



<https://hao-ai-lab.github.io/dsc204a-w24/>

DSC 204A: Scalable Data Systems

Winter 2024

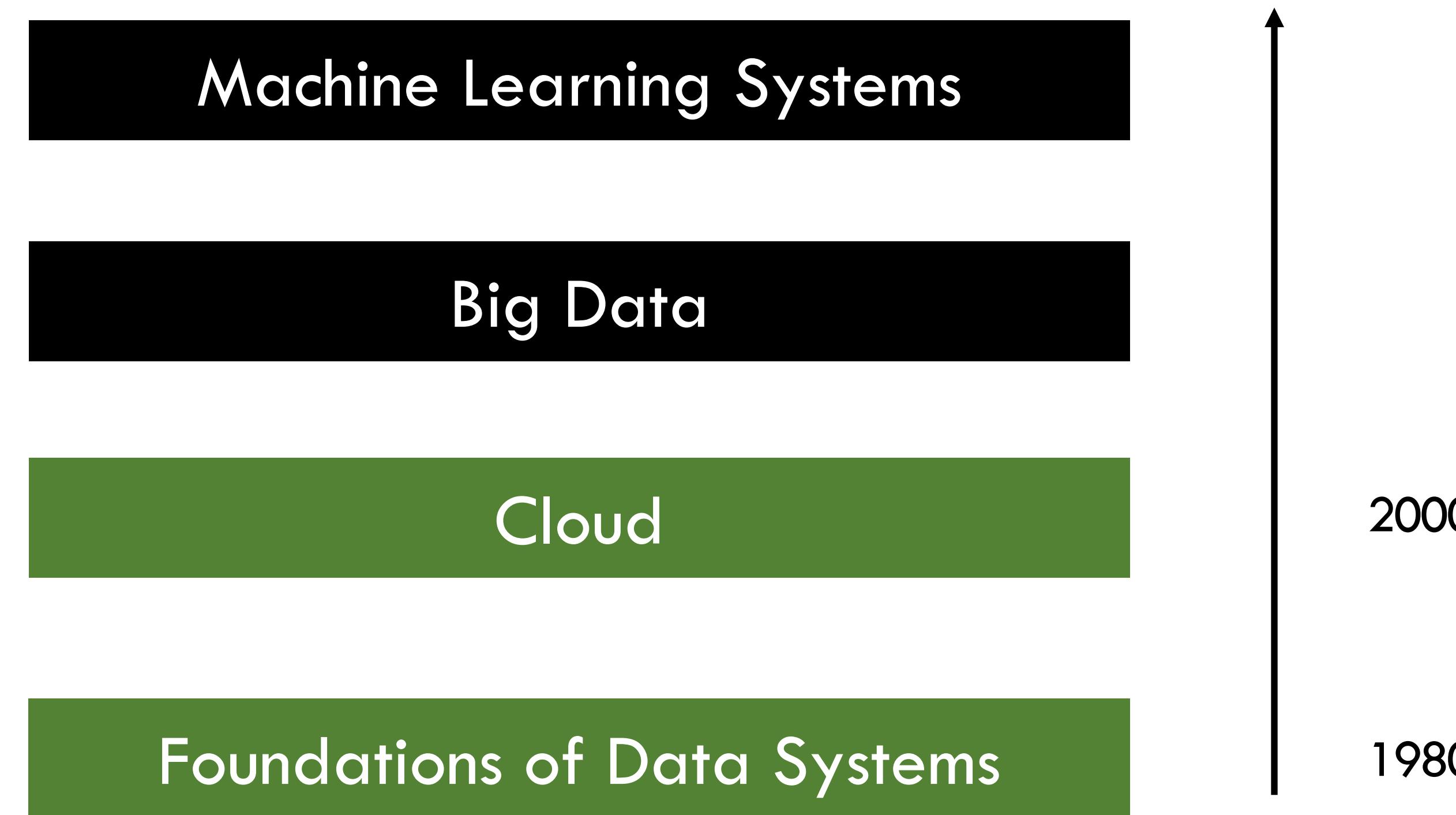
Machine Learning Systems

Big Data

Cloud

Foundations of Data Systems

Where We Are



Today's topic

- Collective communication
 - Connection between distributed SGD and collective comm
- Communication Model: $\alpha + n\beta, \beta = \frac{1}{B}$
 - Small Message size ($n \rightarrow 0$): α dominates, emphasize latency
 - Large Message Size ($n \rightarrow +\infty$): $n\beta$ dominate, emphasize bandwidth utilization

Recap: Minimum Spanning Tree Algorithm

Reduce(-to-one)

$$\log(p)(\alpha + n\beta + n\gamma)$$

Scatter

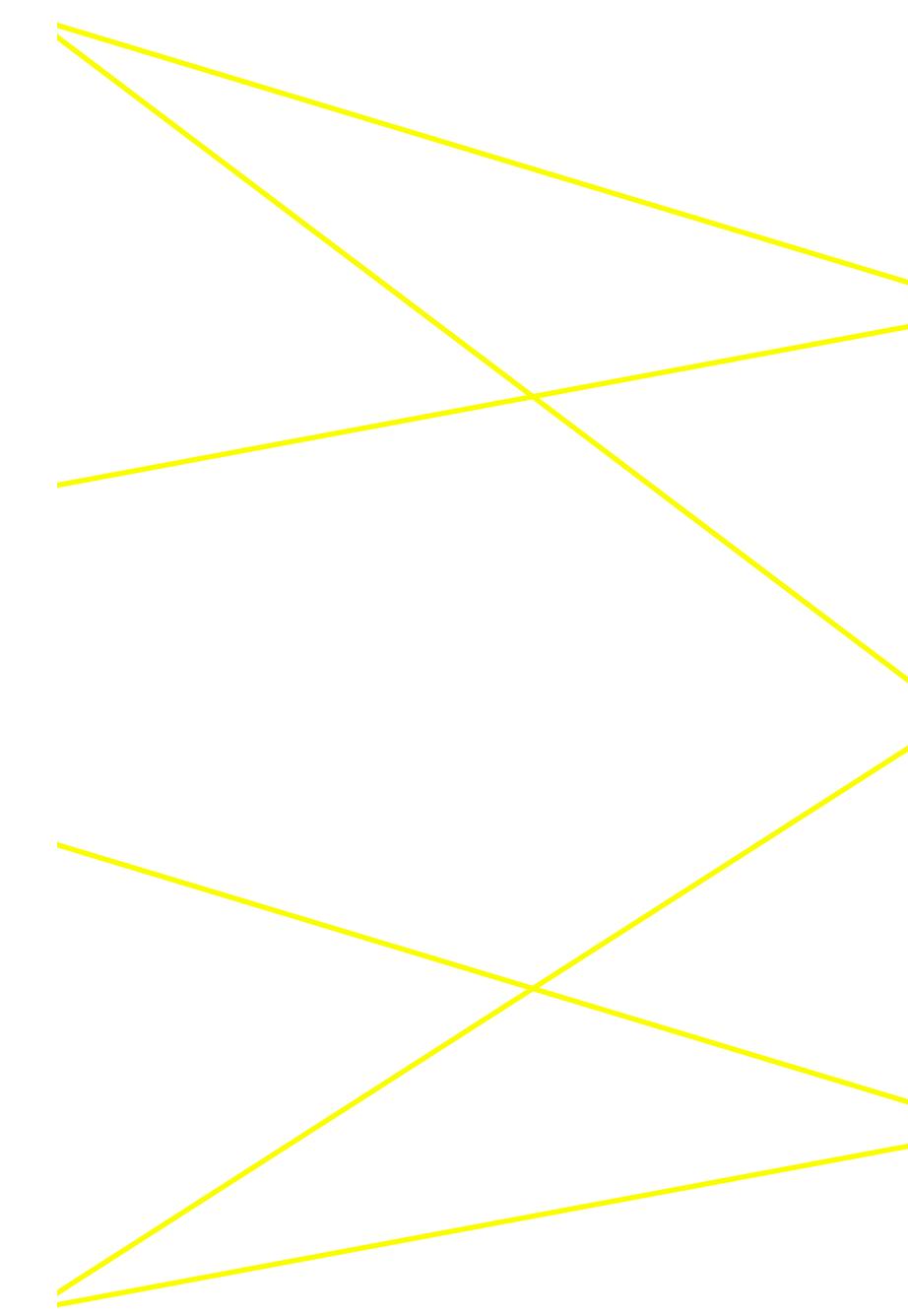
$$\log(p)\alpha + \frac{p-1}{p}n\beta$$

Gather

$$\log(p)\alpha + \frac{p-1}{p}n\beta$$

Broadcast

$$\log(p)(\alpha + n\beta)$$



Reduce-scatter

$$2\log(p)\alpha + \log(p)n(\beta + \gamma) + \frac{p-1}{p}n\beta$$

Allreduce

$$2\log(p)\alpha + \log(p)n(2\beta + \gamma)$$

Allgather

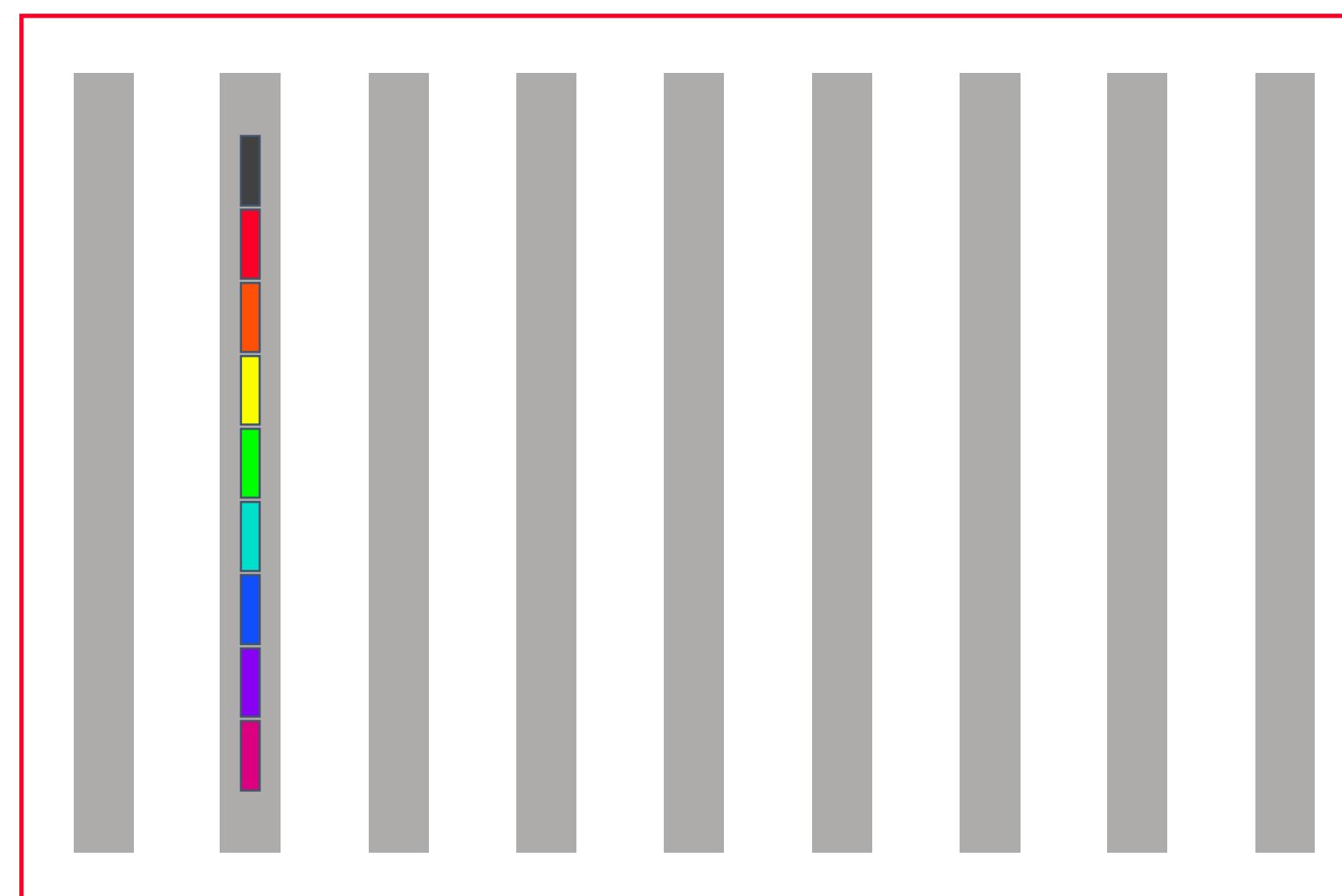
$$2\log(p)\alpha + \log(p)n\beta + \frac{p-1}{p}n\beta$$

Pros and Cons of MST algorithms

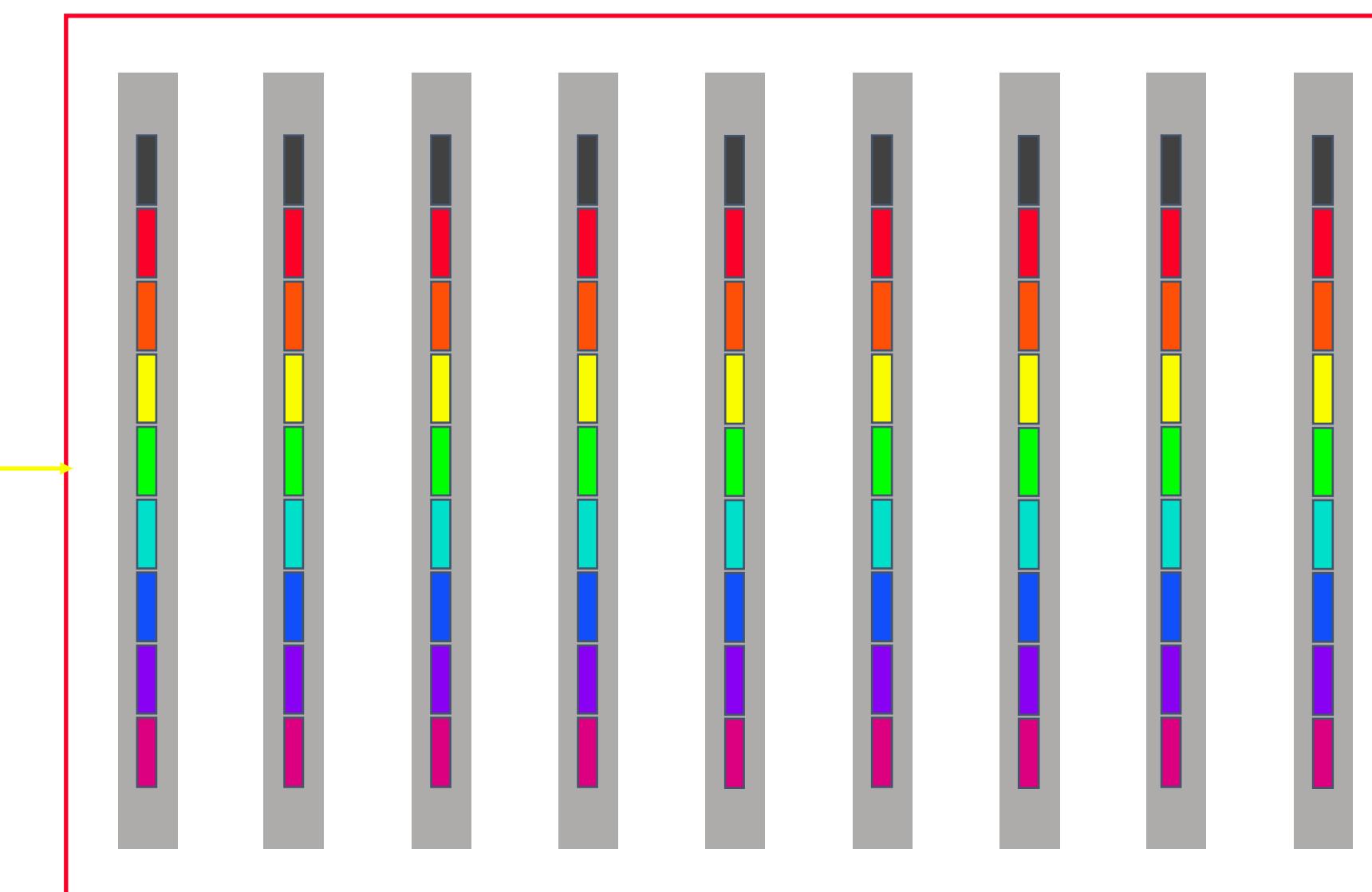
- Emphasize **low latency**
 - MST-based algorithm is latency-optimal
 - **How to prove?** (Taking broadcast as an example)
- Problem of Minimum Spanning Tree Algorithm?
 - Some links are idle

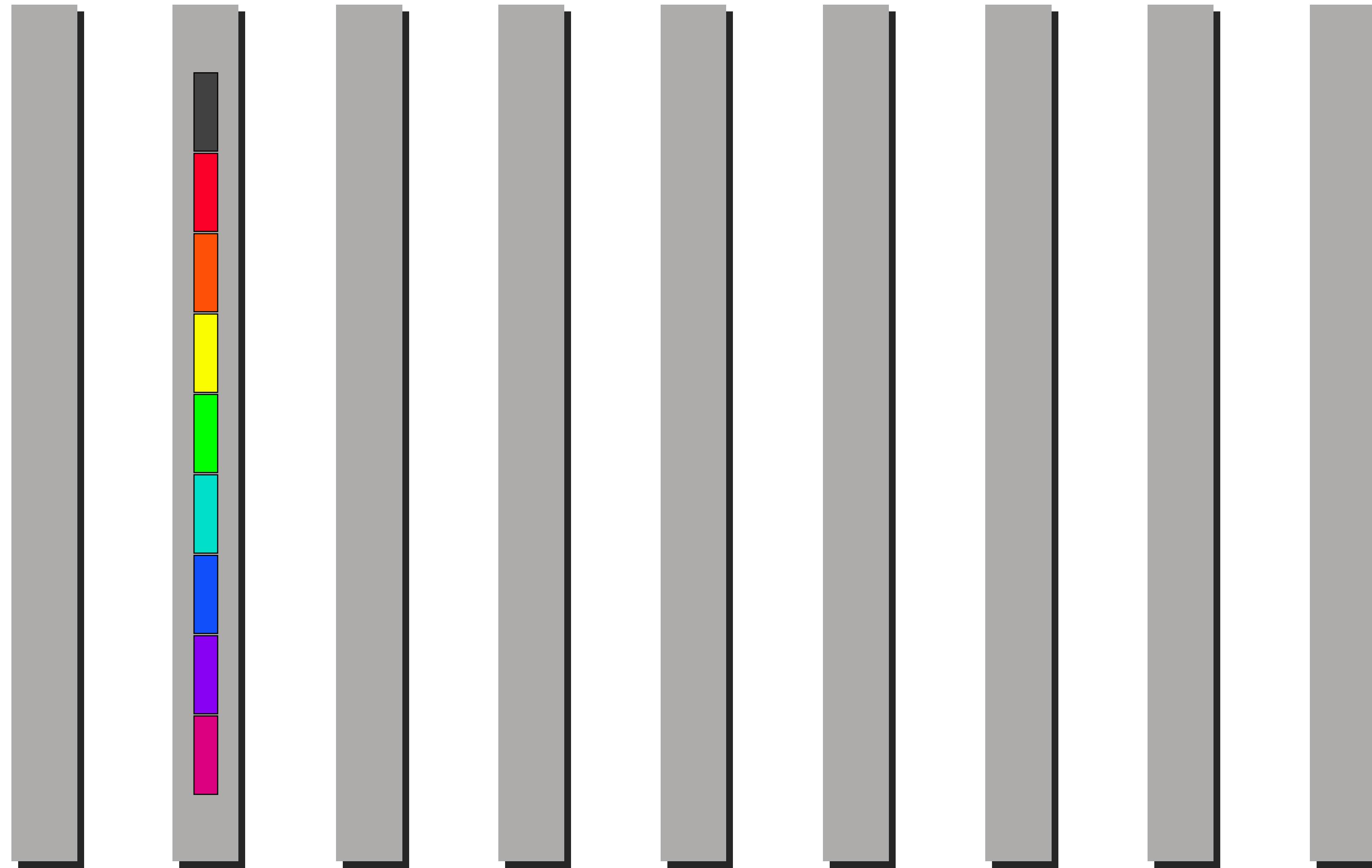
Broadcast

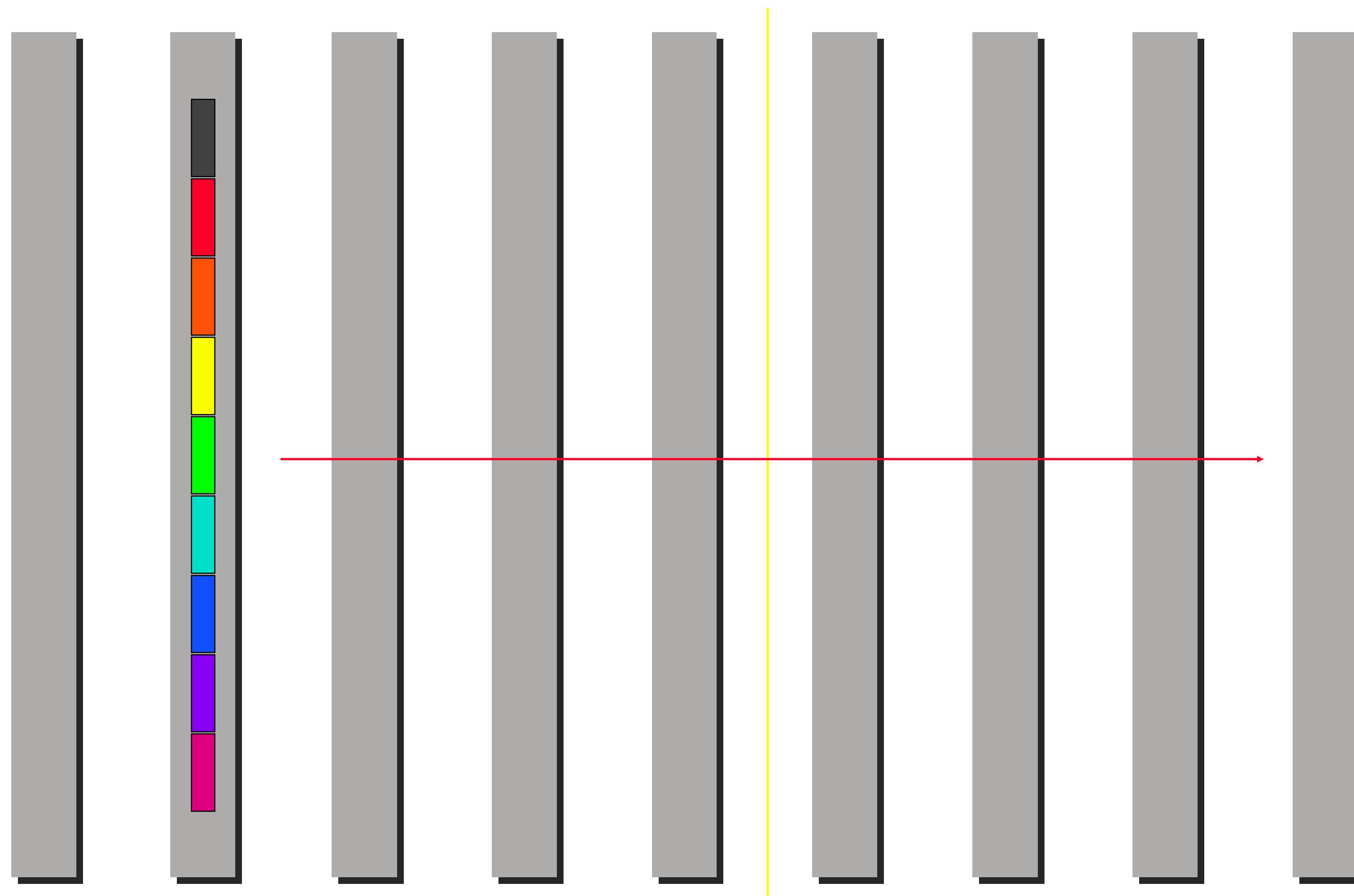
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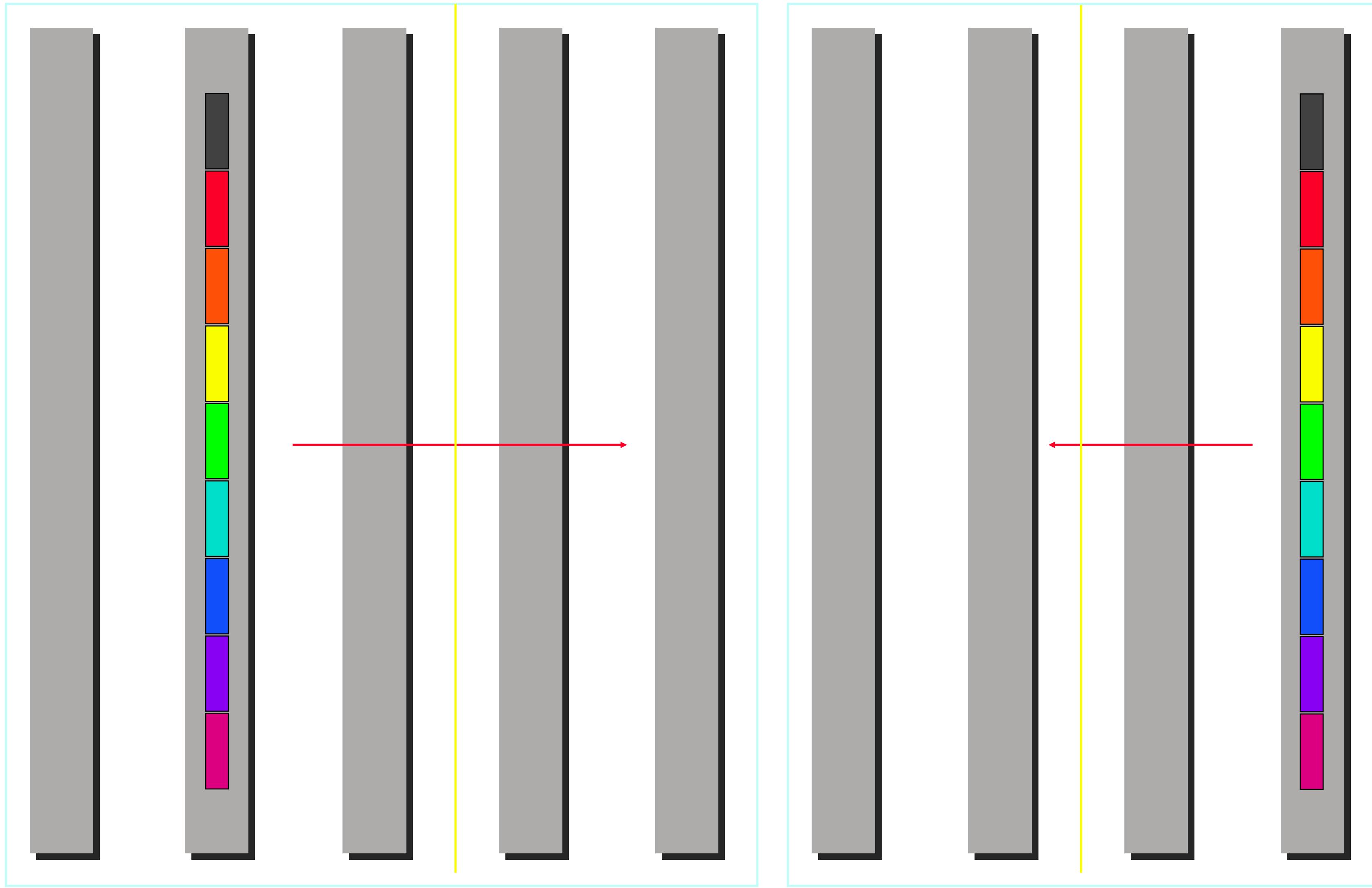


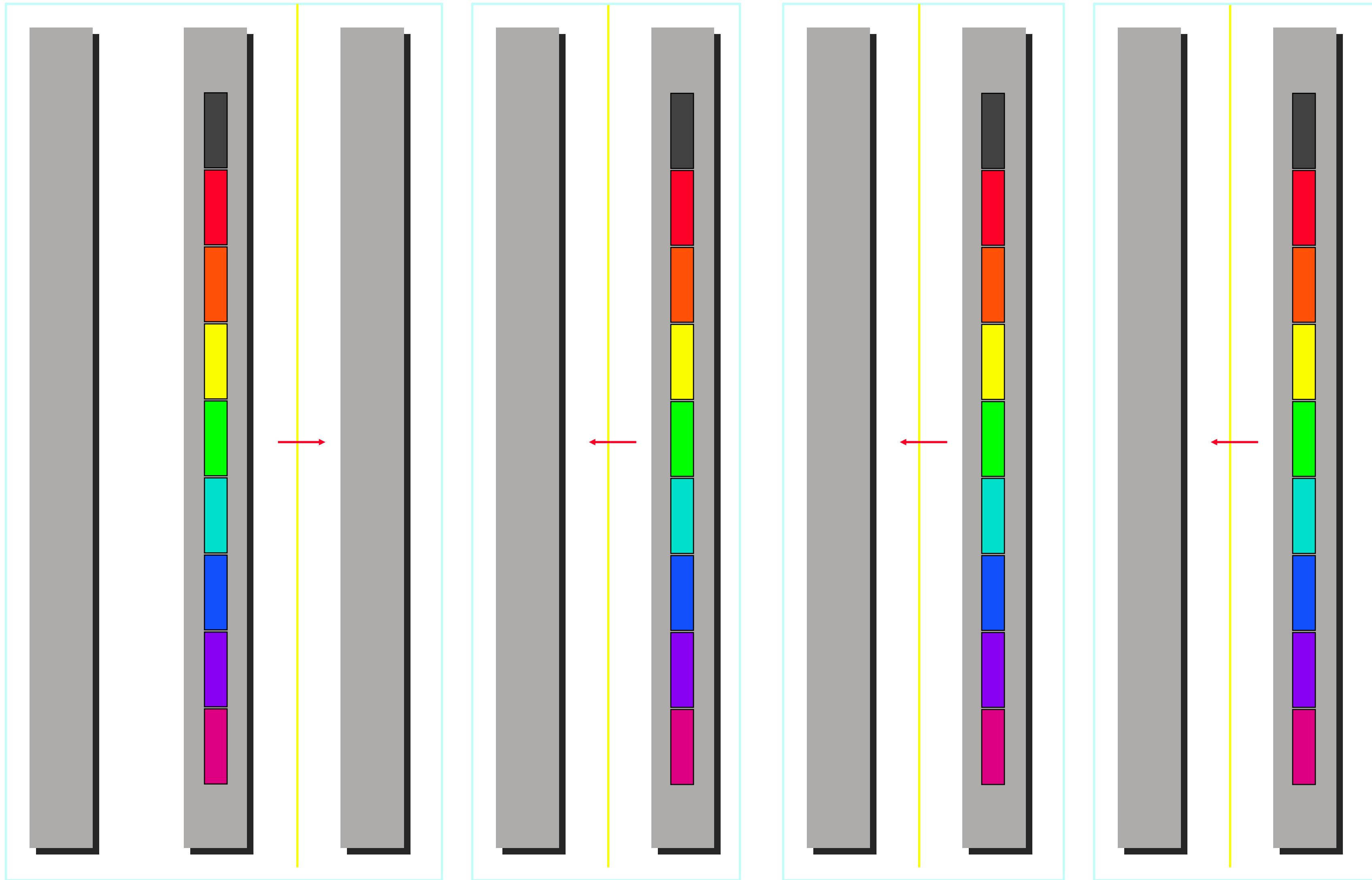
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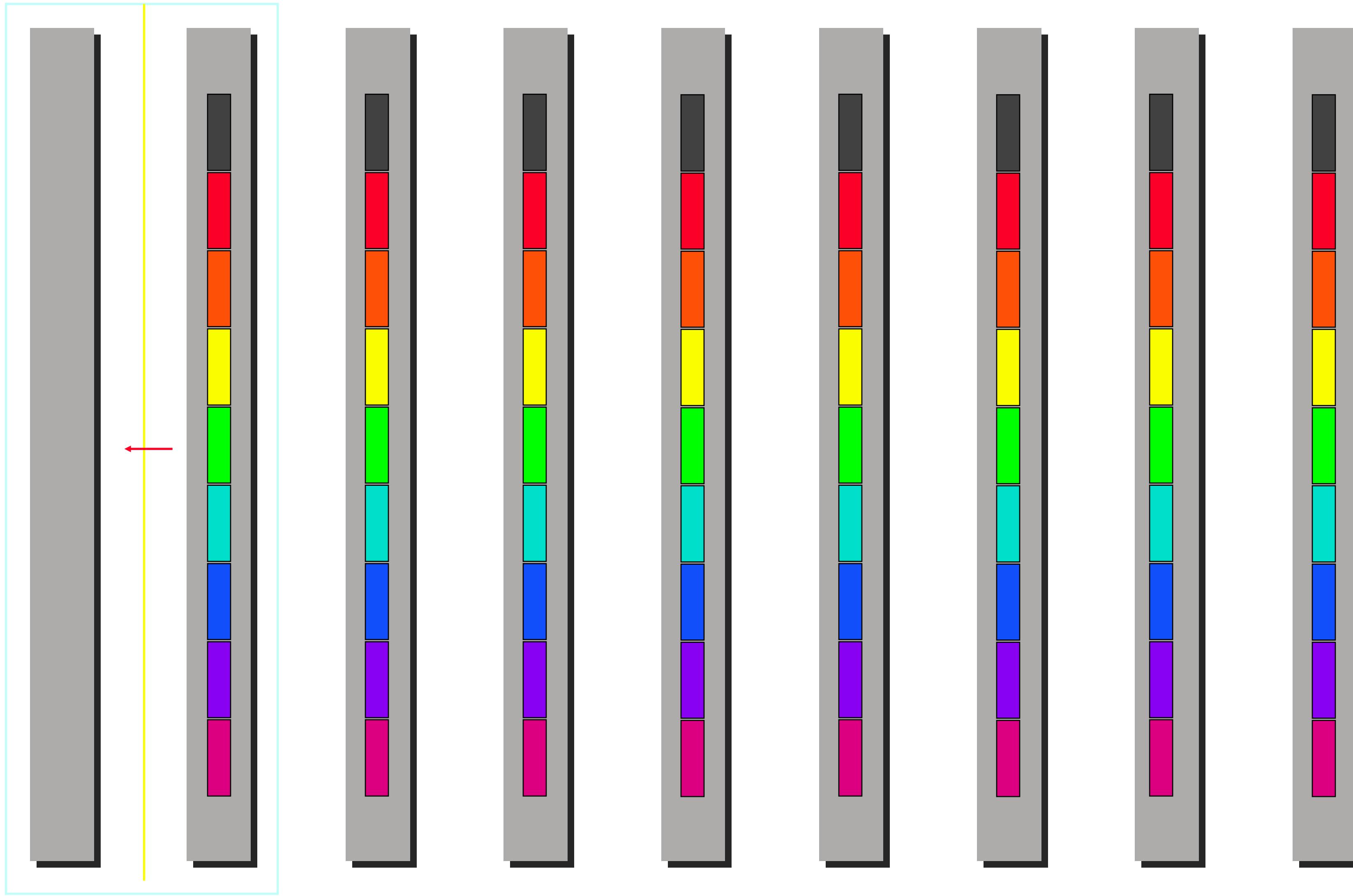


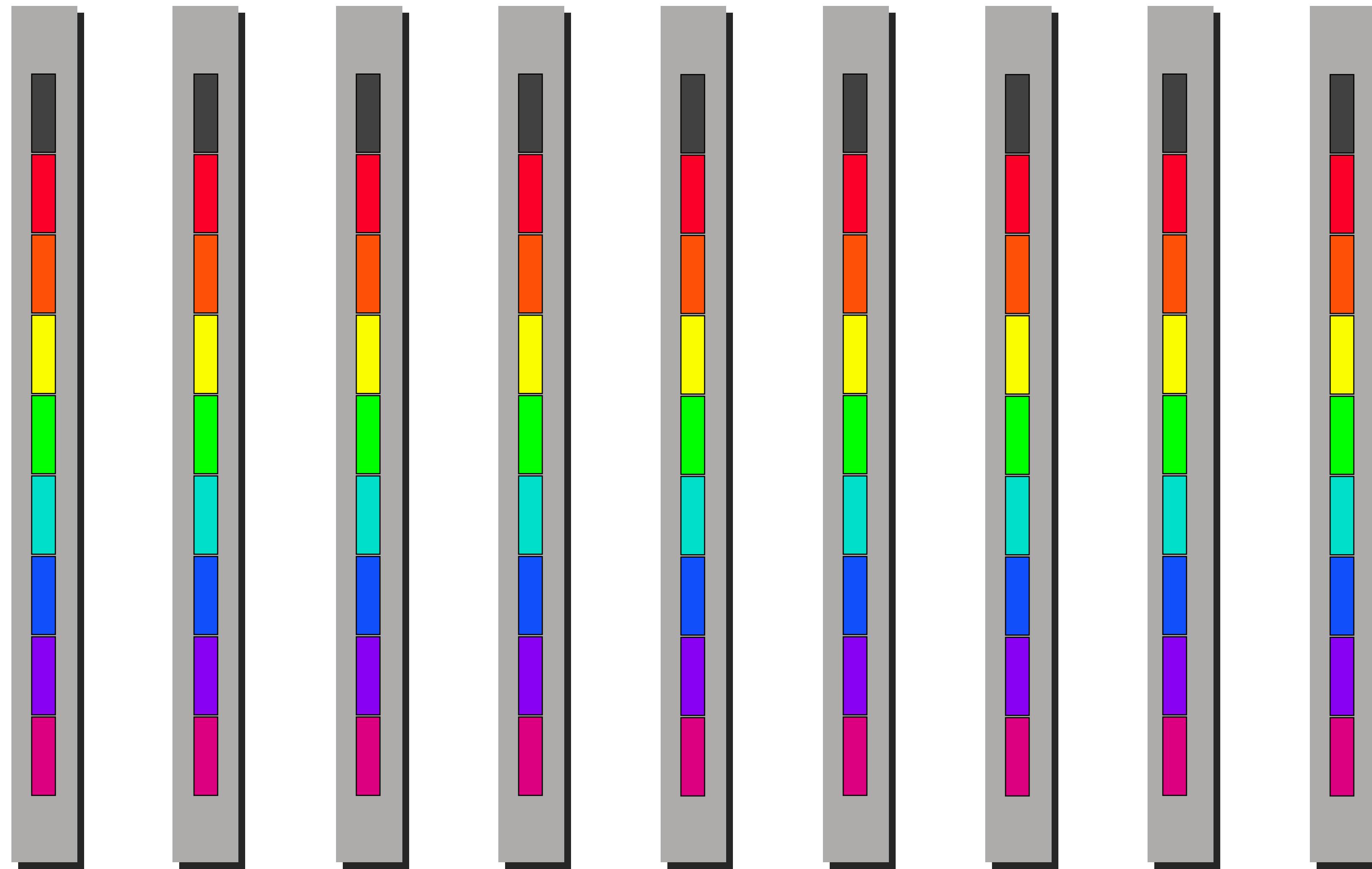












Large Message

Communication Model: $\alpha + n\beta, \beta = \frac{1}{B}$

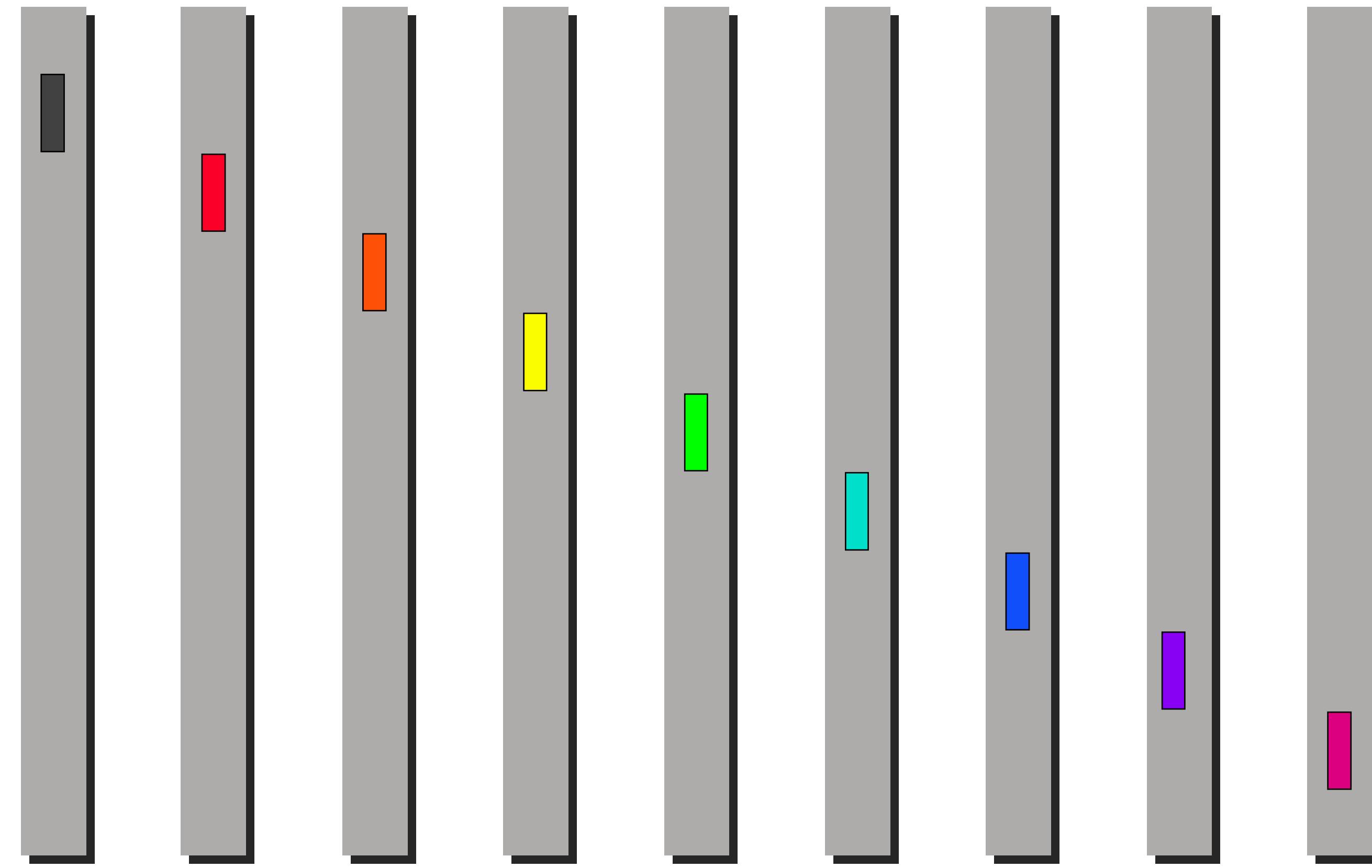
- The second term dominates – we want to minimize the second term
- We want to utilize the bandwidth as much as possible

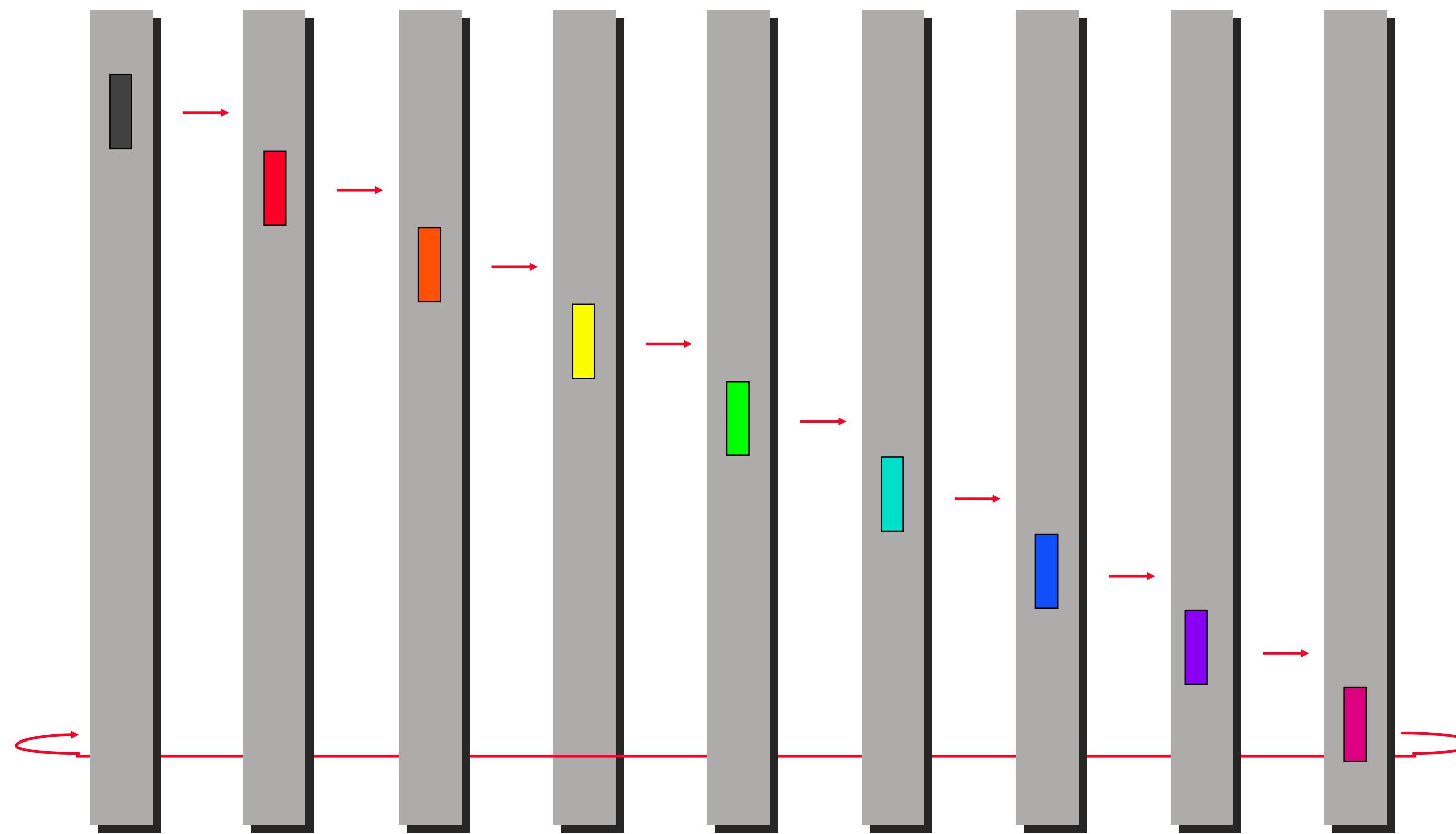
Long vector building blocks

- We will show how the following building blocks:
 - collect/distributed combine
 - scatter/gathercan be implemented using “bucket” algorithms while attaining
 - minimal cost due to length of vectors
 - implementation for arbitrary numbers of nodes
 - no network conflicts
- NOTICE: scatter and gather already satisfy these conditions

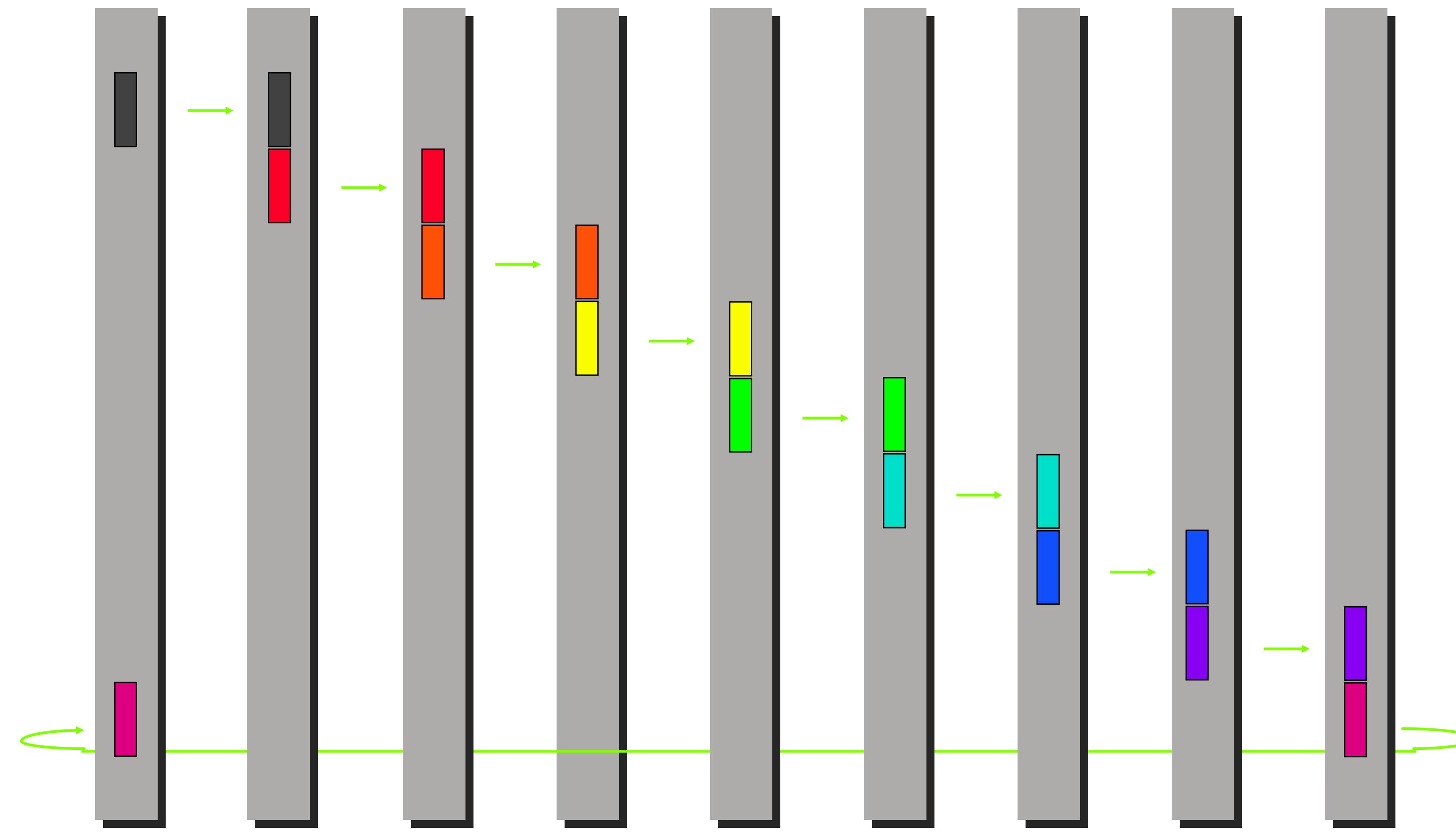
General principles

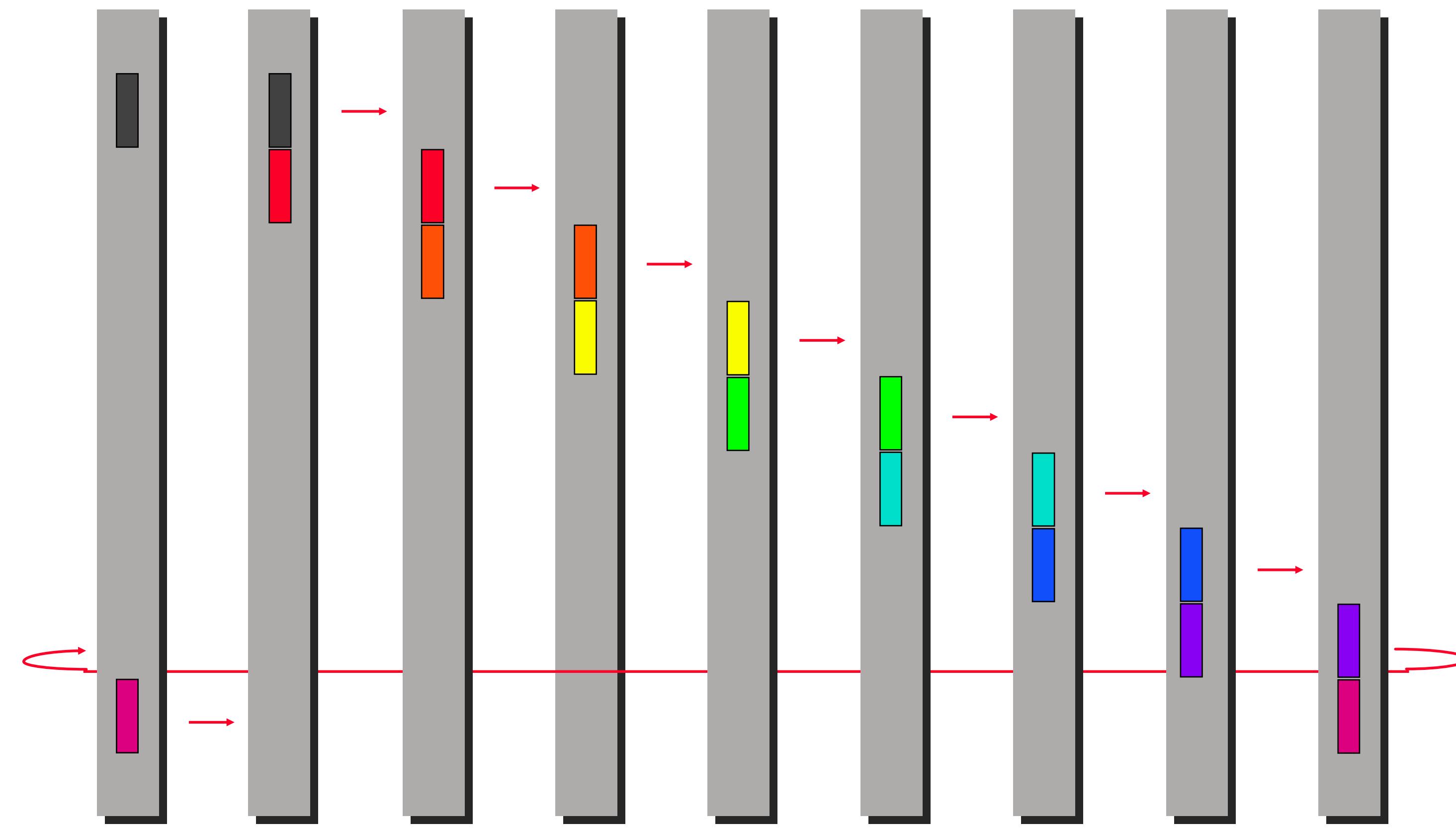
- Use all the links between every two nodes
- A logical ring can be embedded in a physical linear array with worm-hole routing, since the “wrap-around” message doesn’t conflict

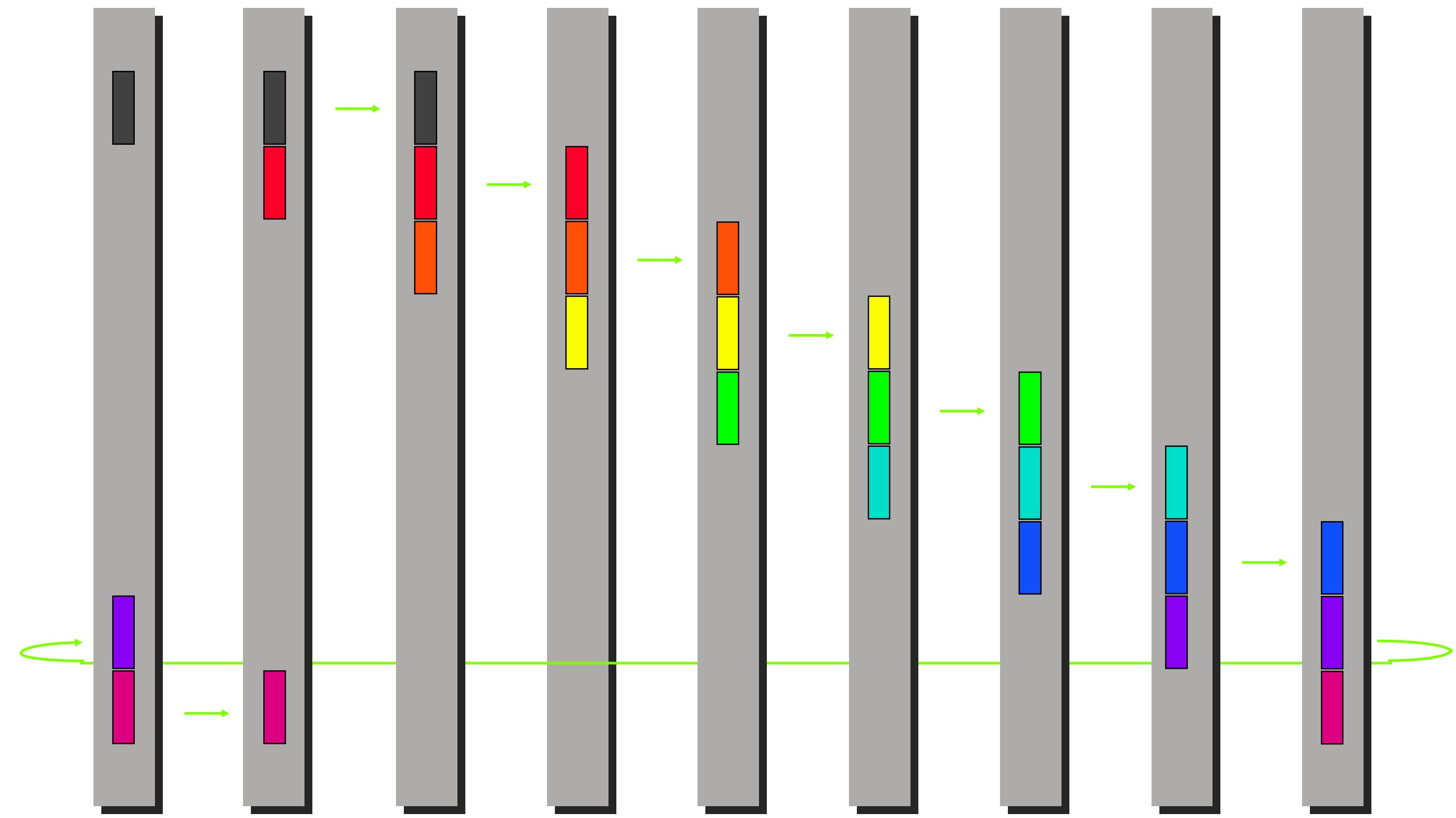




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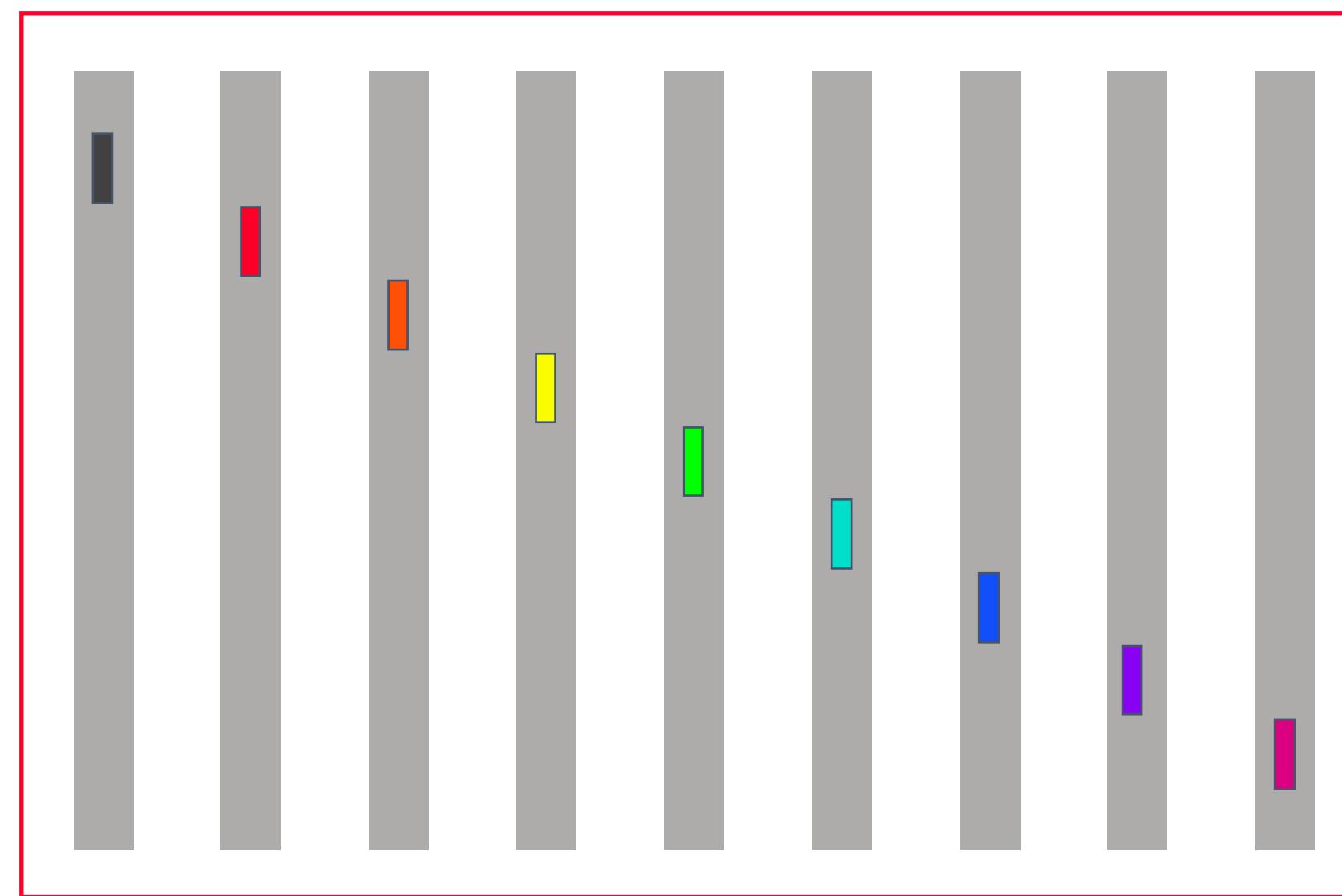
General principles

Ring algorithm has the following advantages

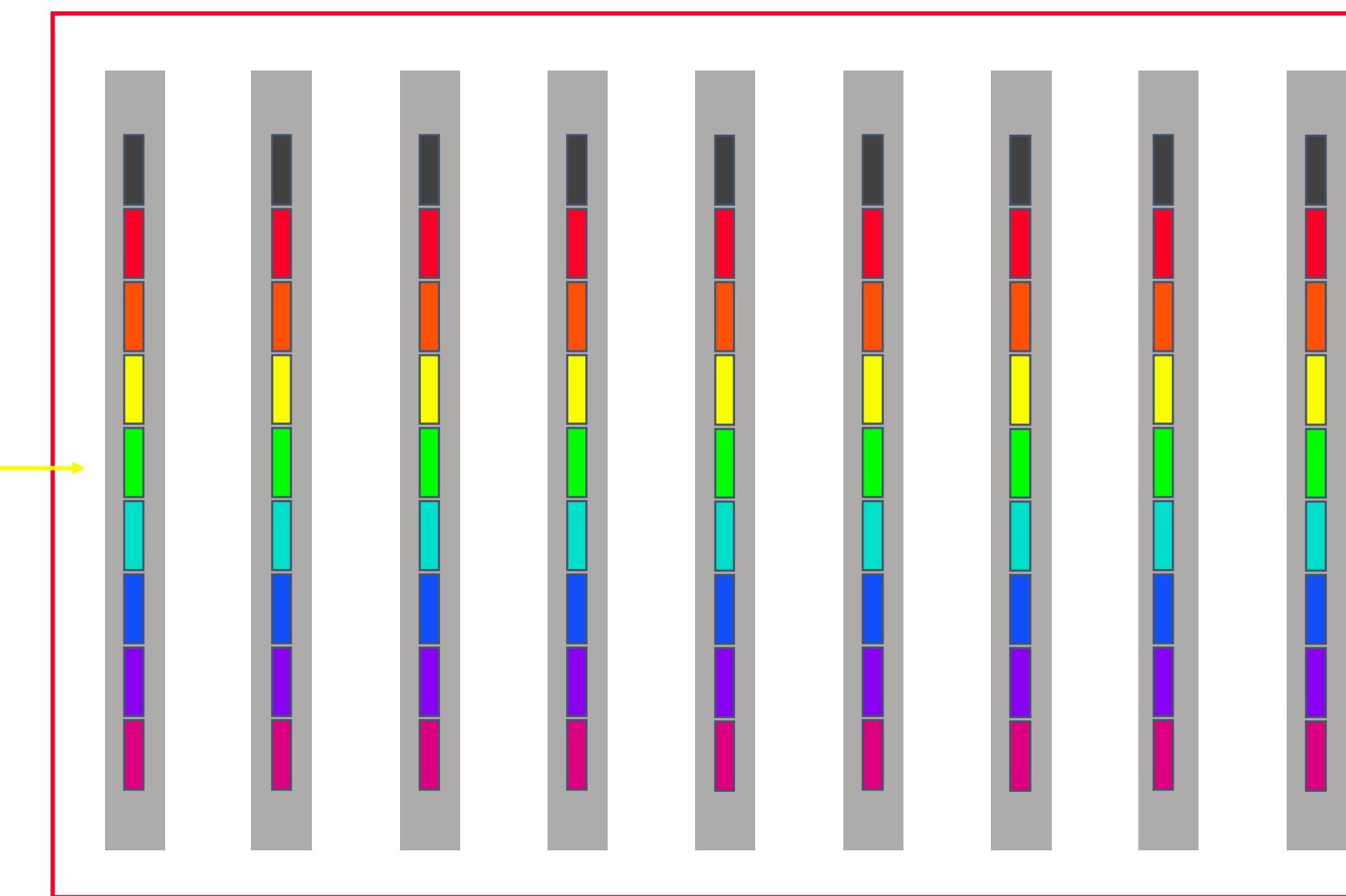
- Fully utilize the bandwidth (bandwidth optimal)
- implementation for arbitrary numbers of node

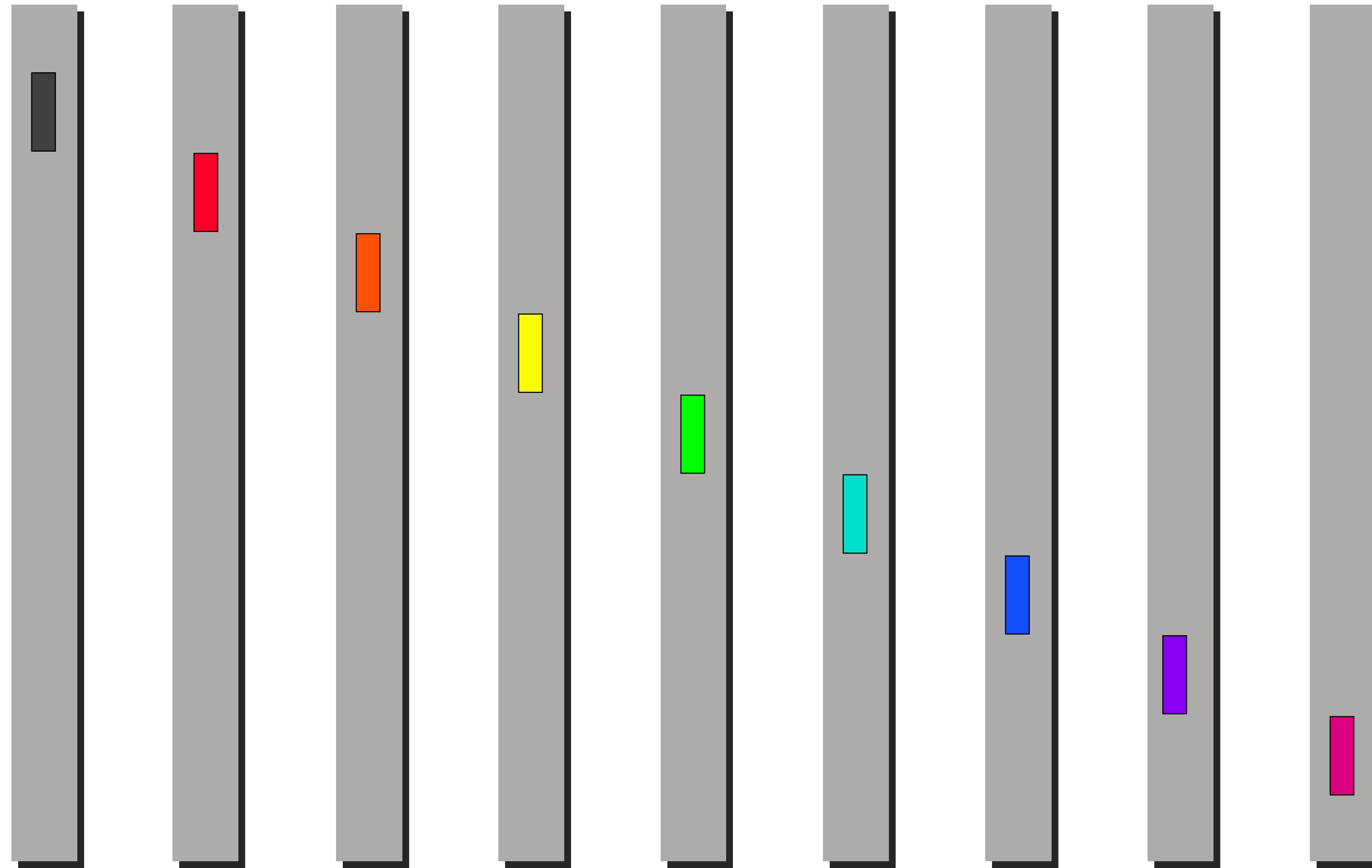
Allgather

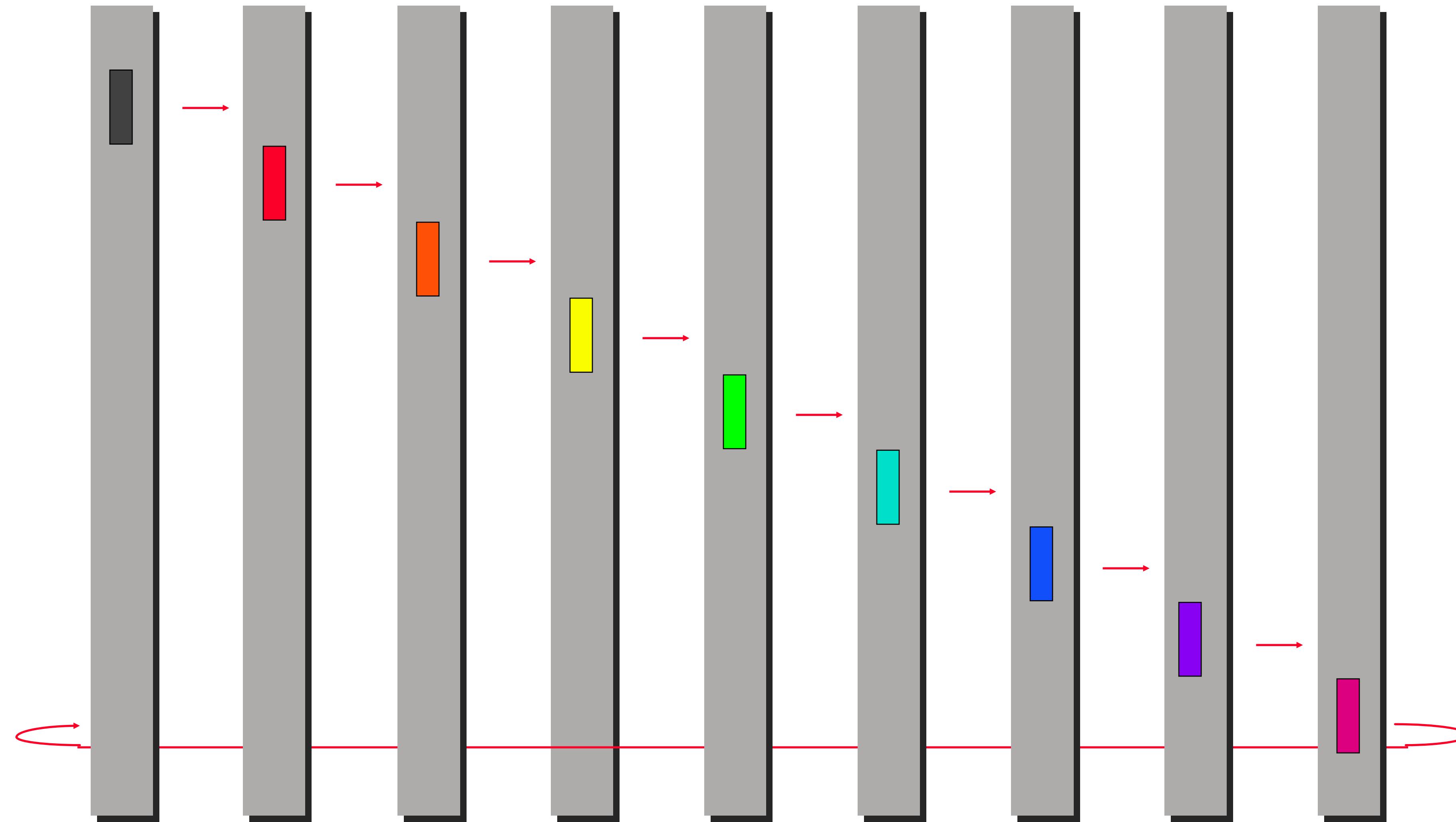
Before

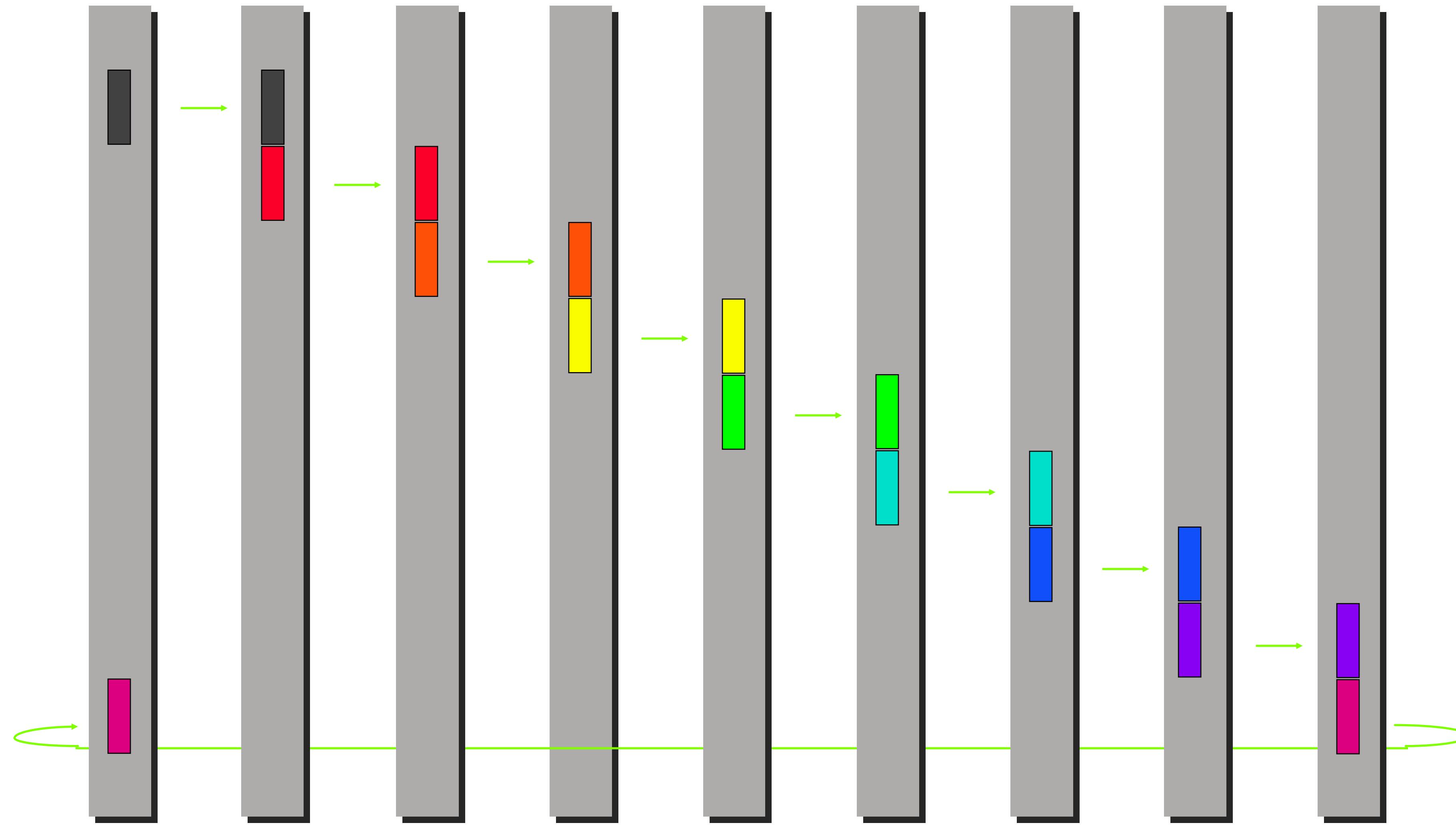


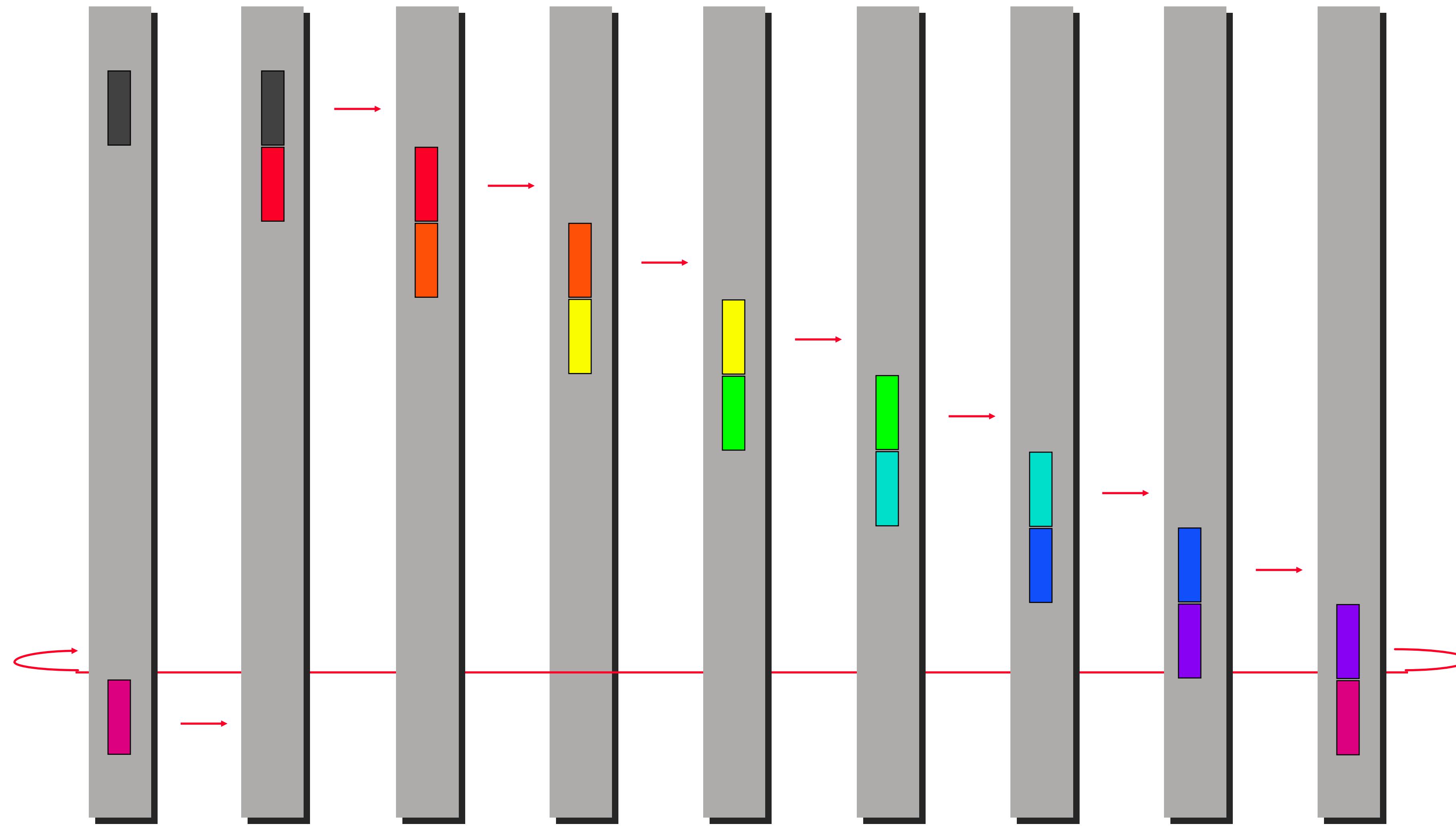
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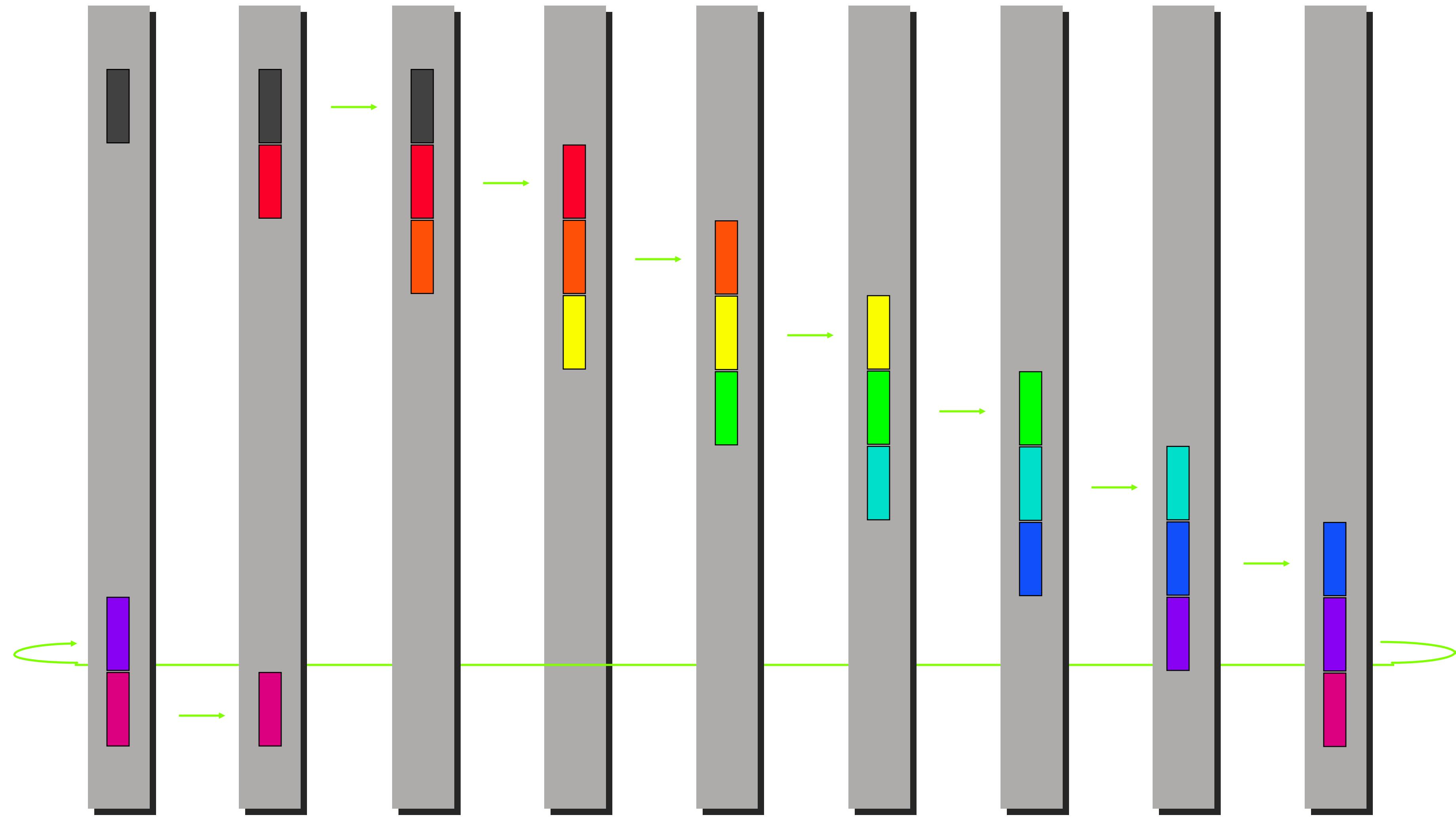


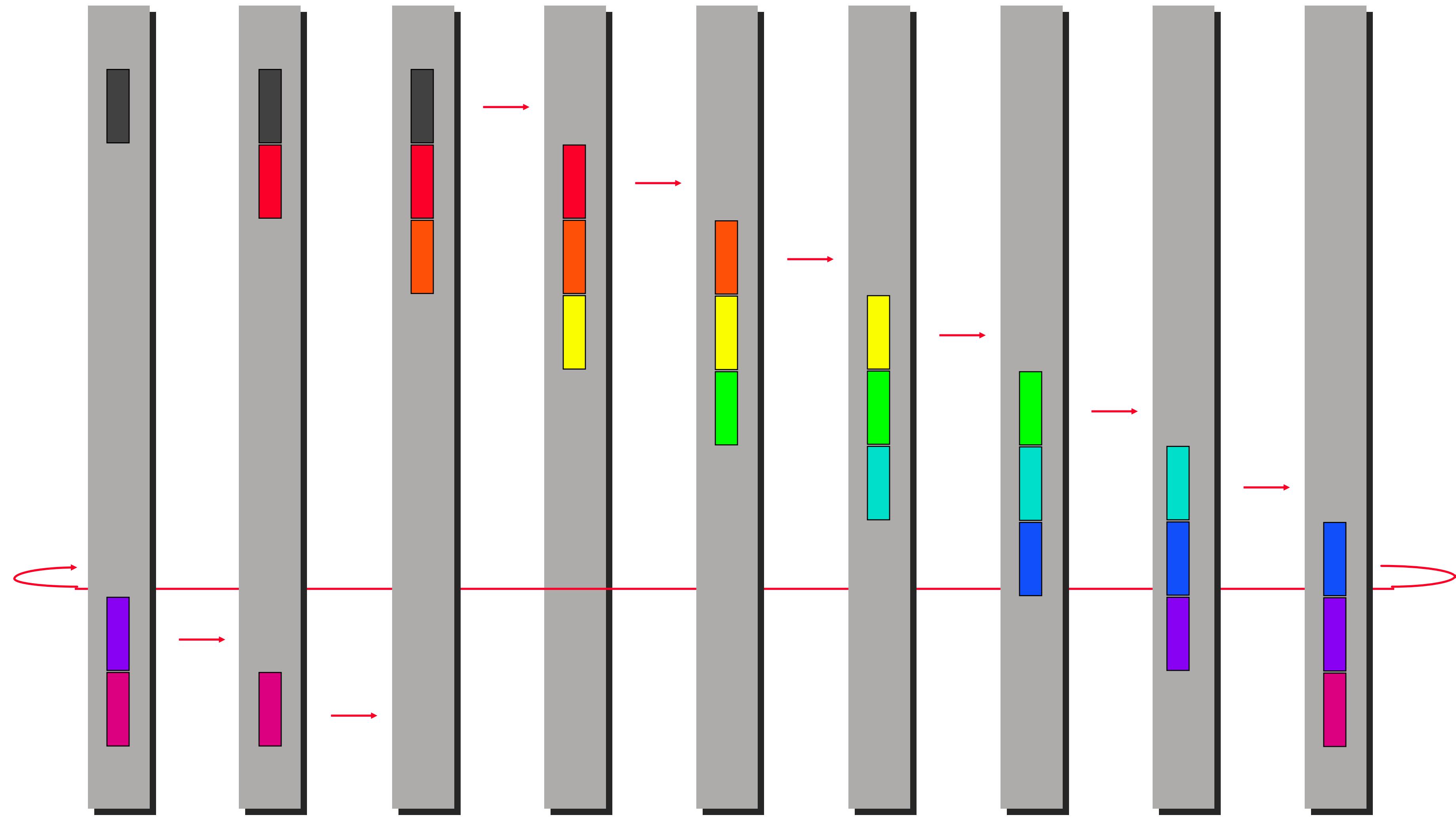


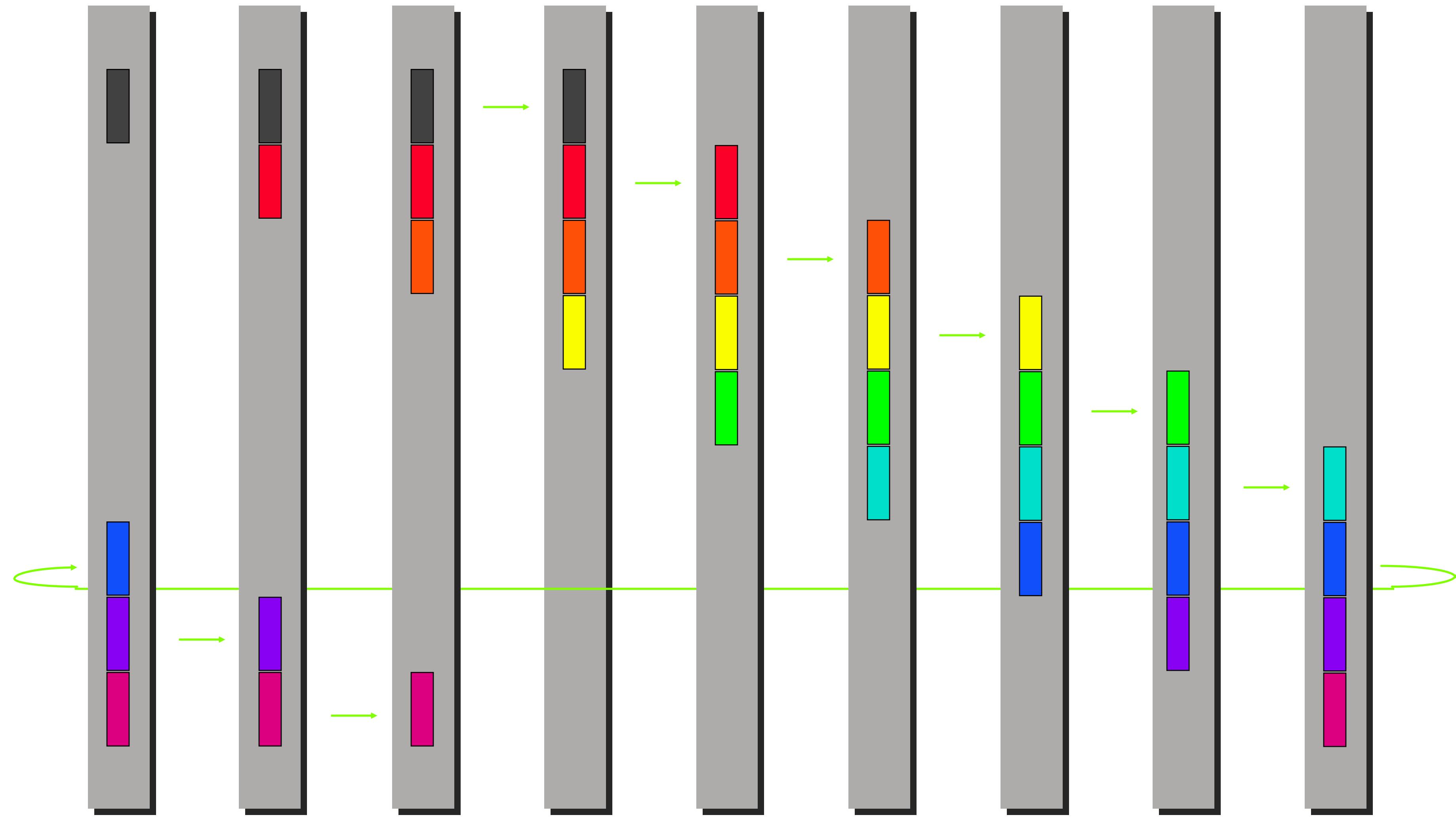


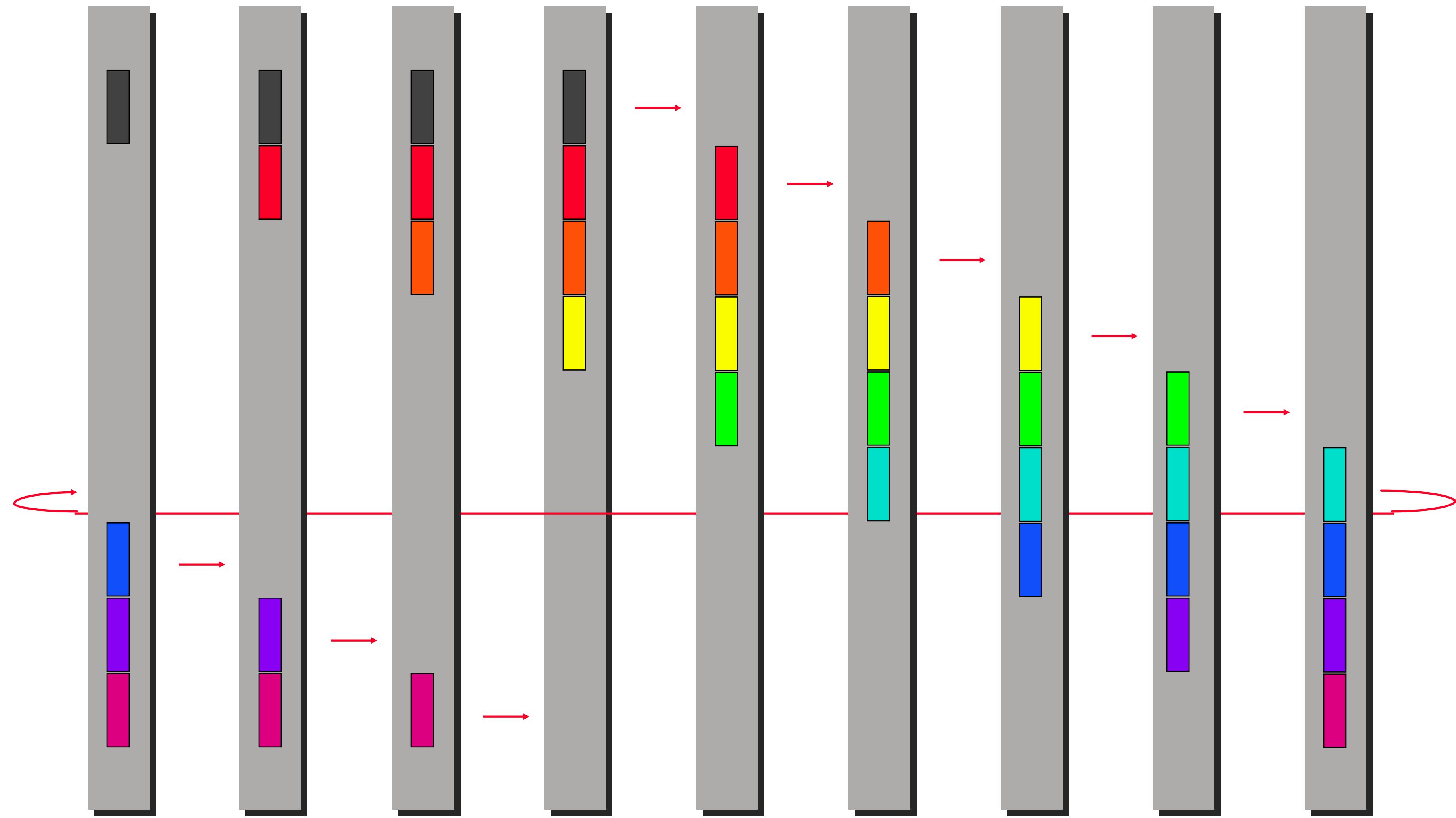


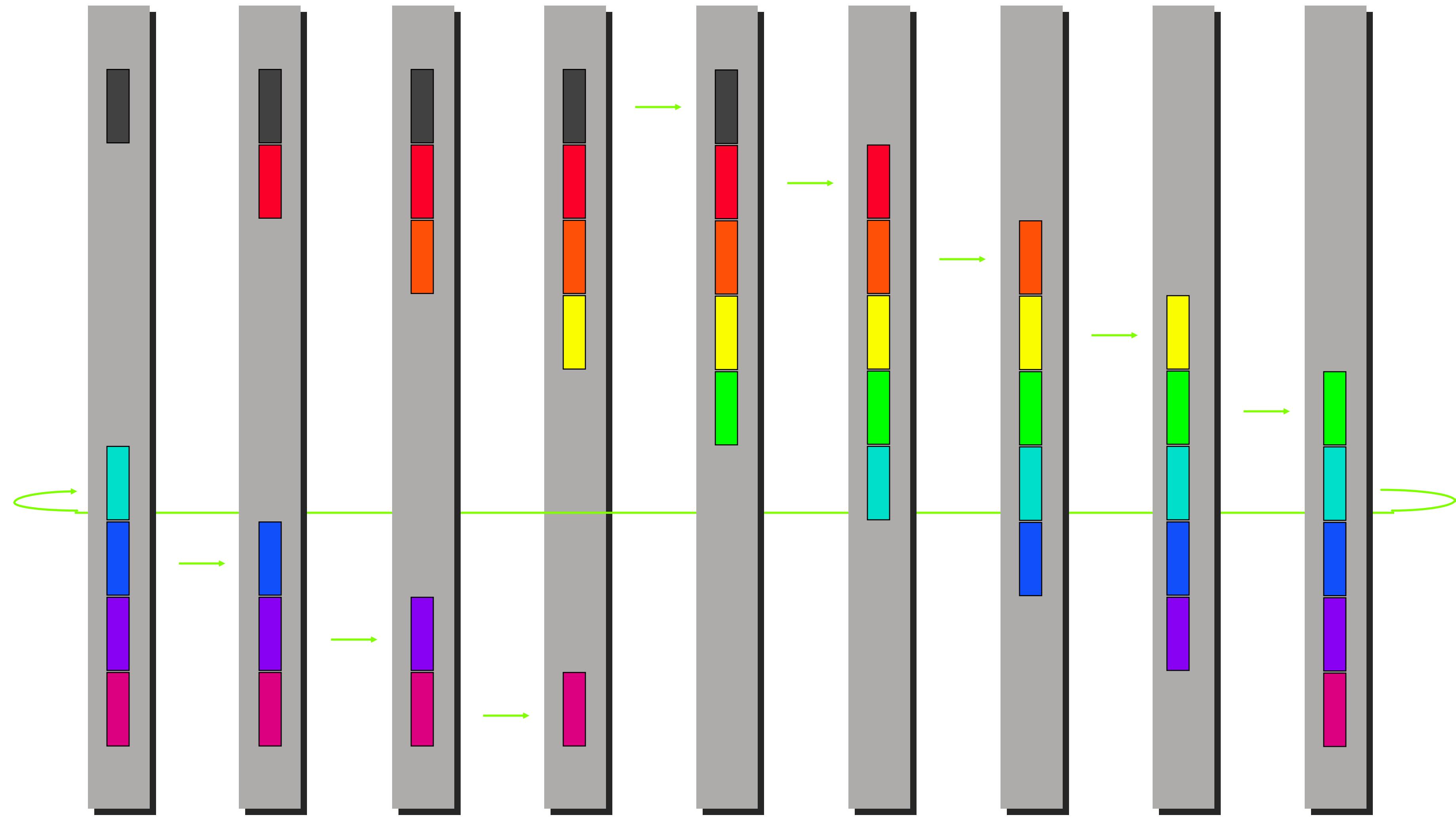


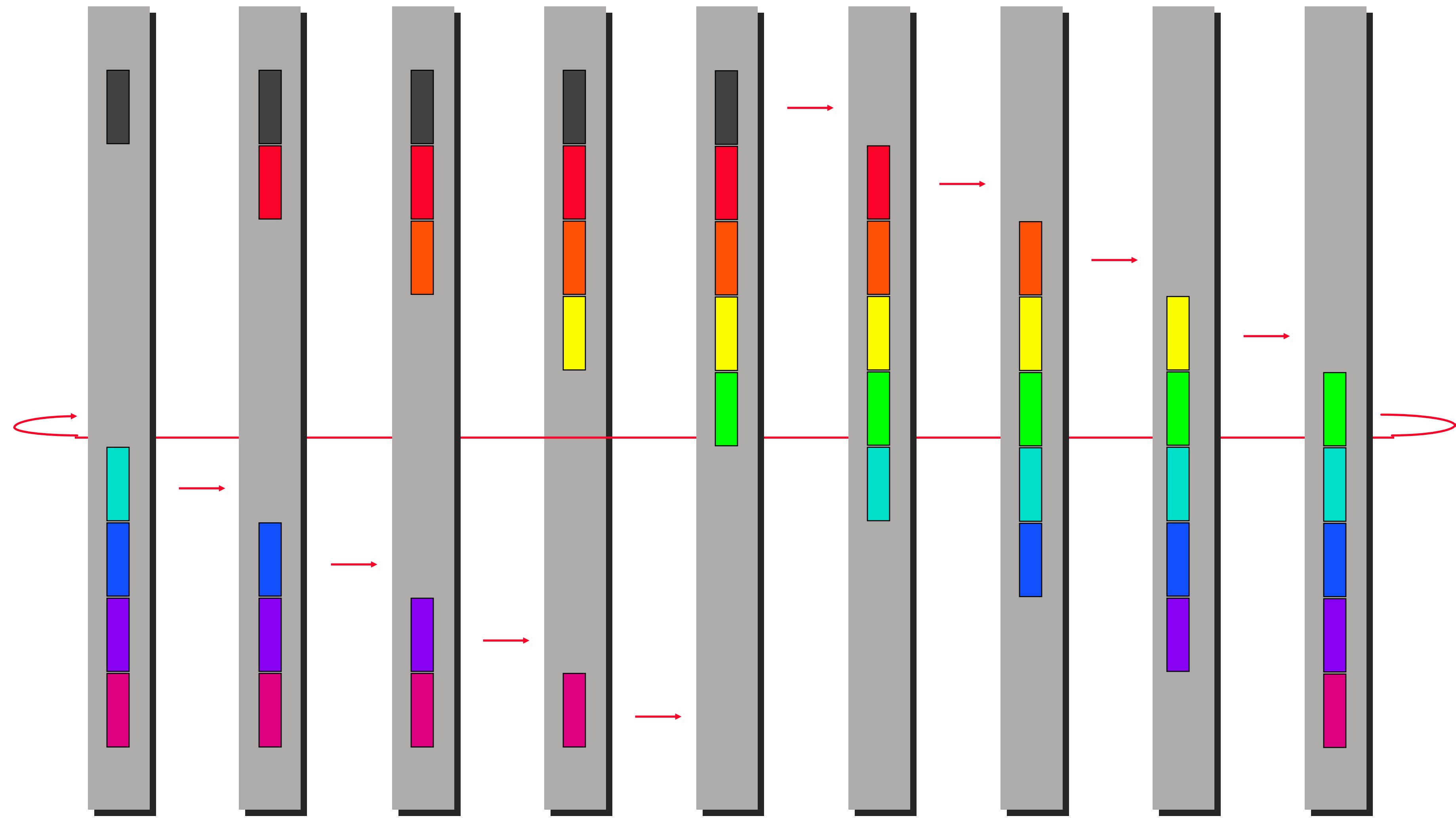


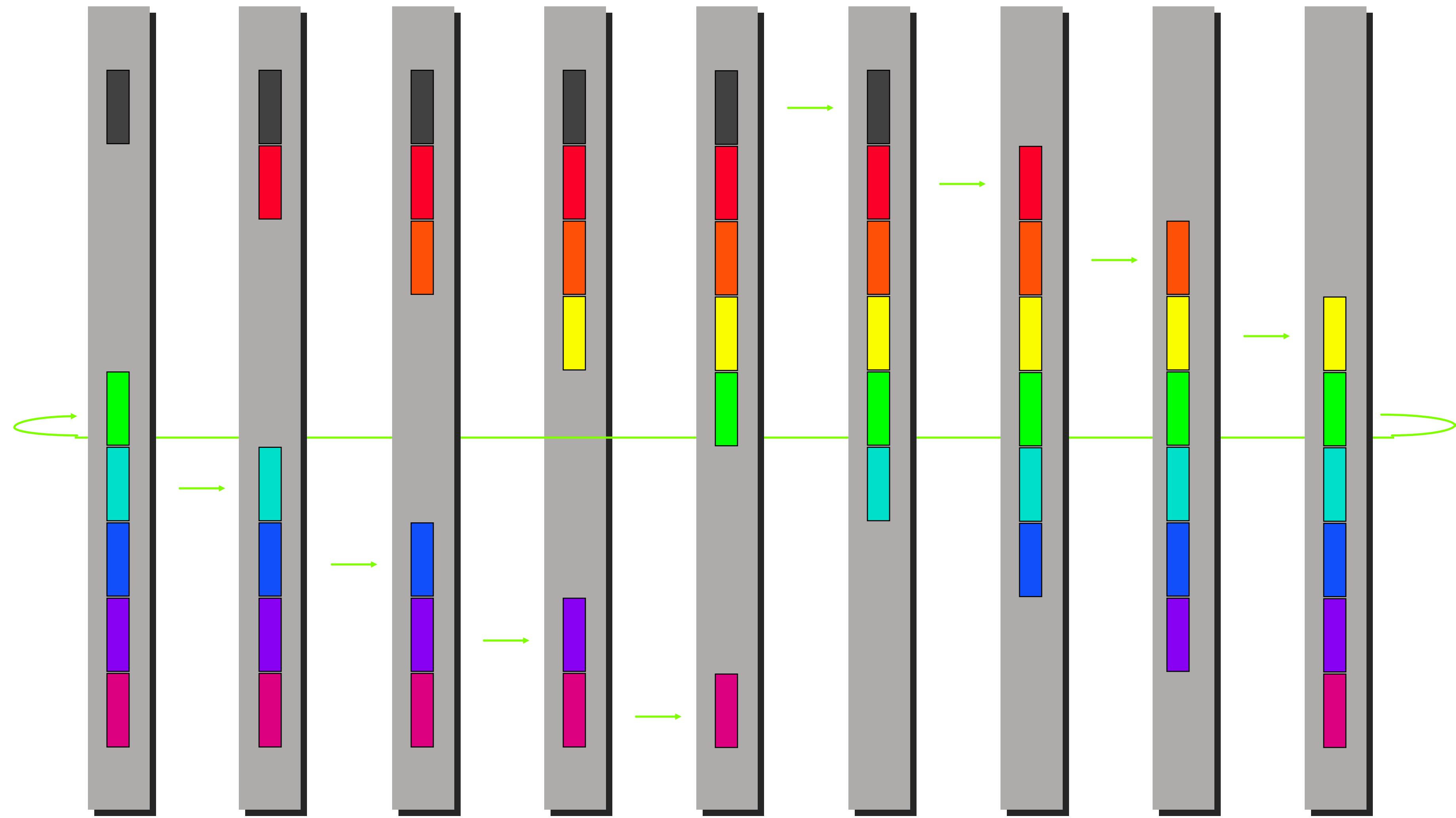


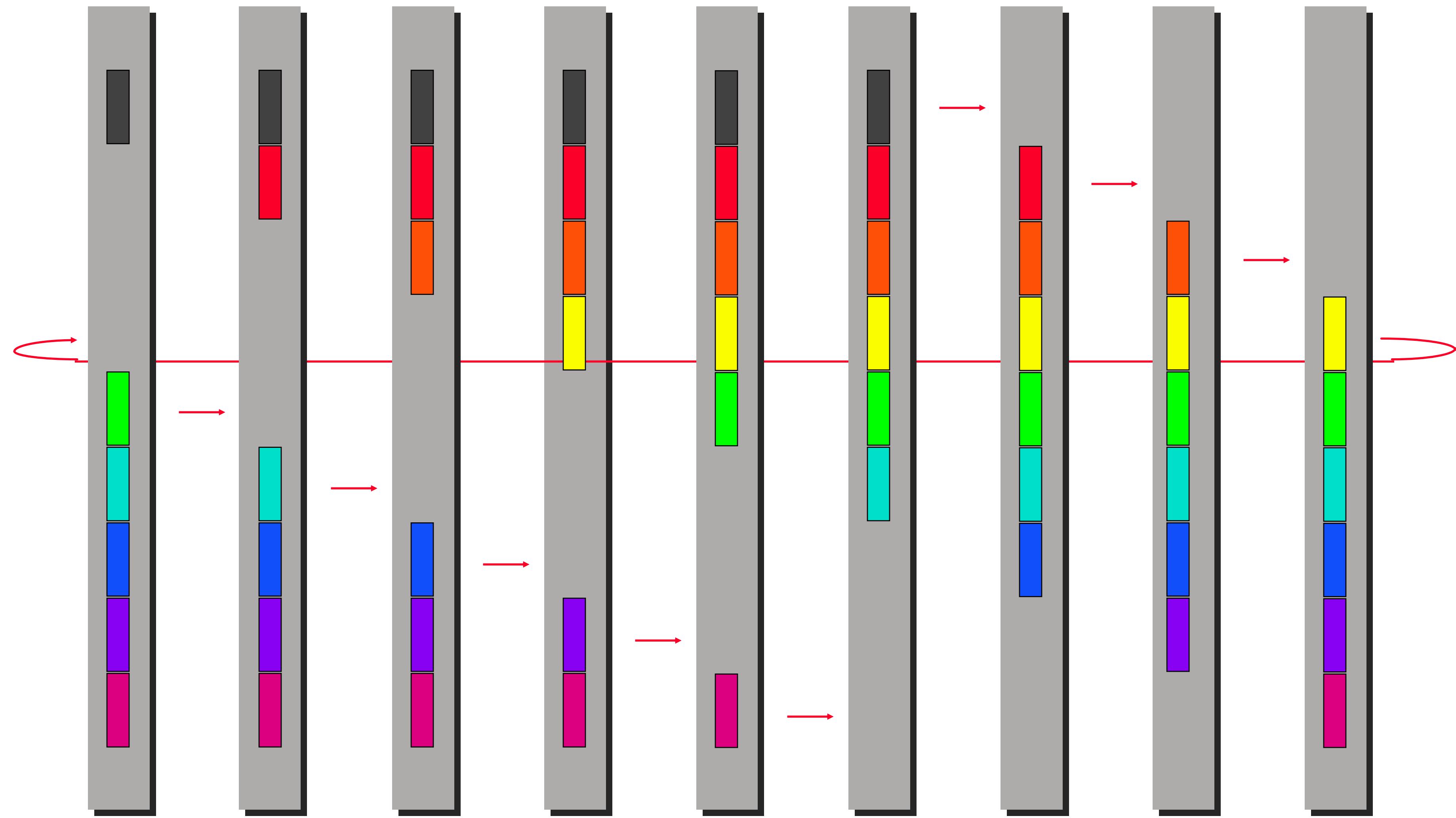


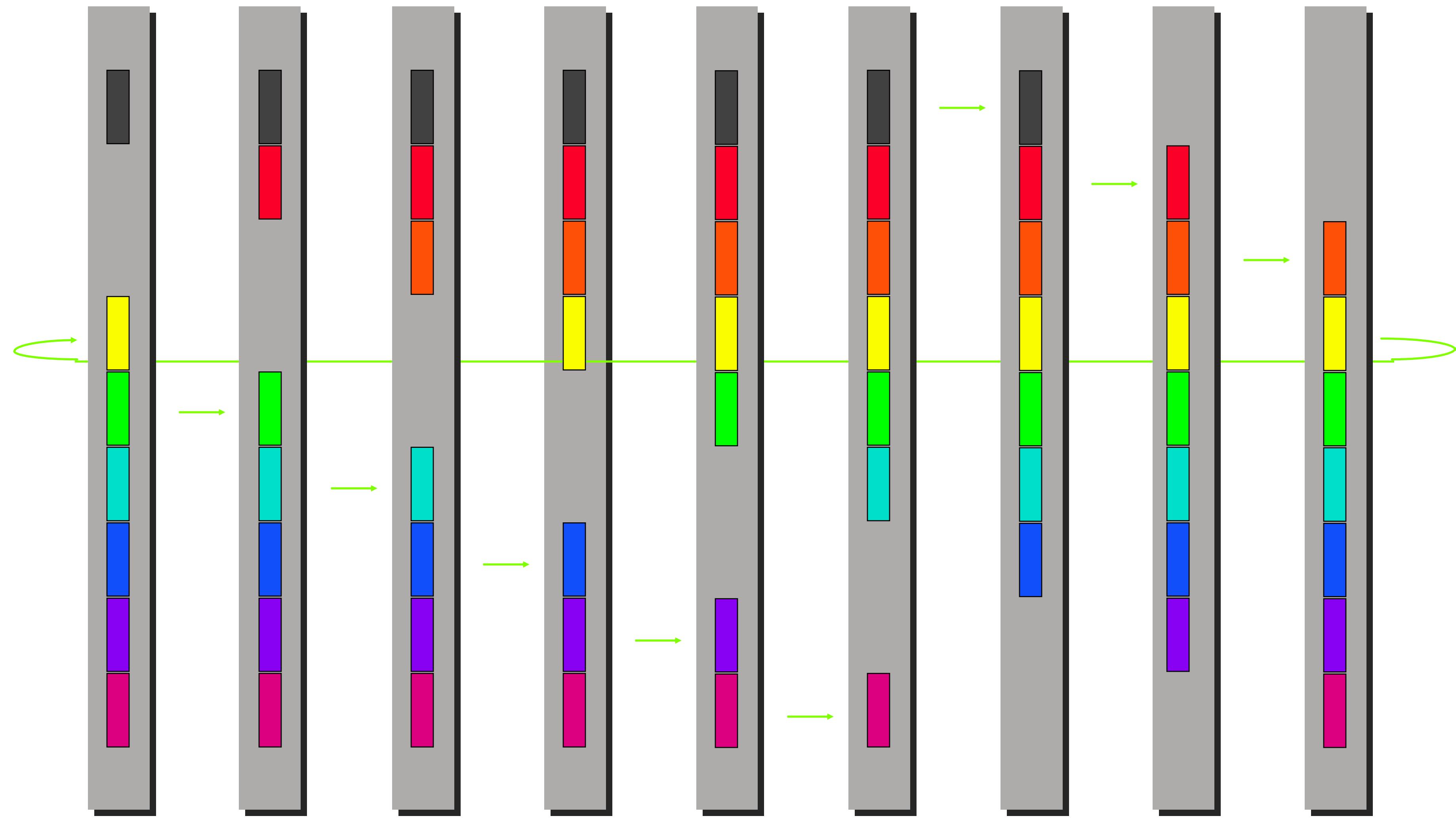


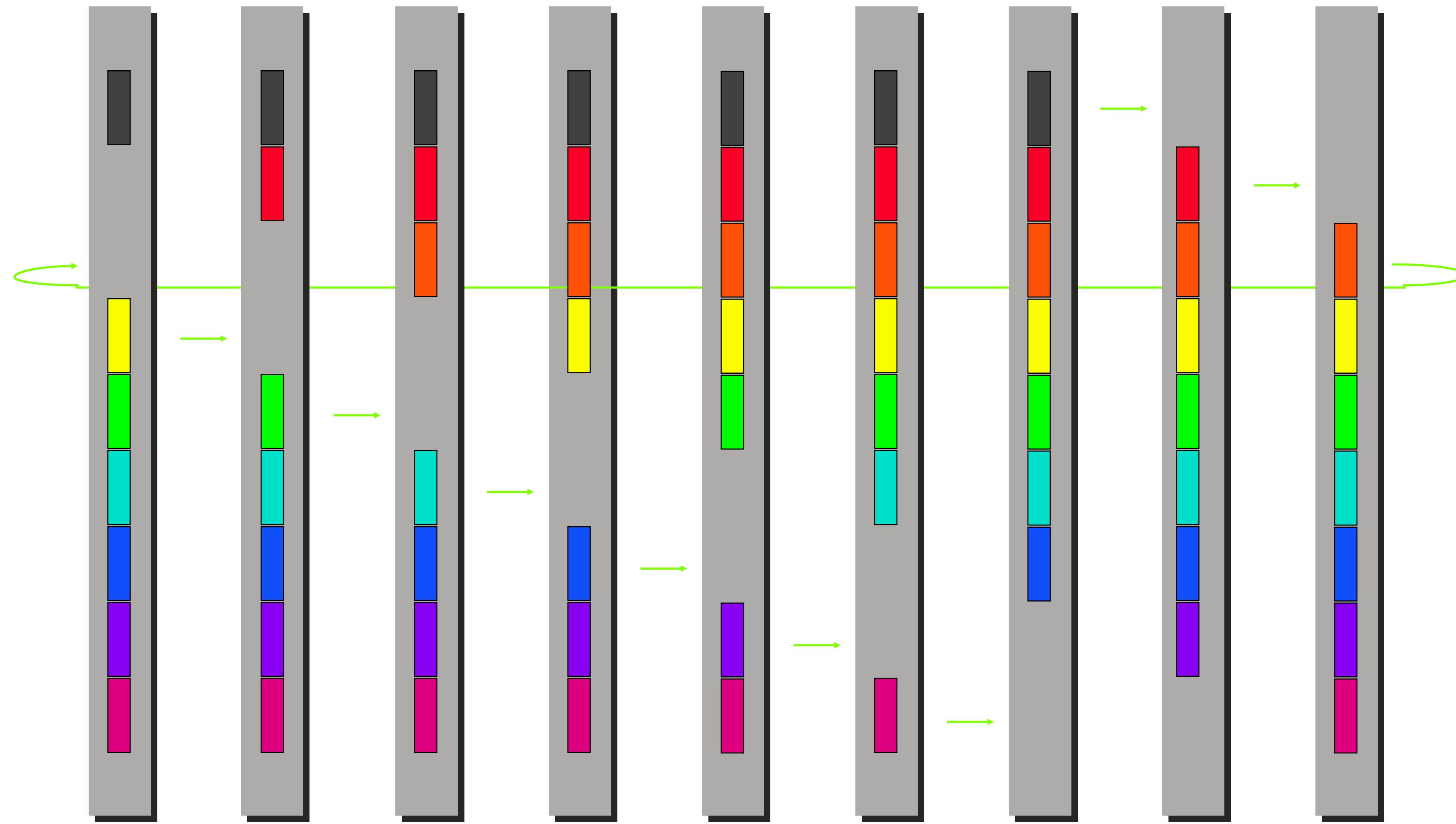


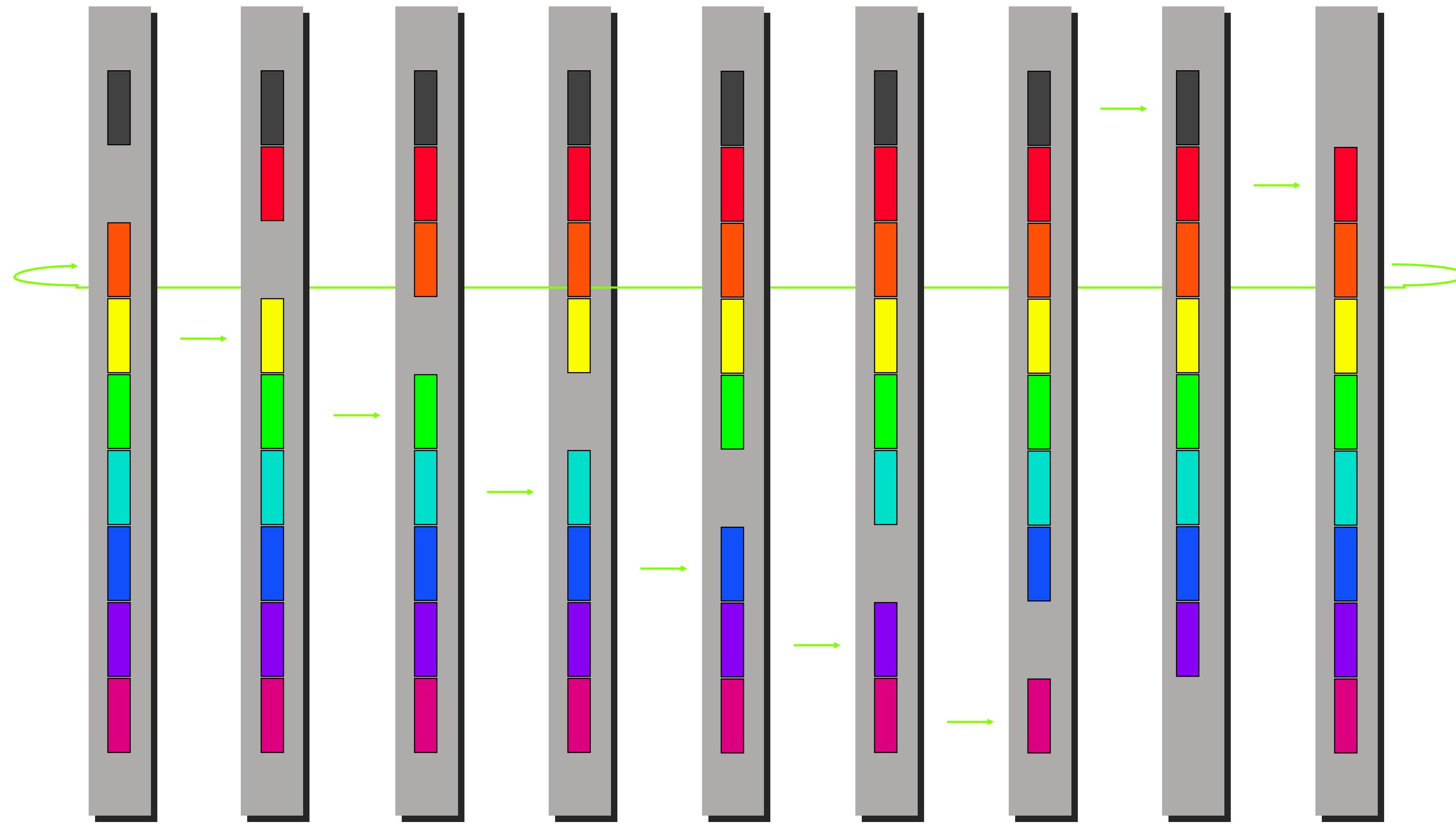


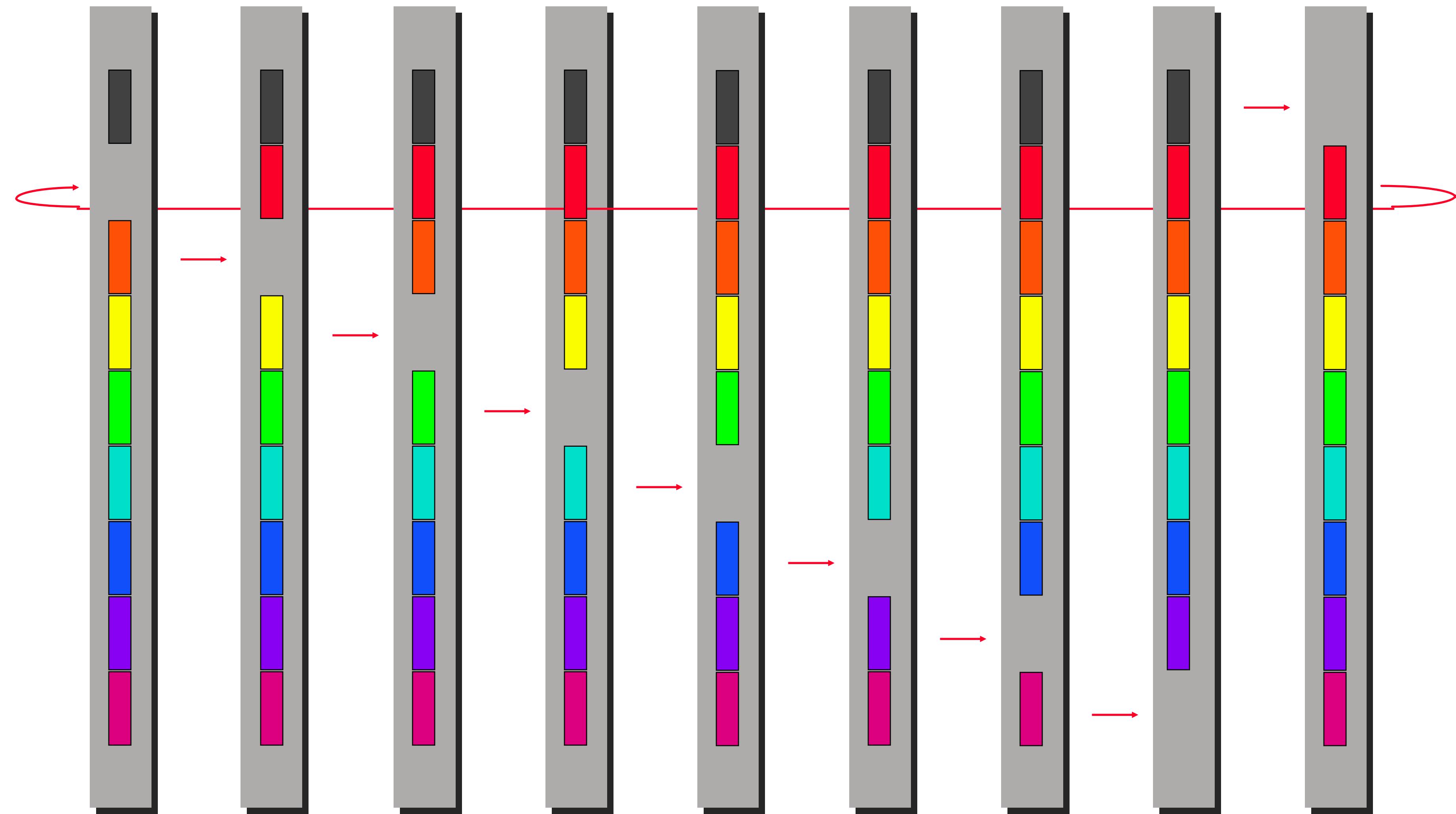


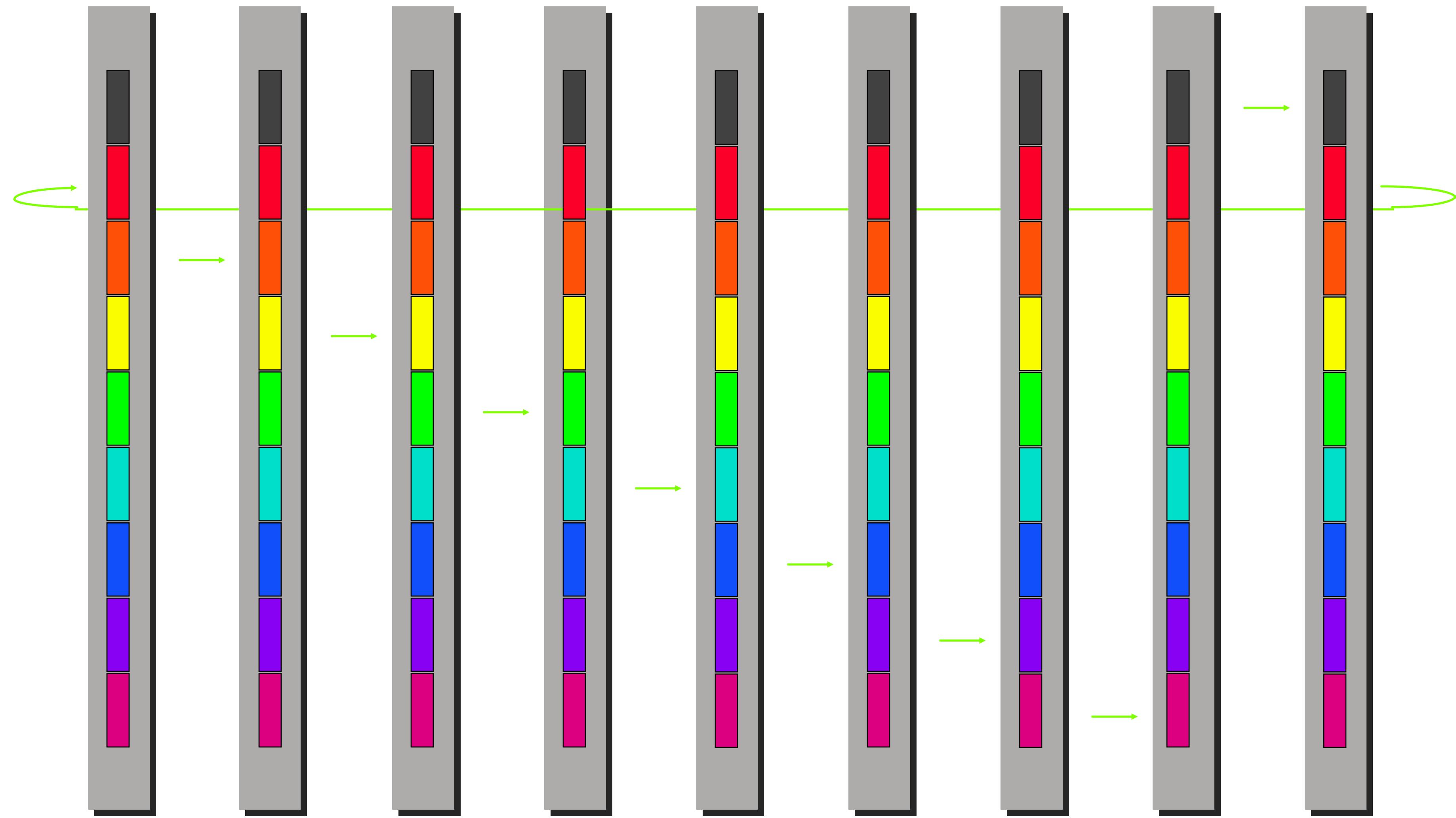


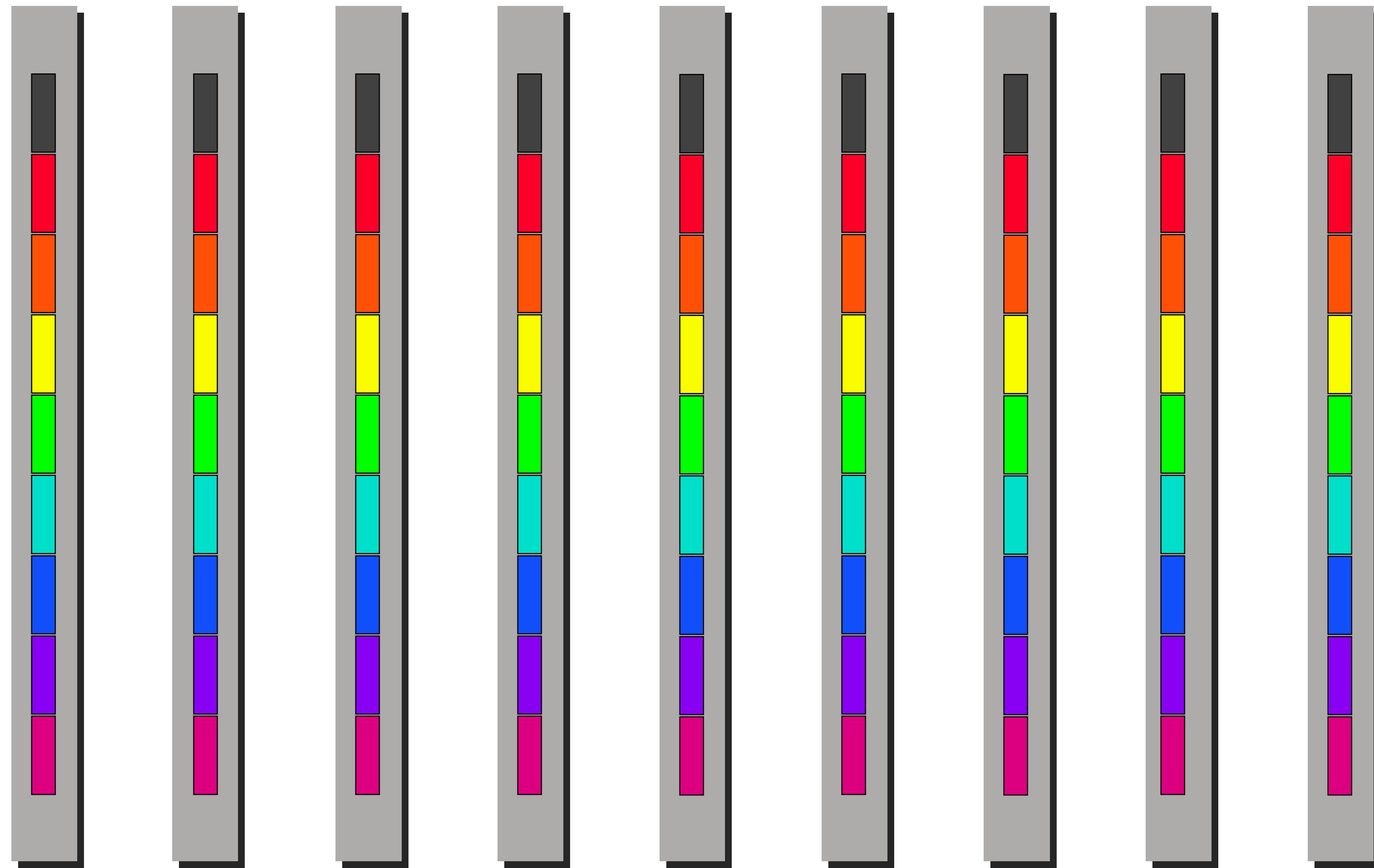










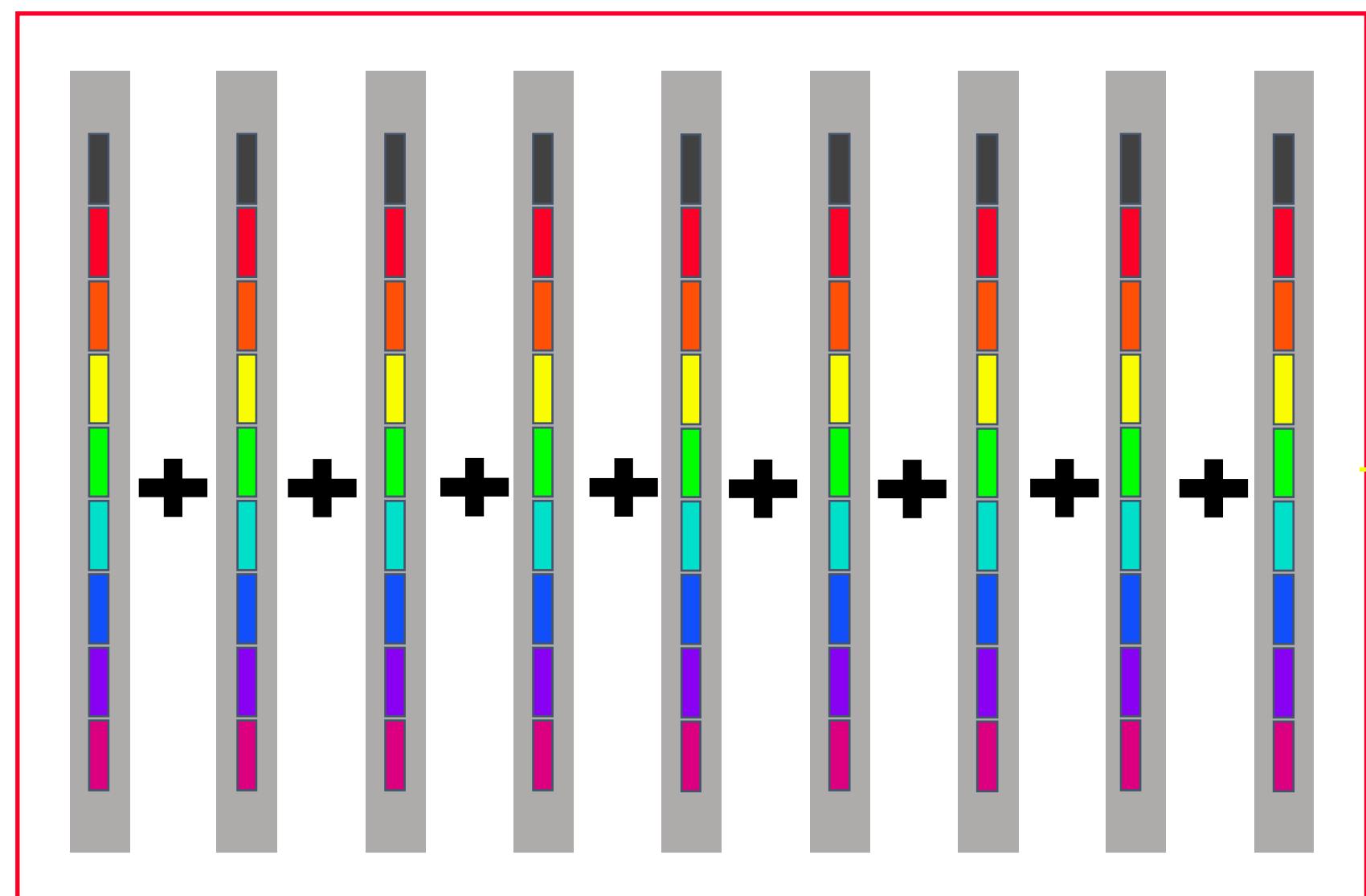


Cost of bucket Allgather

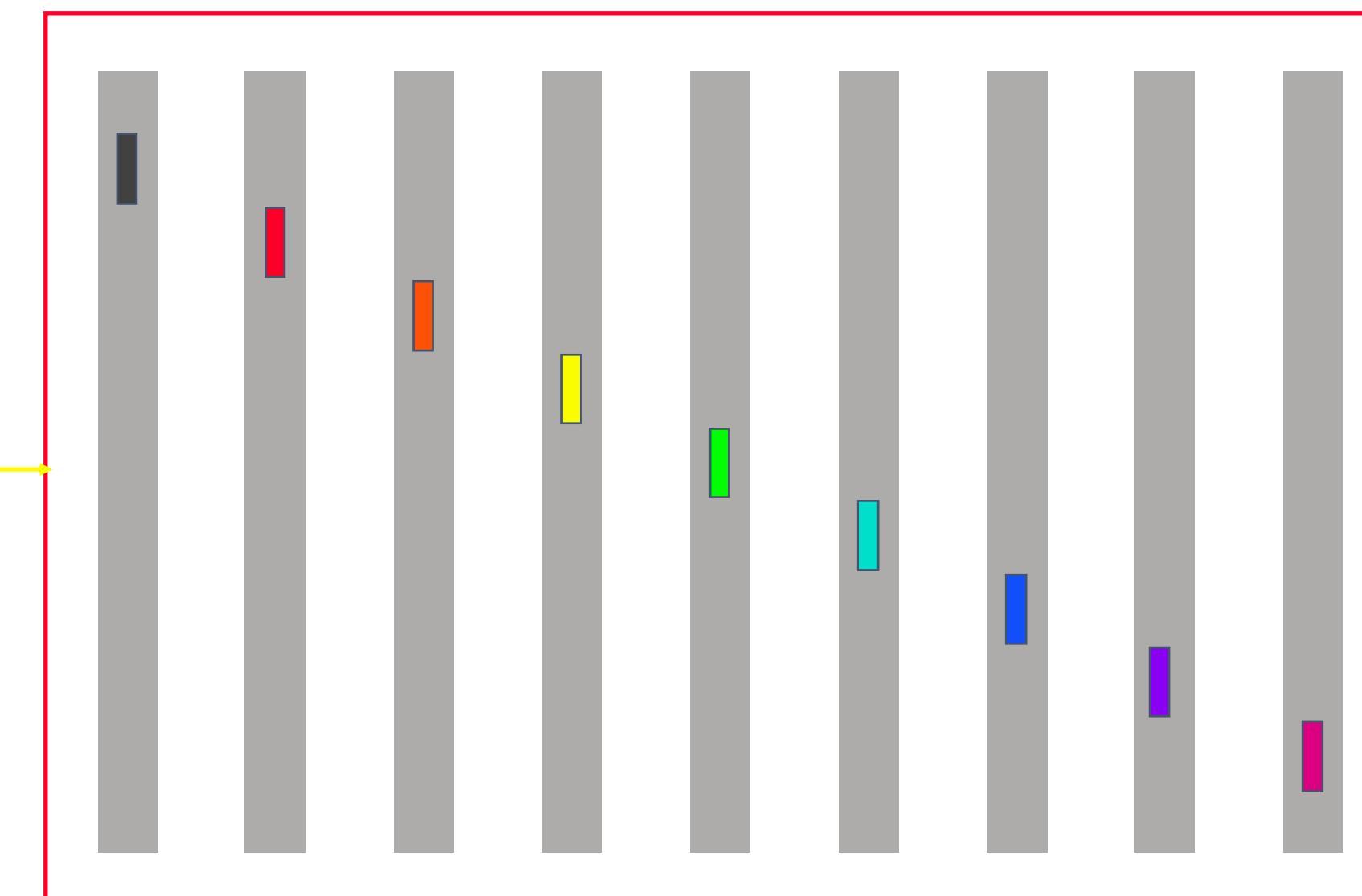
$$\text{number of steps} \quad (p-1) \left(\alpha + \frac{n}{p} \beta \right) = \text{cost per steps}$$
$$(p-1)\alpha + \frac{p-1}{p} n\beta$$

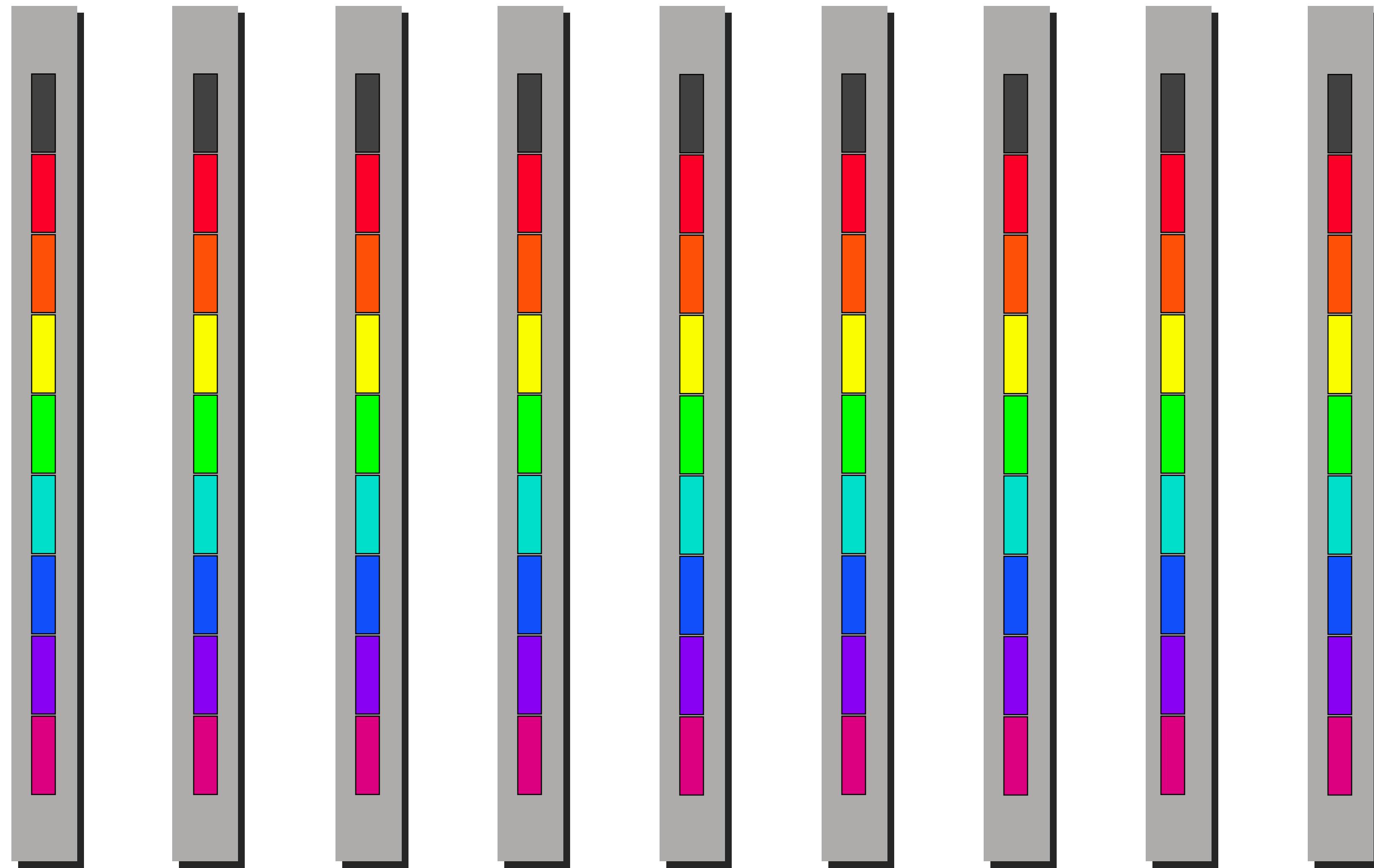
Reduce-scatter

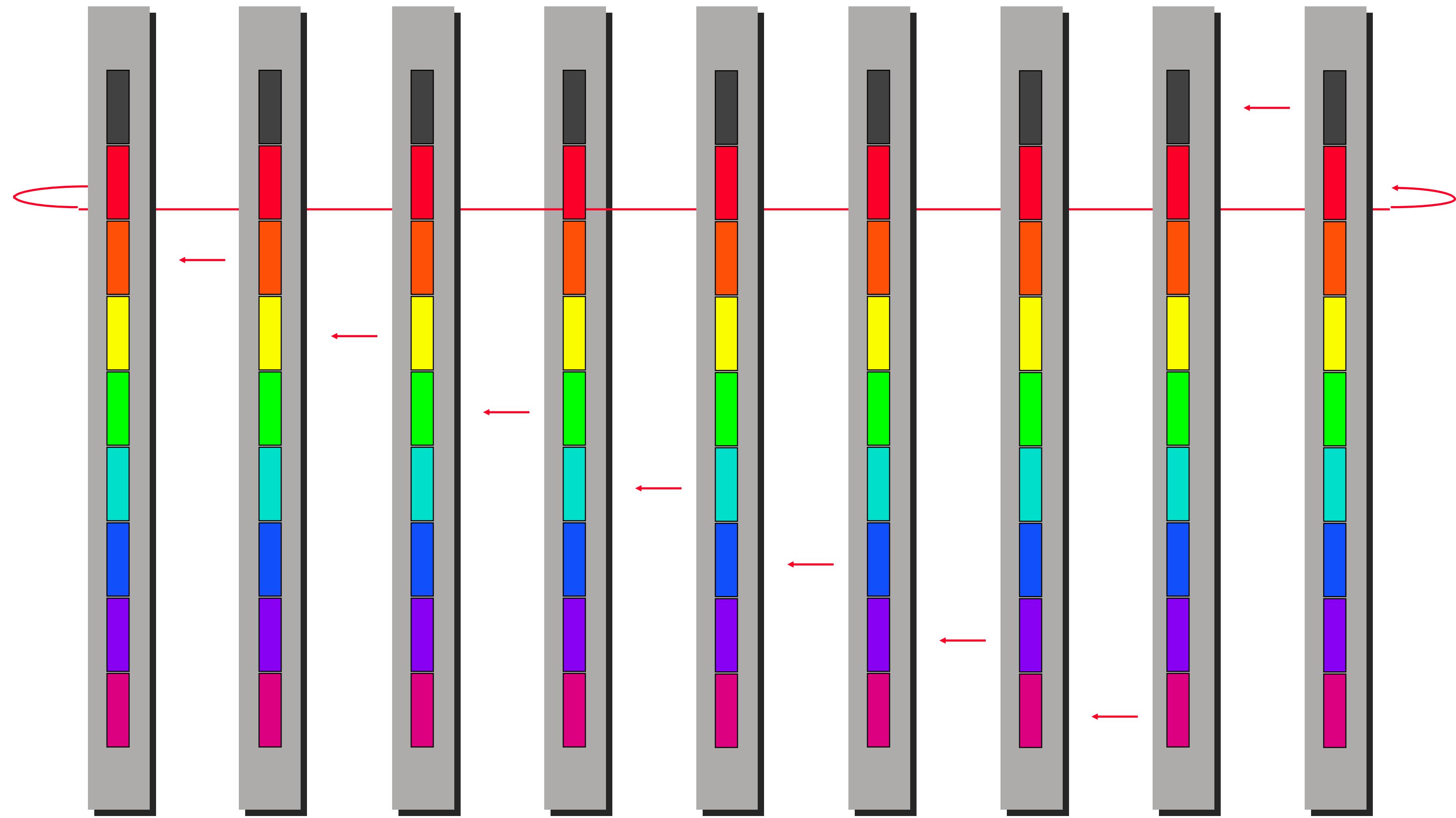
Before

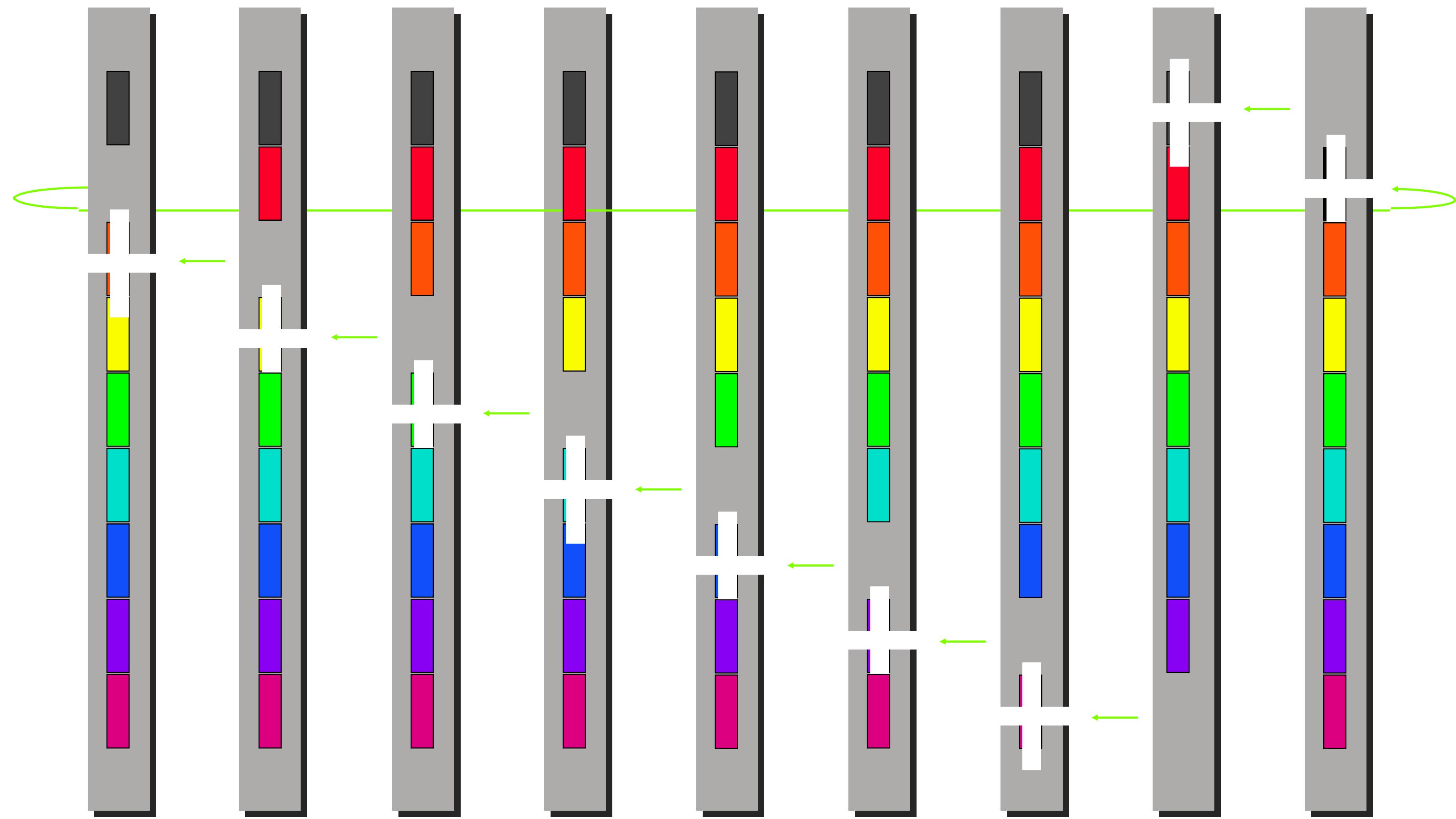


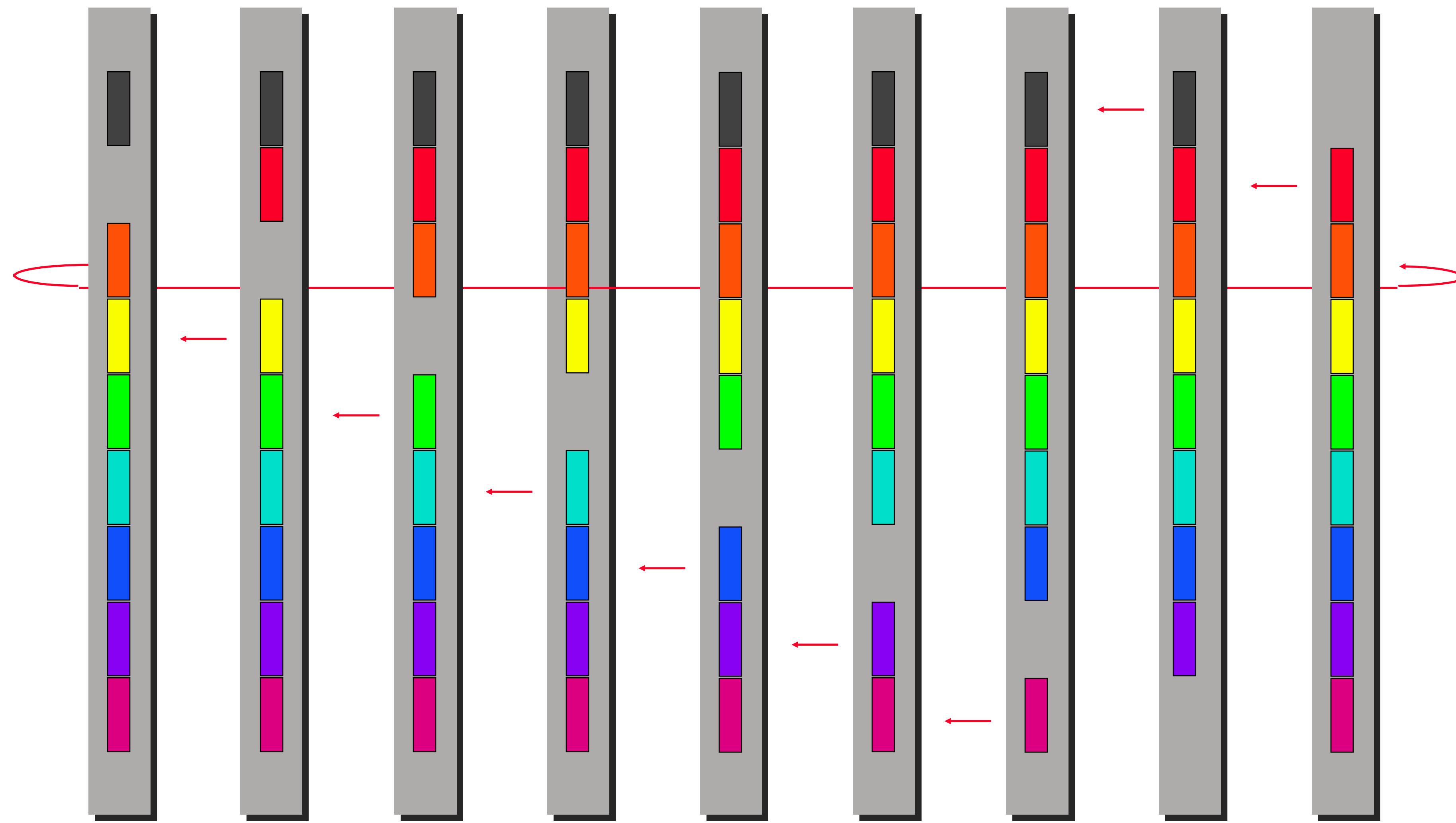
After

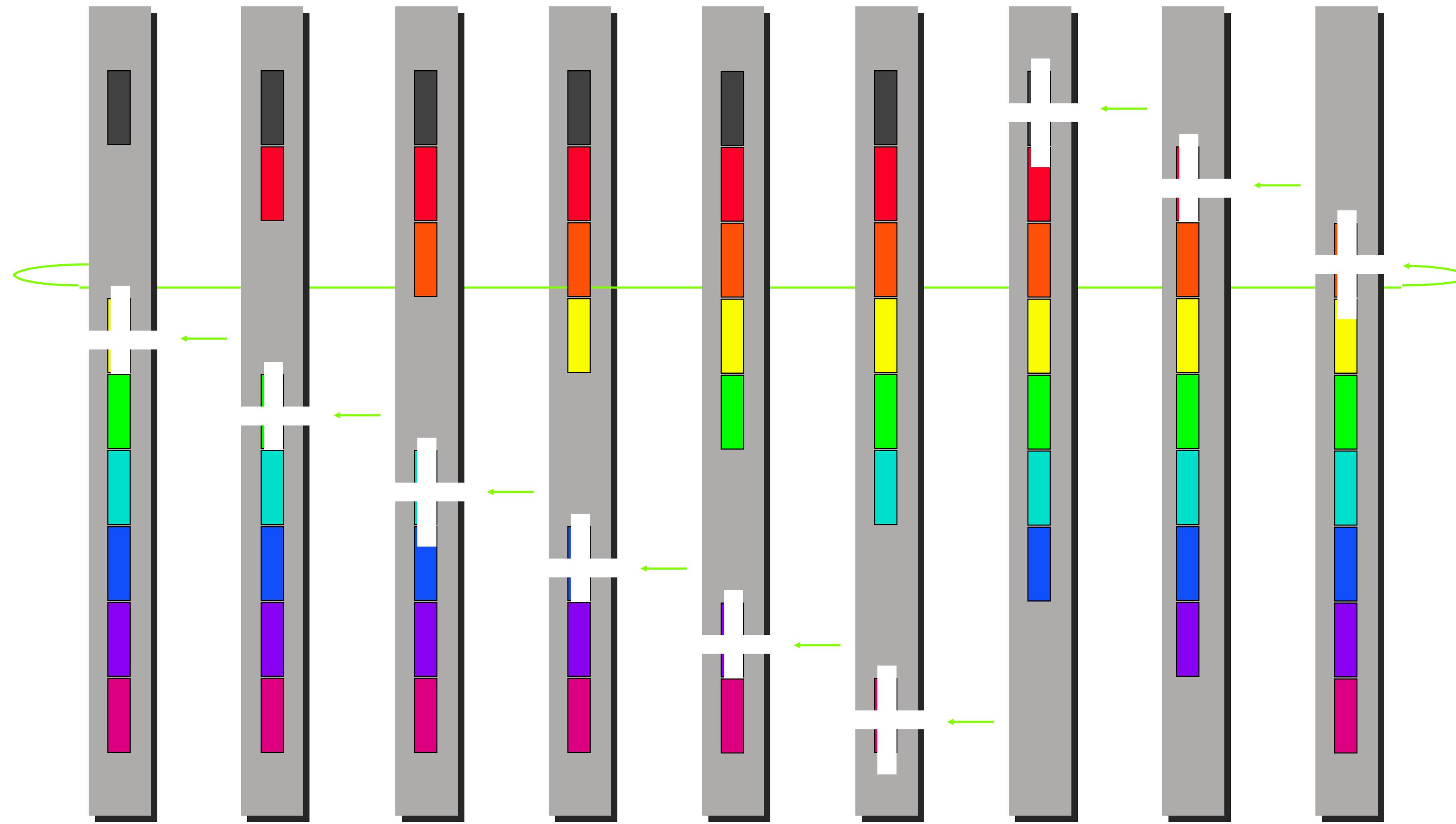


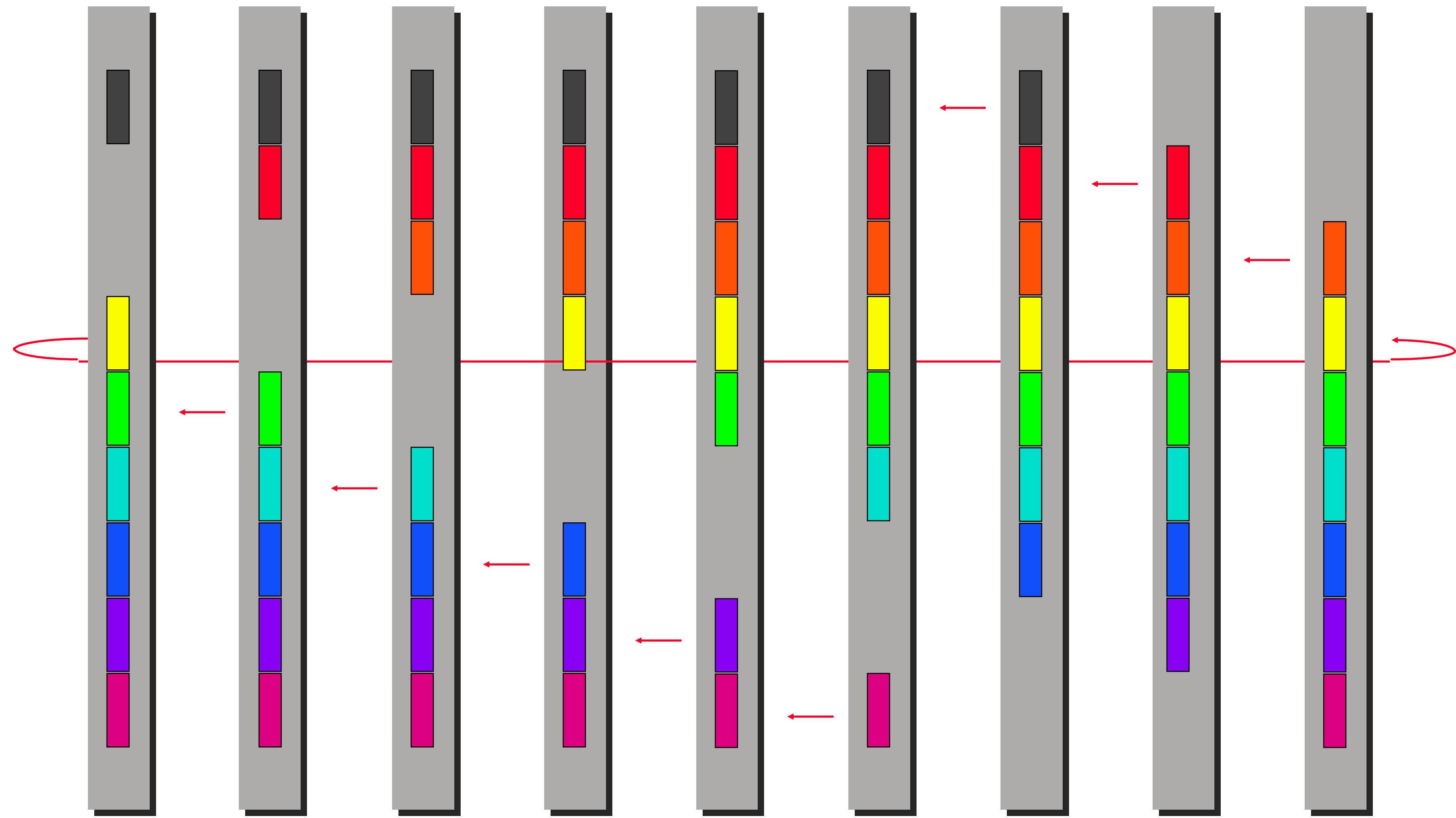


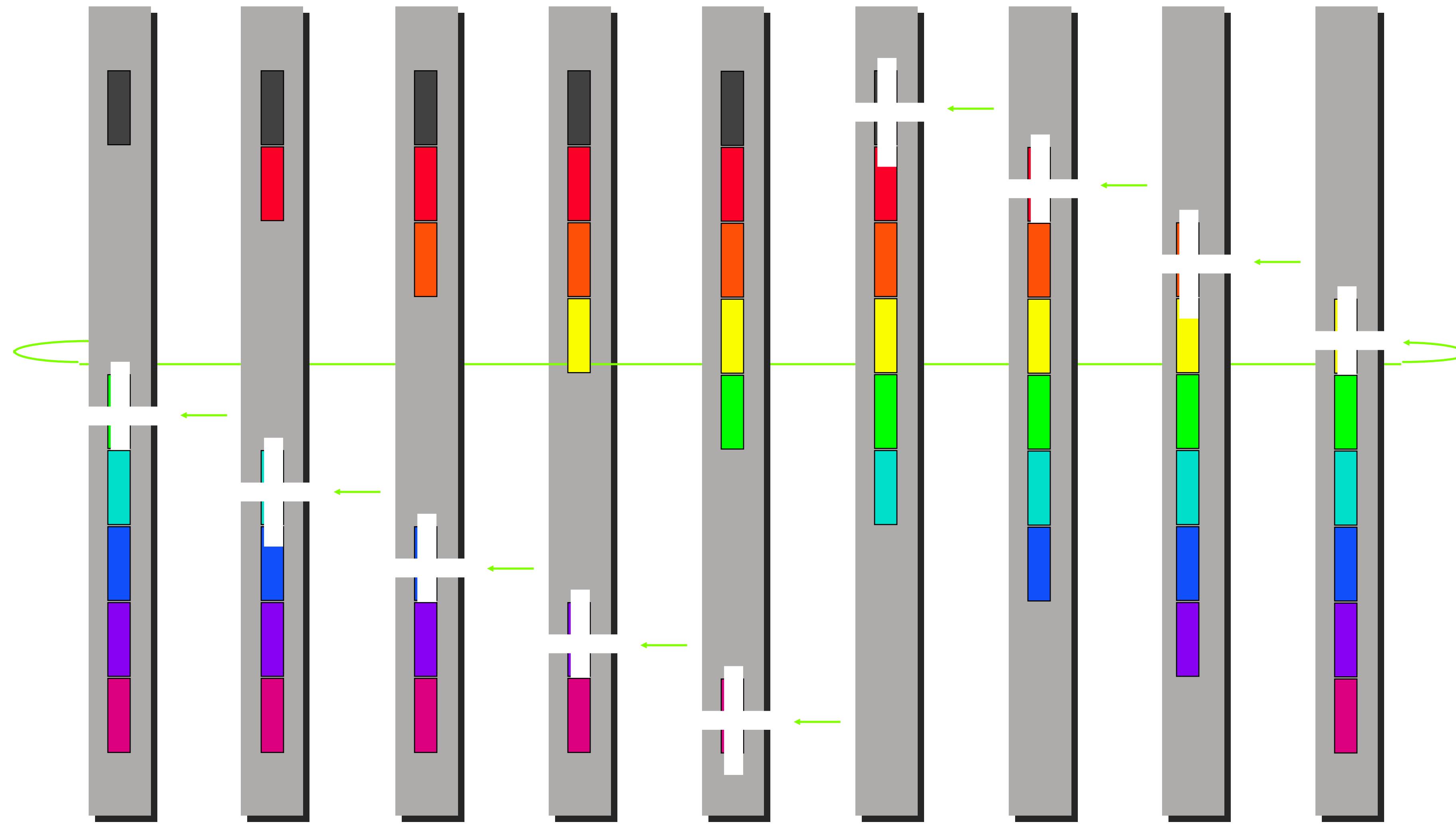


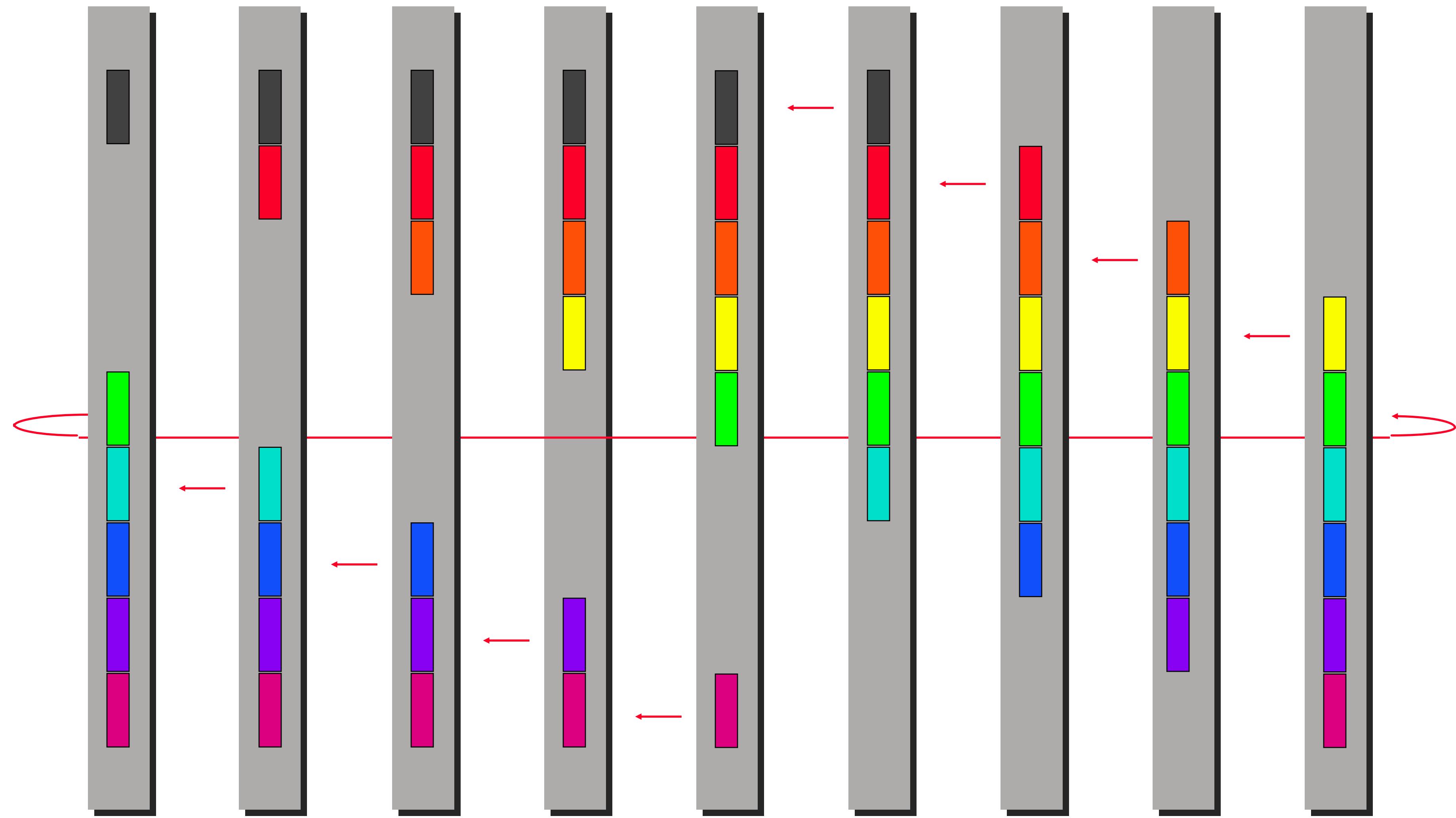


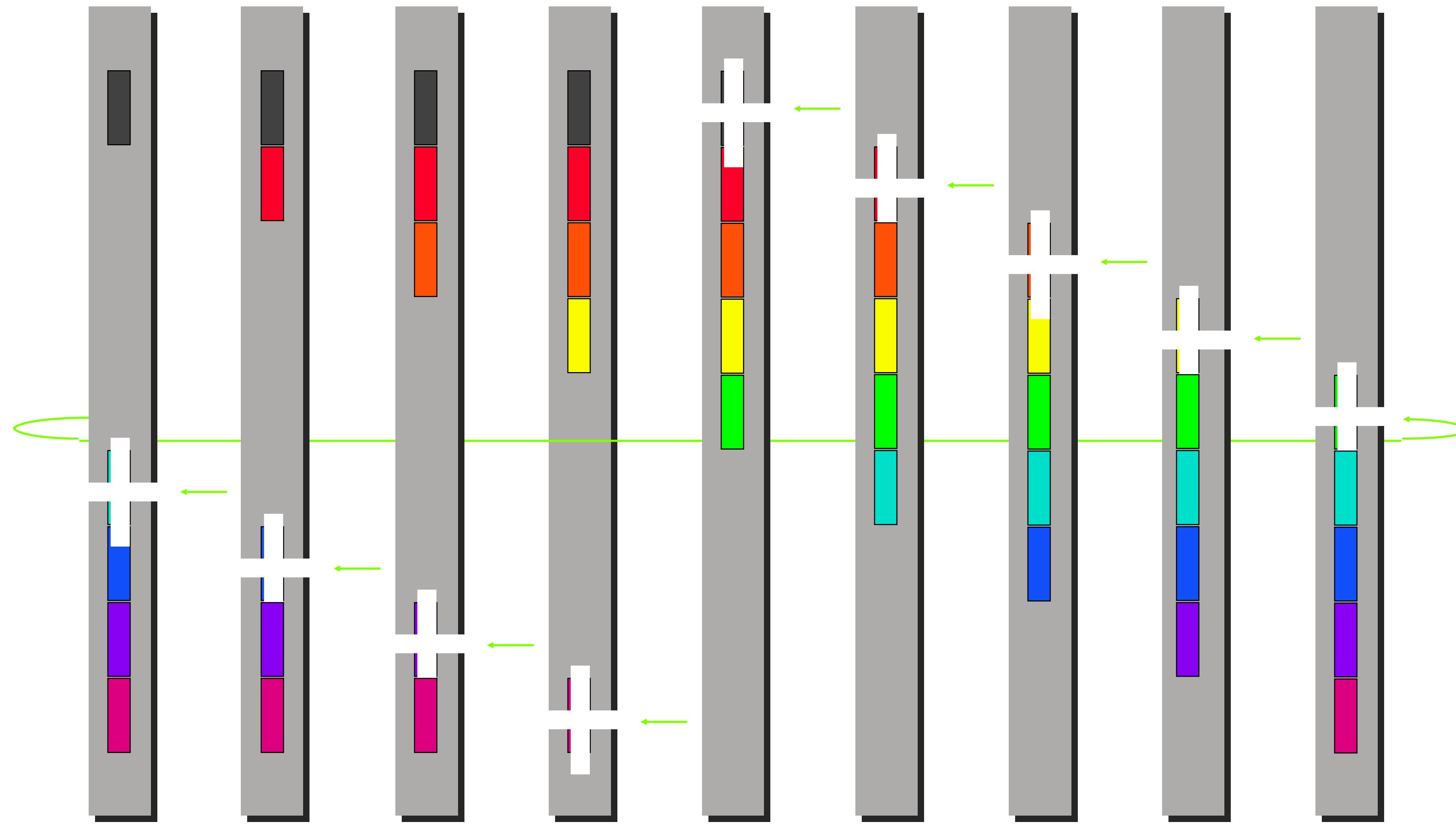


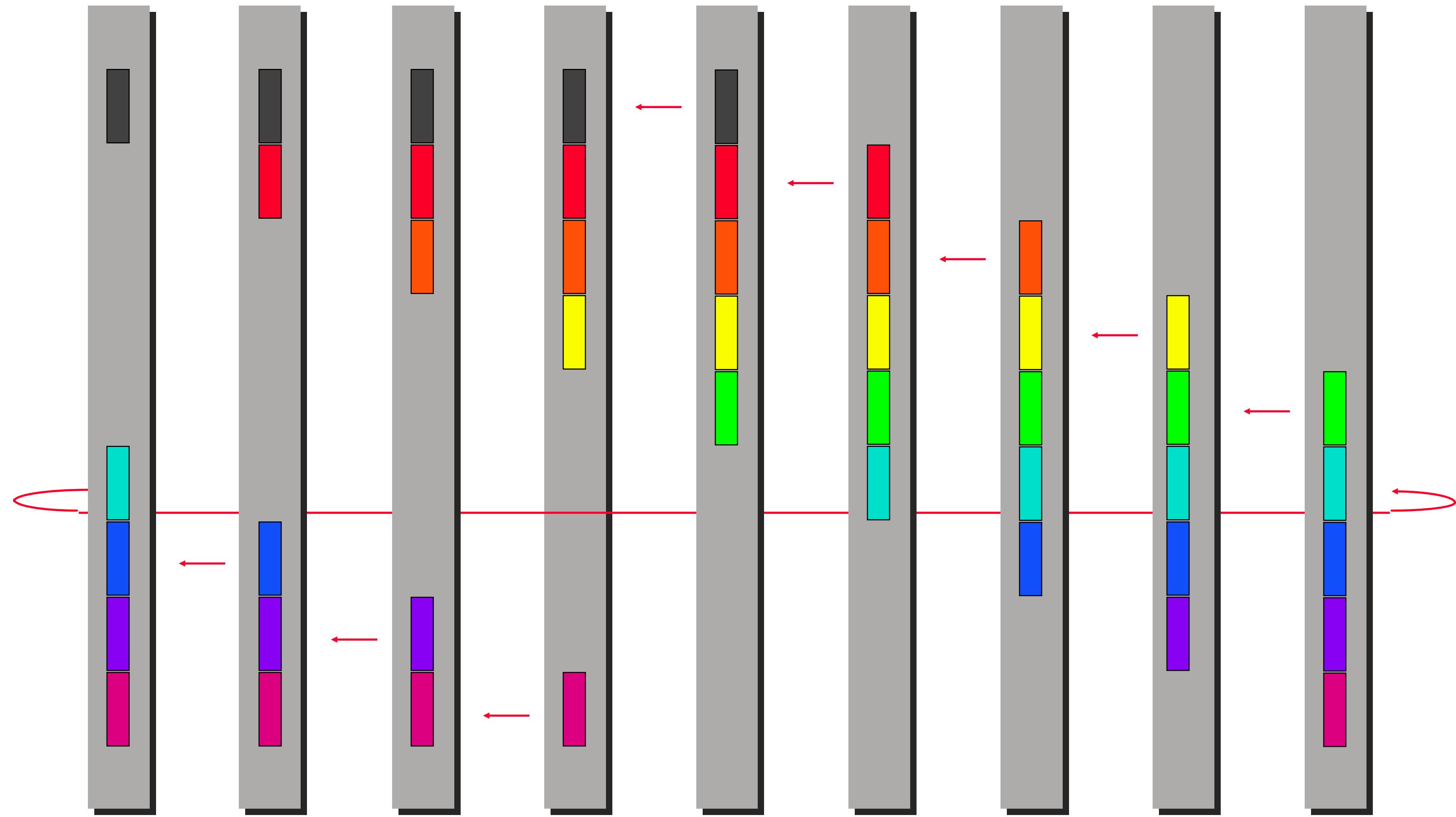


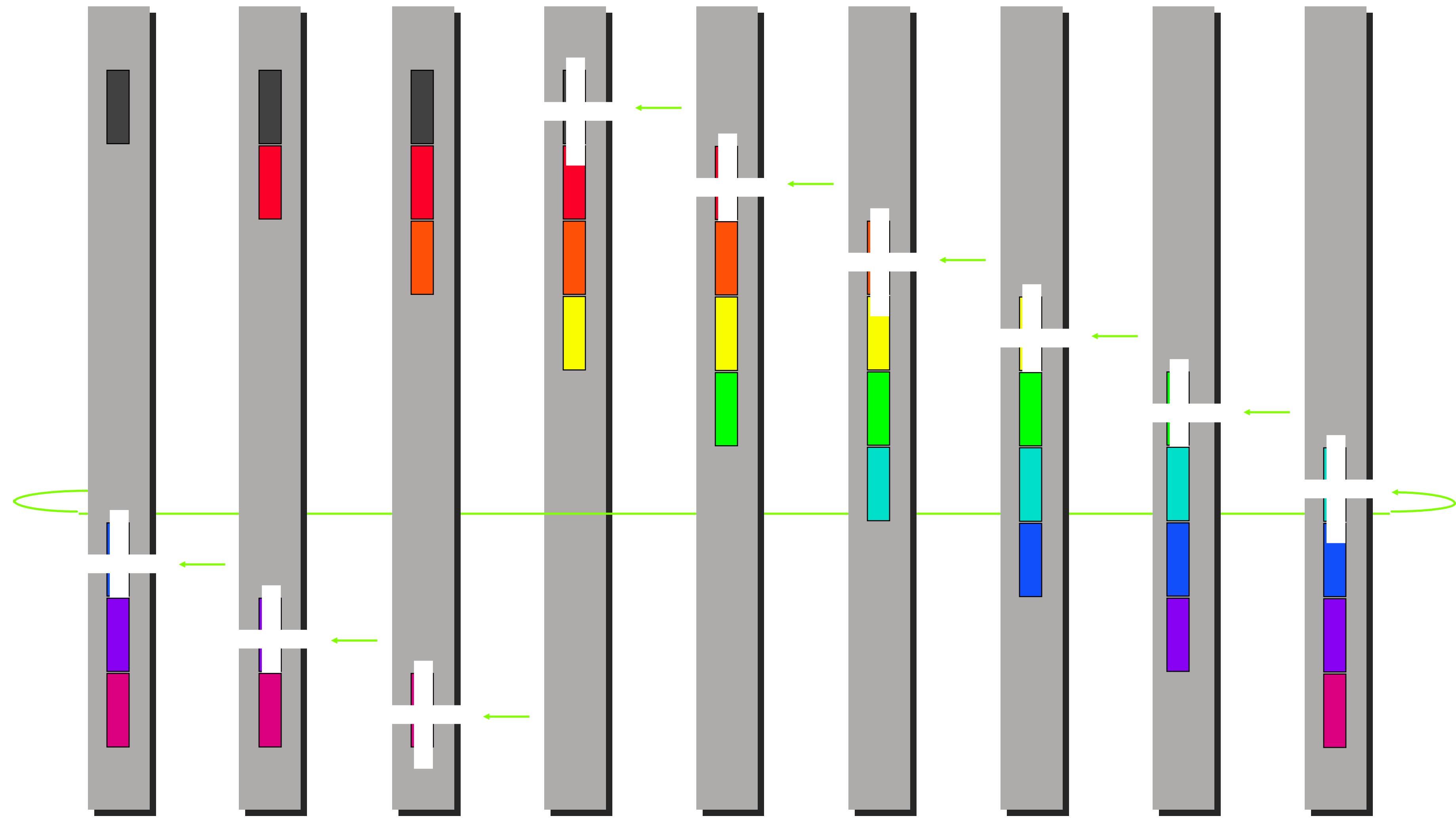


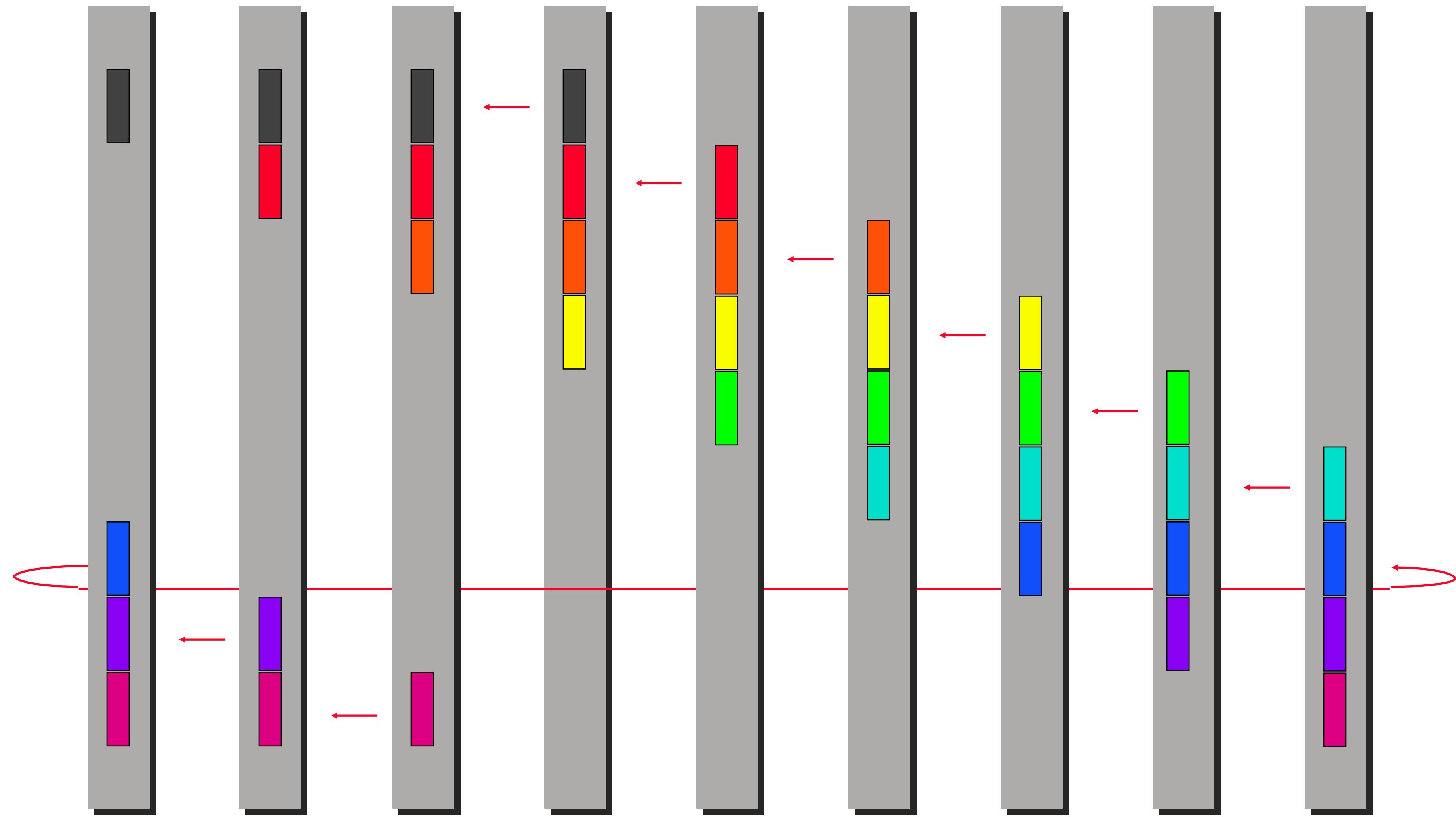


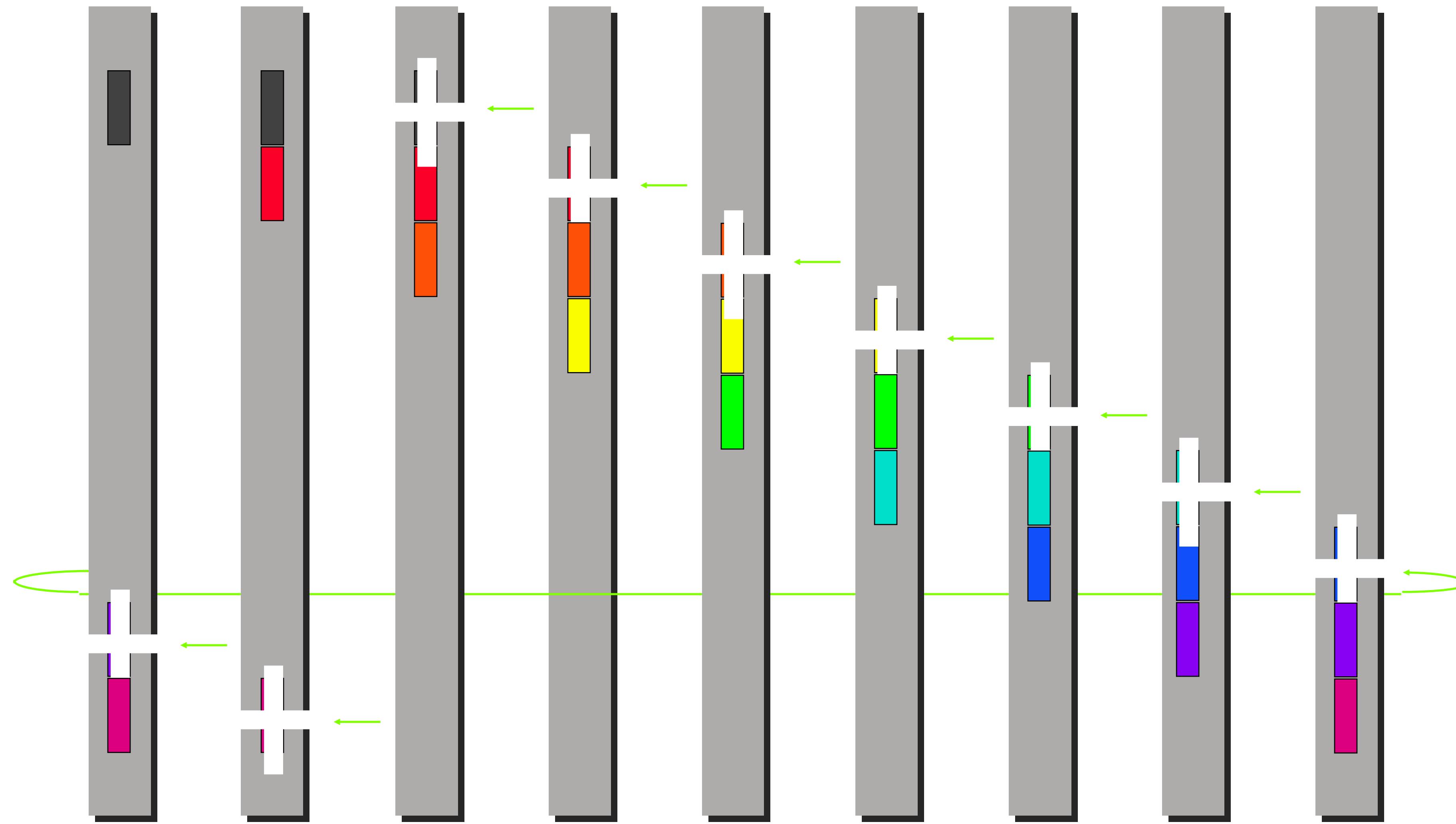


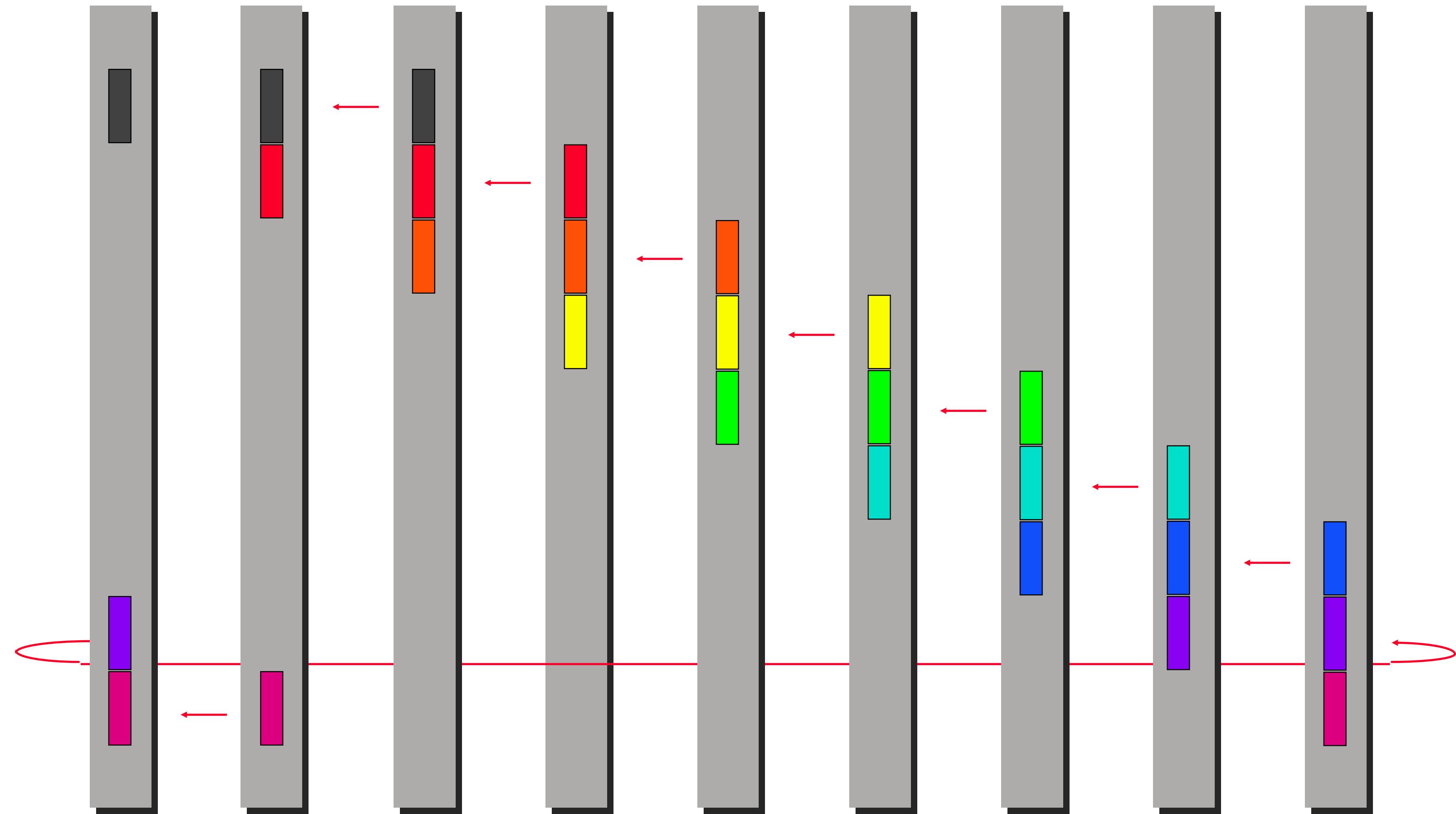


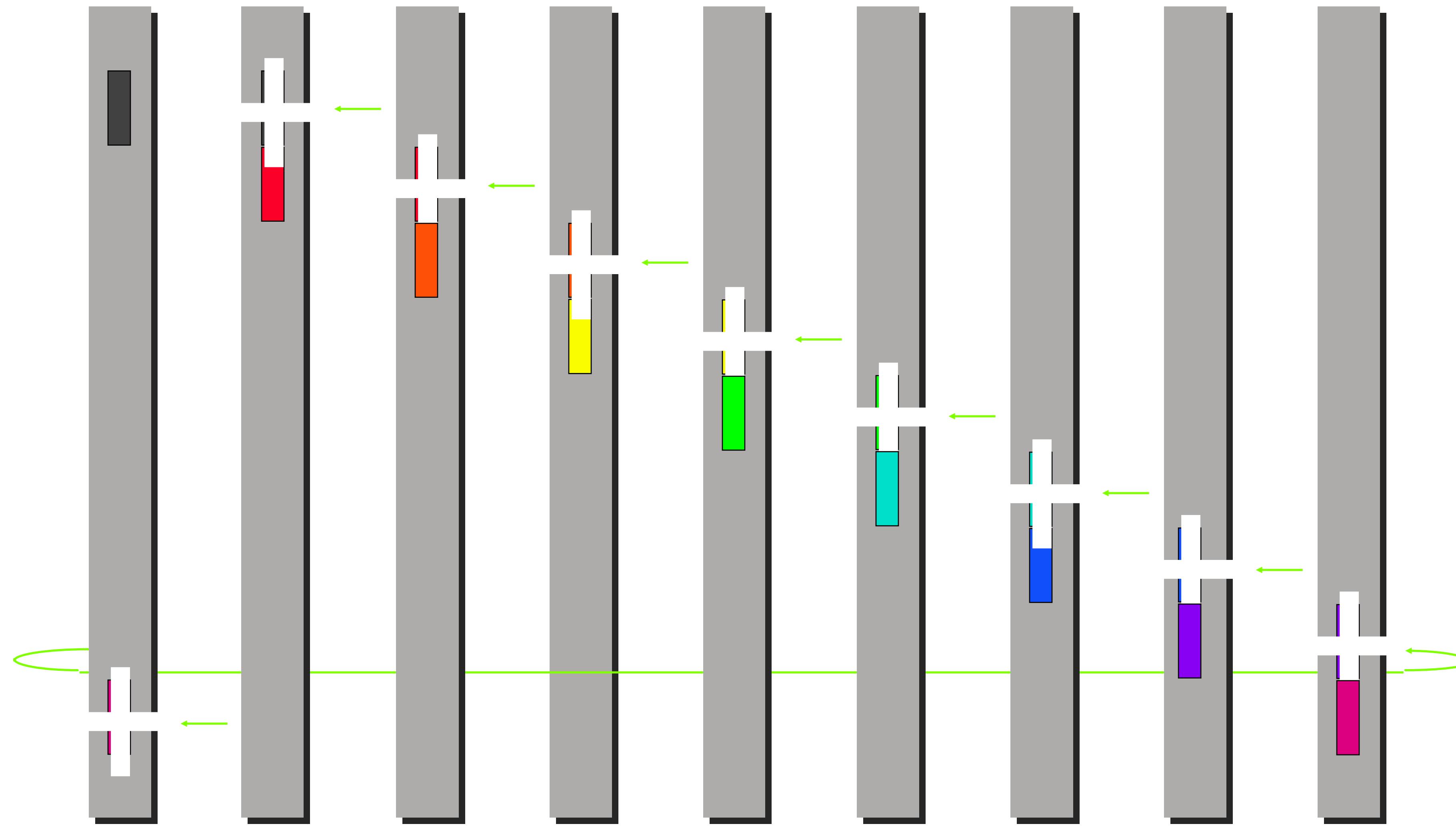


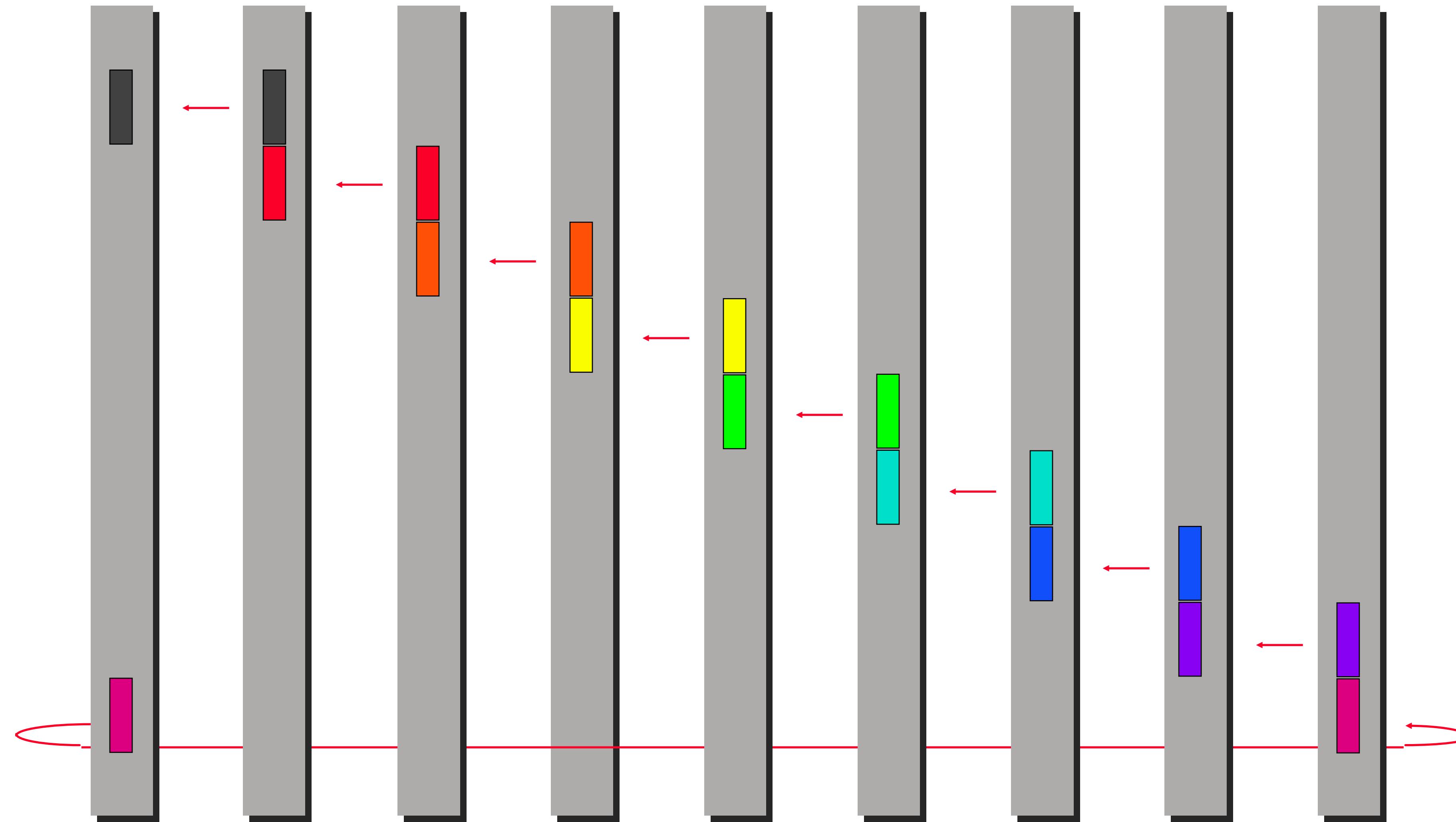


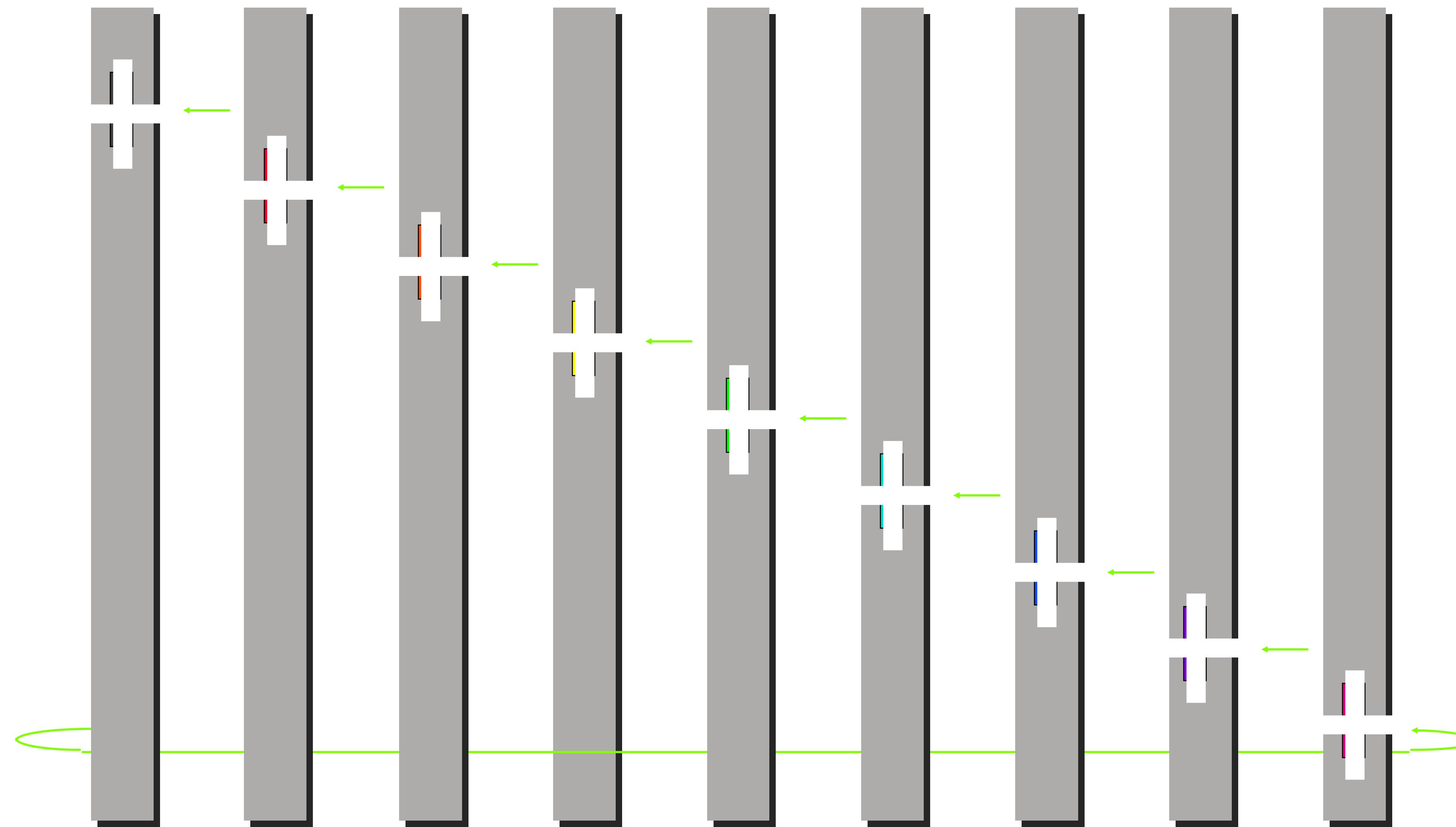


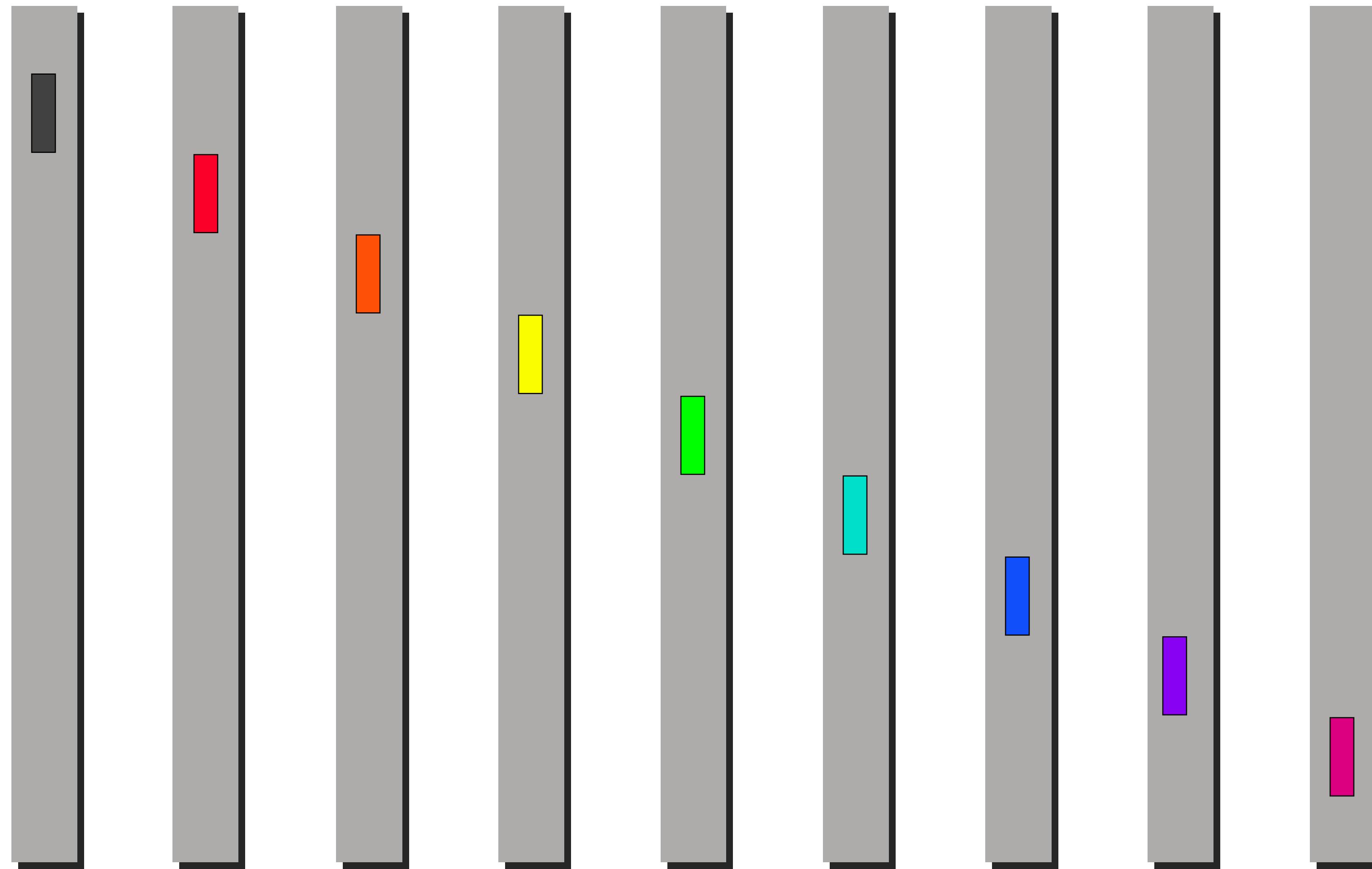




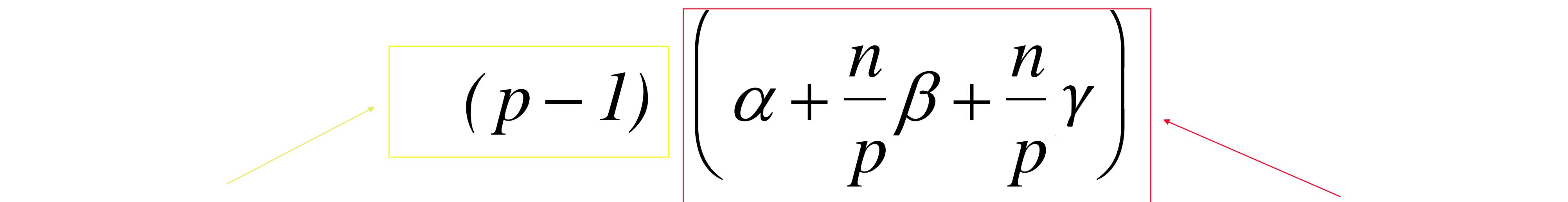








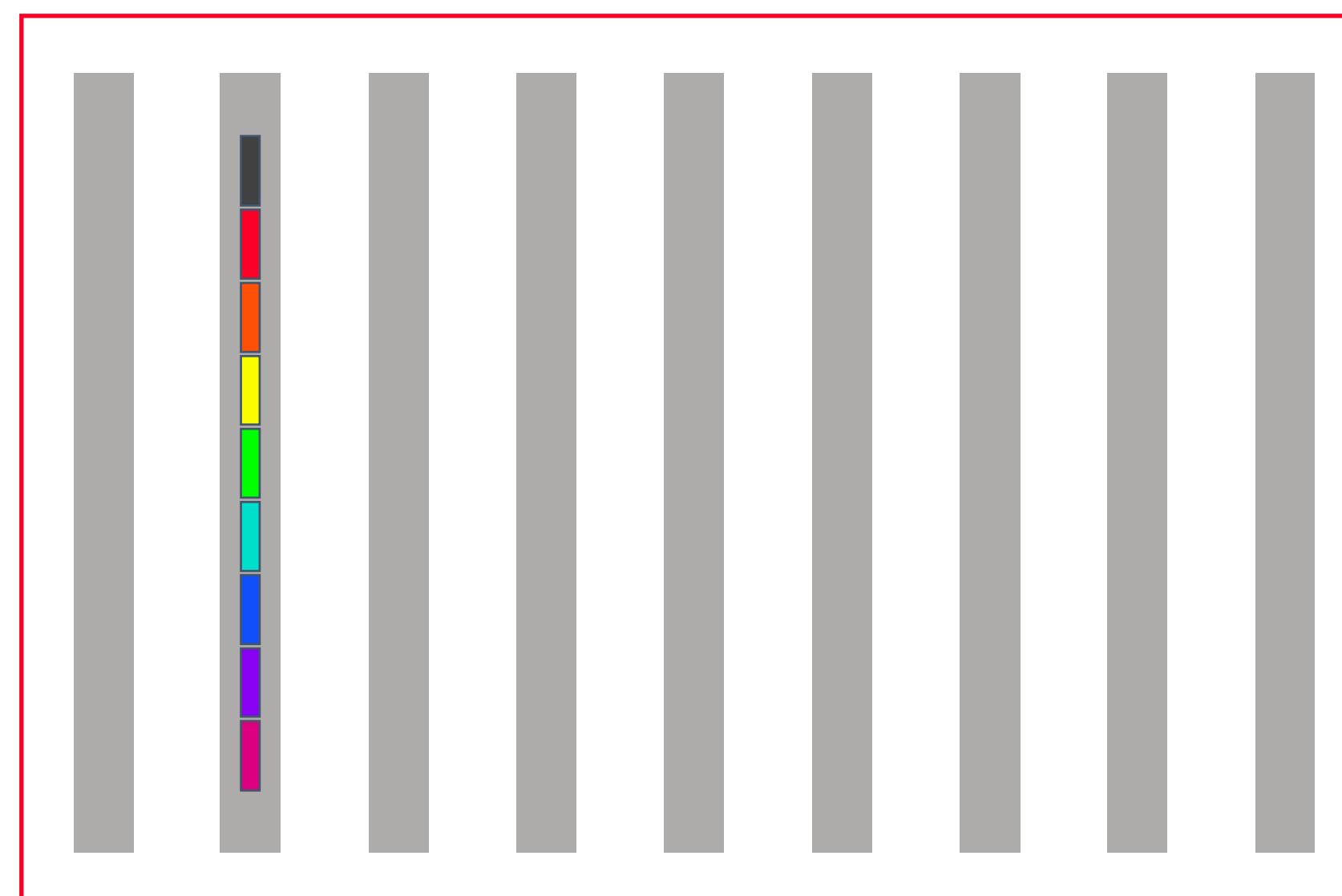
Cost of bucket distributed combine

$$\text{number of steps} \quad (p-1) \left(\alpha + \frac{n}{p} \beta + \frac{n}{p} \gamma \right) = \text{cost per steps}$$
$$(p-1)\alpha + \frac{p-1}{p} n\beta + \frac{p-1}{p} n\gamma$$


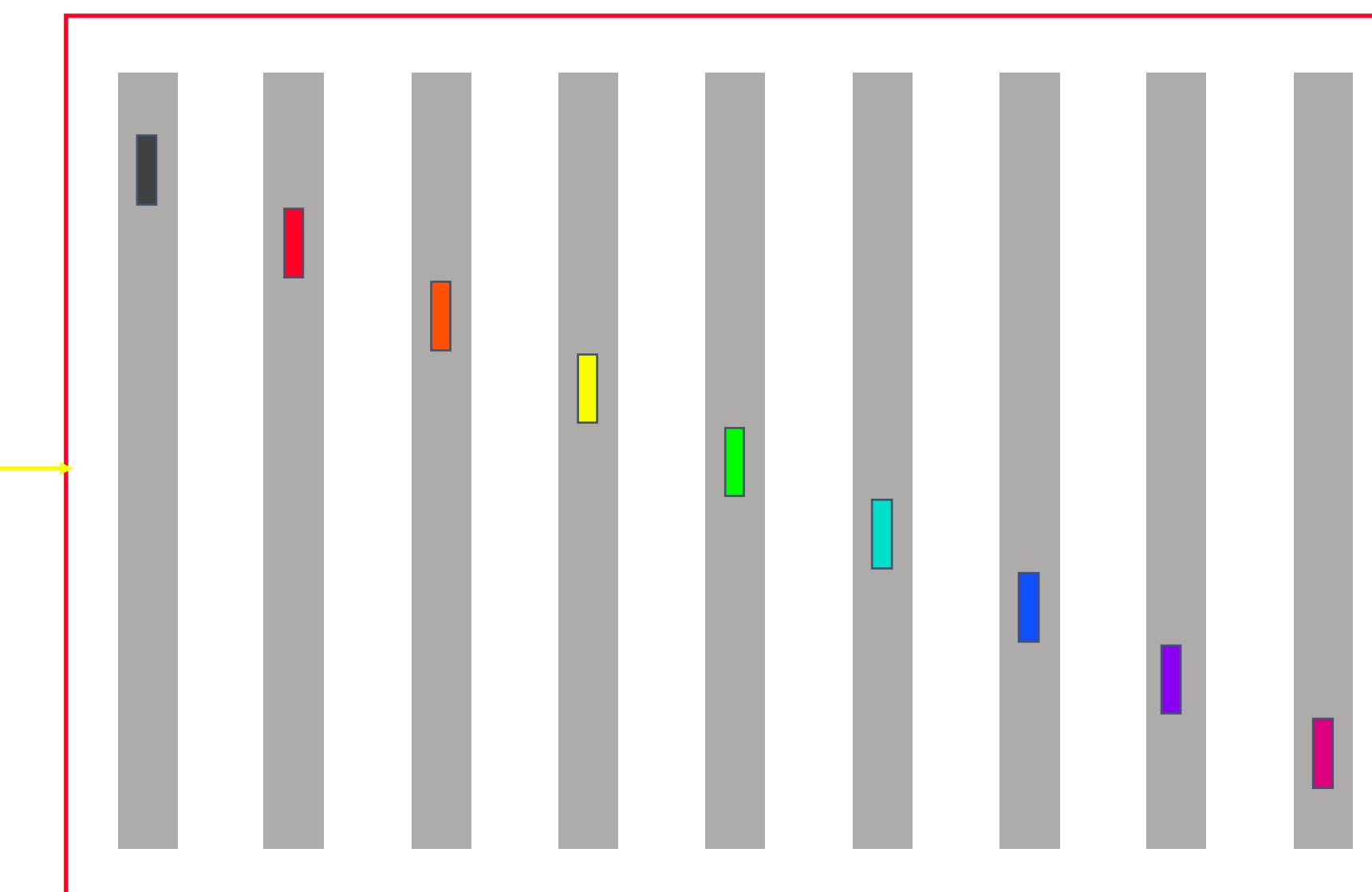
Scatter: Can Ring Be Better?

Notice: Scatter as implemented before using MST was optimal in Bandwidth as well (How to Prove?)

Before



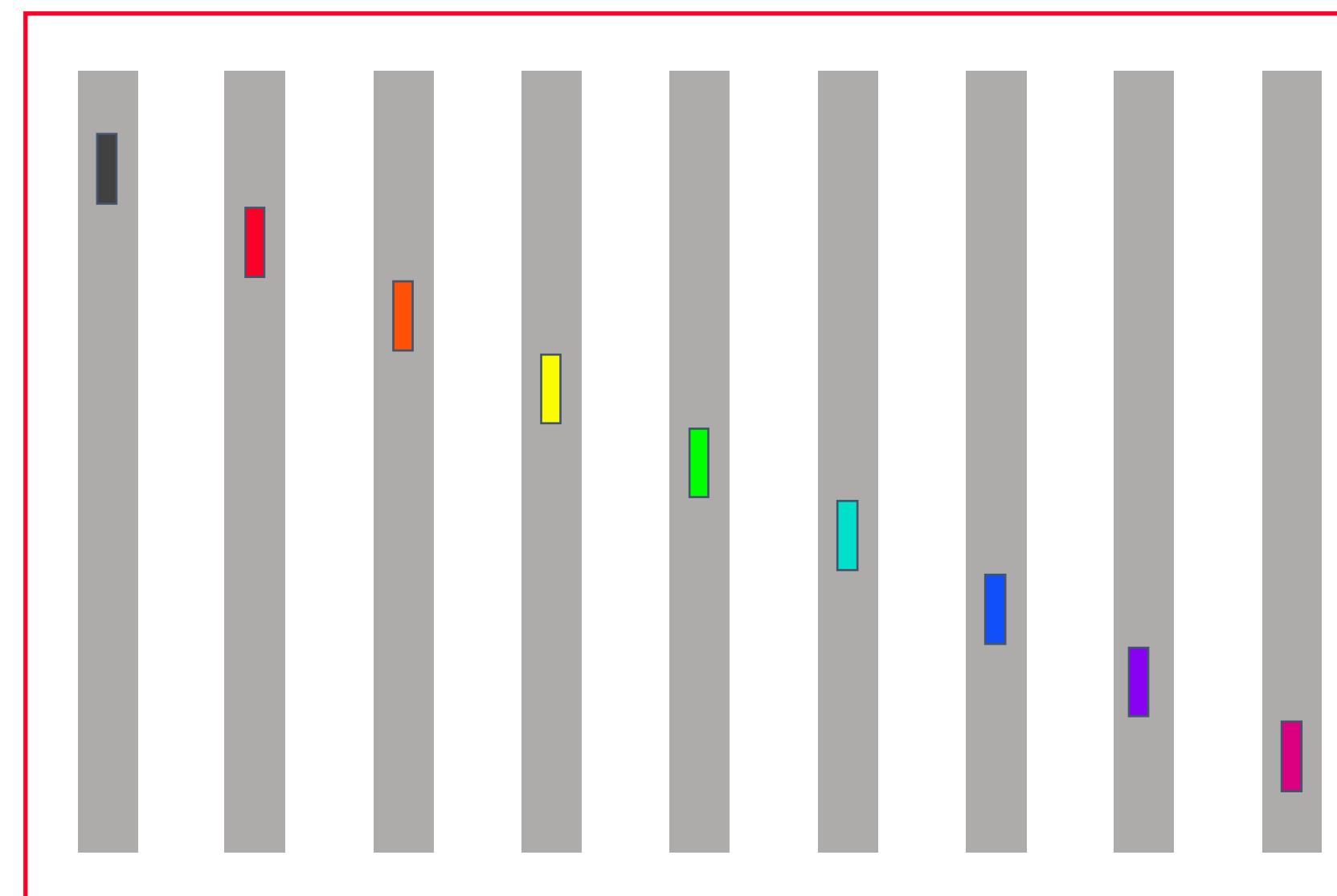
After



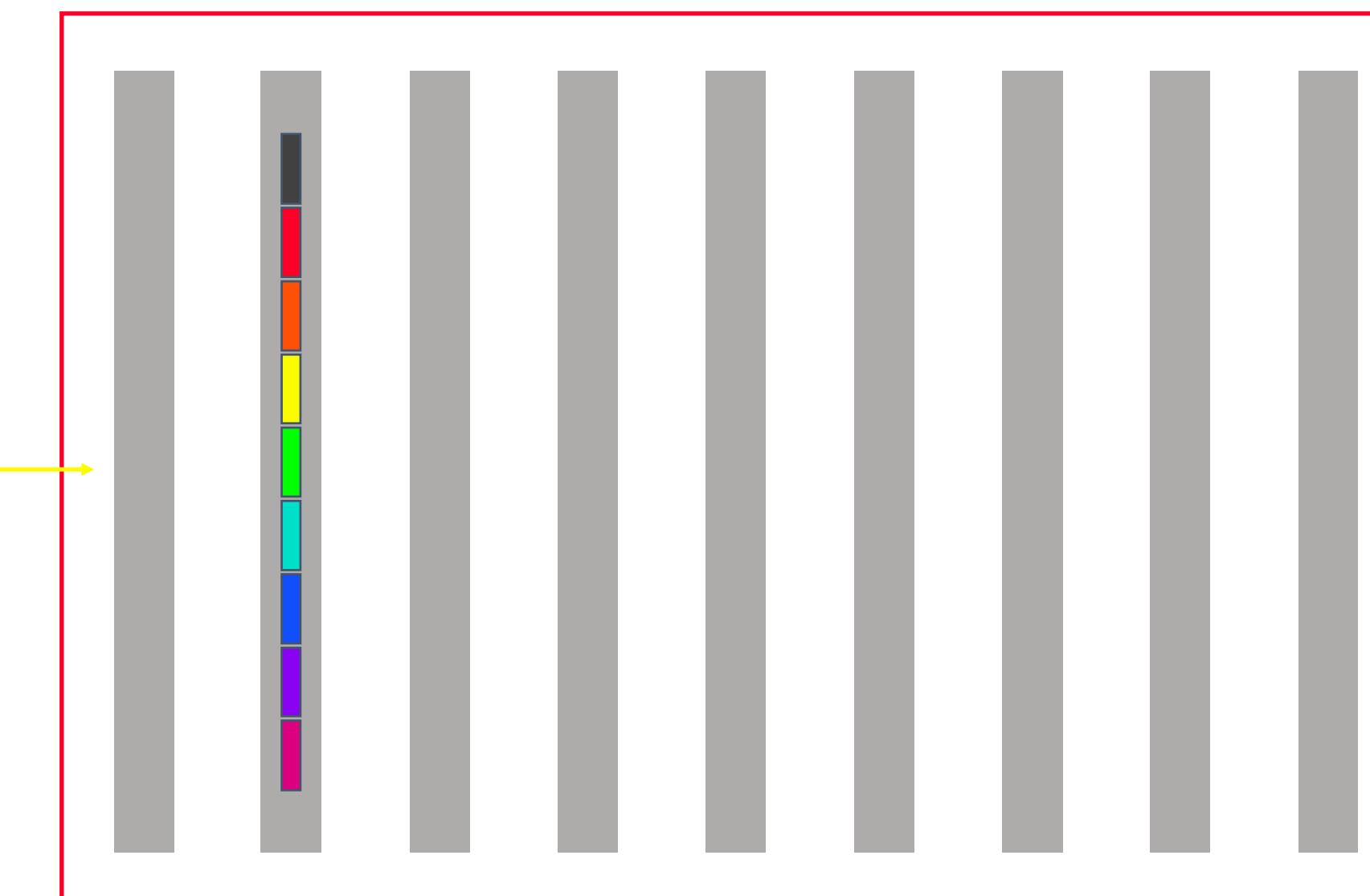
Gather

Notice: Gather as implemented before using MST was optimal in bandwidth as well (how to prove?)

Before

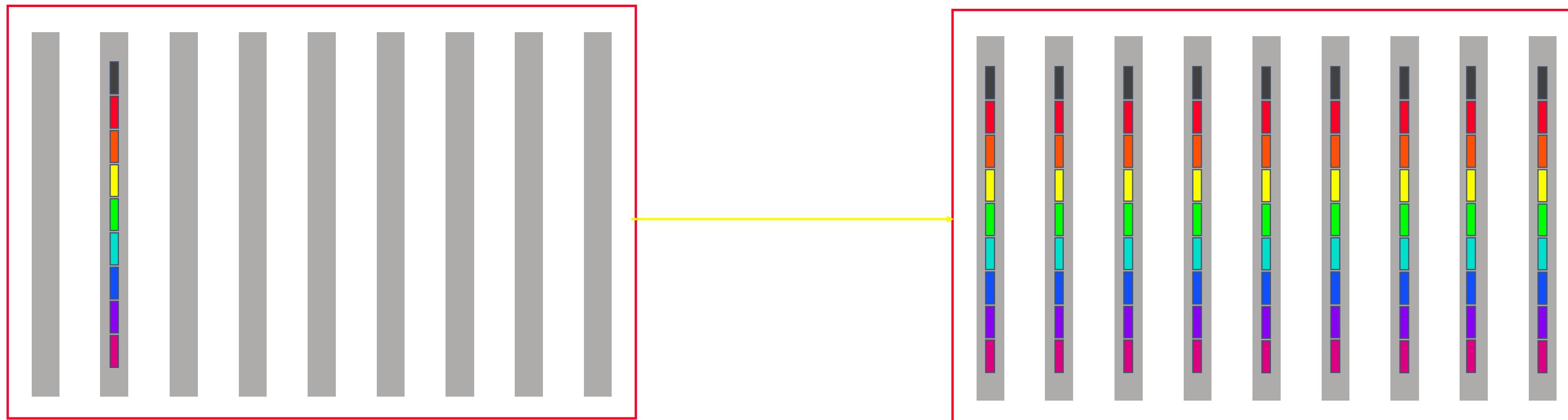


After

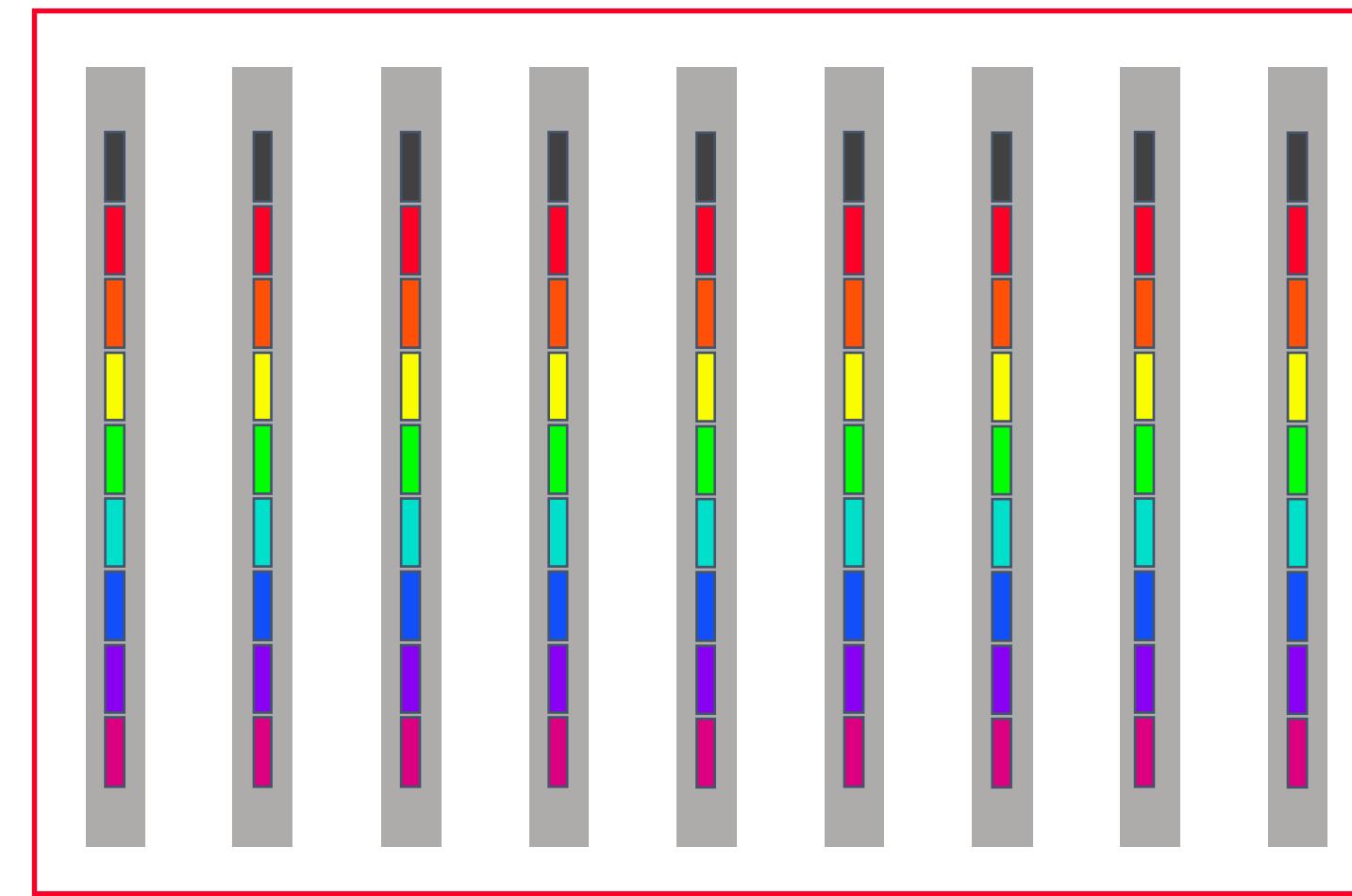
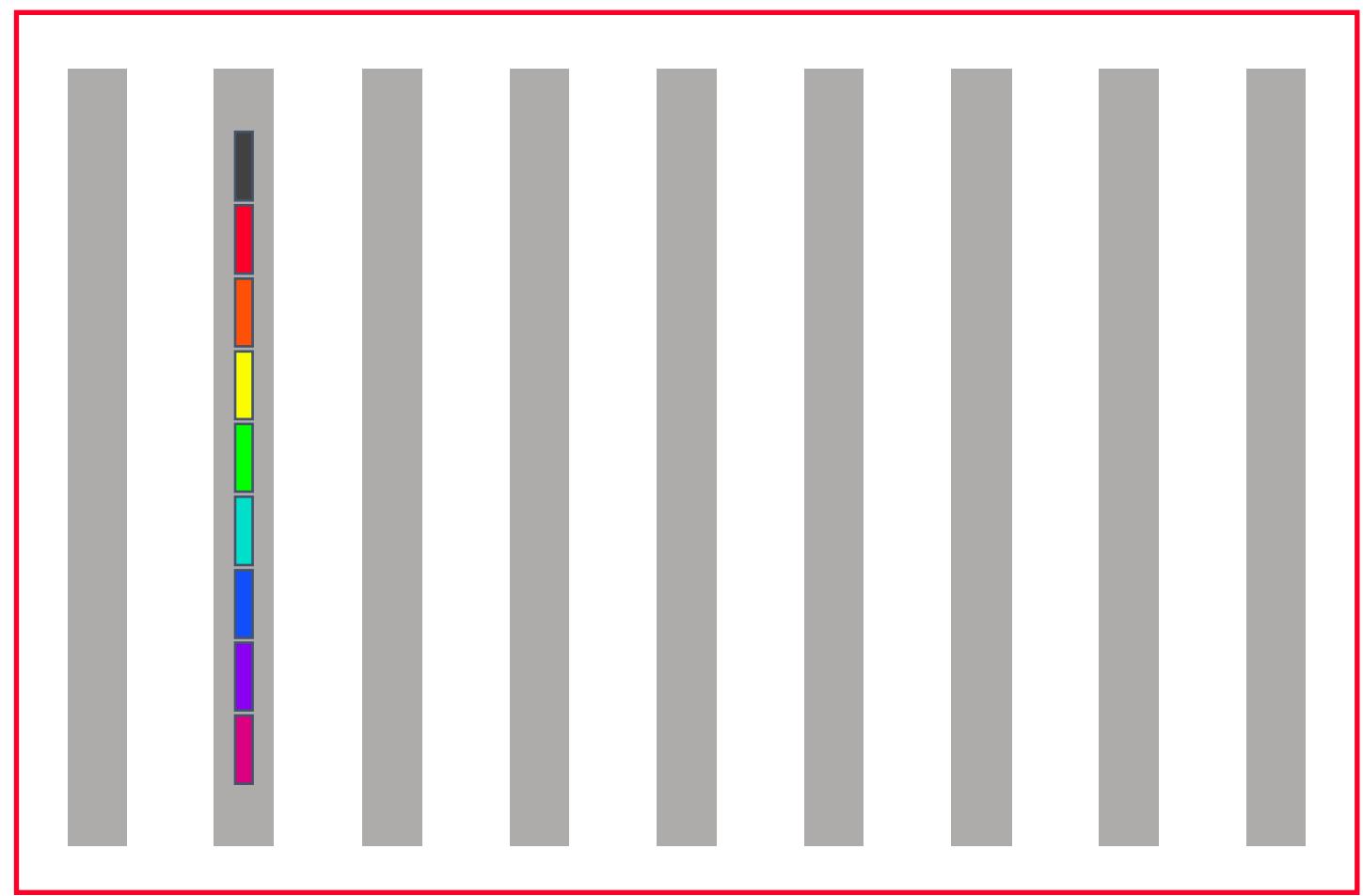


Using the building blocks

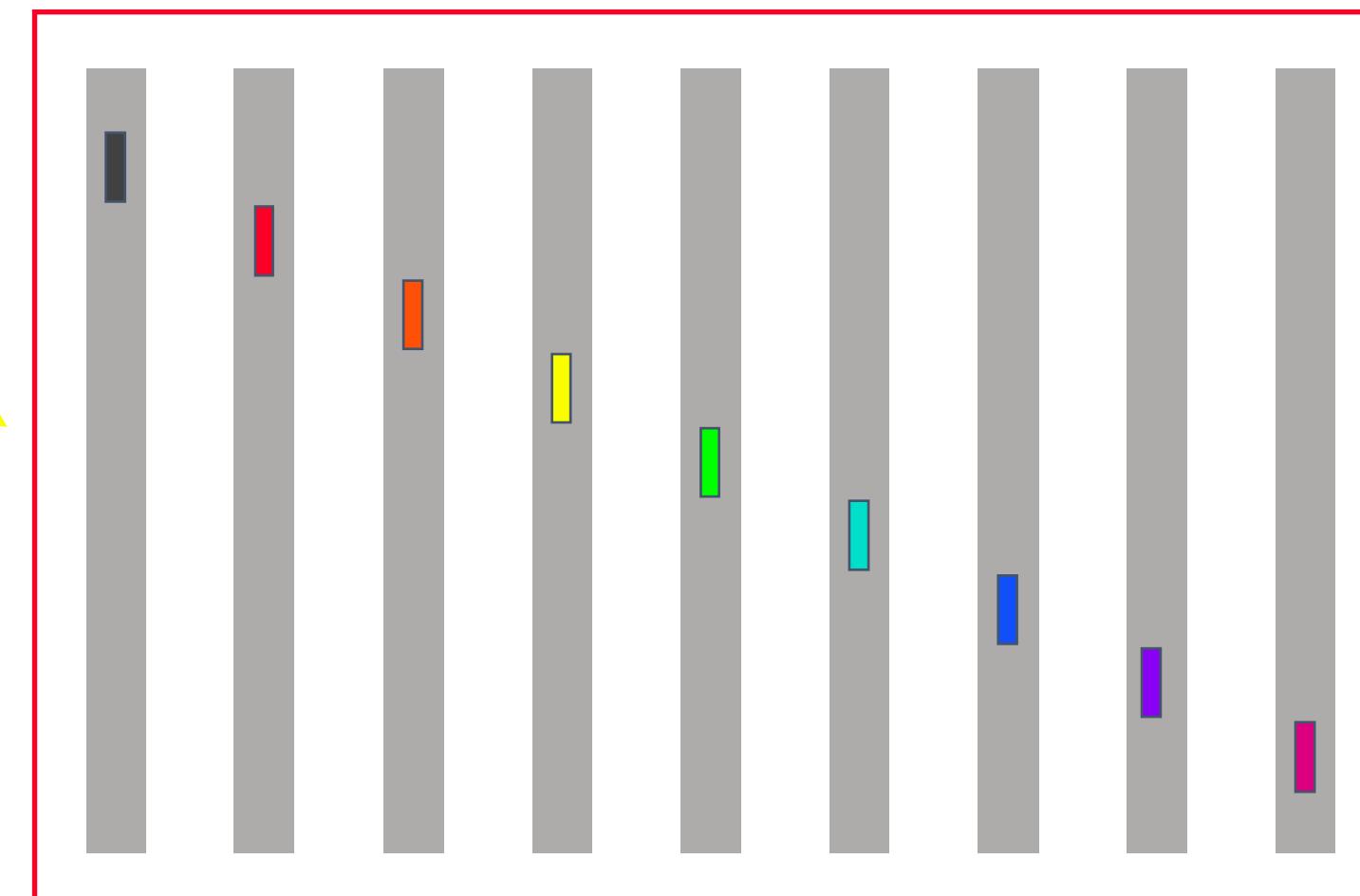
Broadcast (Large Message)



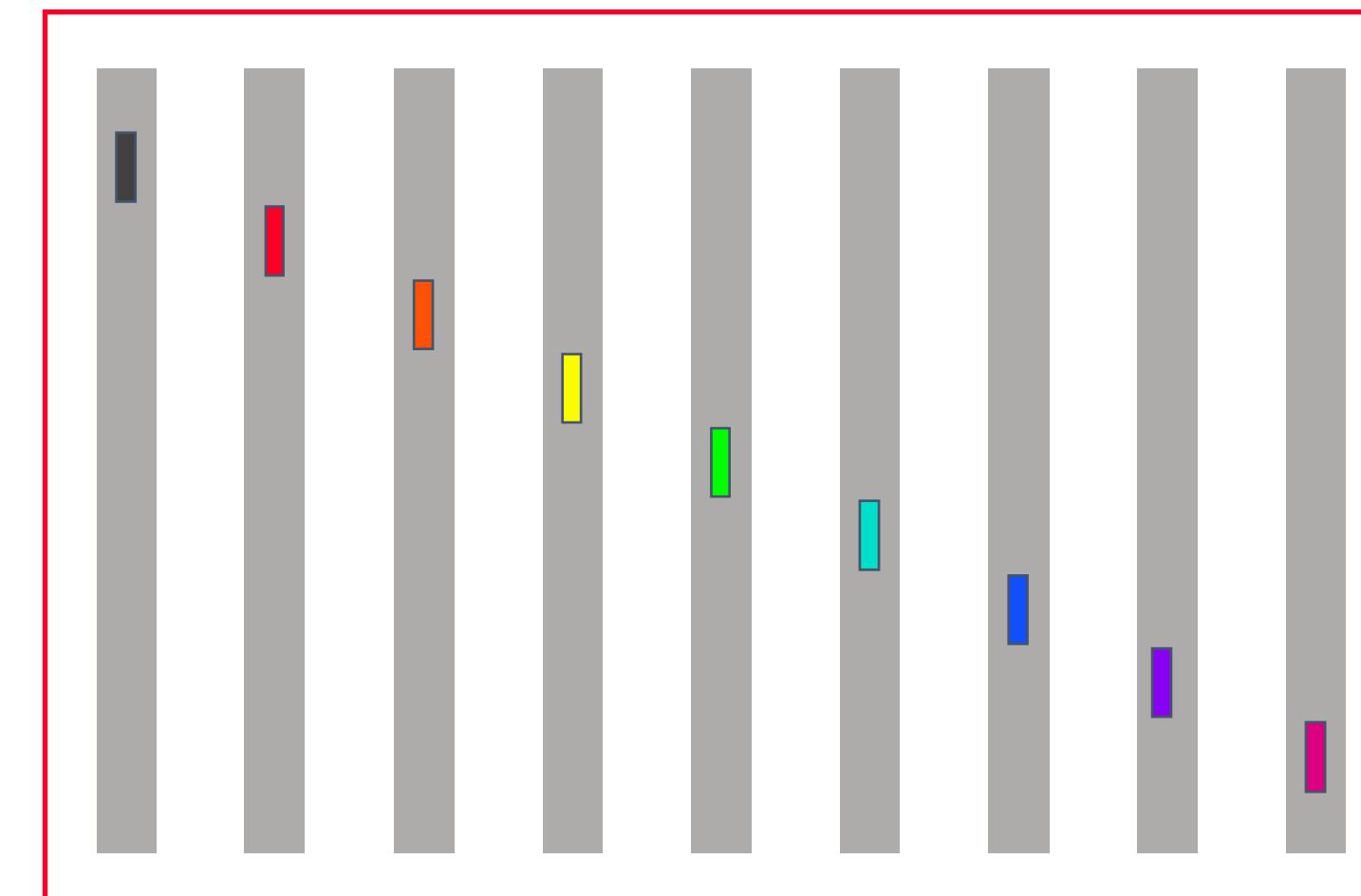
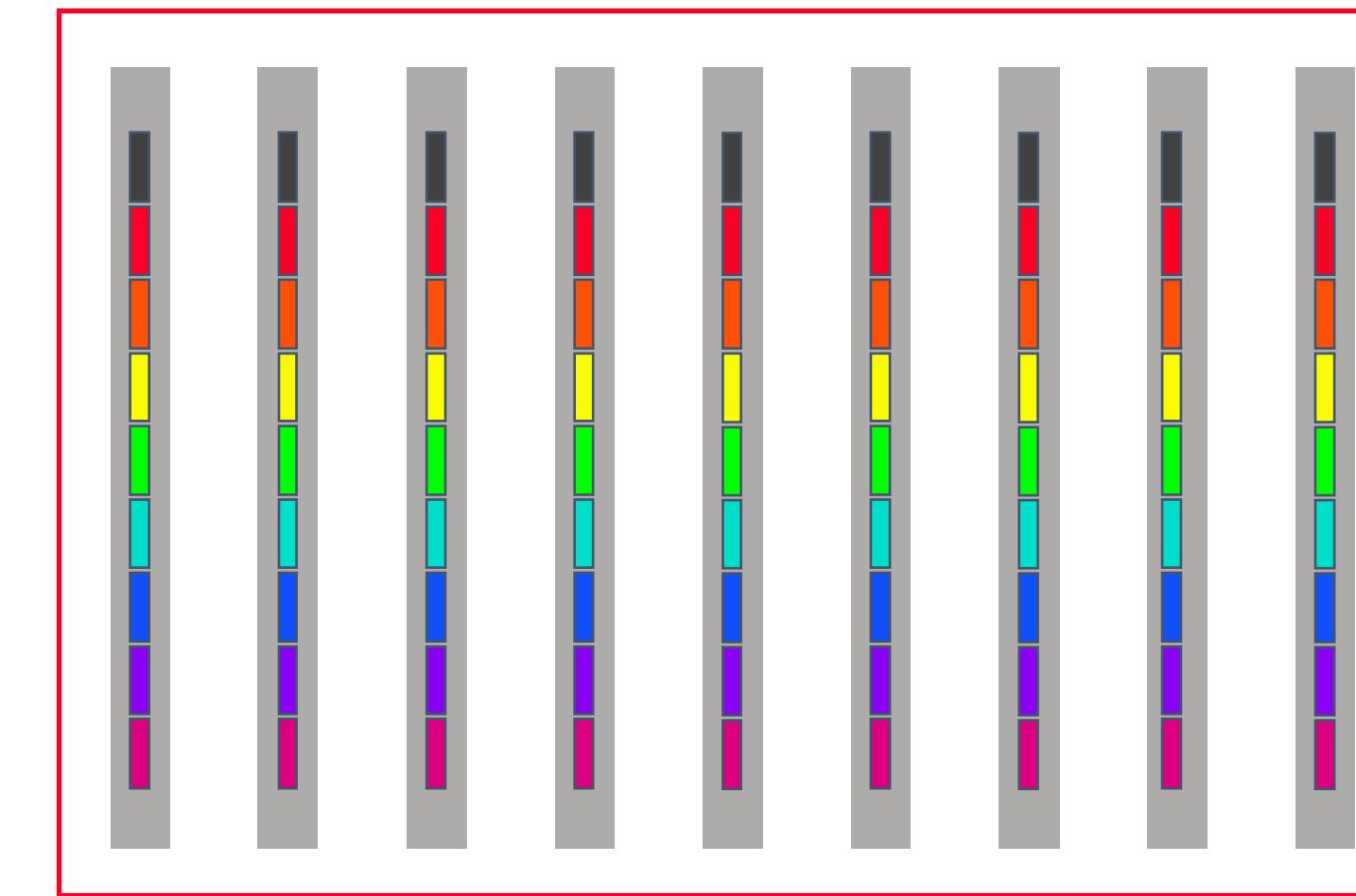
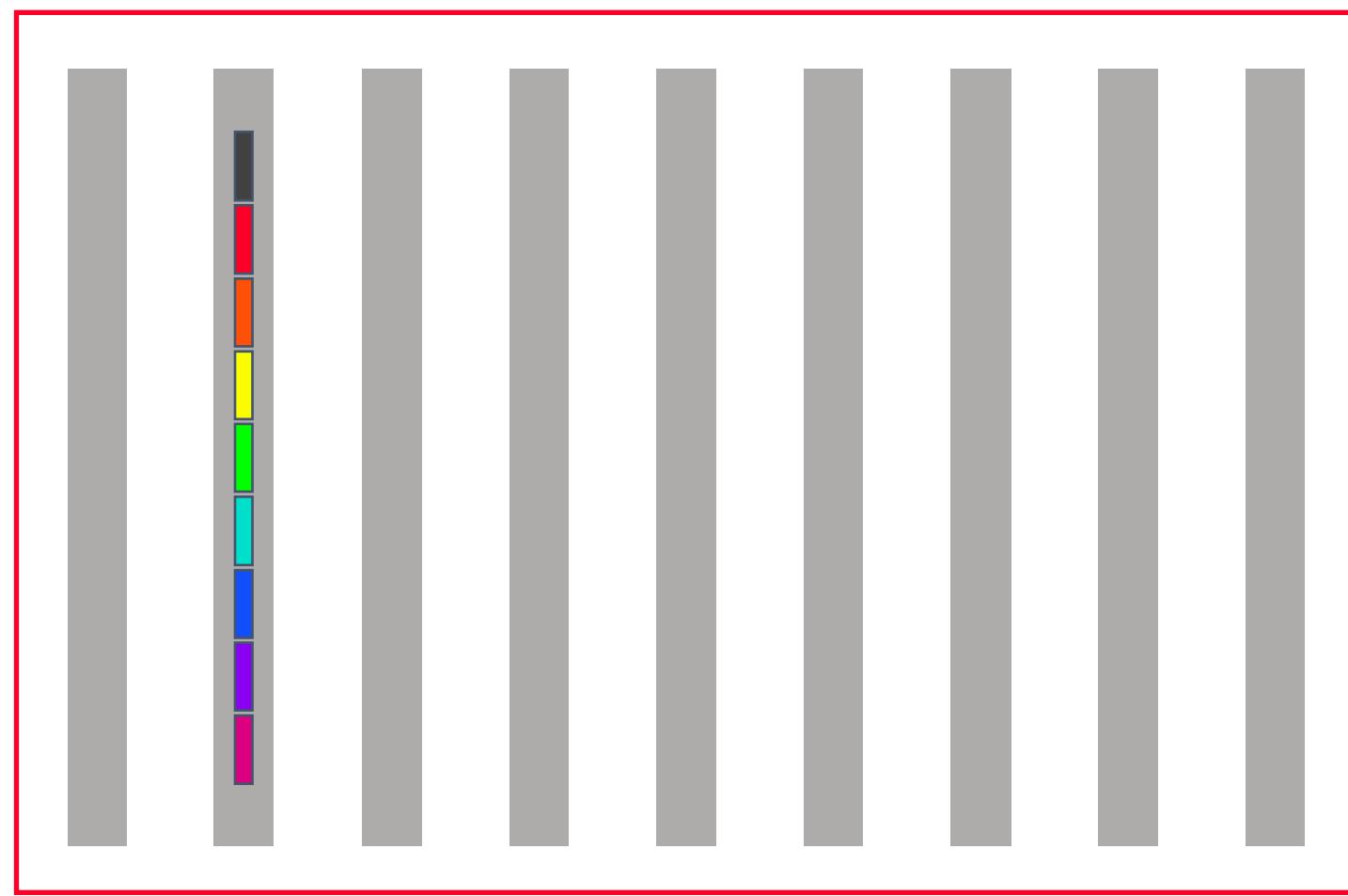
Broadcast (long vector)



Scatter



Broadcast (long vector)



Allgather

Cost of scatter/allgather broadcast

- Assumption: power of two number of nodes

scatter

$$\log(p)\alpha + \frac{p-1}{p}n\beta$$

allgather

$$(p-1)\alpha + \frac{p-1}{p}n\beta$$

$$(\log(p) + p - 1)\alpha + 2\frac{p-1}{p}n\beta$$

Cost of scatter/allgather broadcast

- Assumption: power of two number of nodes

scatter

$$\log(p)\alpha + \frac{p-1}{p}n\beta$$

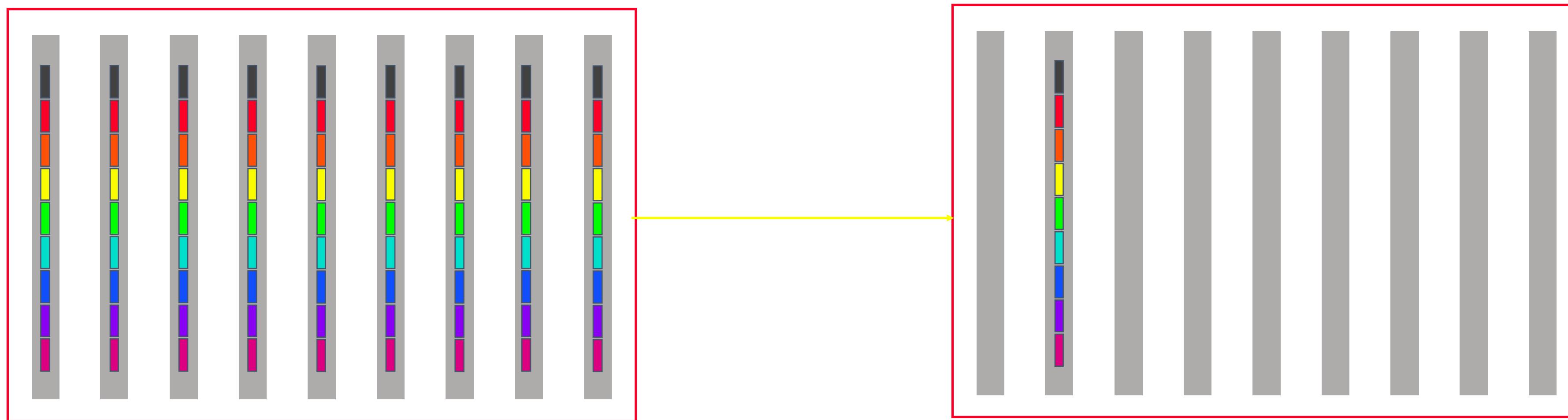
allgather

$$(p-1)\alpha + \frac{p-1}{p}n\beta$$

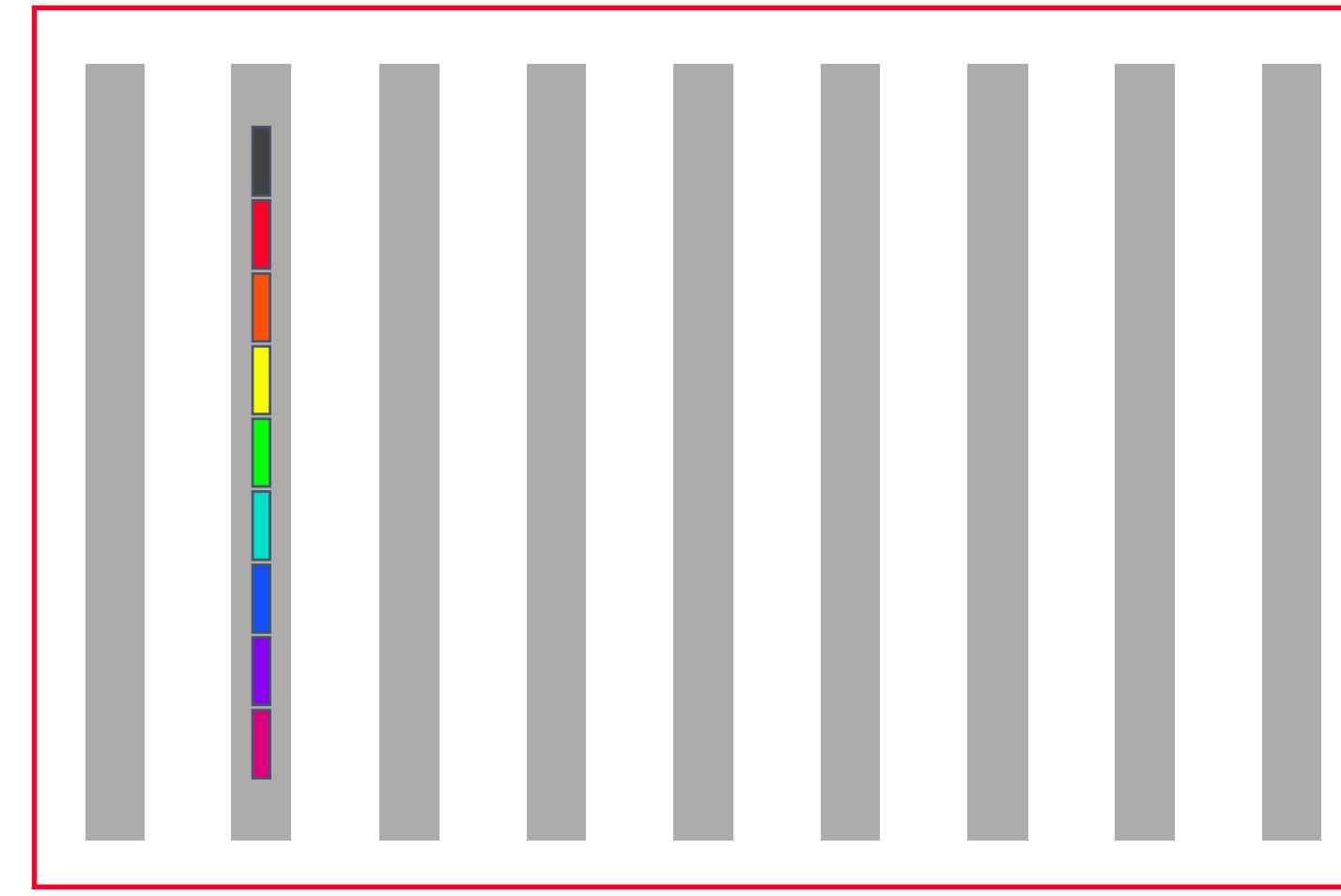
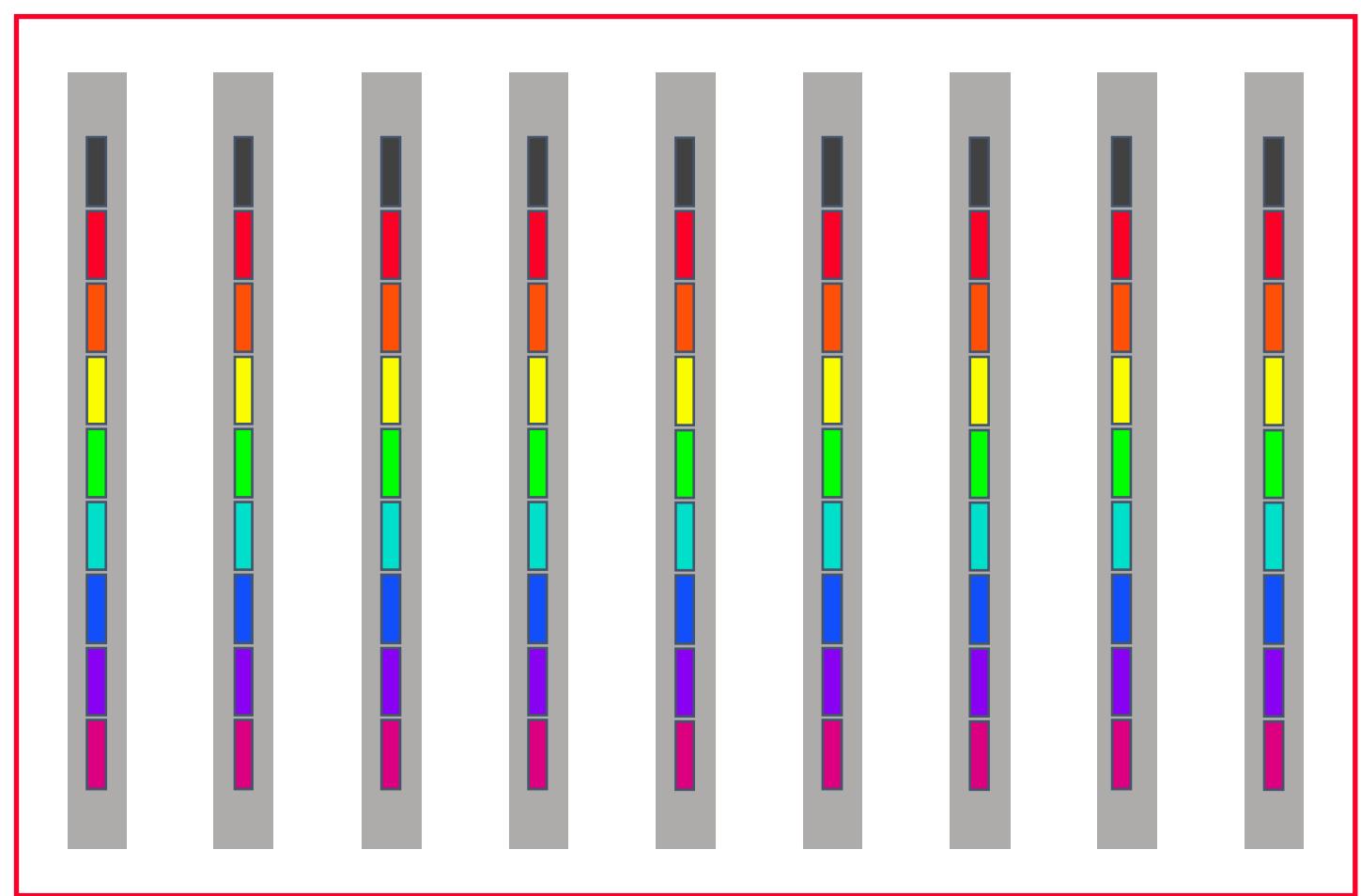
$$(\log(p) + p - 1)\alpha + 2\frac{p-1}{p}n\beta$$

Vs. MST broadcast: $\lceil \log(p) \rceil (\alpha + n\beta)$

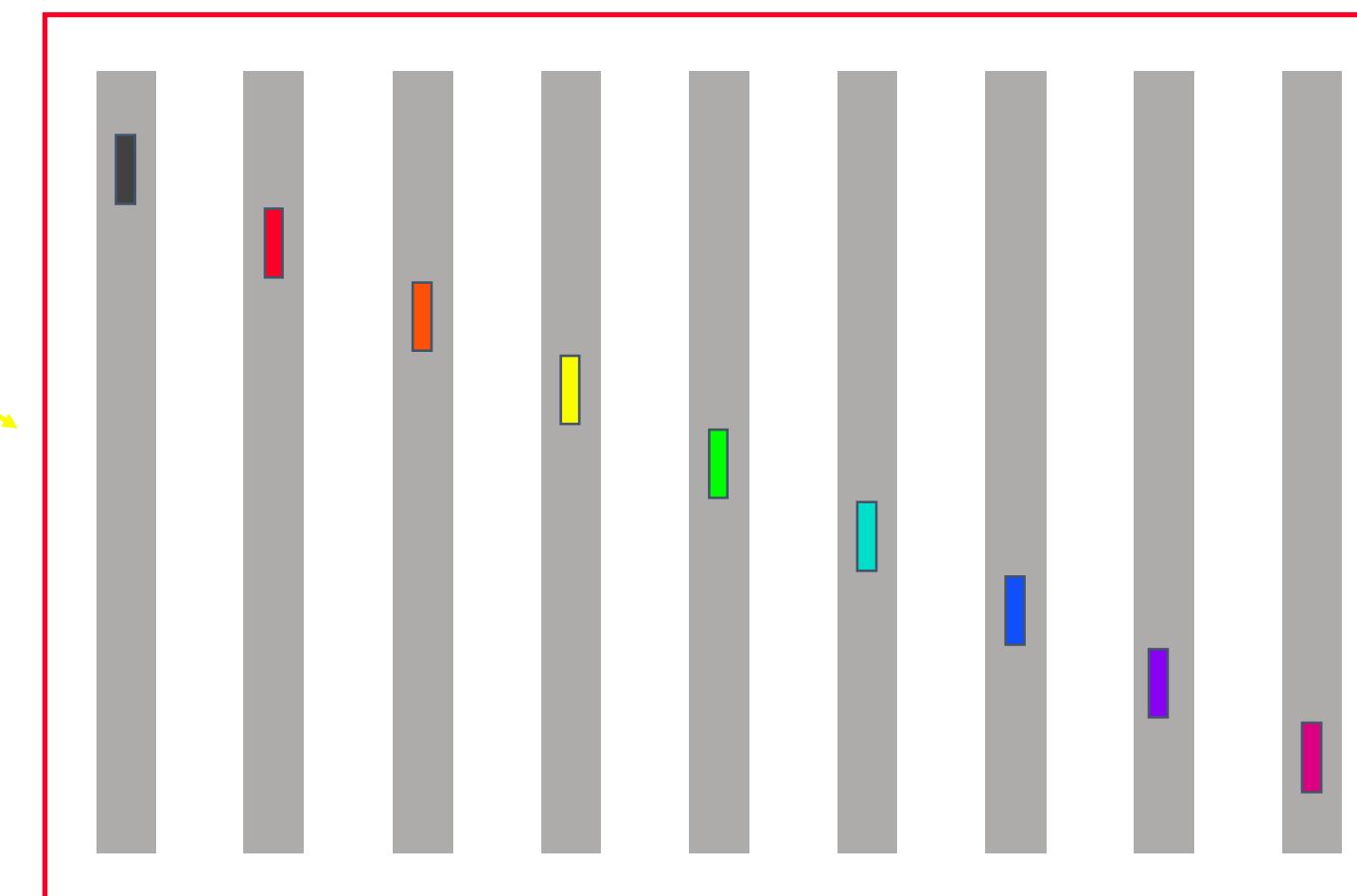
Reduce(-to-one) (long vector)



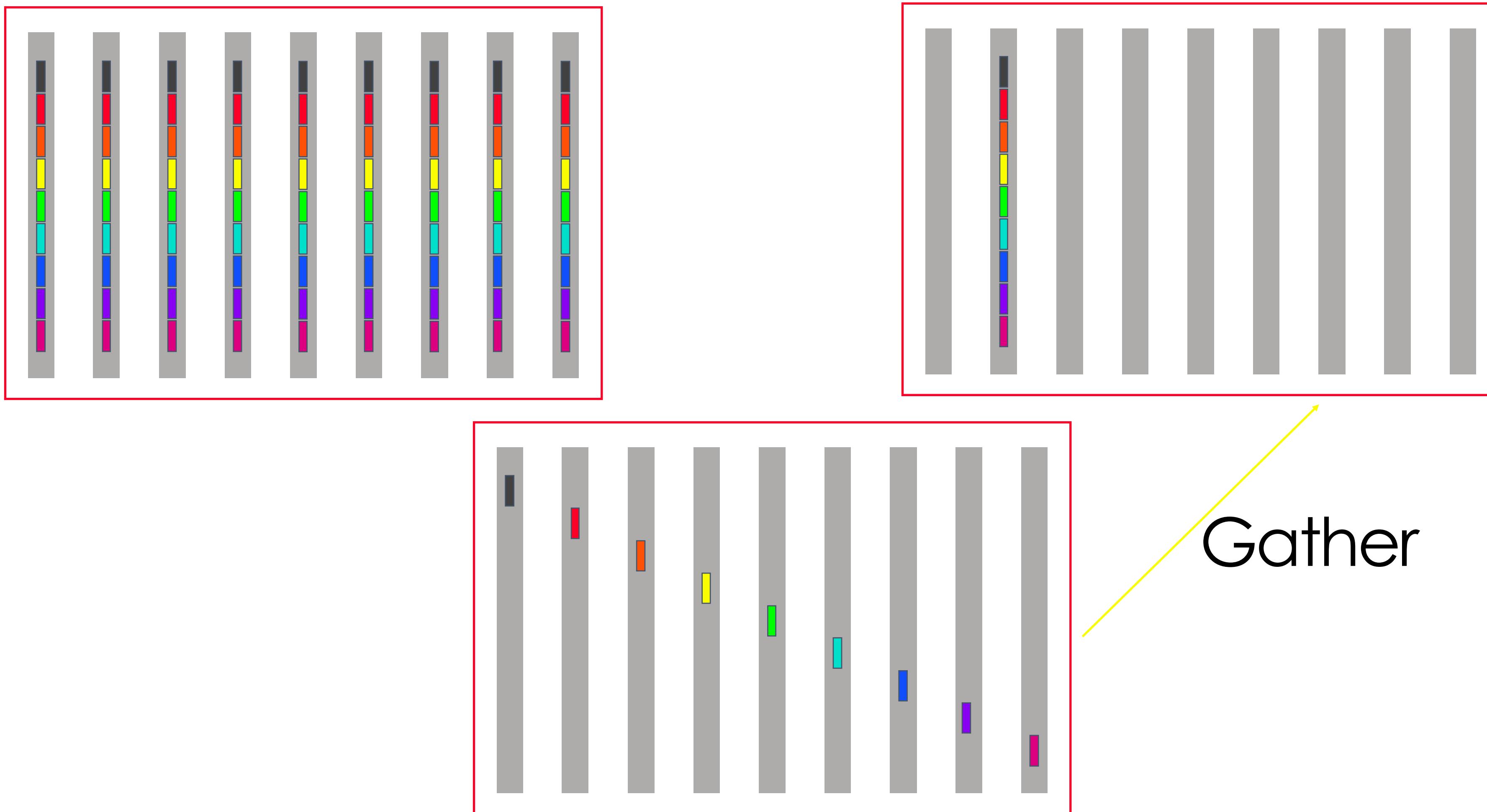
Reduce (long vector)



Reduce-scatter



Combine-to-one (long vector)



Cost of Reduce-scatter/Gather Reduce(-to-one)

- Assumption: power of two number of nodes

$$\frac{\text{Reduce-scatter } (p-1)\alpha + \frac{p-1}{p}n\beta + \frac{p-1}{p}n\gamma}{\text{gather } \log(p)\alpha + \frac{p-1}{p}n\beta}$$

$$(log(p) + p - 1)\alpha + 2\frac{p-1}{p}n\beta + \frac{p-1}{p}n\gamma$$

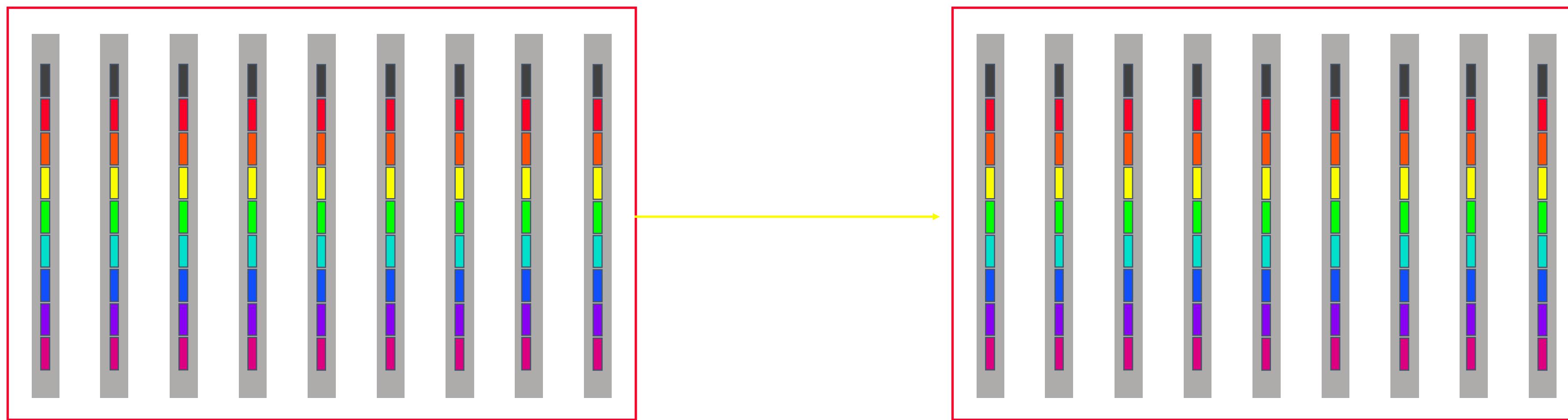
Cost of Reduce-scatter/Gather Reduce(-to-one)

- Assumption: power of two number of nodes

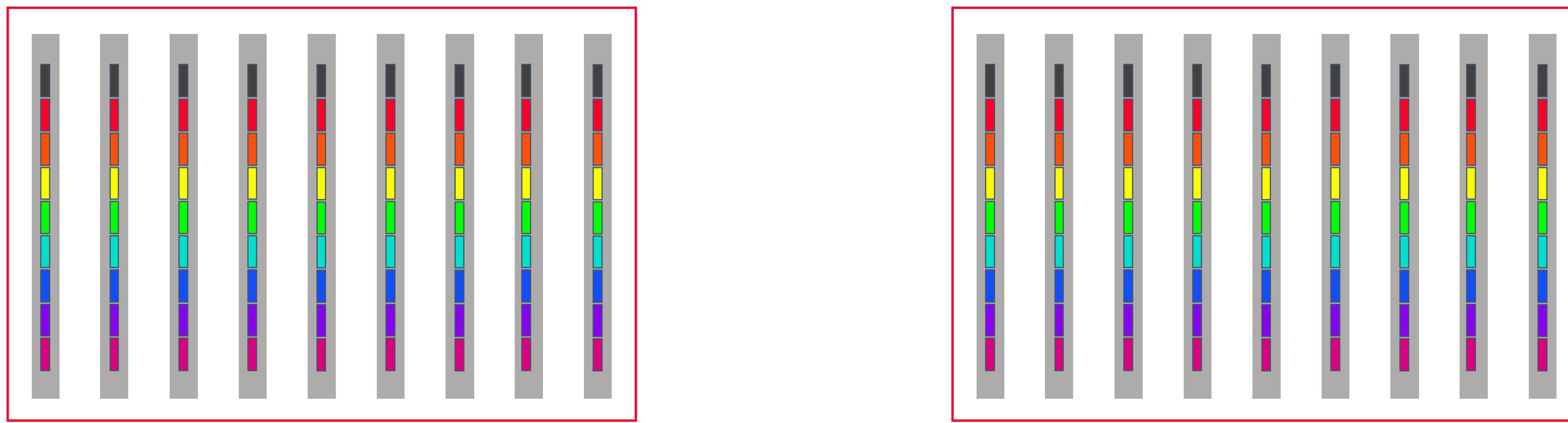
$$\begin{array}{l} \text{Reduce-scatter } (p-1)\alpha + \frac{p-1}{p} n\beta + \frac{p-1}{p} n\gamma \\ \text{gather } \log(p)\alpha + \frac{p-1}{p} n\beta \\ \hline (\log(p) + p - 1)\alpha + 2\frac{p-1}{p} n\beta + \frac{p-1}{p} n\gamma \end{array}$$

Vs. MST reduce: $\lceil \log(p) \rceil (\alpha + n\beta + n\gamma)$

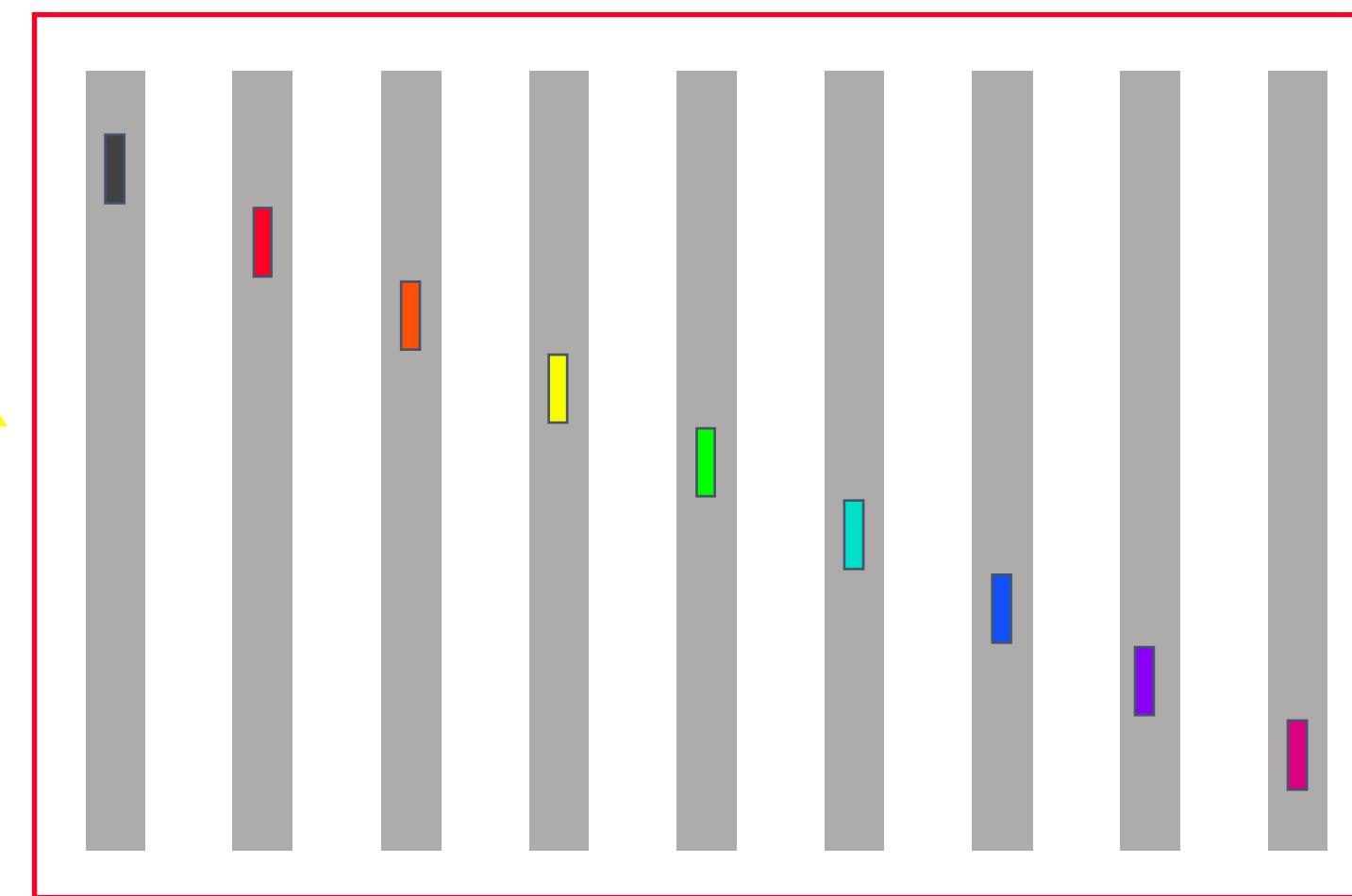
Allreduce (Large Message)



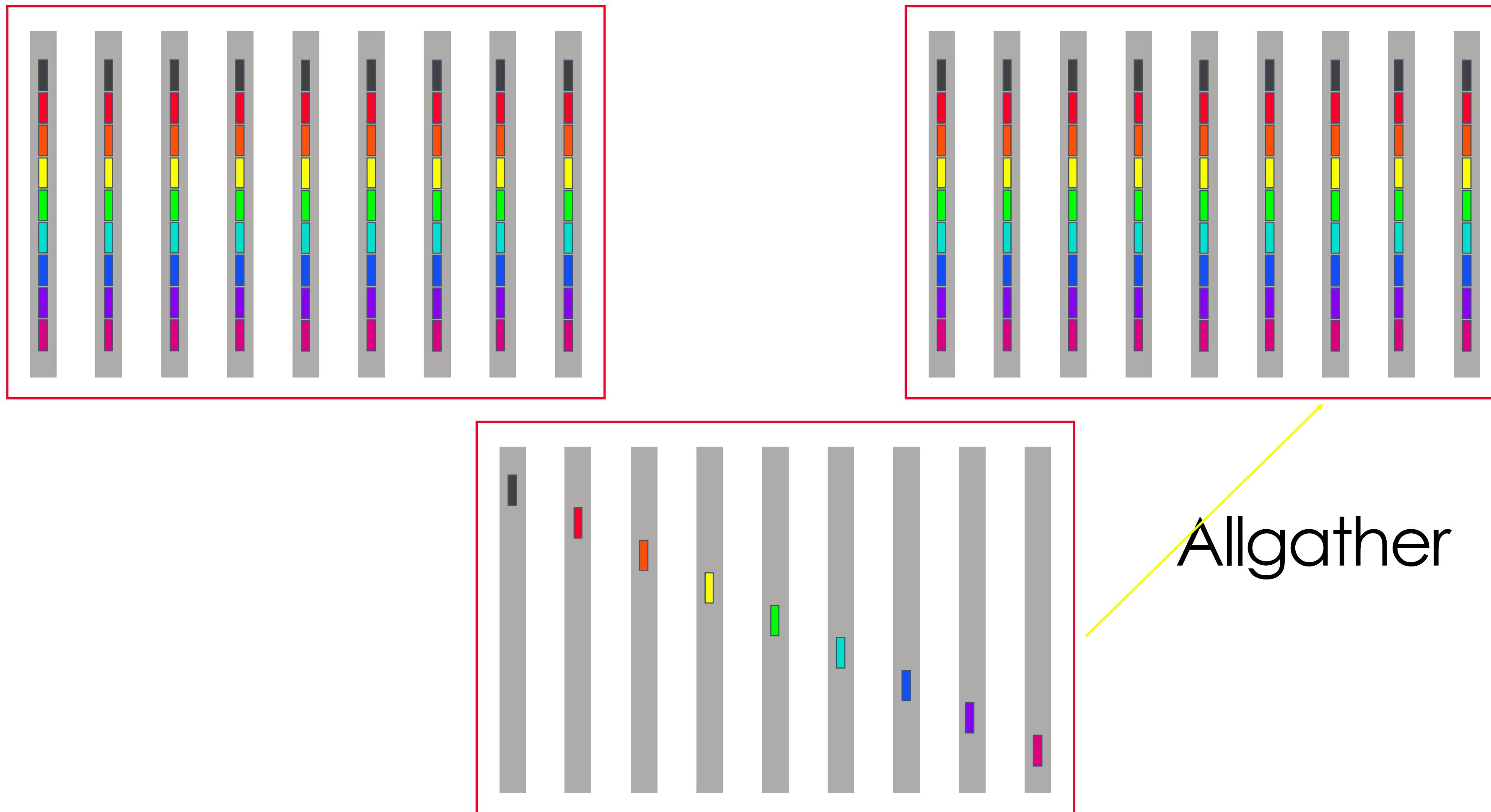
Allreduce (Large Message)



Reduce-scatter



Allreduce (long vector)



Cost of Reduce-scatter/Allgather Allreduce

- Assumption: power of two number of nodes

$$\frac{\text{Reduce-scatter}((p-1)\alpha + \frac{p-1}{p}n\beta + \frac{p-1}{p}n\gamma)}{\text{Allgather}((p-1)\alpha + \frac{p-1}{p}n\beta) / 2(p-1)\alpha + 2\frac{p-1}{p}n\beta + \frac{p-1}{p}n\gamma}$$

Cost of Reduce-scatter/Allgather Allreduce

- Assumption: power of two number of nodes

$$\frac{\text{Reduce-scatter} \quad (p-1)\alpha + \frac{p-1}{p}n\beta + \frac{p-1}{p}n\gamma}{\text{Allgather} \quad (p-1)\alpha + \frac{p-1}{p}n\beta} = \frac{2(p-1)\alpha + 2\frac{p-1}{p}n\beta + \frac{p-1}{p}n\gamma}{2(p-1)\alpha + 2\frac{p-1}{p}n\beta + \frac{p-1}{p}n\gamma}$$

Vs. Reduce-broadcast
allreduce

$$2\log(p)\alpha + 2\log(p)n\beta + \log(p)n\gamma$$

Recap

Reduce-scatter

$$(p-1)\alpha + \frac{p-1}{p}n(\beta + \gamma)$$

Scatter

$$\log(p)\alpha + \frac{p-1}{p}n\beta$$

Gather

$$\log(p)\alpha + \frac{p-1}{p}n\beta$$

Allgather

$$(p-1)\alpha + \frac{p-1}{p}n\beta$$

Reduce(-to-one)

Allreduce

Broadcast

Recap

Reduce-scatter
 $(p-1)\alpha + \frac{p-1}{p}n(\beta + \gamma)$

Scatter
 $\log(p)\alpha + \frac{p-1}{p}n\beta$

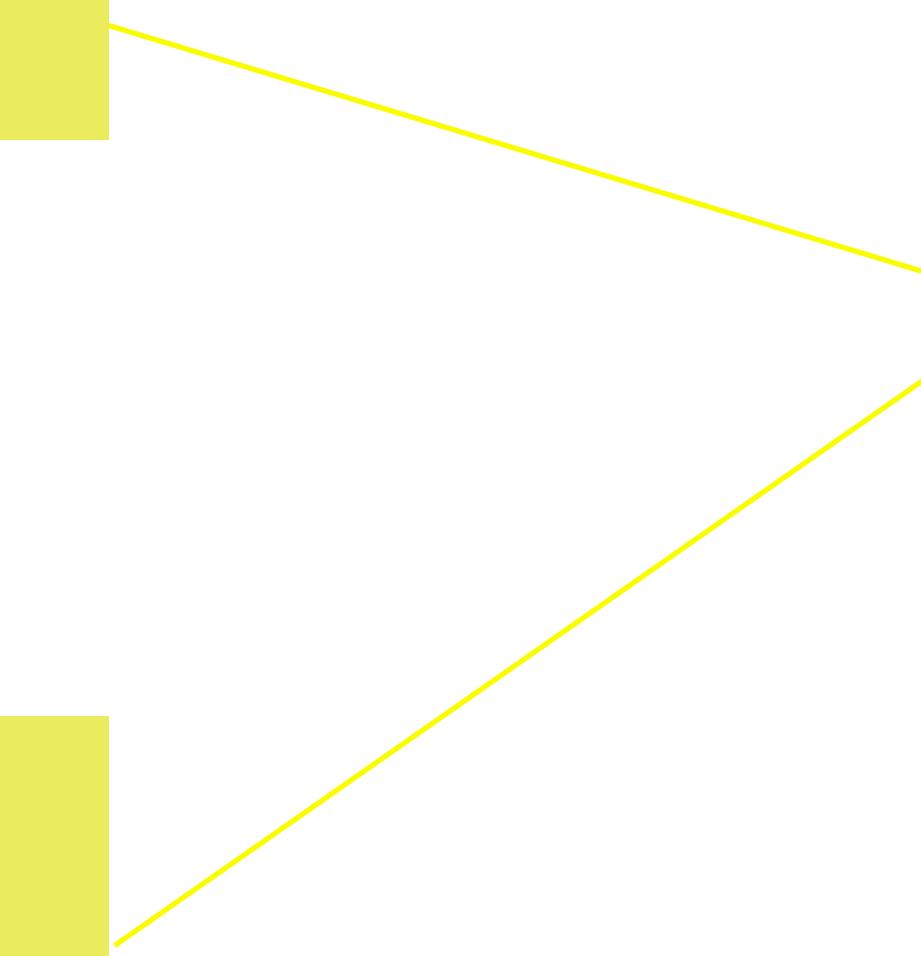
Gather
 $\log(p)\alpha + \frac{p-1}{p}n\beta$

Allgather
 $(p-1)\alpha + \frac{p-1}{p}n\beta$

Reduce(-to-one)
 $(p - 1 + \log(p))\alpha + \frac{p-1}{p}n(2\beta + \gamma)$

Allreduce

Broadcast



Recap

Reduce-scatter

$$(p-1)\alpha + \frac{p-1}{p}n(\beta + \gamma)$$

Scatter
 $\log(p)\alpha + \frac{p-1}{p}n\beta$

Gather

$$\log(p)\alpha + \frac{p-1}{p}n\beta$$

Allgather
 $(p-1)\alpha + \frac{p-1}{p}n\beta$

Reduce(-to-one)

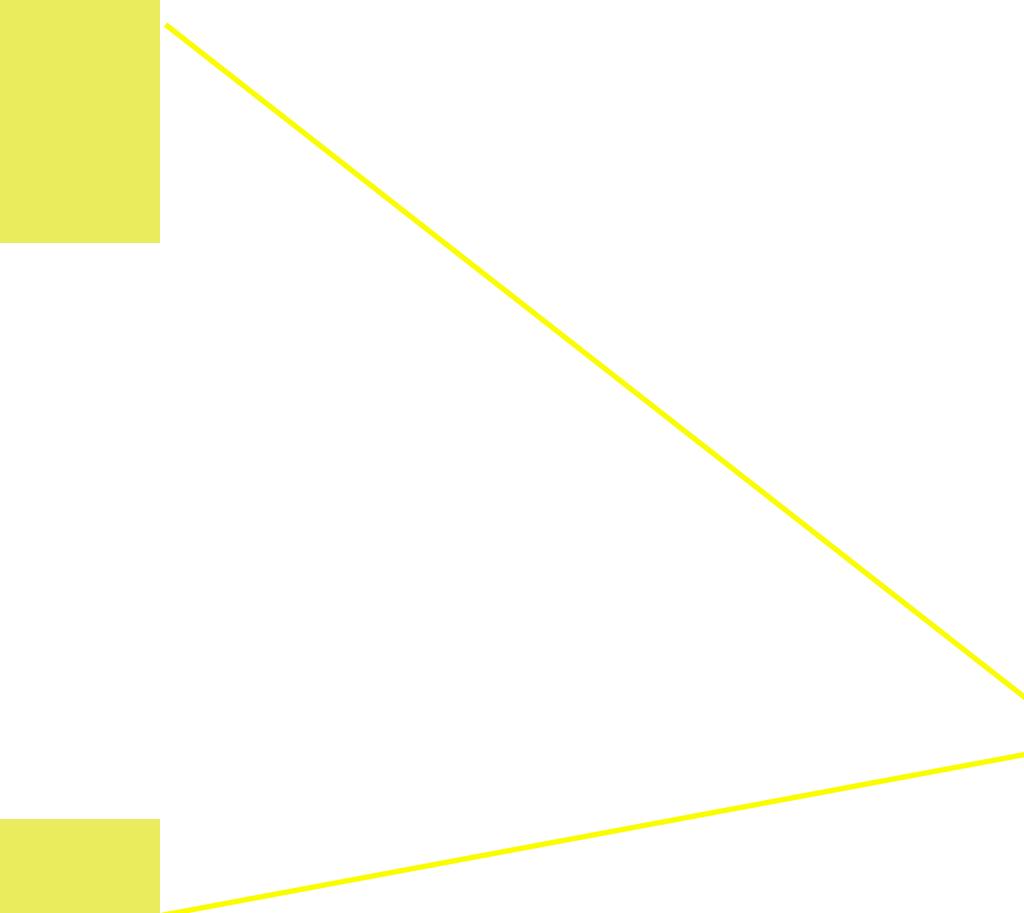
$$(p - 1 + \log(p))\alpha + \frac{p-1}{p} n(2\beta + \gamma)$$

Allreduce

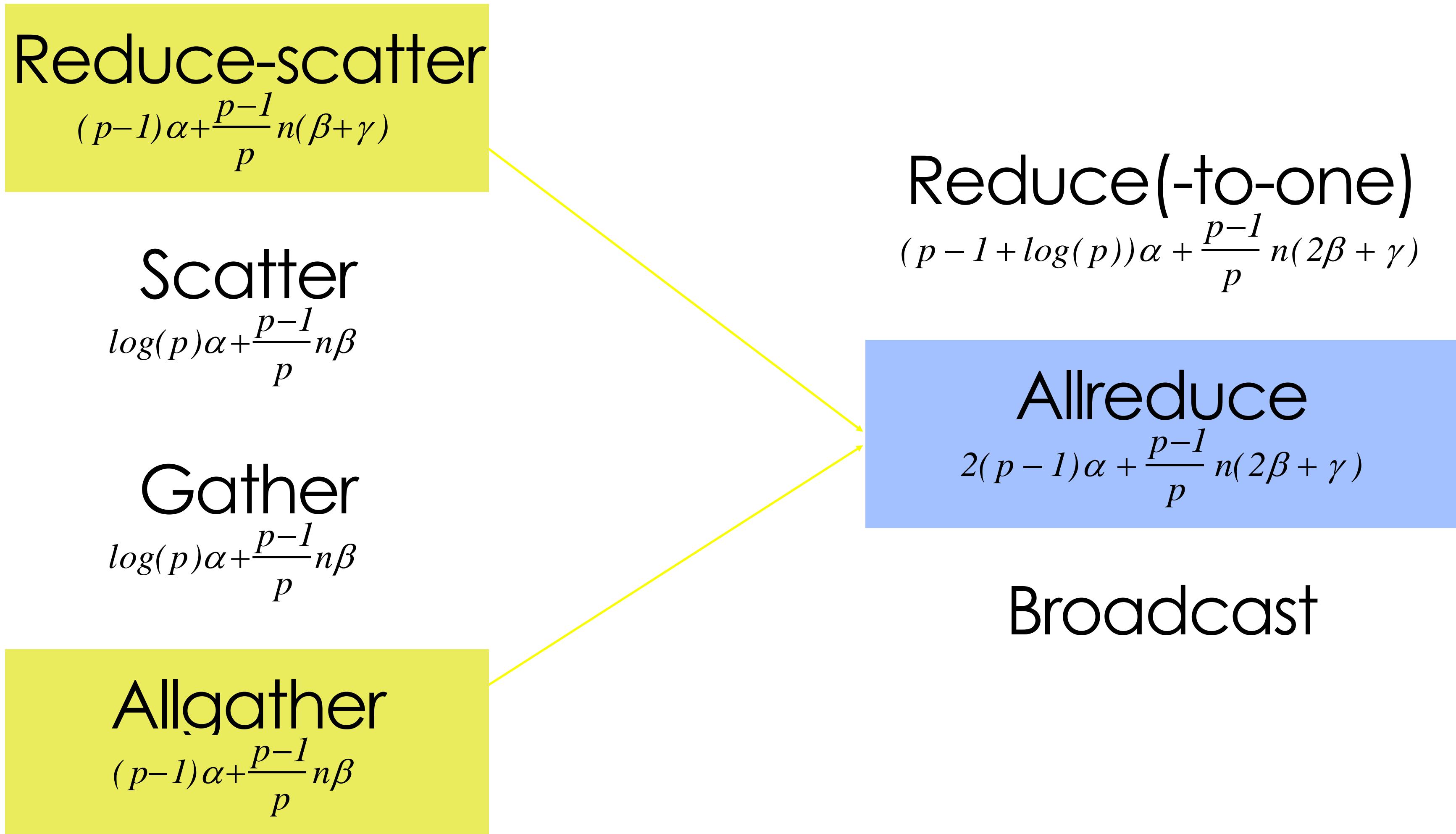
$$2(p - 1)\alpha + \frac{p-1}{p} n(2\beta + \gamma)$$

Broadcast

$$(\log(p) + p - 1)\alpha + 2\frac{p-1}{p} n\beta$$



Recap



Recap

Reduce-scatter

$$(p-1)\alpha + \frac{p-1}{p}n(\beta + \gamma)$$

Scatter

$$\log(p)\alpha + \frac{p-1}{p}n\beta$$

Gather

$$\log(p)\alpha + \frac{p-1}{p}n\beta$$

Allgather

$$(p-1)\alpha + \frac{p-1}{p}n\beta$$



Reduce(-to-one)

$$(p - 1 + \log(p))\alpha + \frac{p-1}{p} n(2\beta + \gamma)$$

Allreduce

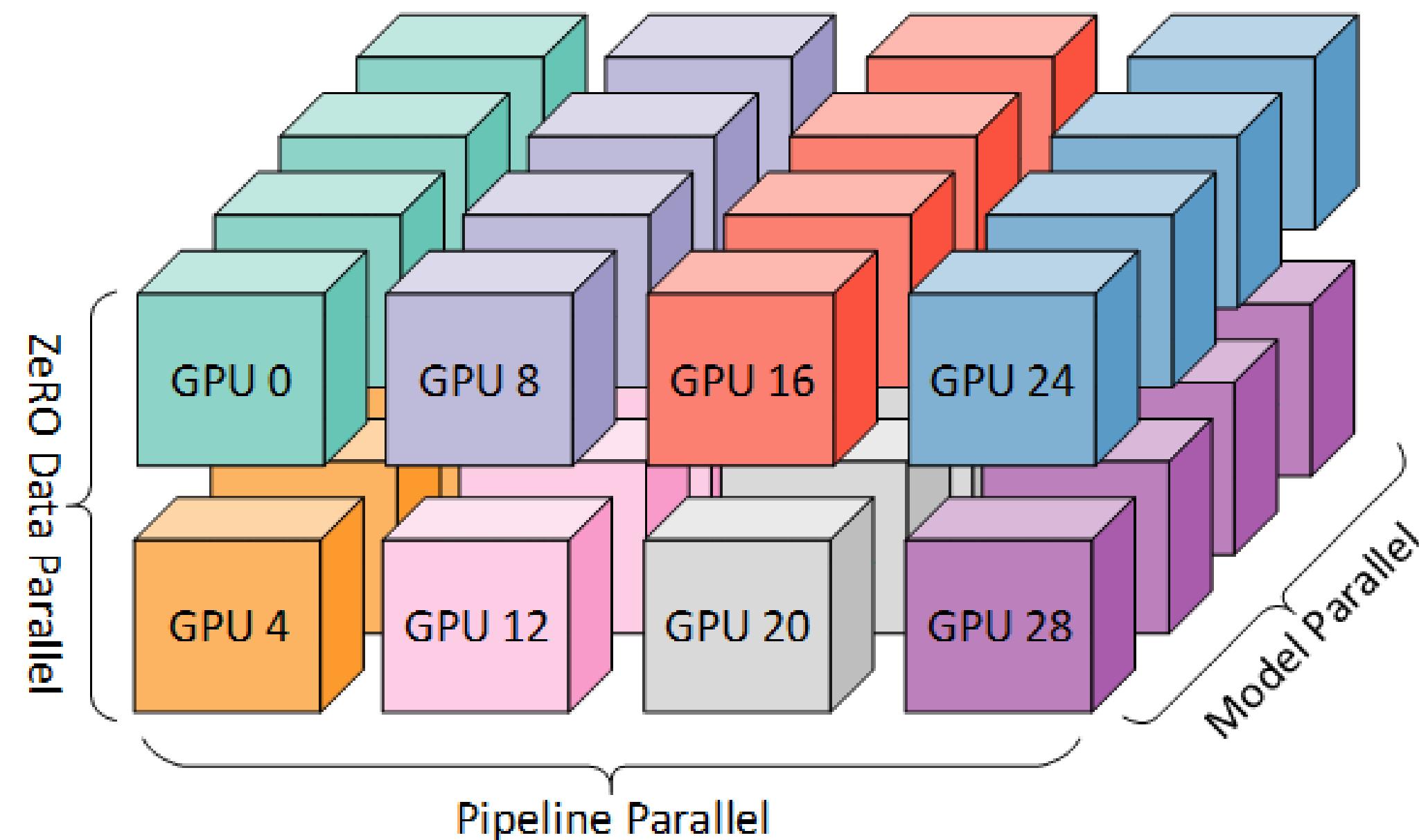
$$2(p - 1)\alpha + \frac{p-1}{p} n(2\beta + \gamma)$$

Broadcast

$$(\log(p) + p - 1)\alpha + 2\frac{p-1}{p} n\beta$$

A More Complicate Case

- Real Cluster to train ChatGPT:
 - If using GPU: 2D Mesh
 - If using TPU: 3D Mesh, see figure below



Example: 2D Broadcast



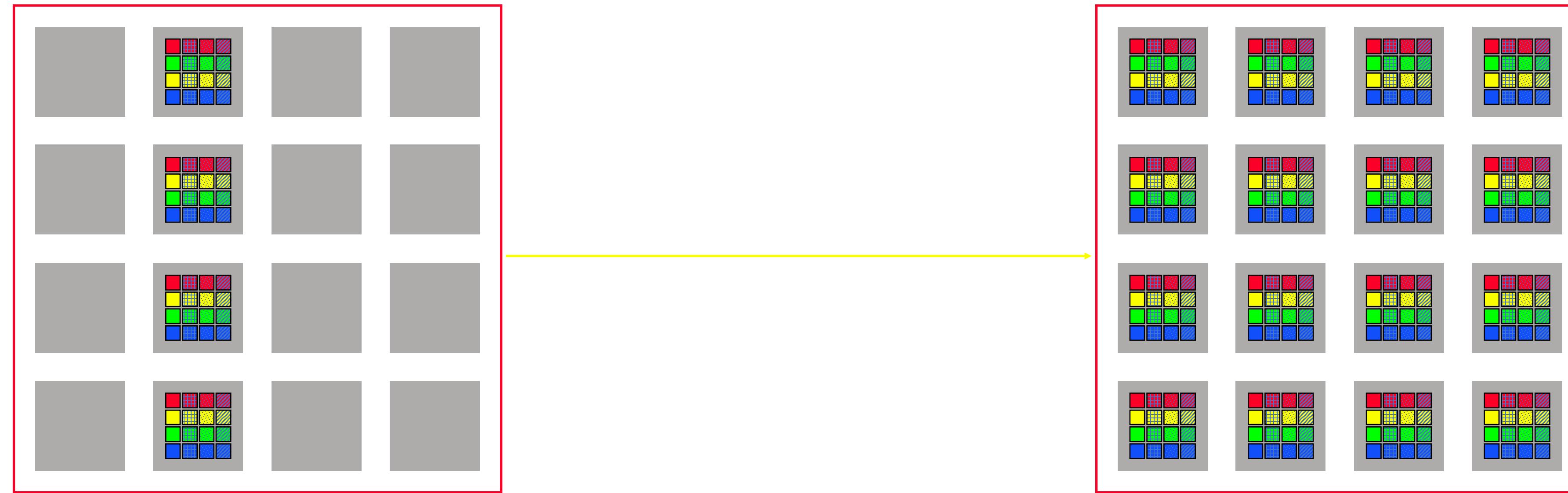
- Idea: Use 1D to compose 2

Example: 2D Broadcast



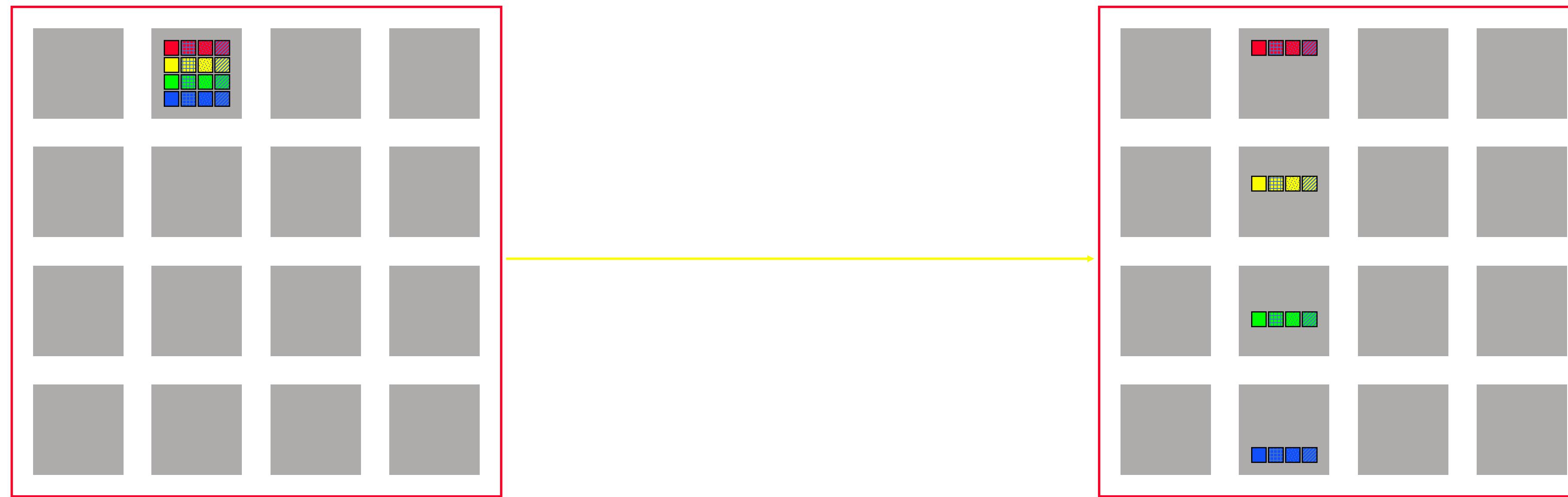
- Idea: Use 1D to compose 2
- Option 1:
 - **MST broadcast in column**

Example: 2D Broadcast



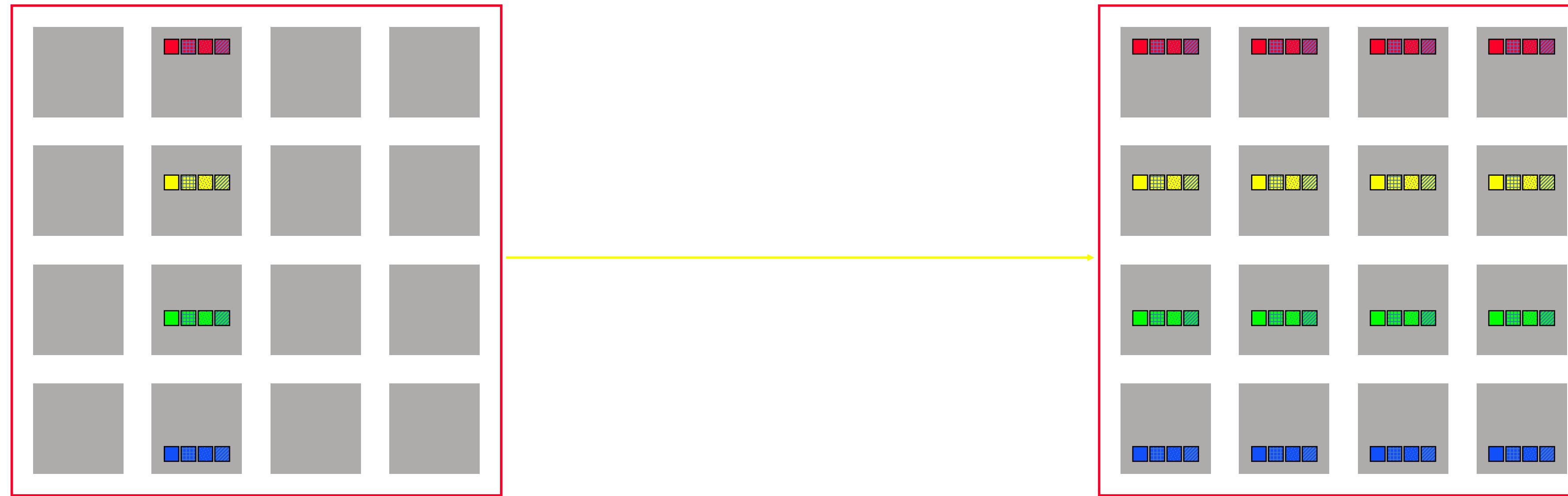
- Option 1:
 - MST broadcast in column
 - **MST broadcast in rows**

Example: 2D Broadcast



- Option 2:
 - Scatter in column

Example: 2D Broadcast



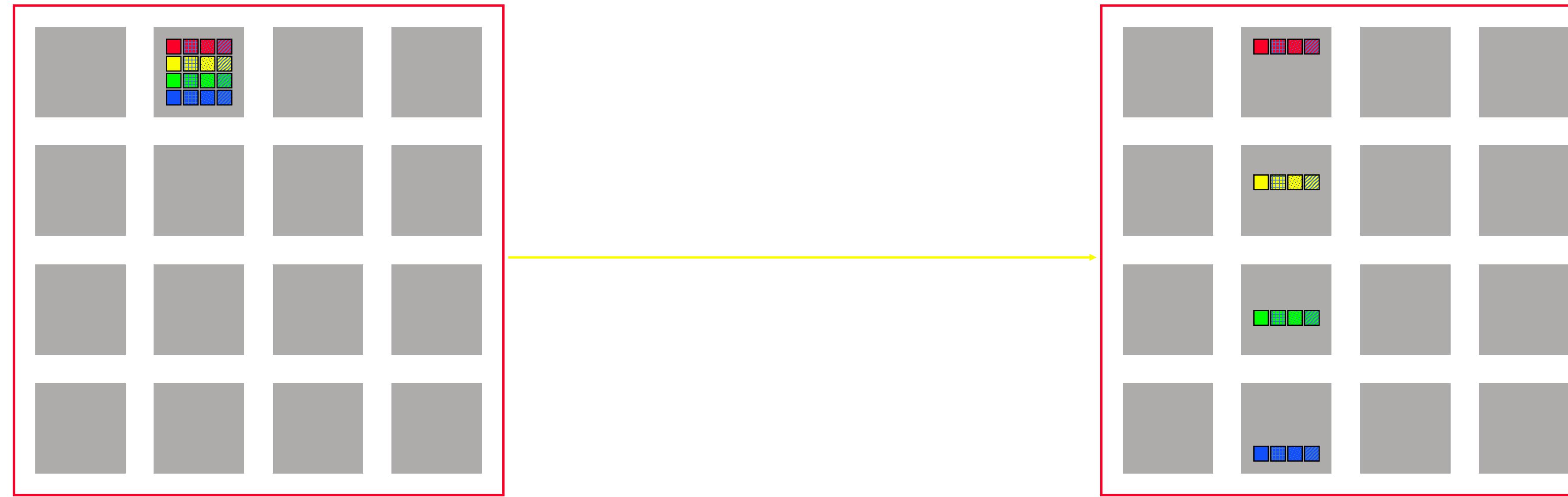
- Option 2:
 - Scatter in column
 - MST broadcast in rows

Example: 2D Broadcast



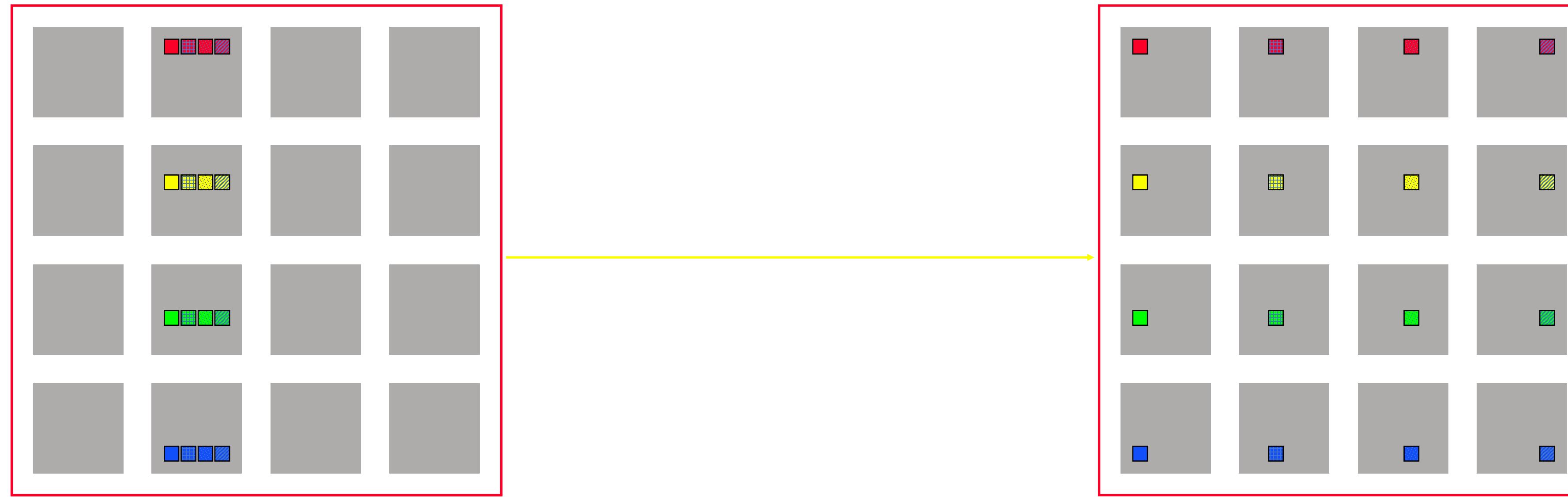
- Option 2:
 - Scatter in column
 - MST broadcast in rows
 - Allgather in columns

Example: 2D Broadcast



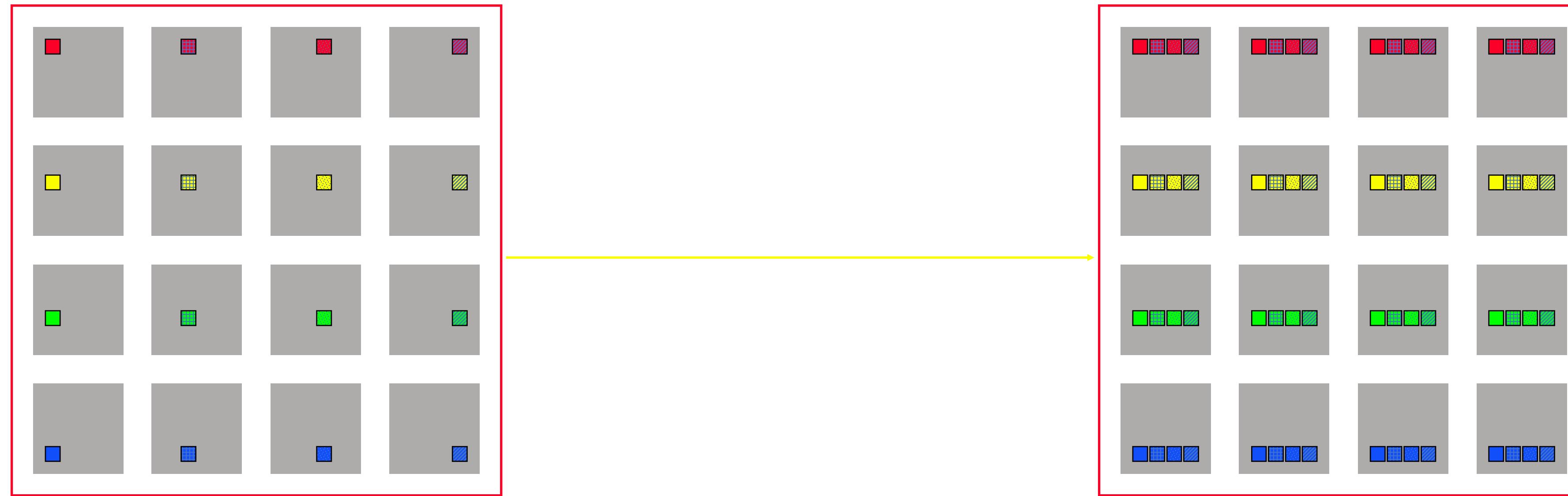
- Option 3:
 - Scatter in column
 - Scatter in rows
 - Allgather in rows
 - Allgather in columns

Example: 2D Broadcast



- Option 3:
 - Scatter in column
 - Scatter in rows

Example: 2D Broadcast



- Option 3:
 - Scatter in column
 - Scatter in rows
 - Allgather in rows

Example: 2D Broadcast



- Option 3:
 - Scatter in column
 - Scatter in rows
 - Allgather in rows
 - Allgather in columns

Cost comparison

- Option 1:
 - MST broadcast in column
 - MST broadcast in rows
- Option 2:
 - Scatter in column
 - MST broadcast in rows
 - Allgather in columns
- Option 3:
 - Scatter in column
 - Scatter in rows
 - Allgather in rows
 - Allgather in columns

$$\frac{\log(c)\alpha + \log(c)n\beta}{\log(r)\alpha + \log(r)n\beta}$$
$$\frac{\log(p)\alpha + \log(p)n\beta}{\log(p)\alpha + \log(p)n\beta}$$

Cost comparison

- Option 1:
 - MST broadcast in column
 - MST broadcast in rows
- **Option 2:**
 - **Scatter in column**
 - **MST broadcast in rows**
 - **Allgather in columns**
- Option 3:
 - Scatter in column
 - Scatter in rows
 - Allgather in rows
 - Allgather in columns

$$\frac{\log(c)\alpha + \frac{c-1}{c}n\beta}{(\log(p) + c - 1)\alpha + \left(2\frac{c-1 + \log(r)}{c}\right)n\beta}$$
$$\log(r)\alpha + \log(r)\frac{n}{c}\beta$$
$$(c-1)\alpha + \frac{c-1}{c}n\beta$$

Cost comparison

- Option 1:
 - MST broadcast in column
 - MST broadcast in rows
- Option 2:
 - Scatter in column
 - MST broadcast in rows
 - Allgather in columns
- **Option 3:**
 - **Scatter in column**
 - **Scatter in rows**
 - **Allgather in rows**
 - **Allgather in columns**

$$\begin{aligned} & \log(c)\alpha + \frac{c-1}{c}n\beta \\ & \log(r)\alpha + \frac{r-1}{r}\frac{n}{c}\beta \\ & (r-1)\alpha + \frac{r-1}{r}\frac{n}{c}\beta \\ \hline & (c-1)\alpha + \frac{c-1}{c}n\beta \\ \hline & (\log(p) + r + c - 2)\alpha + 2\frac{p-1}{p}n\beta \end{aligned}$$

Cost comparison

- Option 1:
 - MST broadcast in column
 - MST broadcast in rows
 - Option 2:
 - Scatter in column
 - MST broadcast in rows
 - Allgather in columns
 - Option 3:
 - Scatter in column
 - Scatter in rows
 - Allgather in rows
 - Allgather in columns
- $\log(p)\alpha + \log(p)n\beta$
- $(\log(p) + c - 1)\alpha + \left(2\frac{c - 1 + \log(r)}{c}\right)n\beta$
- $(\log(p) + r + c - 2)\alpha + 2\frac{p - 1}{p}n\beta$

Summary and Question

- MST -> when alpha dominates
- Ring -> when $n * \beta$ dominates
- 2D can be composed using 1D, 3D can be composed using 2D,
- ...
- Latency / Bandwidth trade-offs

Recap

- Q1: Which collective primitive maps to the distributed SGD gradient synchronization step?
- Q2: How many messages do we need to transfer over the network for a single iteration of GPT-3 SGD update assuming 8-gpu parallelism?
- Q3: For Q2, assuming 1D mesh, should we use MST or Ring?

Collective Pros

- A set of structured / well-defined communication primitives
- Extremely well-optimized
- Beautiful math, easy to analyze, and easy to understand its performance

Collective Cons

- Lack of Fault Tolerance
 - What if one node (in the ring) is dead?
- Requires Homogeneity
 - What if one node computes slower than all other nodes?
 - What if one link has lower bandwidth than the other node?

Real Cluster:

- Need Fault tolerance
- Heterogeneous hardware setup

Next Topics

- This week 2 classes: Data base + Cloud Storage
 - Delta from previous year offering:
 - we skip a substantial part of relational database
 - spend more time on networking, HPC, and ML
- Next week: Parallelism and Big Data processing
 - We will come back to study how we address the problem of Collectives