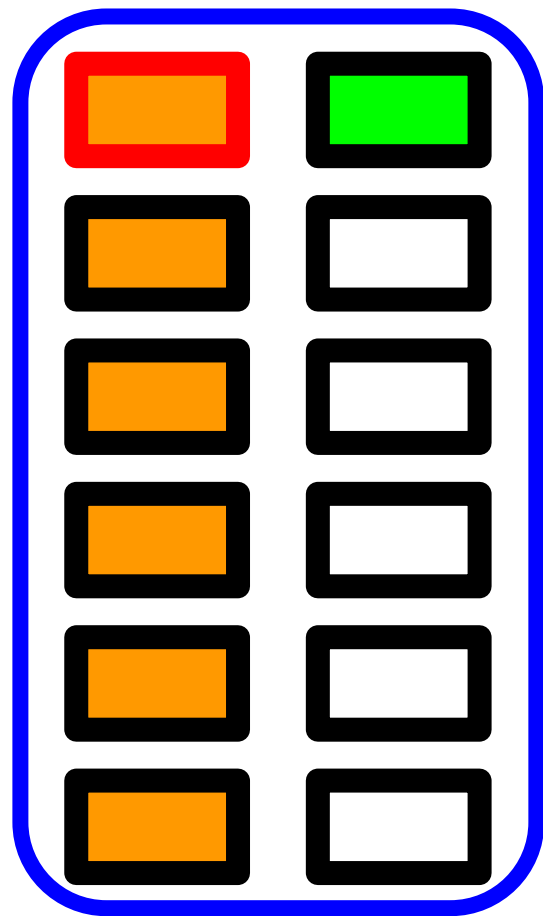
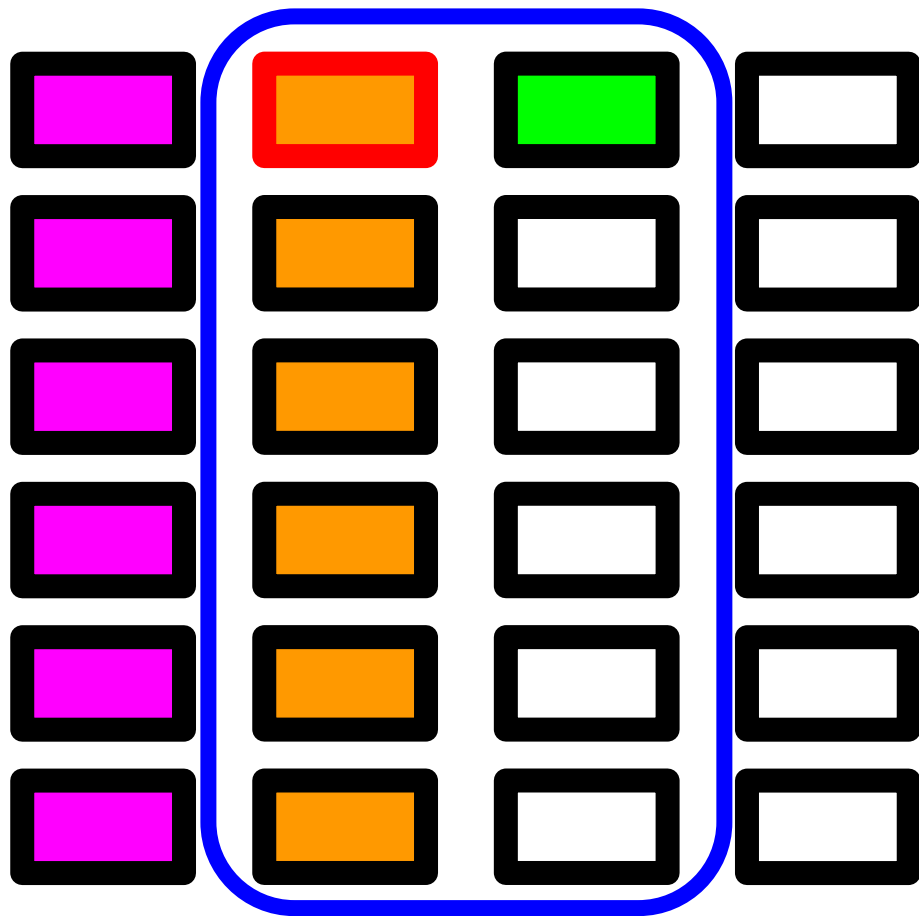
	0	1	2	3	j
0	0	0	0	0	
1	0	0	0	0	
2	0	0	0	0	
3	0	0	0	0	
4	0	7	7	7	
5	0	7	7	7	
6	0	7	7	7	
7	0	7	7	7	
8	0	14	14	14	
9	0	14	14	14	
10	0	14	14	14	
11	0	14	14	14	
12	0	21	21	21	
13	0	21	21	21	
14	0	21	21	21	
Capacity					

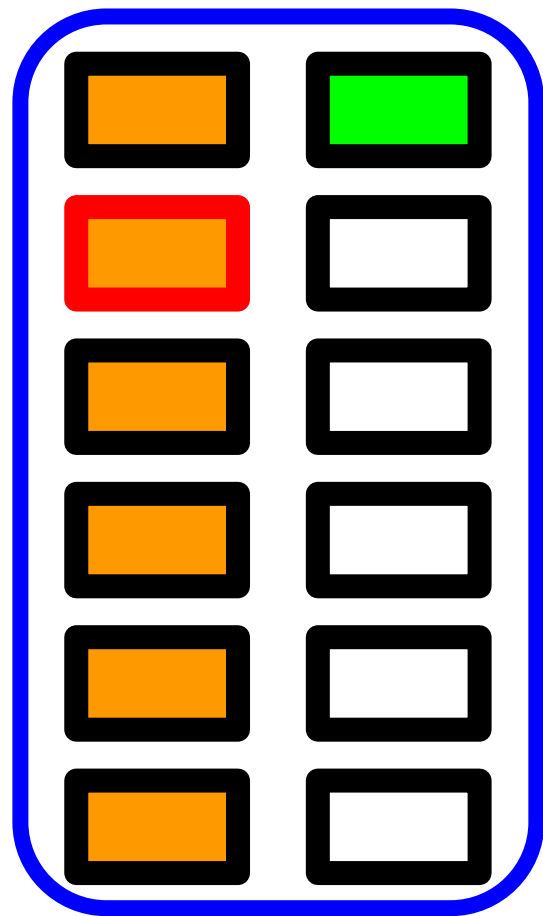
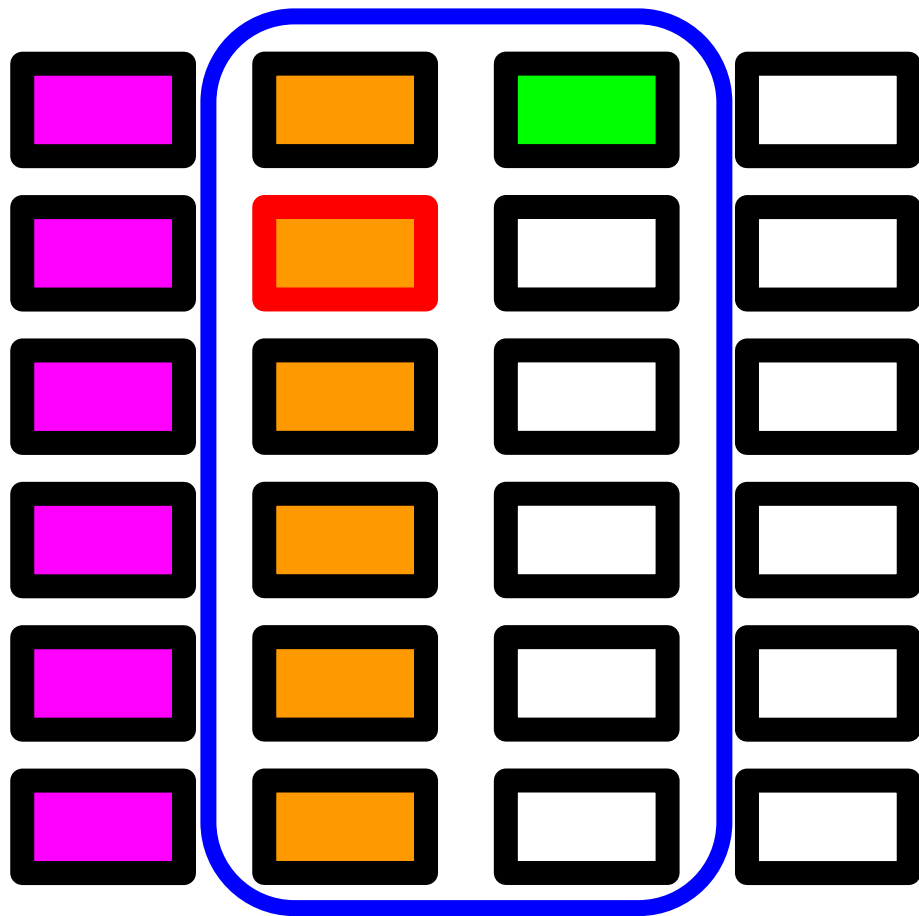


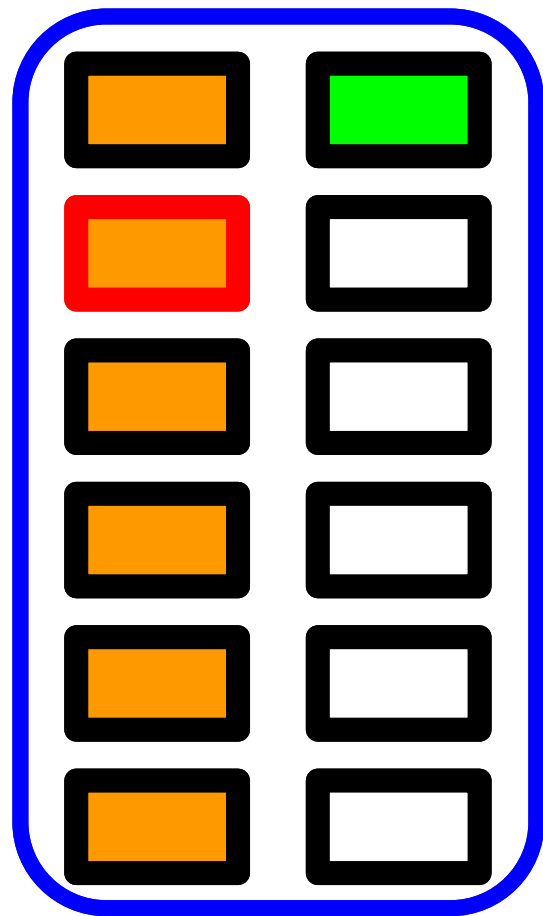
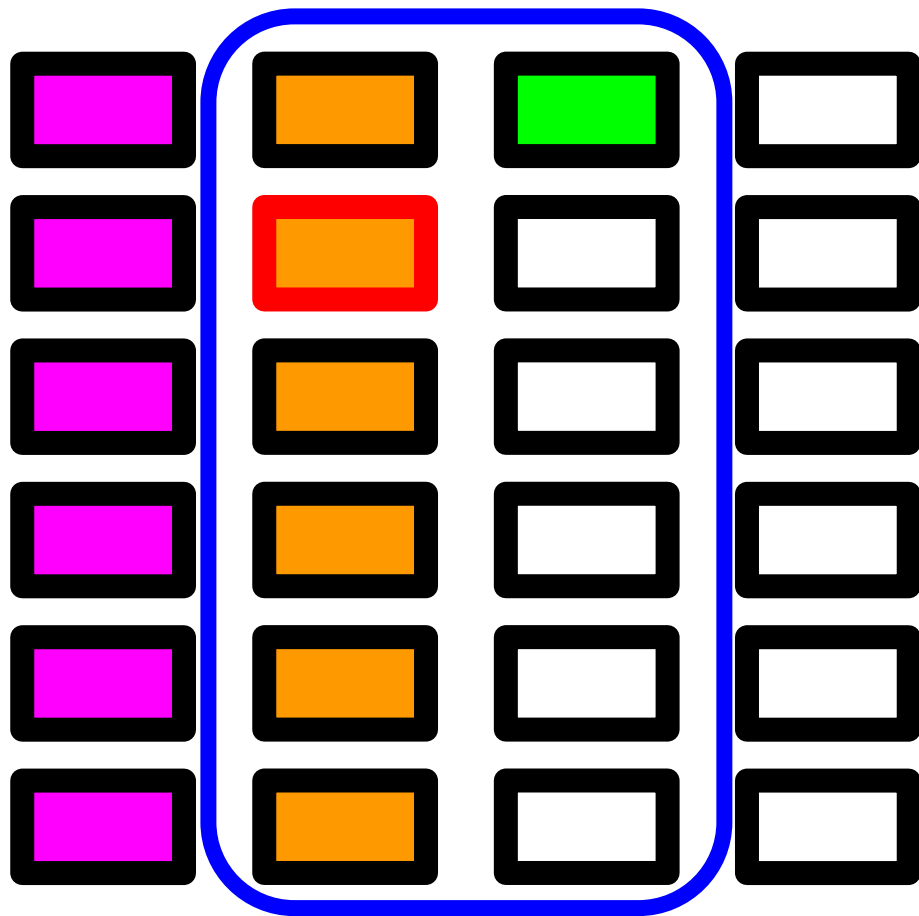
	2	3	j
0	0	0	
1	0	0	
2	0	0	
3	0	0	
4	7	7	
5	7	7	
6	7	7	
7	7	7	
8	14	14	
9	14	14	
10	14	14	
11	14	14	
12	21	21	
13	21	21	
14	21	21	
Capacity			

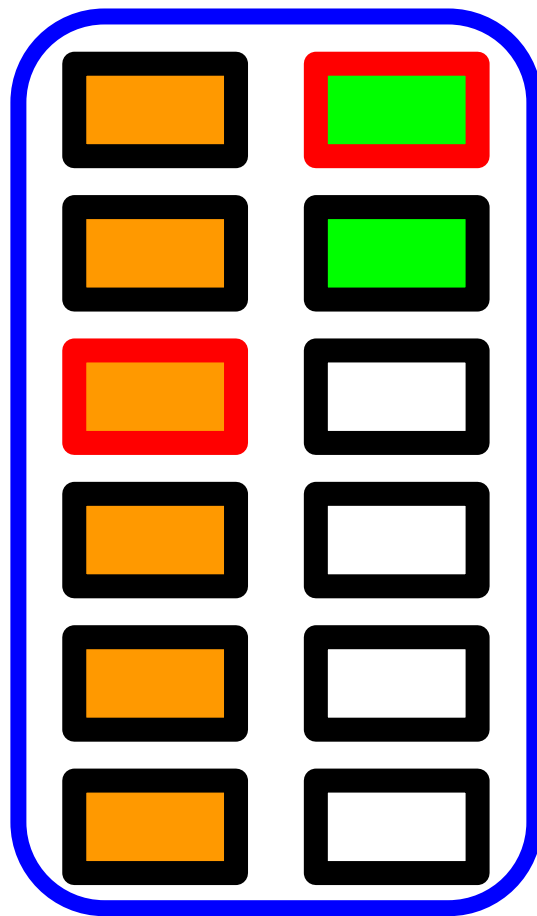
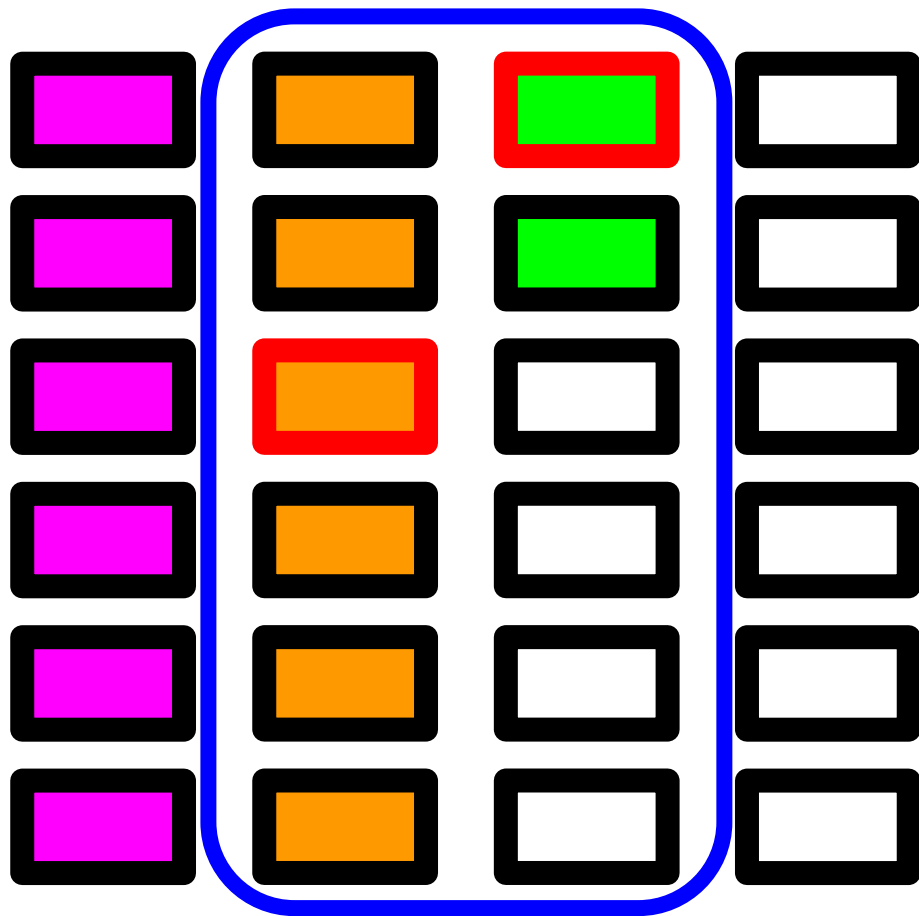


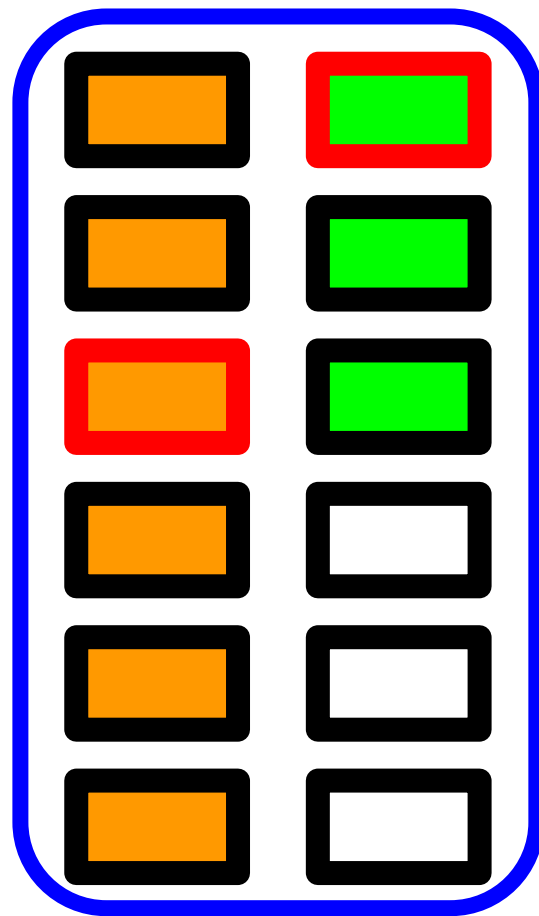
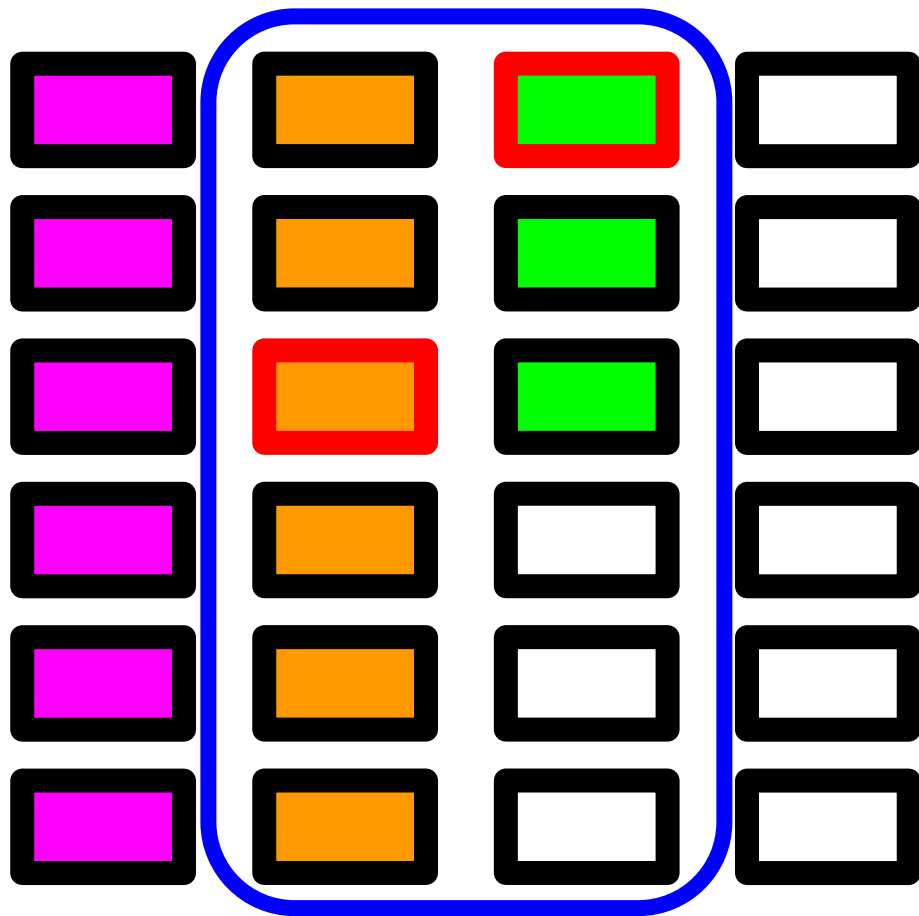


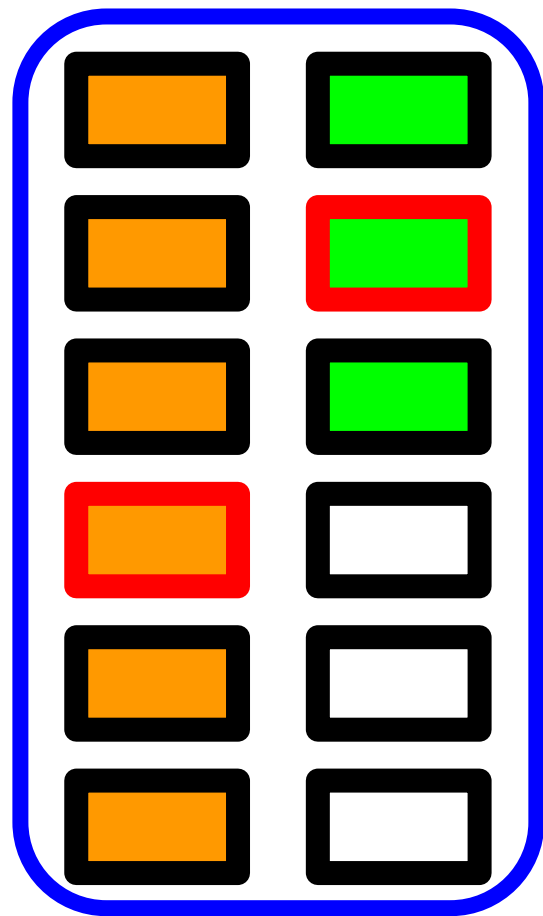
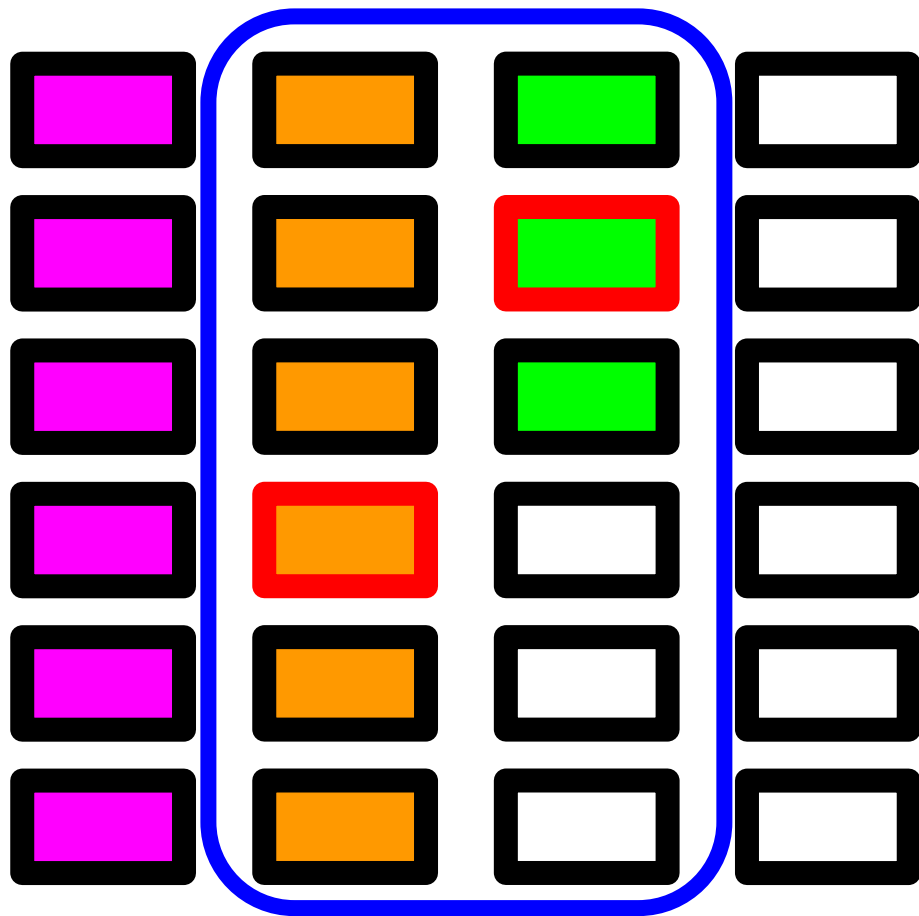


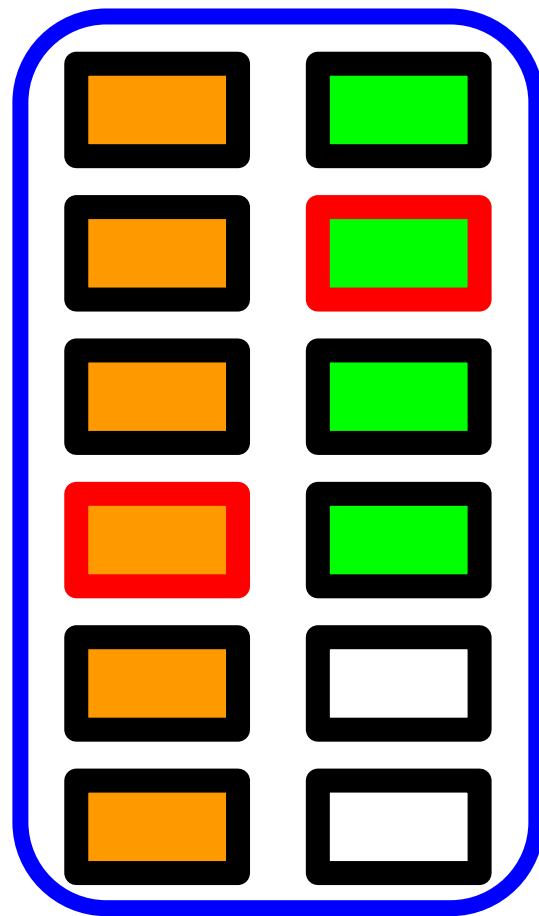
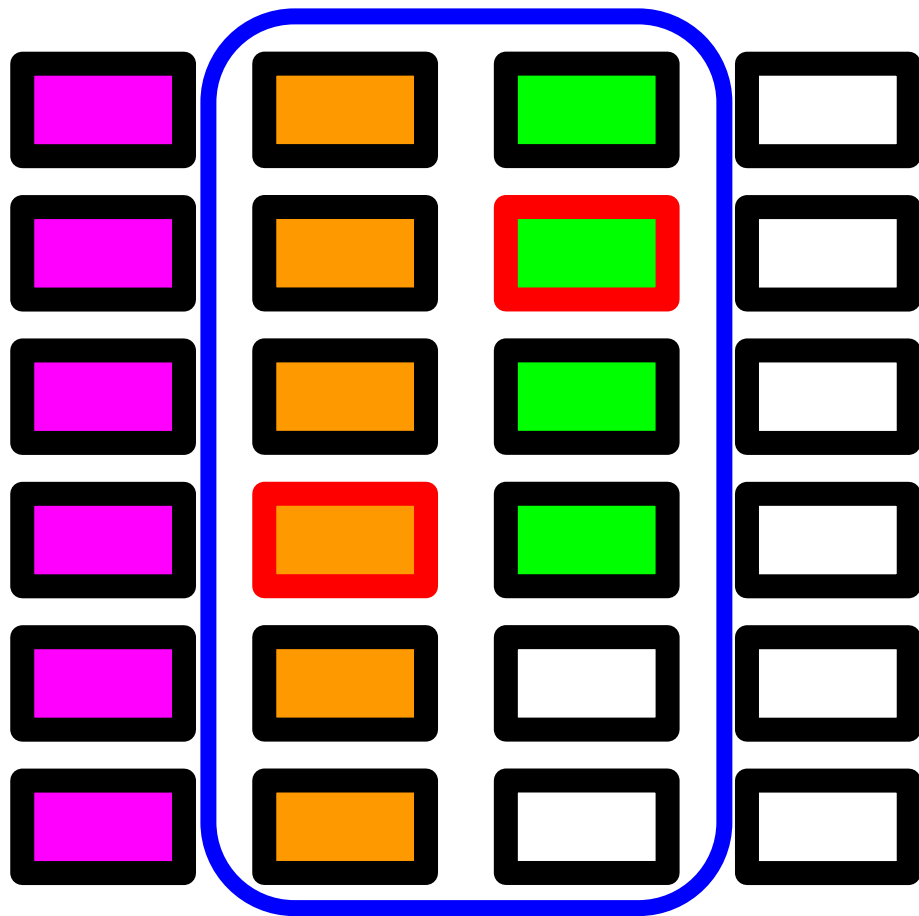


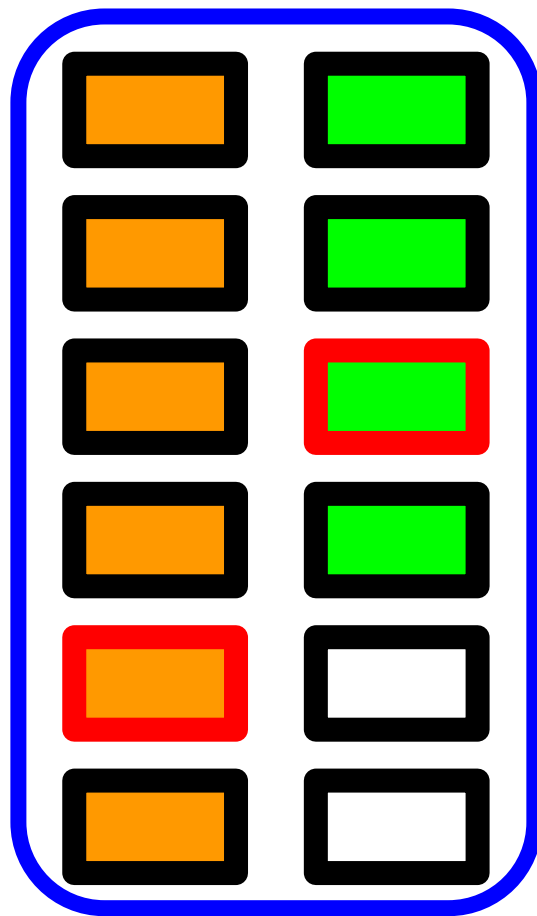
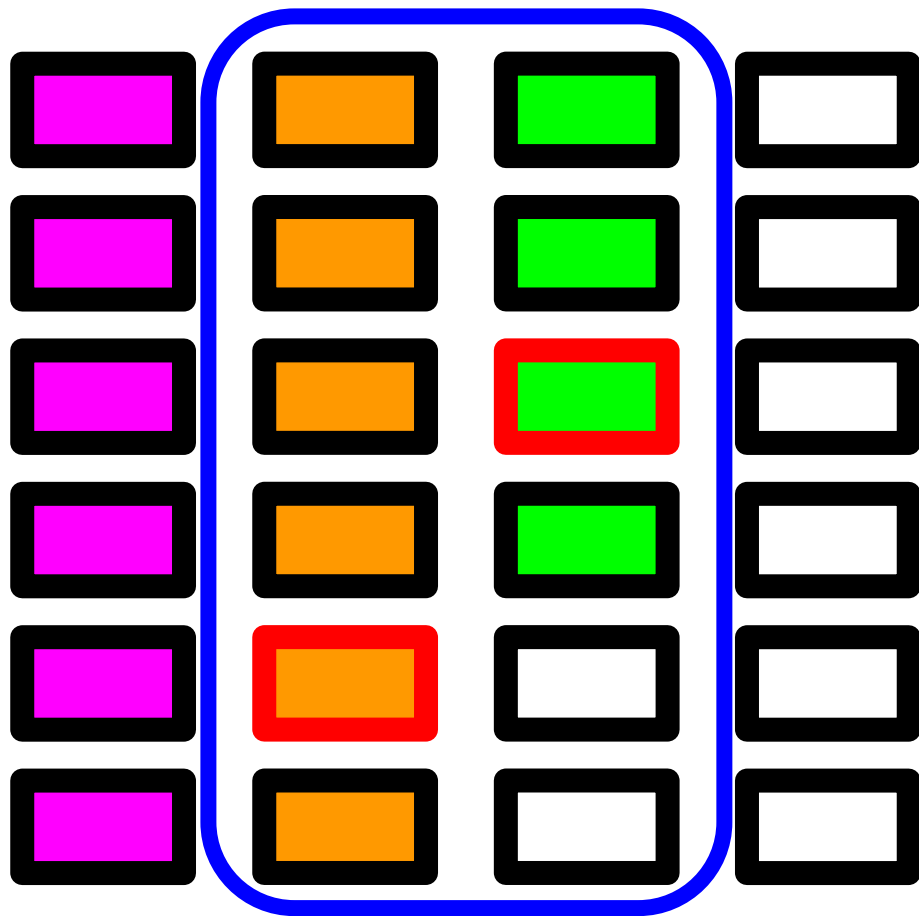


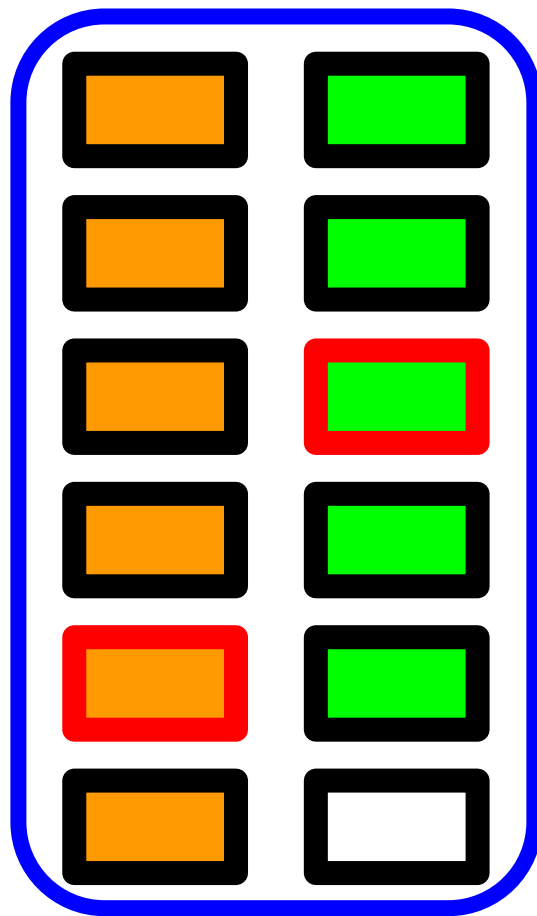
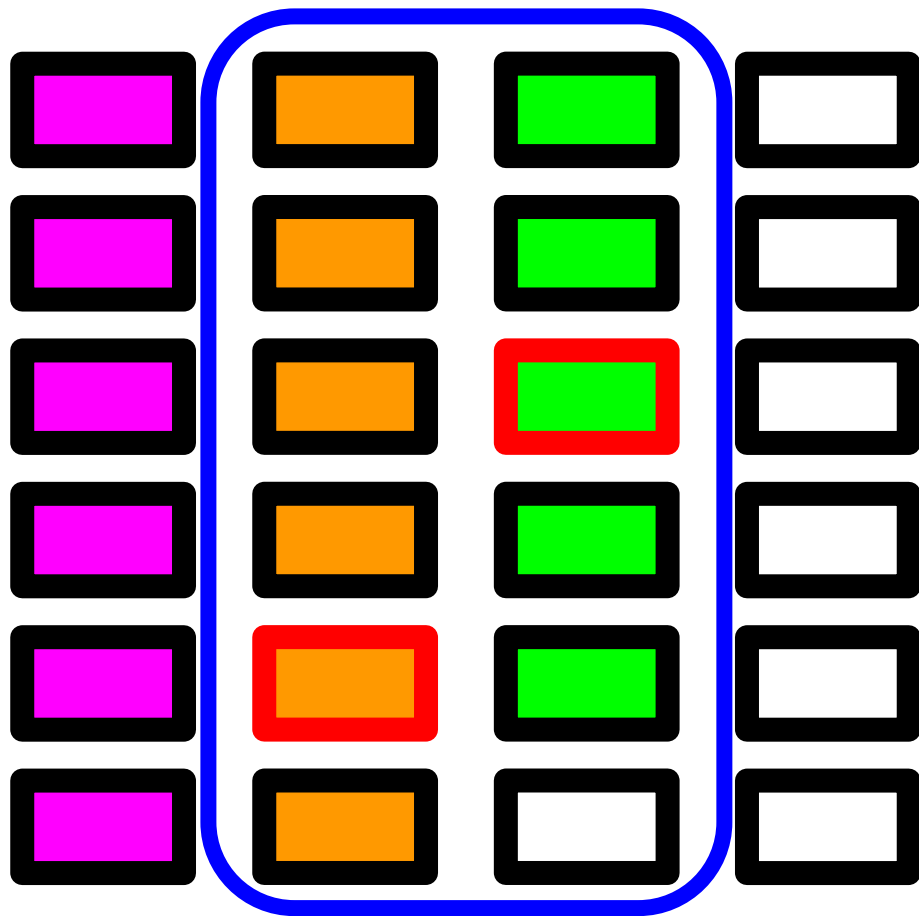


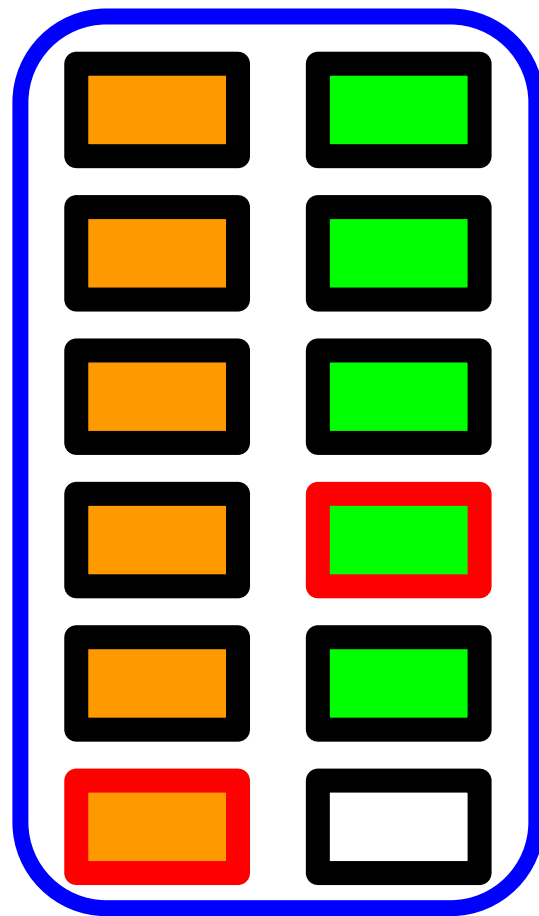
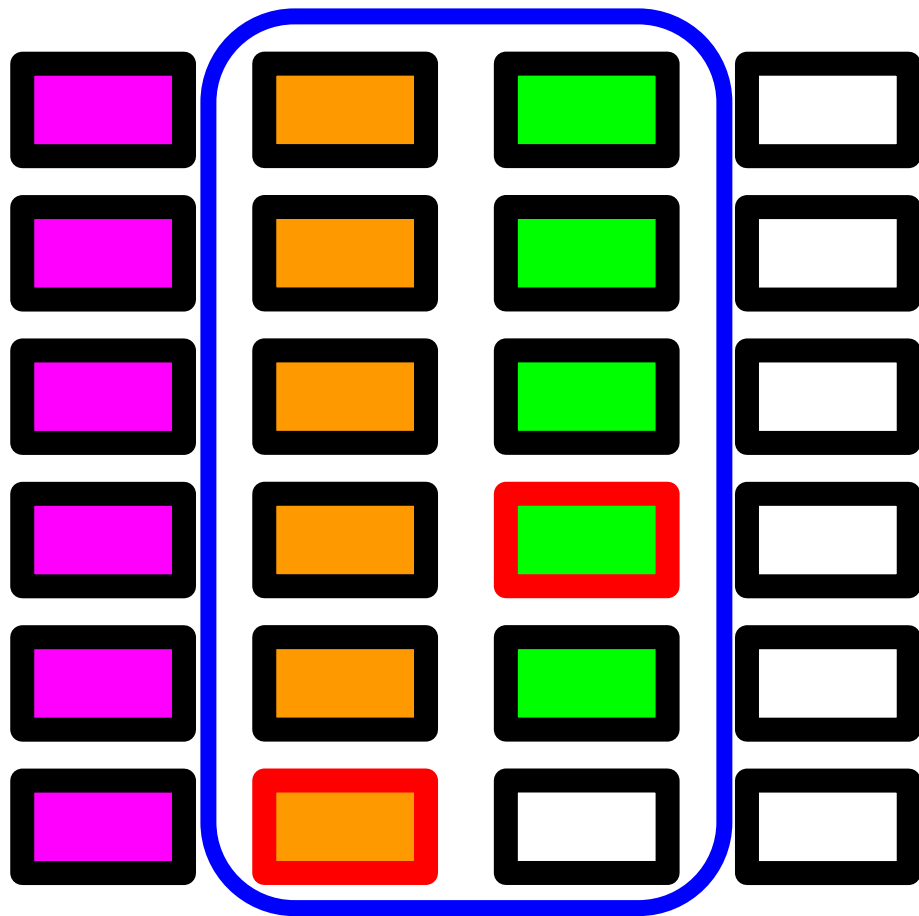


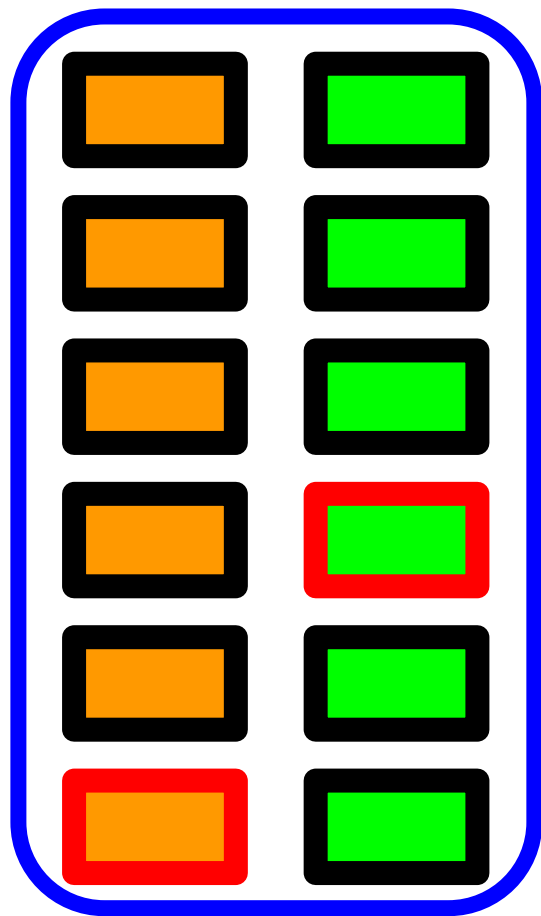
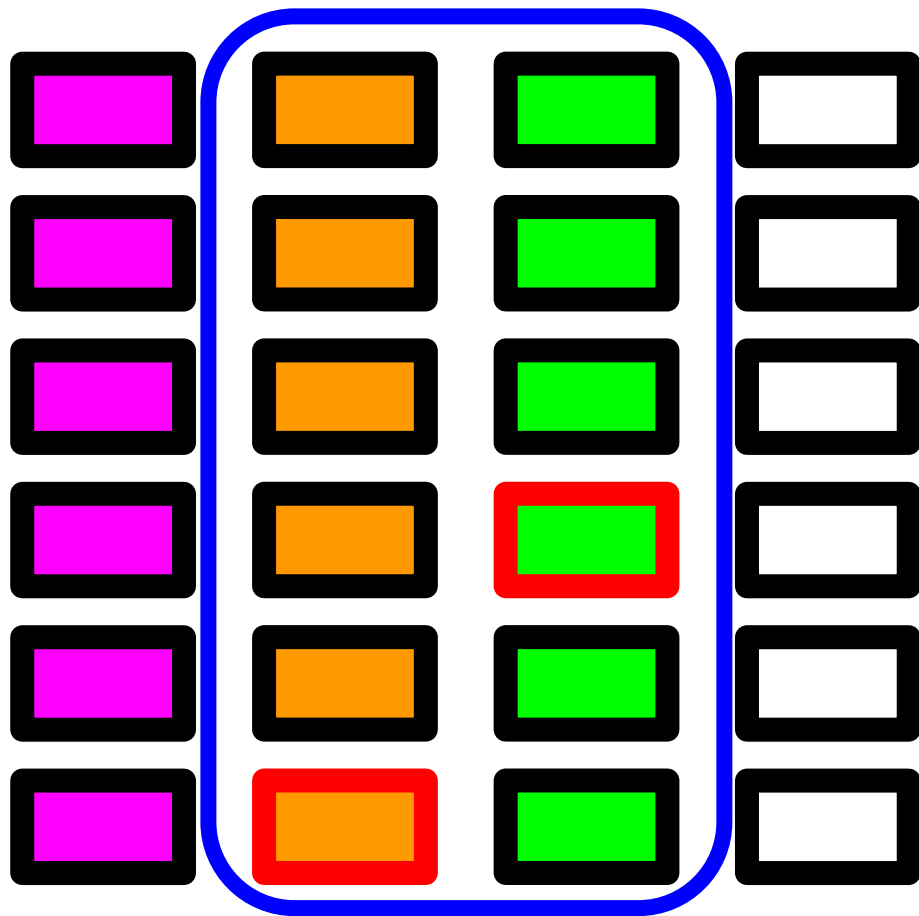


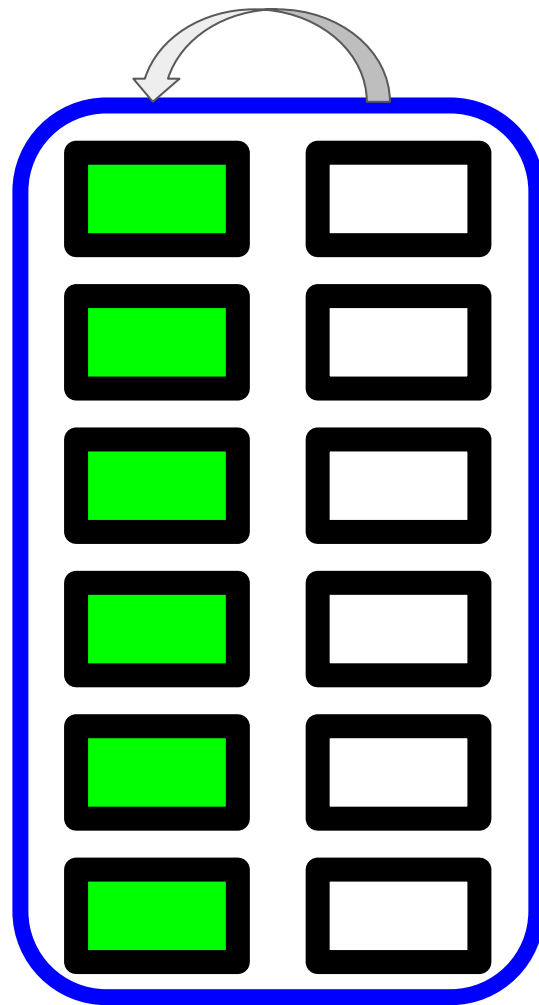
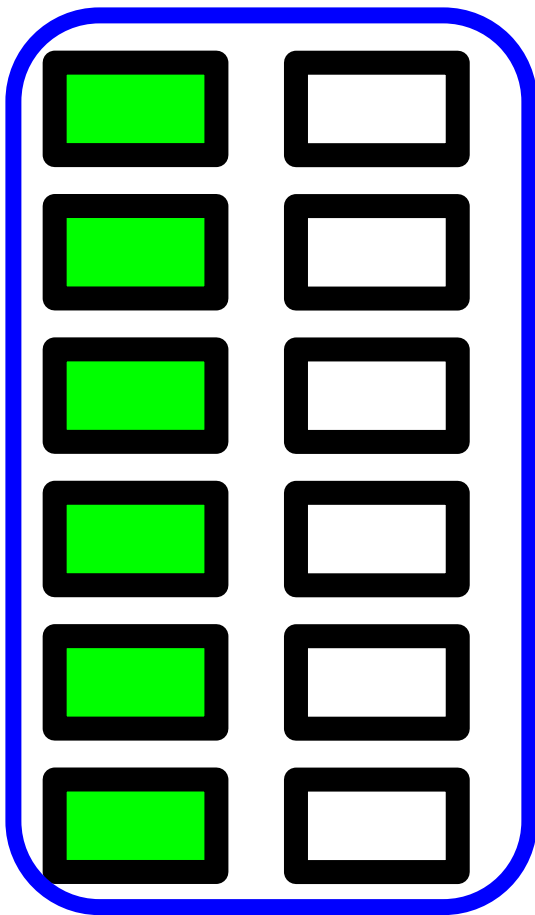
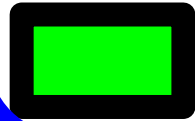
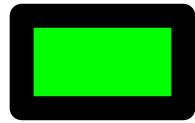
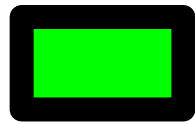
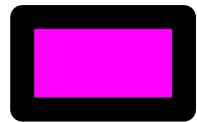




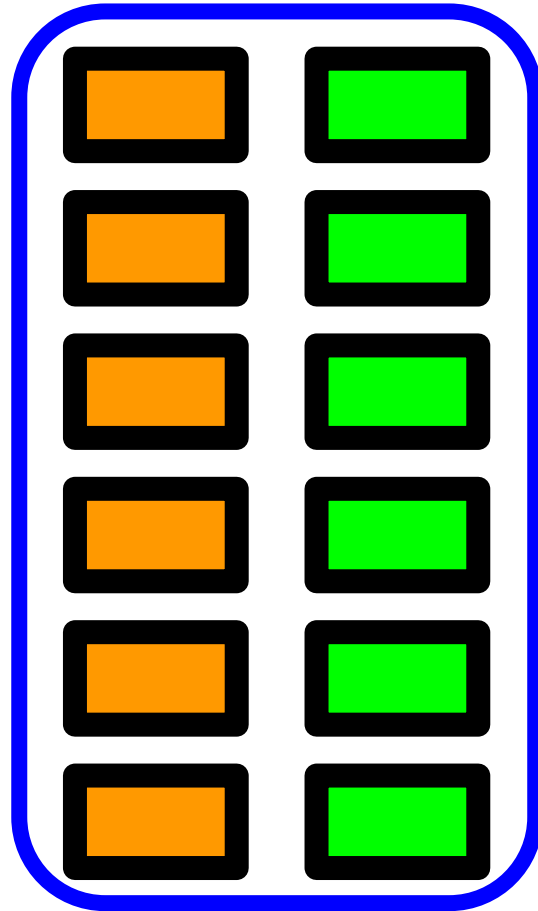


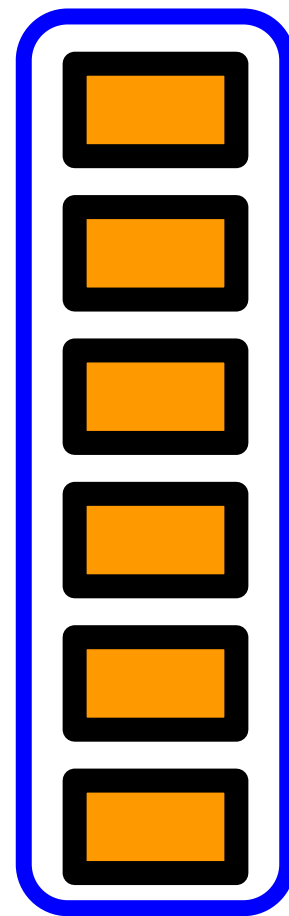
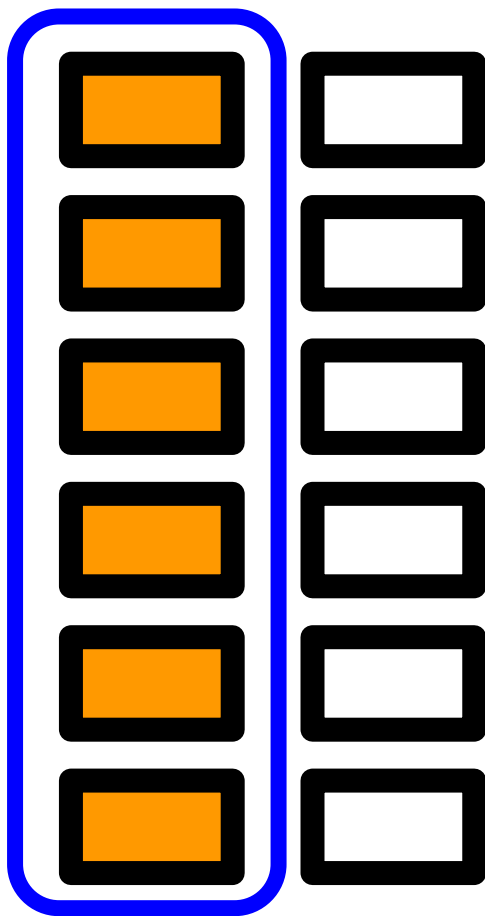


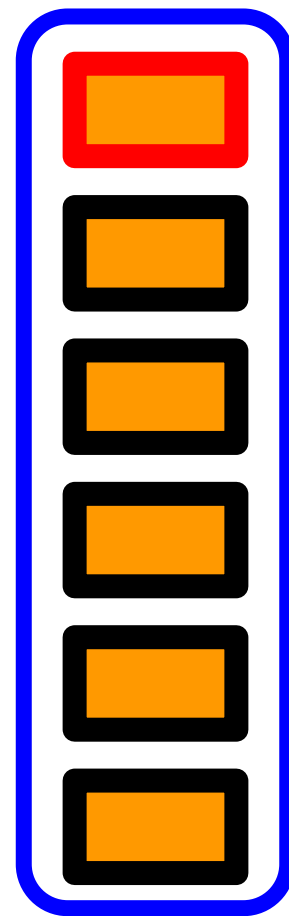
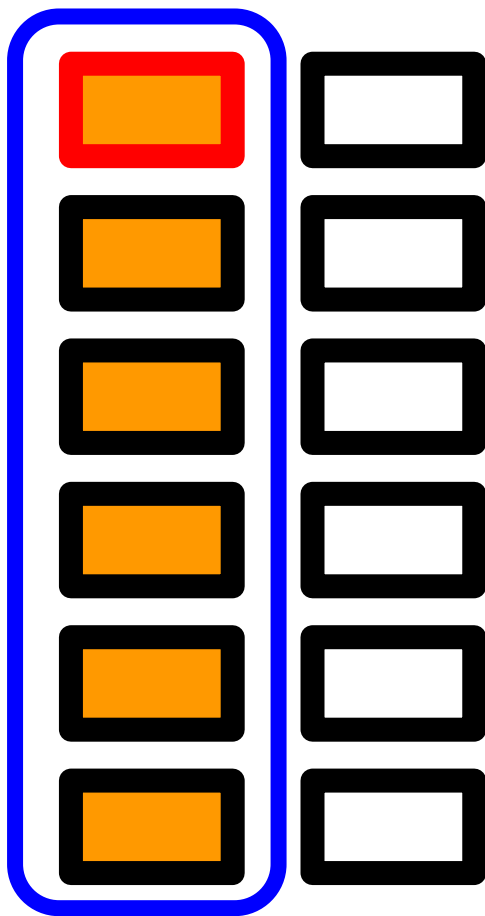


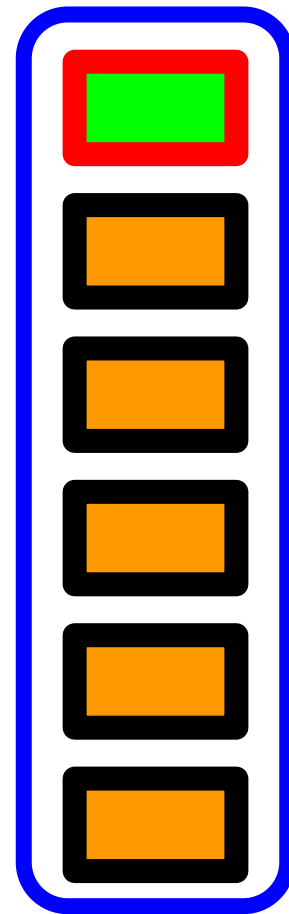
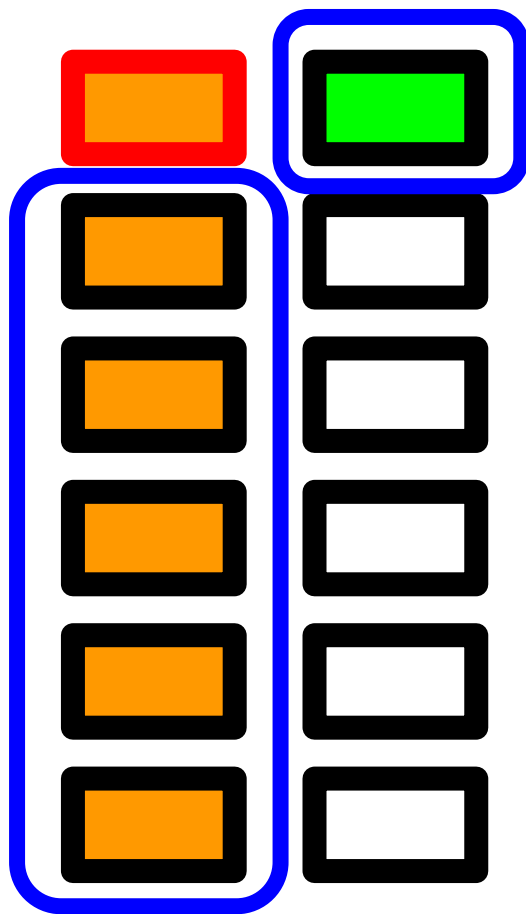


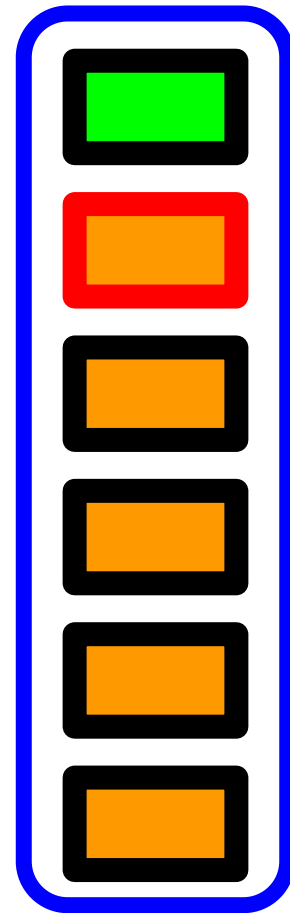
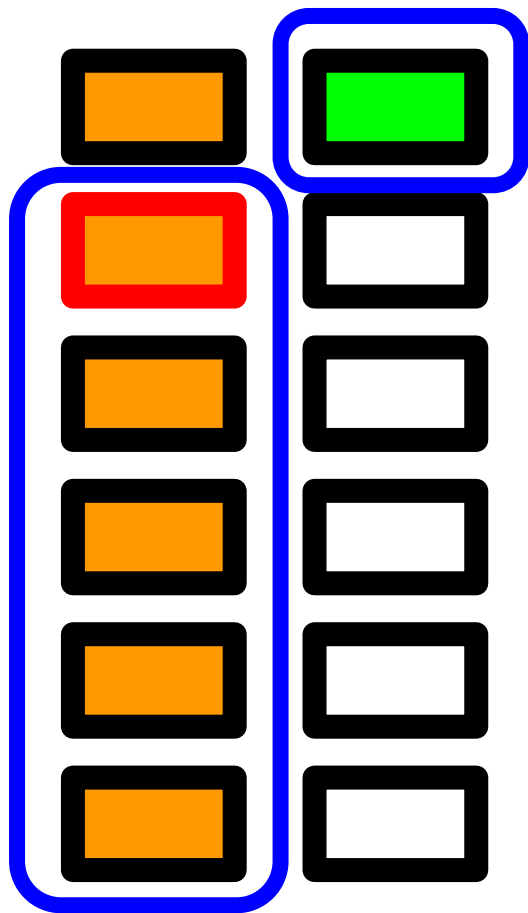
Saves space!
Improves space complexity

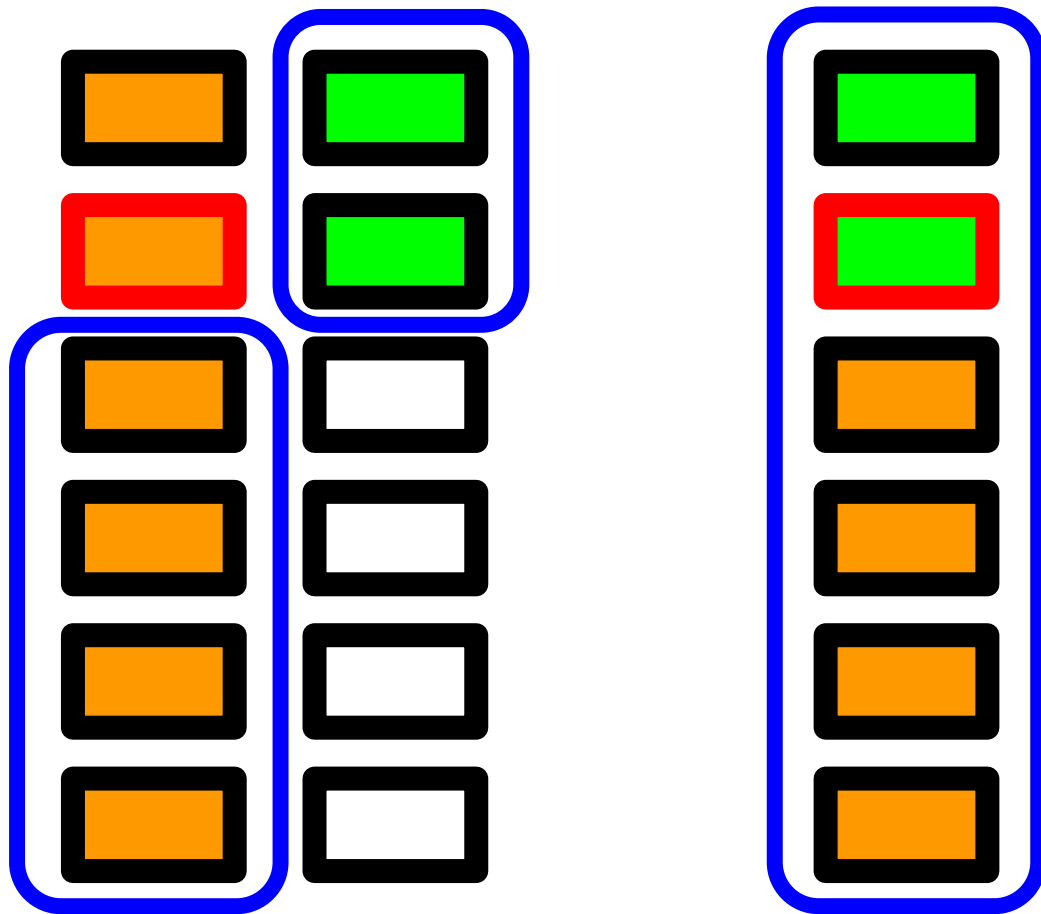


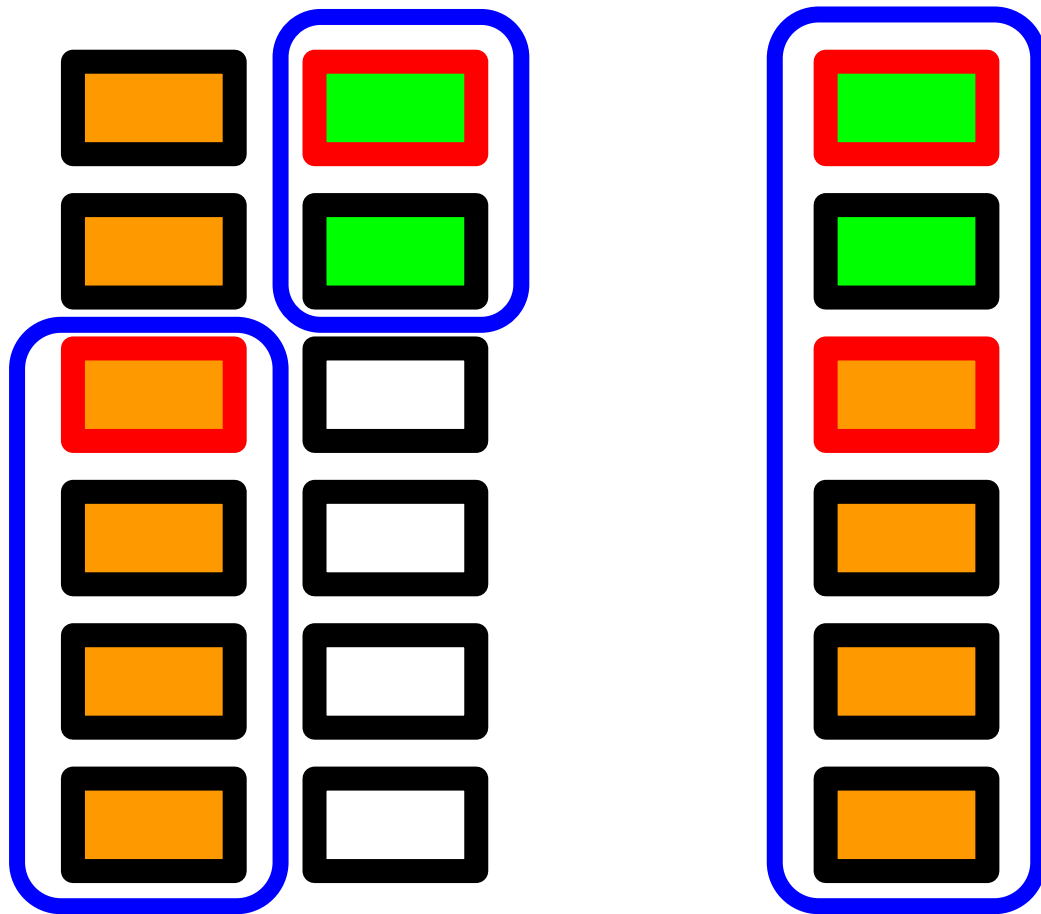


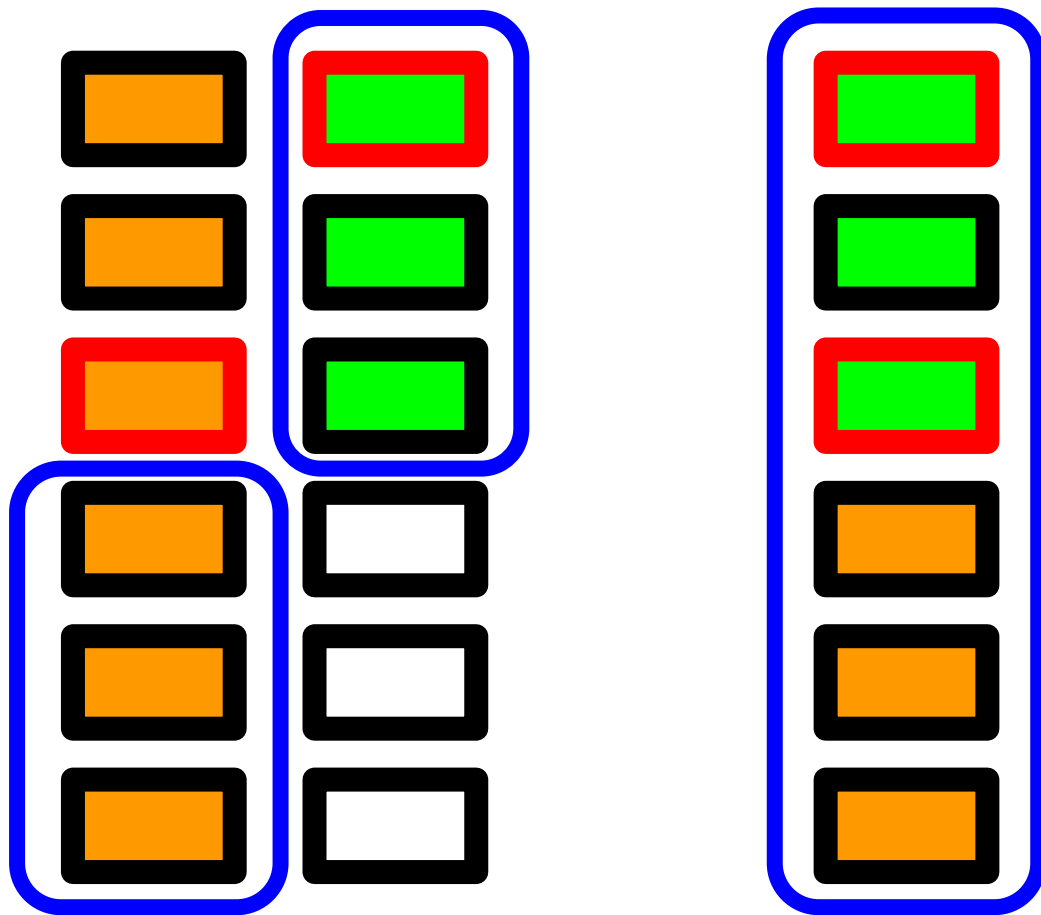


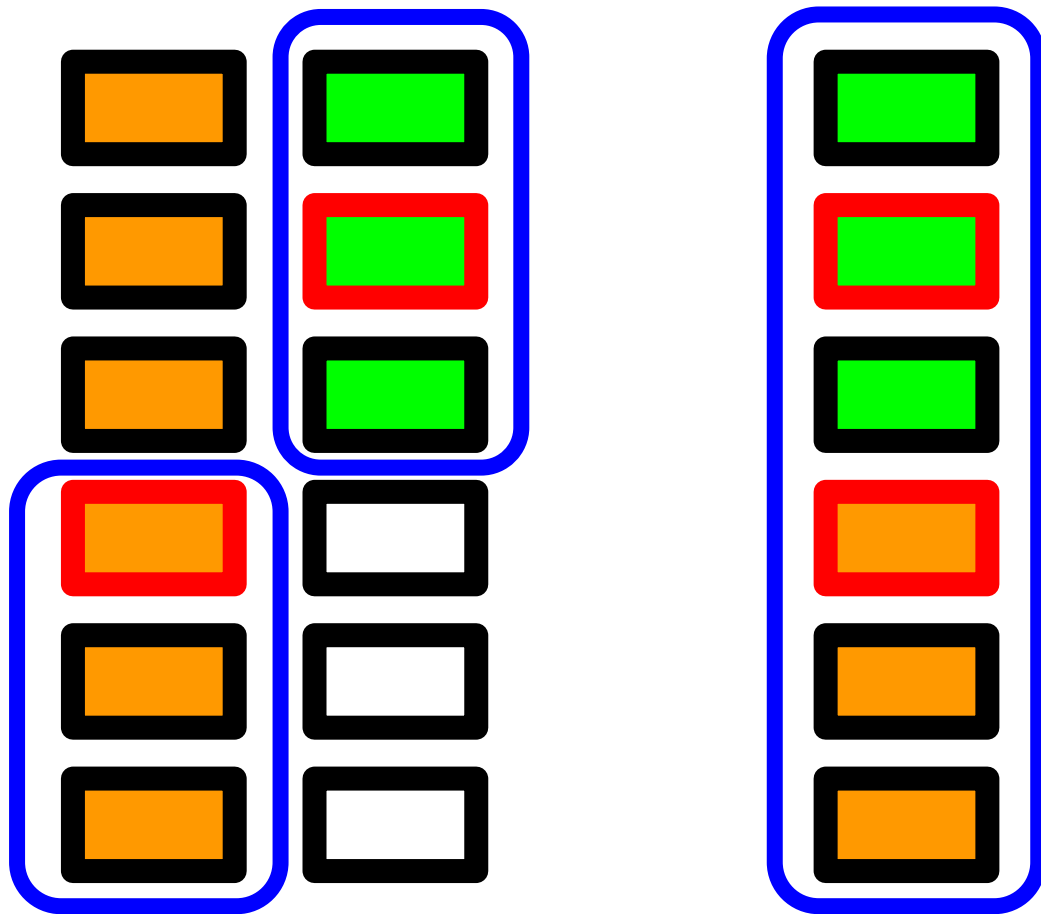


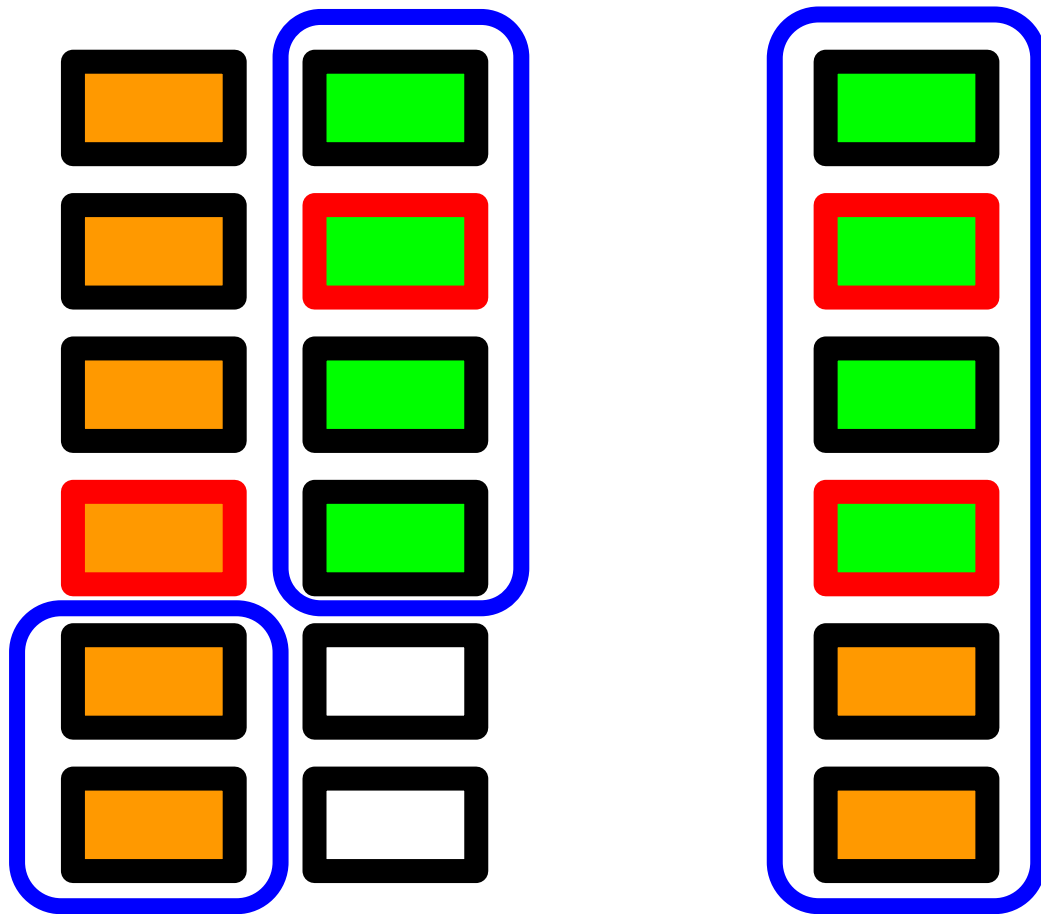


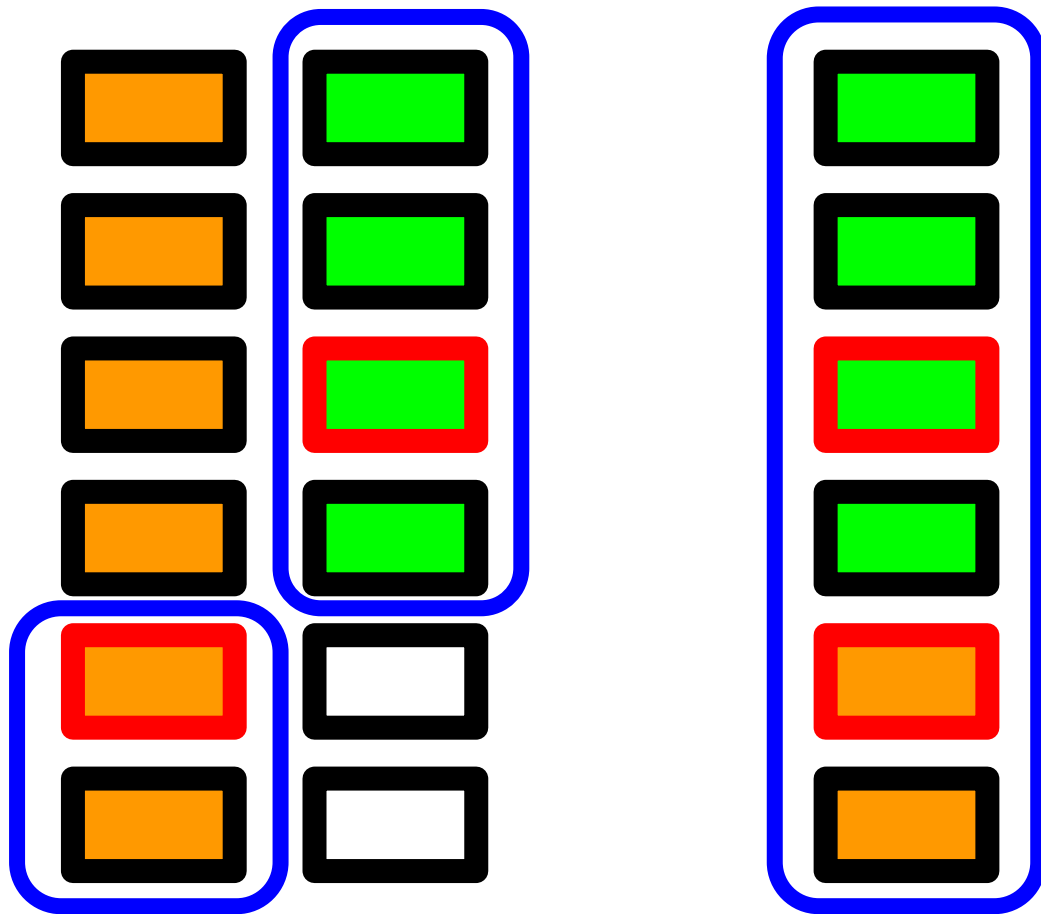


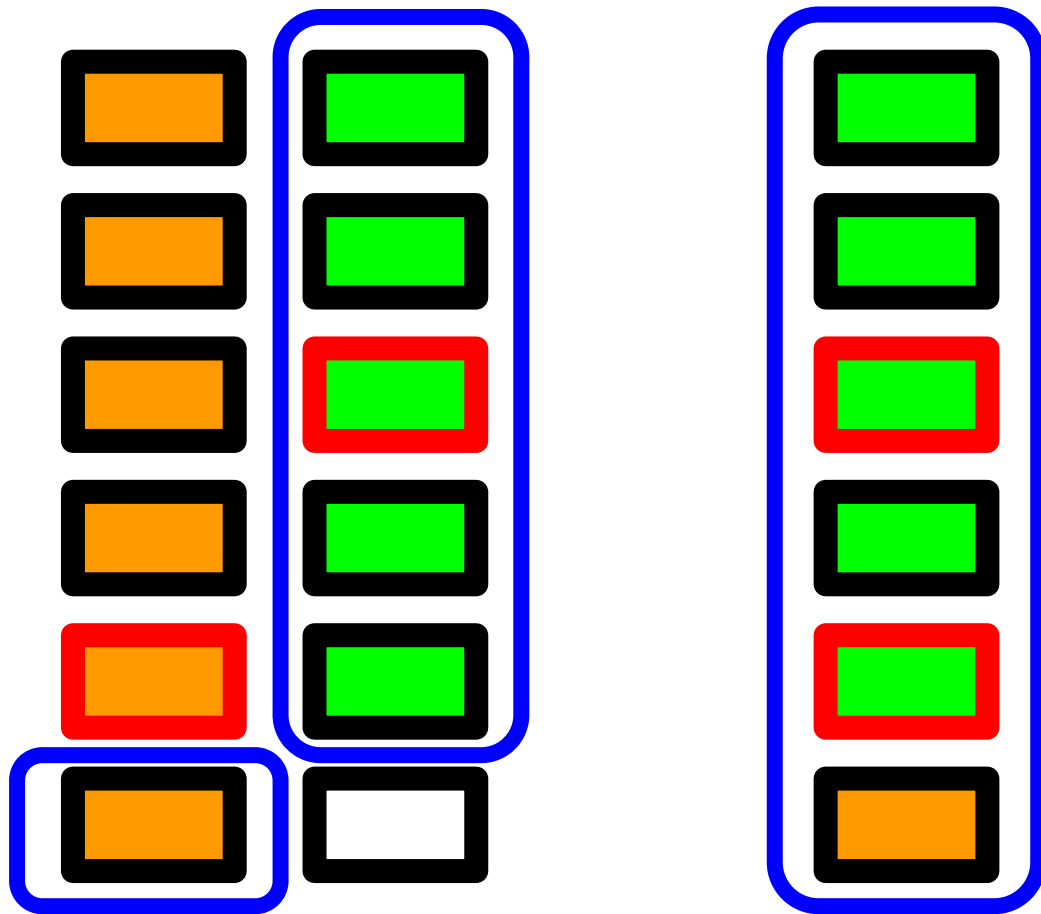


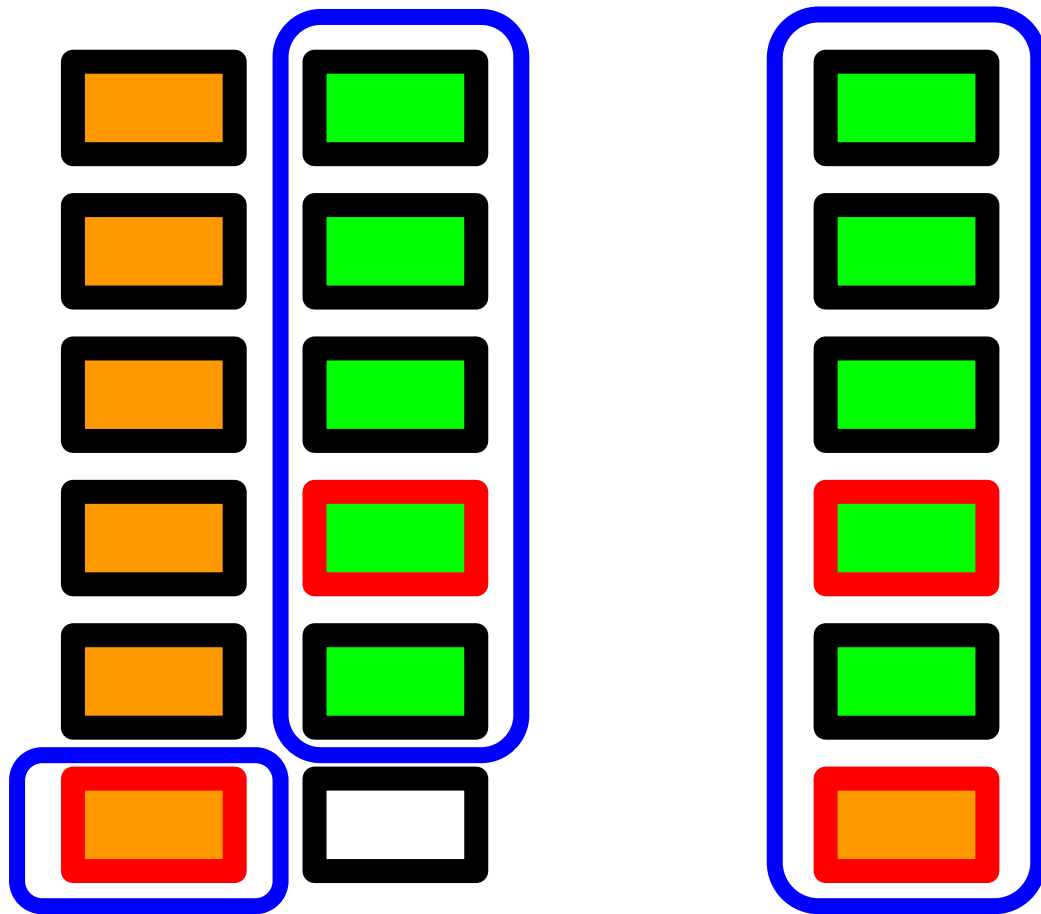


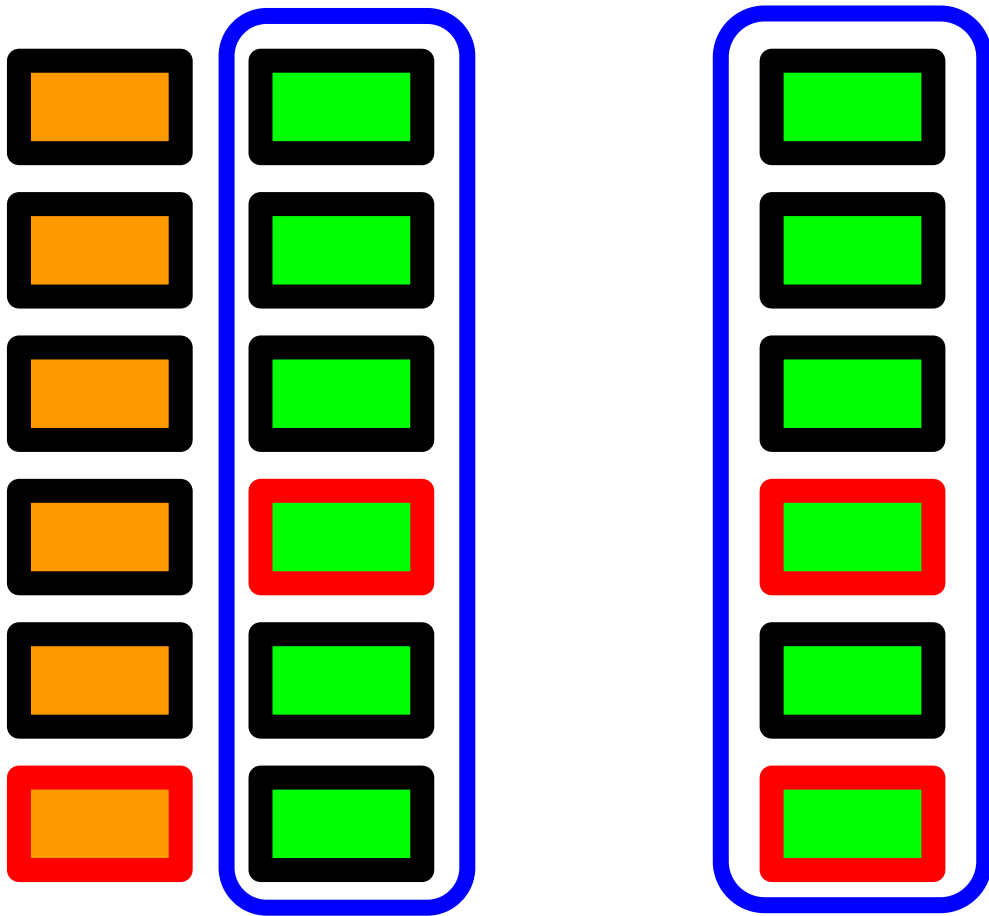




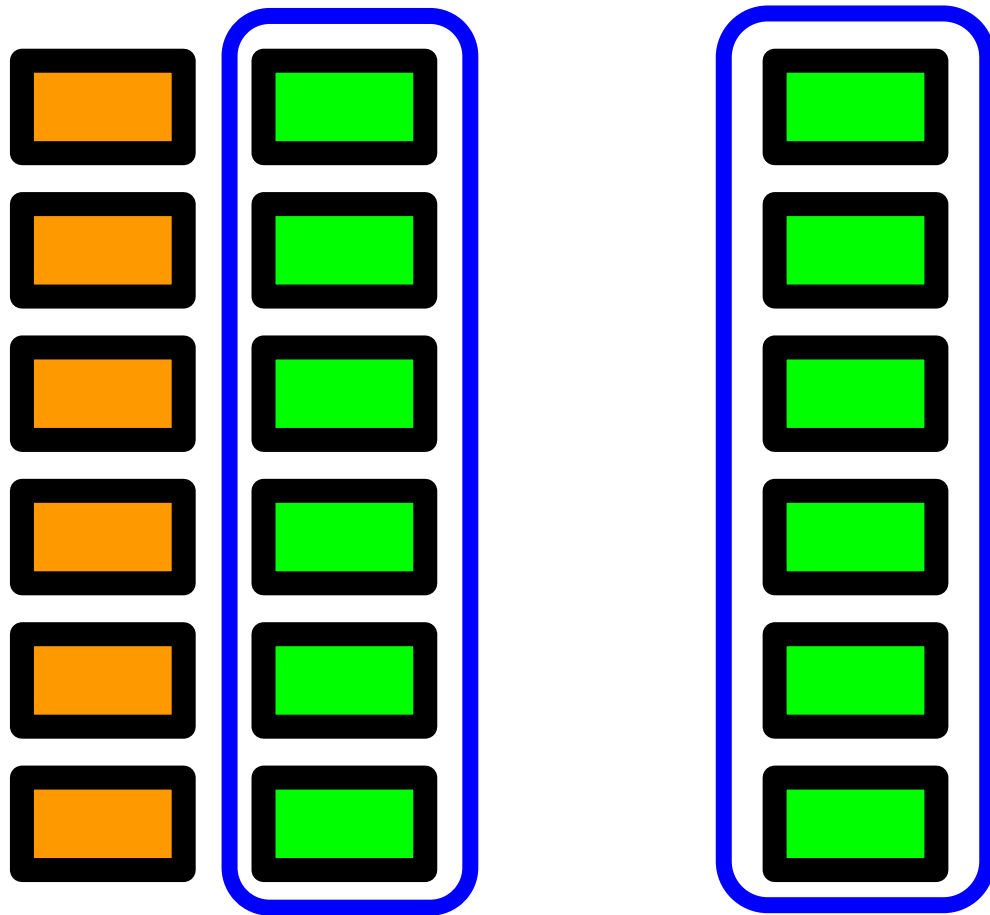








Saves space!
Improves space
complexity



Final Approach

```
int knapsack(const Item *items_begin, const Item *items_end,
            const int capacity) {
    int *f = calloc(capacity + 1, sizeof(int));
    memset(f, 0, (capacity + 1) * sizeof(int));
    for (const Item *item = items_begin; item != items_end; ++item) {
        for (int i = item->weight; i <= capacity; ++i) {
            f[i] = max(f[i], f[i - item->weight] + item->profit);
        }
    }
    int ans = f[capacity];
    return ans;
}
```

Looking closely

	0	1	2
w_i	4	6	8
p_i	7	6	9

```
int knapsack(const Item *items_begin, const Item *items_end,
            const int capacity) {
    int *f = calloc(capacity + 1, sizeof(int));
    memset(f, 0, (capacity + 1) * sizeof(int));
    for (const Item *item = items_begin; item != items_end; ++item) {
        for (int i = item->weight; i <= capacity; ++i) {
            f[i] = max(f[i], f[i - item->weight] + item->profit);
        }
    }
    int ans = f[capacity];
    return ans;
}
```

Choose j^{th} element

$$P(C) = \max_{j=0}^{n-1} \{P(C - w_j) + p_j\}$$

P(14)

P(13)

P(12)

P(11)

P(10)

P(9)

P(8)

P(7)

P(6)

P(5)

P(4)

P(3)

P(2)

P(1)

P(0)