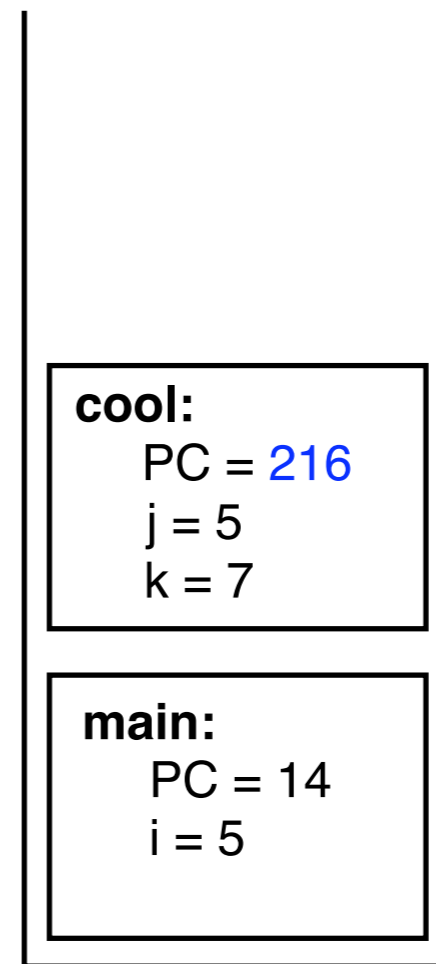


Winter 2016  
COMP-250: Introduction  
to Computer Science

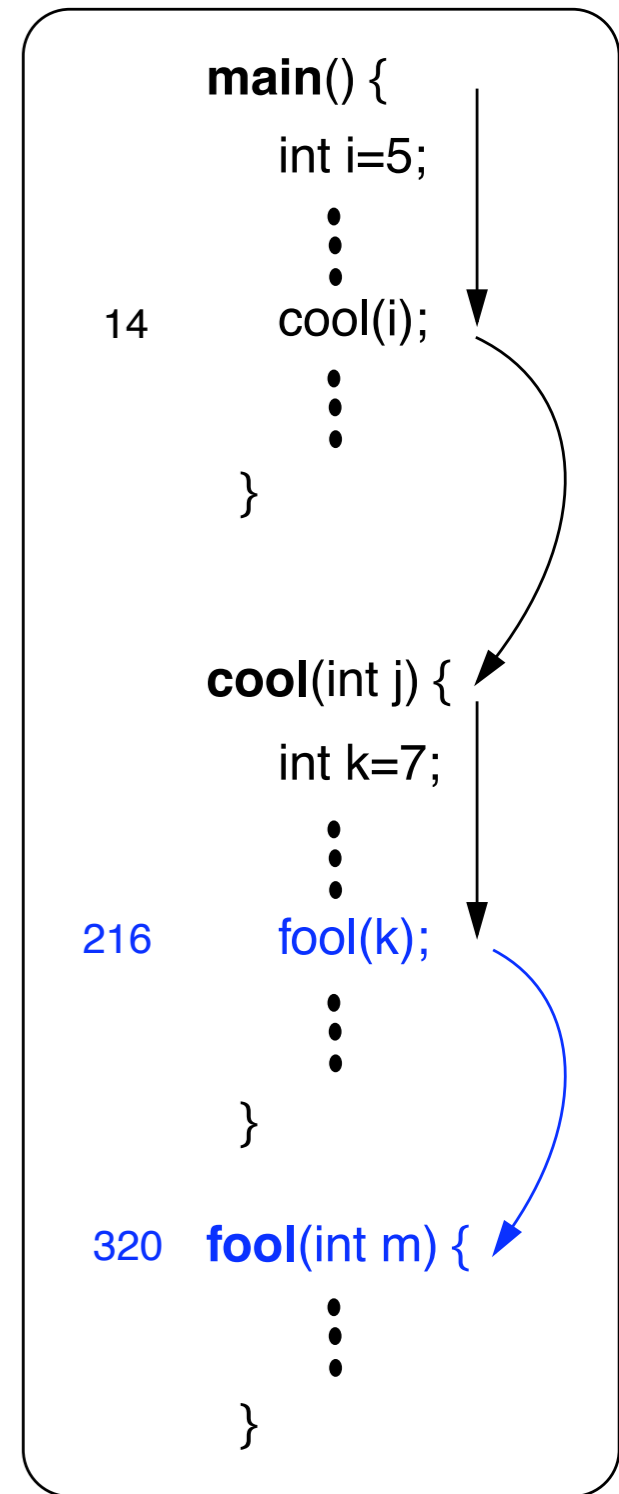
Lecture 8, February 4, 2016

# Stacks in the Java Virtual Machine

- Each process running in a Java program has its own Java Method Stack.
- Each time a method is called, it is pushed onto the stack.
- The choice of a stack for this operation allows Java to do several useful things:
  - Perform recursive method calls
  - Print stack traces to locate an error



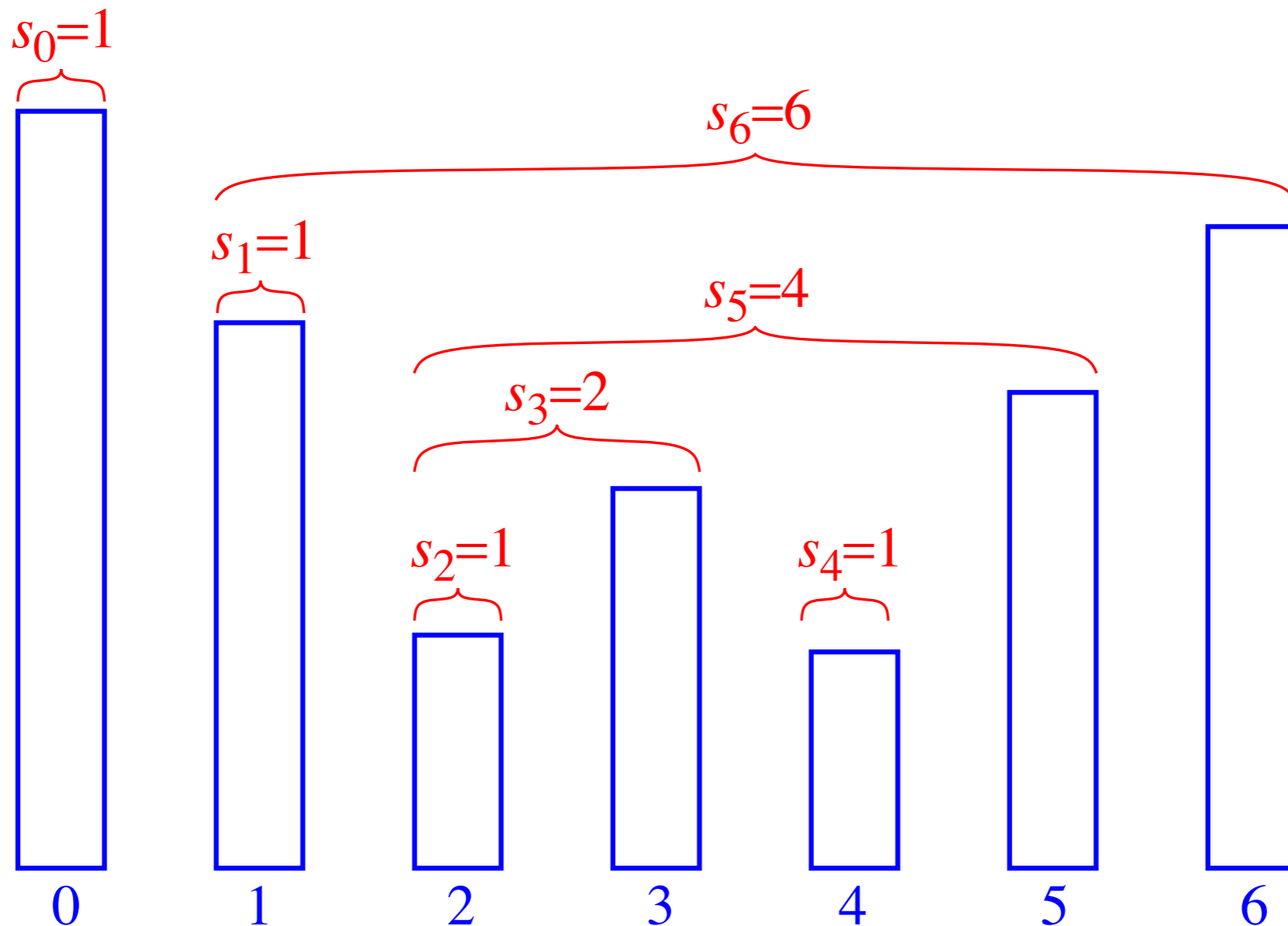
Java Stack



Java Program

# Application: Time Series

- The *span*  $s_i$  of a stock's price on a certain day  $i$  is the maximum number of consecutive days (up to the current day) the price of the stock has been less than or equal to its price on day  $i$ .



# An Inefficient Algorithm

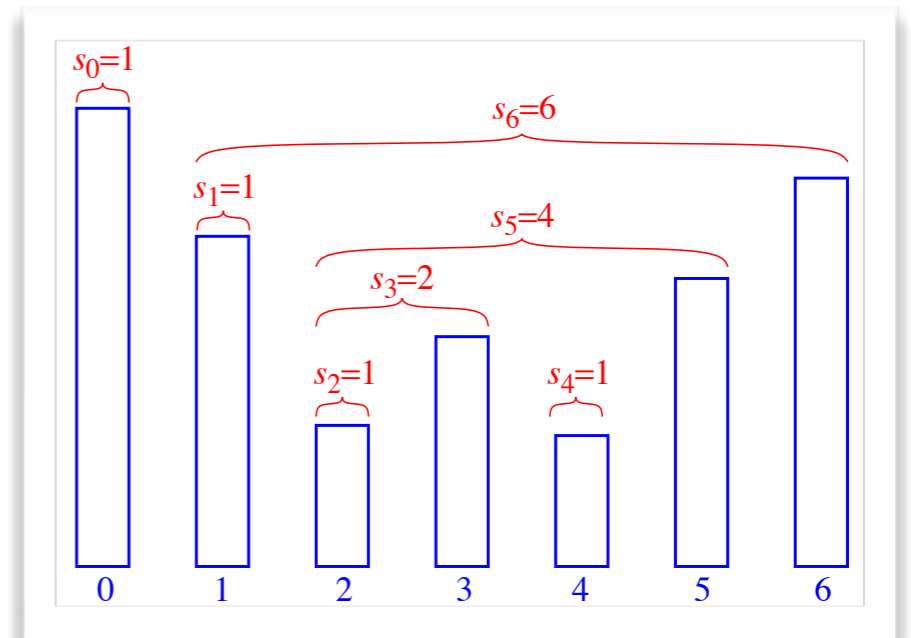
- There is a straightforward way to compute the span of a stock on each of  $n$  days:

**Algorithm** `computeSpans1(P)`:

*Input:* an  $n$ -element array  $P$  of numbers such that  $P[i]$  is the price of the stock on day  $i$

*Output:* an  $n$ -element array  $S$  of numbers such that  $S[i]$  is the span of the stock on day  $i$

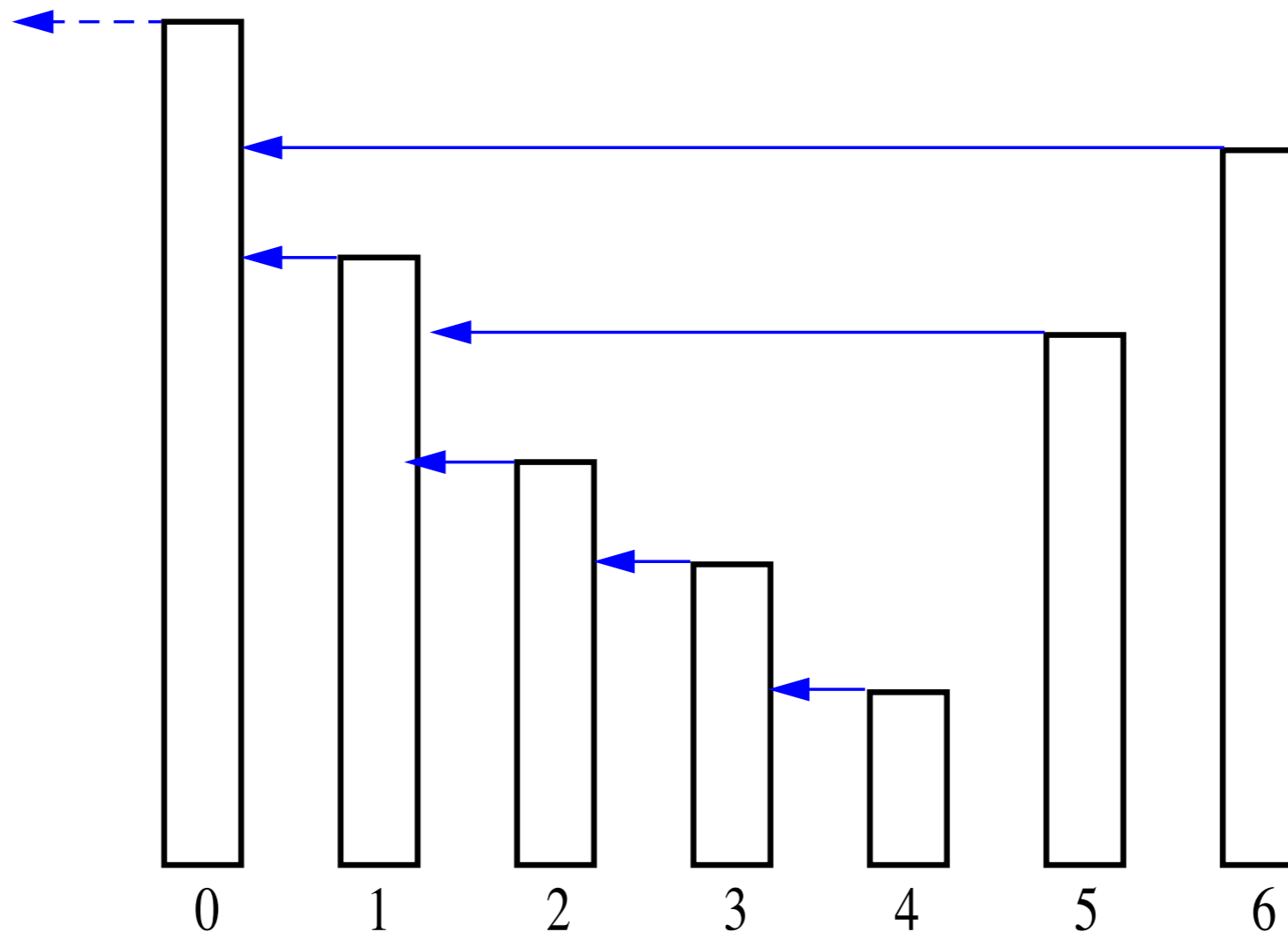
```
for  $i \leftarrow 0$  to  $n - 1$  do  
     $k \leftarrow 0$   
     $done \leftarrow \mathbf{false}$   
    repeat  
        if  $P[i - k] \leq P[i]$  then  
             $k \leftarrow k + 1$   
        else  
             $done \leftarrow \mathbf{true}$   
    until  $(k = i)$  or  $done$   
     $S[i] \leftarrow k$   
return  $S$ 
```



- The running time of this algorithm is (ugh!)  $O(n^2)$ .  
Why?

# A Stack Can Help

- We see that  $s_i$  on day  $i$  can be easily computed if we know the closest day preceding  $i$ , such that the price is greater than on that day than the price on day  $i$ . If such a day exists, let's call it  $h(i)$ , otherwise, we conventionally define  $h(i) = -1$
- The span is now computed as  $s_i = i - h(i)$



We use a *stack* to keep track of  $h(i)$

# An Efficient Algorithm

- The code for our new algorithm:

**Algorithm** computeSpan2( $P$ ):

*Input:* An  $n$ -element array  $P$  of numbers representing stock prices

*Output:* An  $n$ -element array  $S$  of numbers such that  $S[i]$  is the span of the stock on day  $i$

Let  $D$  be an empty stack

**for**  $i \leftarrow 0$  **to**  $n - 1$  **do**

$done \leftarrow$  **false**

**while** **not**( $D.isEmpty()$  **or**  $done$ ) **do**

**if**  $P[i] \geq P[D.top()]$  **then**

$D.pop()$

**else**

$done \leftarrow$  **true**

**if**  $D.isEmpty()$  **then**

$h \leftarrow -1$

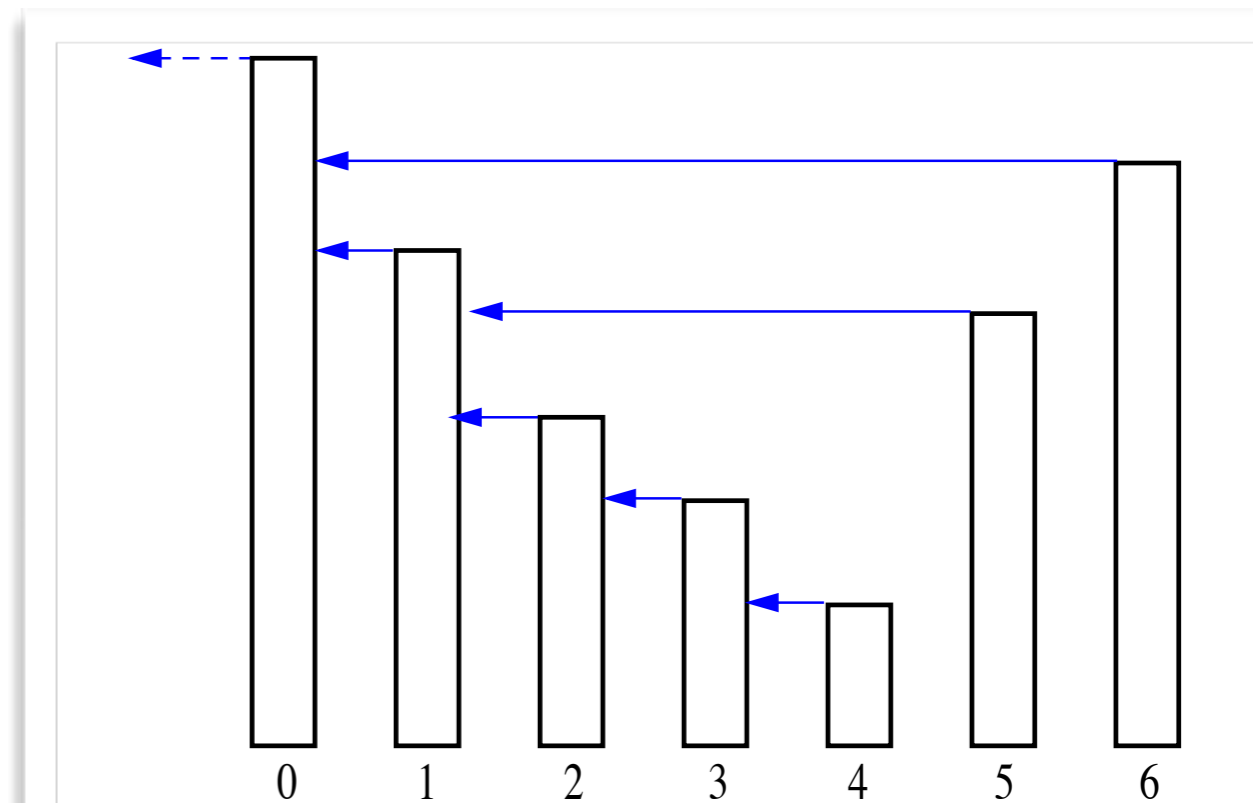
**else**

$h \leftarrow D.top()$

$S[i] \leftarrow i - h$

$D.push(i)$

**return**  $S$



# Queue ADT



# Queue

- A queue differs from a stack in that its insertion and removal routines follows the **first-in-first-out (FIFO)** principle.
- Elements may be inserted at any time, but only the element which has been in the queue the longest may be removed.
- Elements are inserted at the rear (enqueued) and removed from the front (dequeued).



# Queue

- The queue has two fundamental methods:
  - `enqueue(o)`: Inserts object `o` at rear of the queue
  - `dequeue()`: Removes object from front of queue and returns it; **an error occurs if queue is empty.**
- These support methods should also be defined:
  - `size()`: Returns number of objects in the queue
  - `isEmpty()`: Returns a boolean value that indicates whether the queue is empty
  - `front()`: Returns, but not remove, the front object in the queue; **an error occurs if queue is empty.**

# Queue

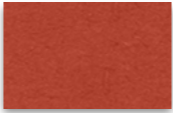
OPERATION	STATE
	-
add(a)	a
add(b)	ab
remove()	b
add(c)	bc
add(d)	bcd
add(e)	bcde
remove()	cde
add(f)	cdef
remove()	def
add(g)	defg

---

	0123456789	head	size
OPERATION			
		0	0
add(a)	a	0	1
add(b)	ab	0	2
remove()	b	1	1
add(c)	bc	1	2
add(d)	bcd	1	3
add(e)	bcde	1	4
remove()	cde	2	3
add(f)	cdef	2	4
remove()	def	3	3
add(g)	defg	3	4

---

---

	0123456 	head	size
OPERATION			
		0	0
add(a)	a	0	1
add(b)	ab	0	2
remove()	b	1	1
add(c)	bc	1	2
add(d)	bcd	1	3
add(e)	bcde	1	4
remove()	cde	2	3
add(f)	cdef	2	4
remove()	def	3	3
add(g)	defg	3	4

---

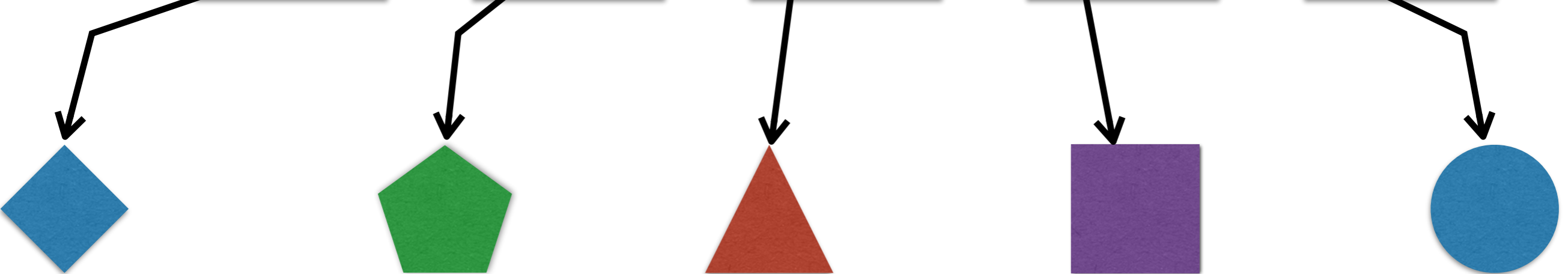
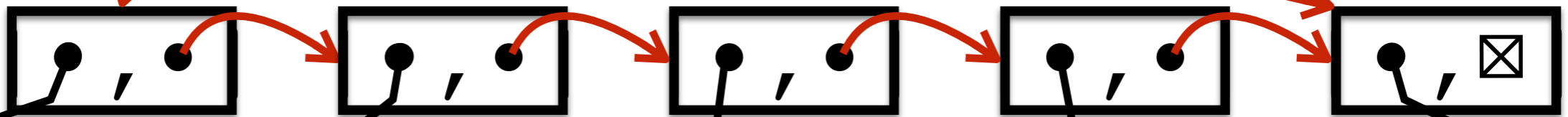
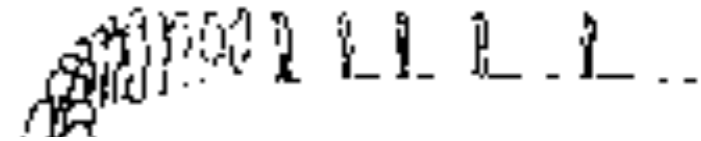
# Queue as List

```
removeFirst(){  
  tmp = head;  
  head = head.next;  
  tmp.next = null;  
  size = size - 1;  
}
```

```
addLast( newNode ){  
  tail.next = newNode;  
  tail = tail.next;  
  size = size + 1;  
}
```

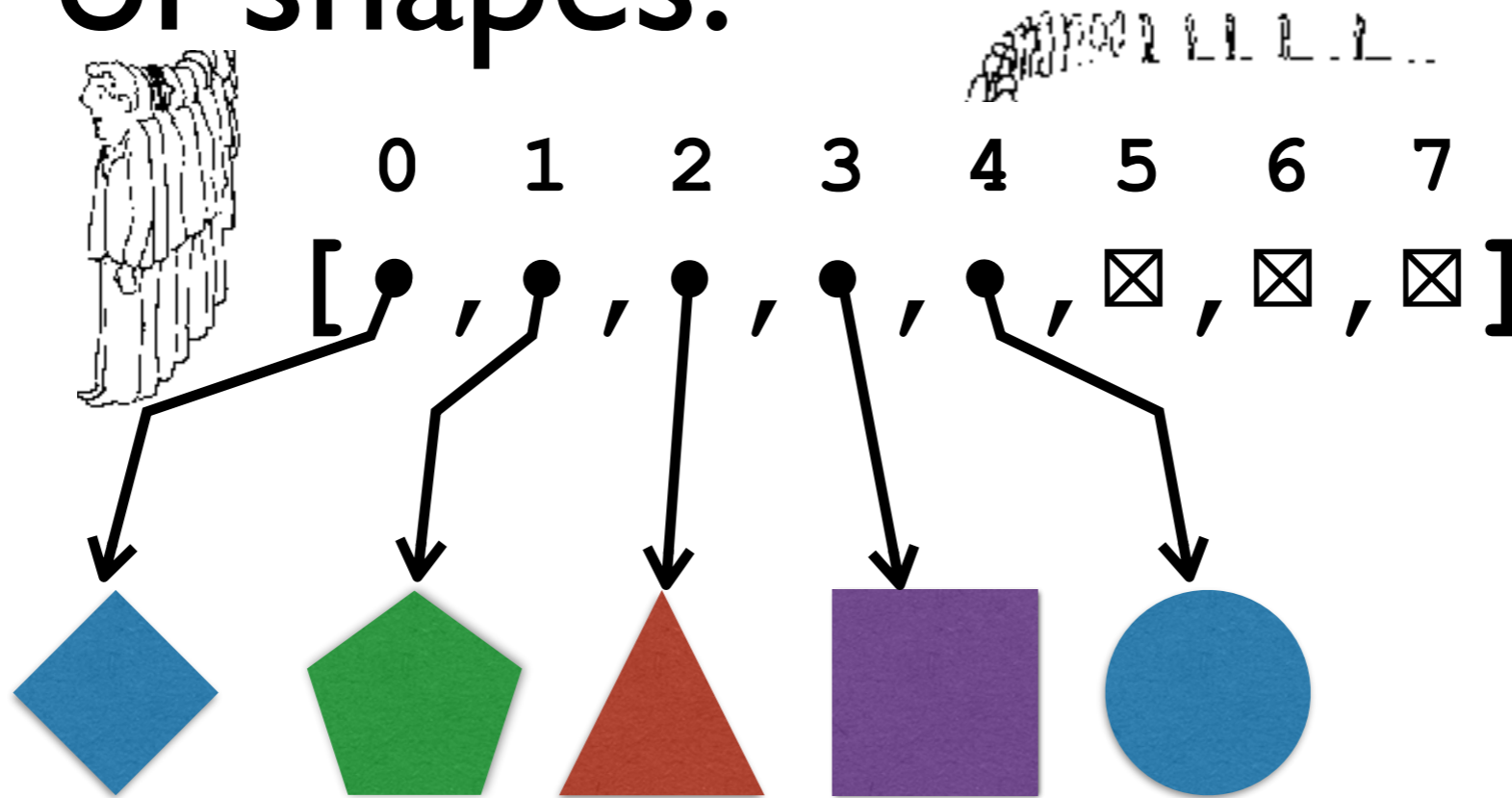


head tail size



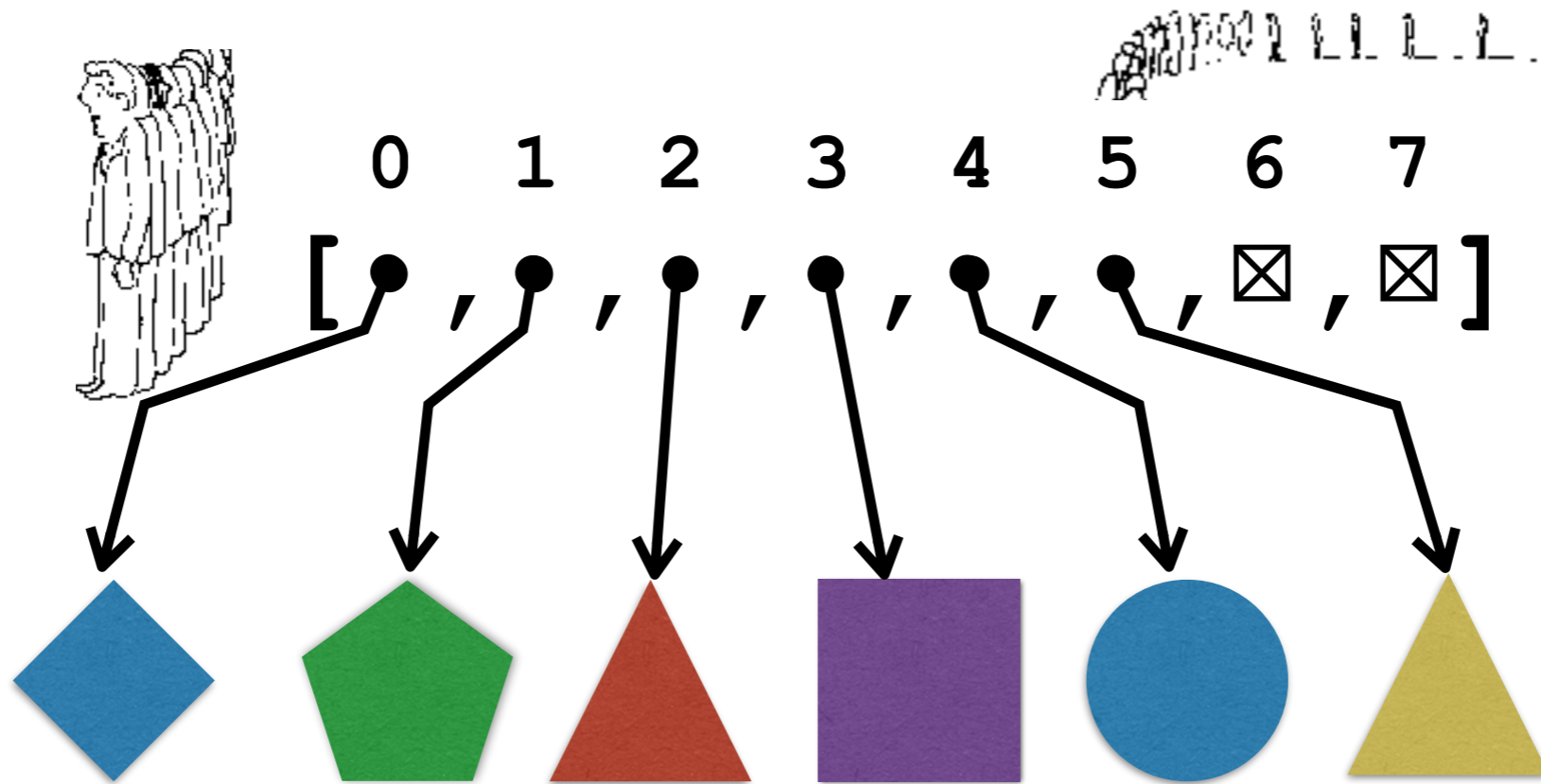
# Queue as Array

Array of shapes:



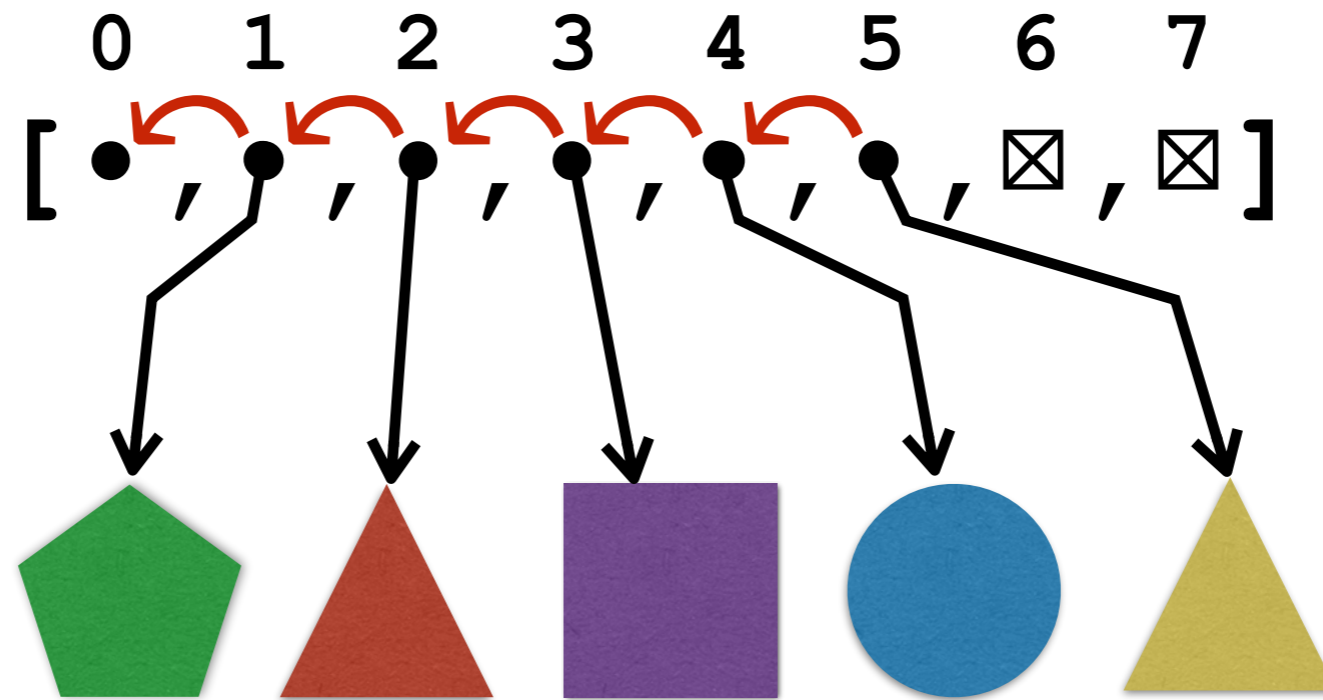
Size=5

# Queue as Array



Size=6

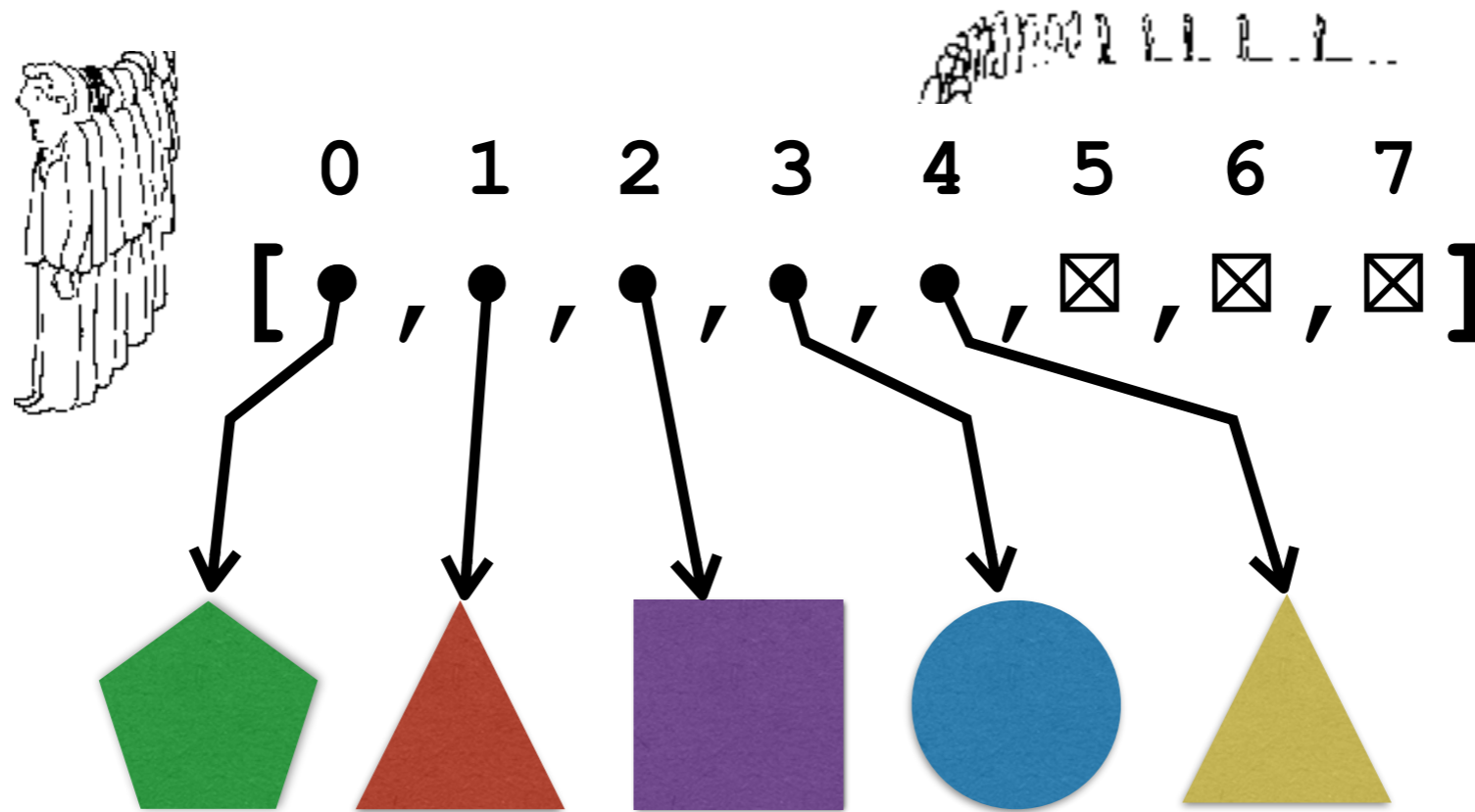
# Queue as Array



Size=5

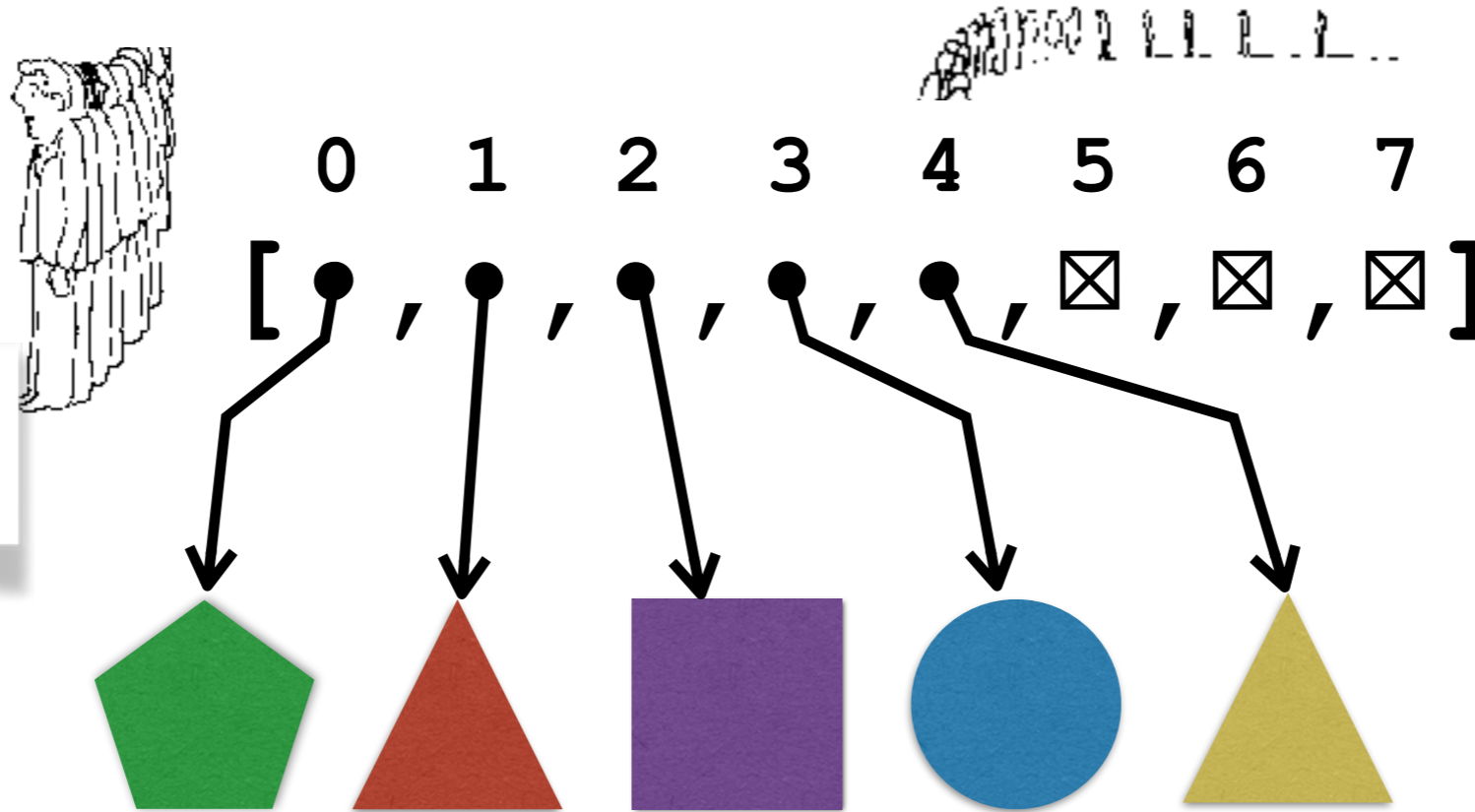


# Queue as Array

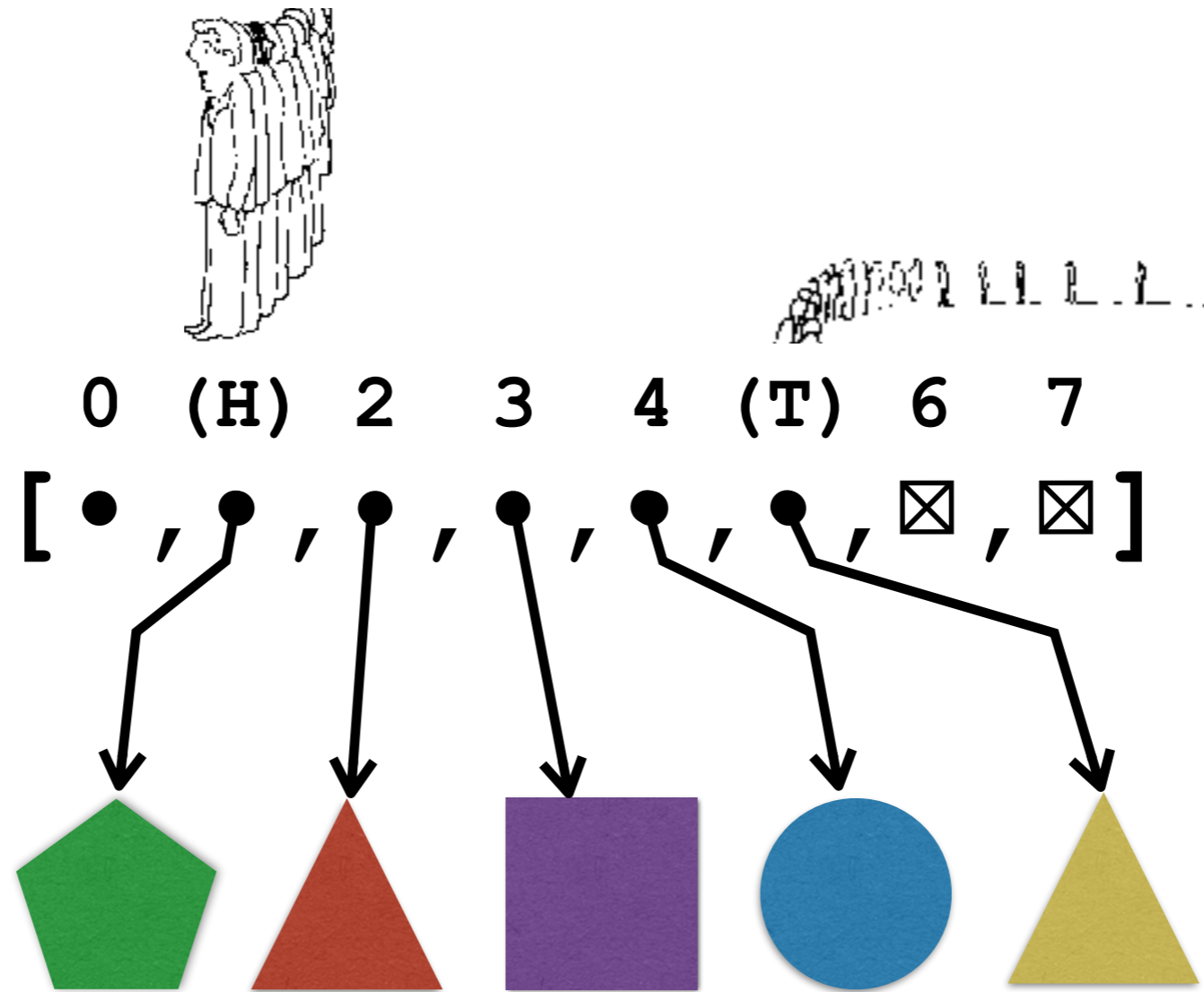


Size=5

# Queue as Array



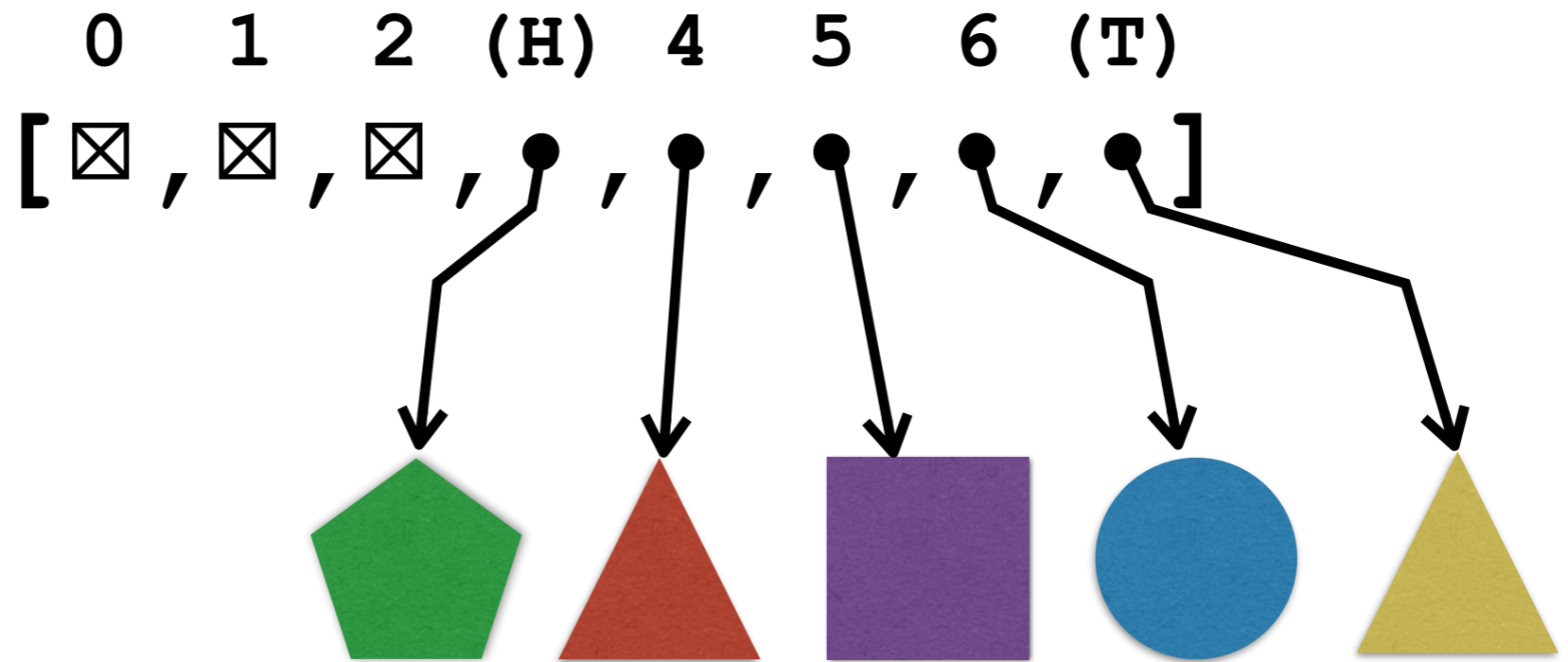
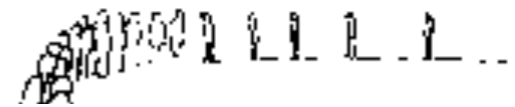
# Queue as Array



head=1, tail=5, (size=5)



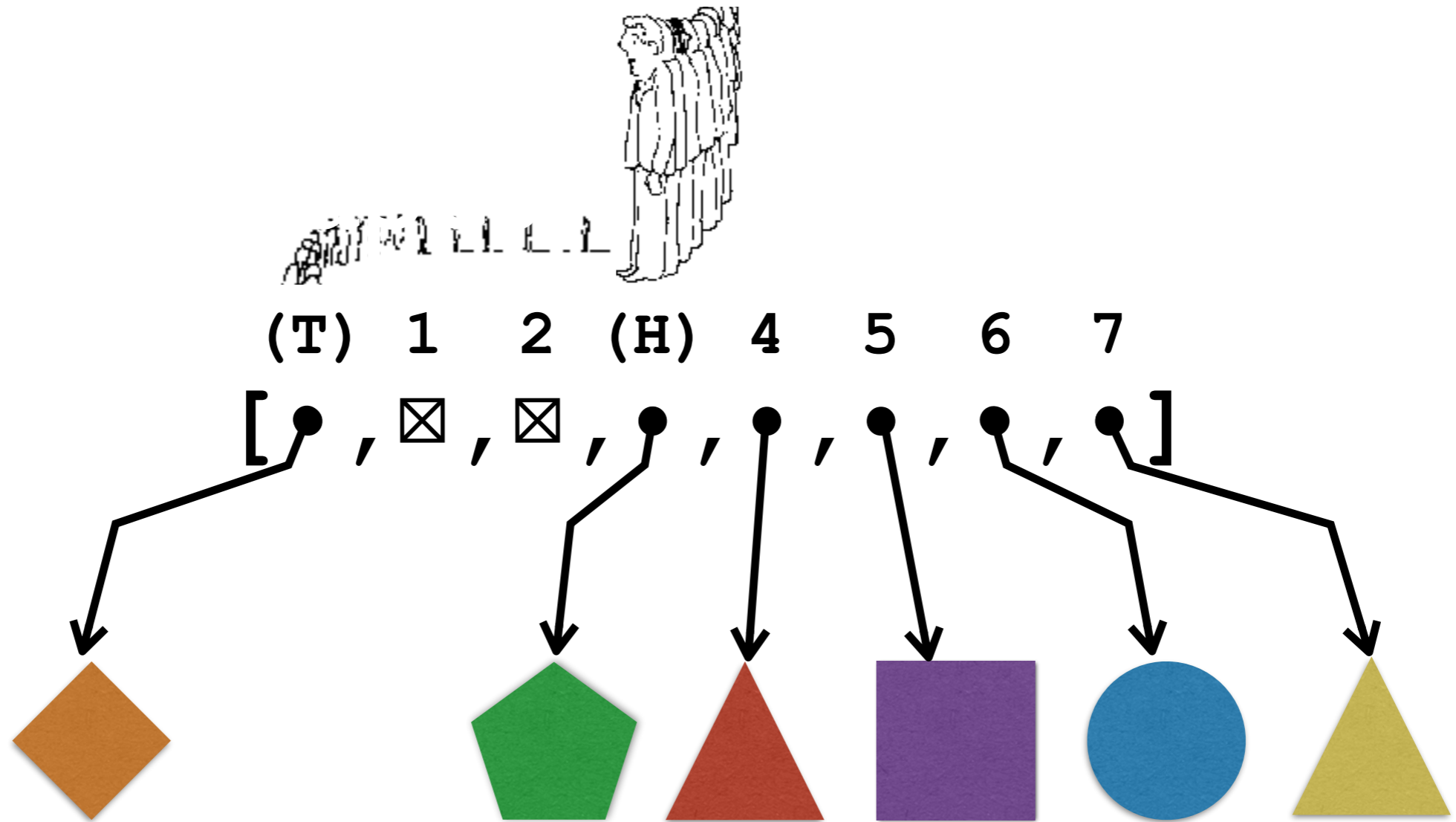
# Queue as Array



FULL ??

head=3, tail=7, (size=5)

# Queue as Array

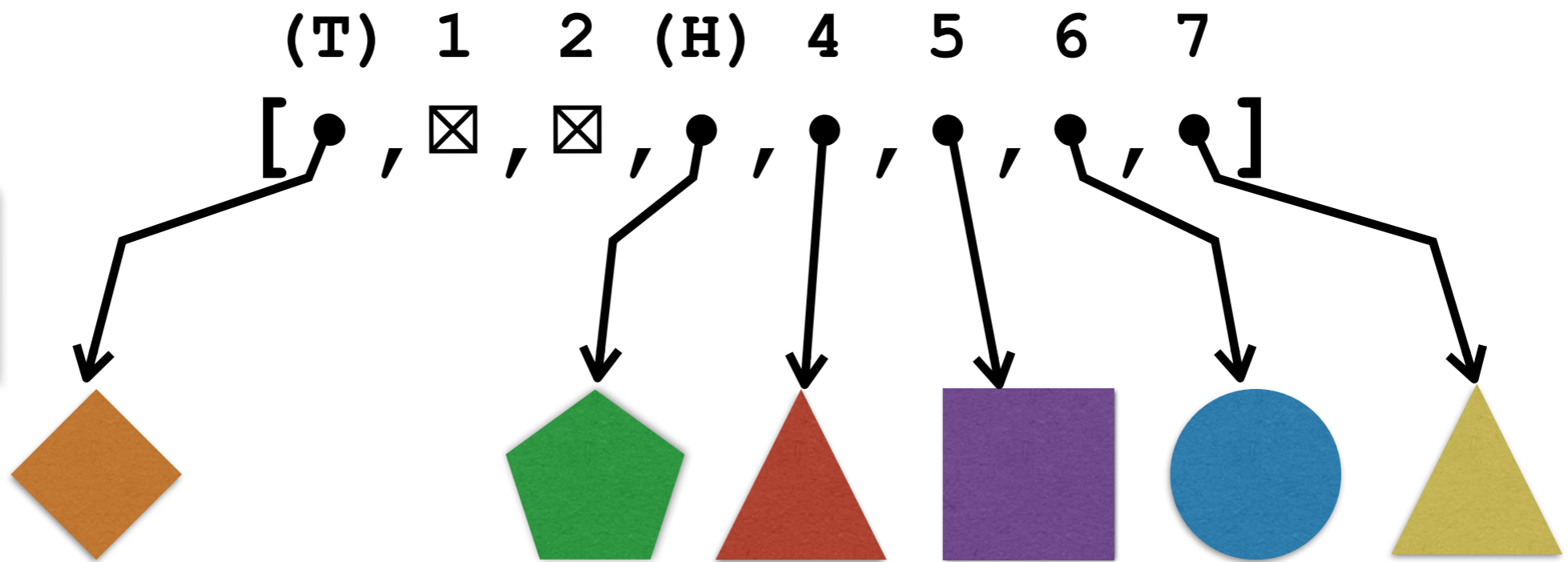


head=3, tail=0, (size=6)

# Queue as Array

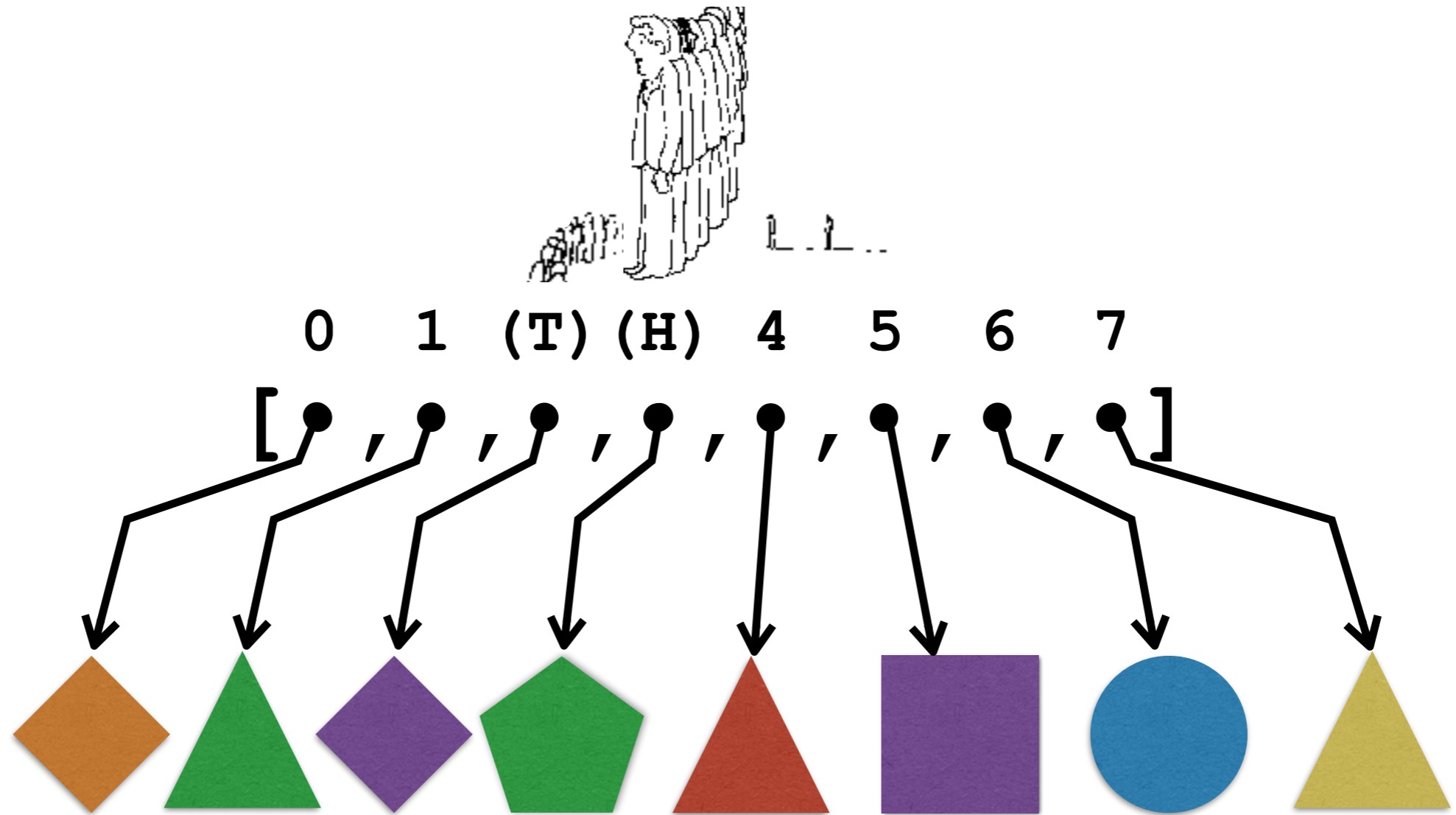


NO !!



head=3, tail=0, (size=6)

# Queue as Array



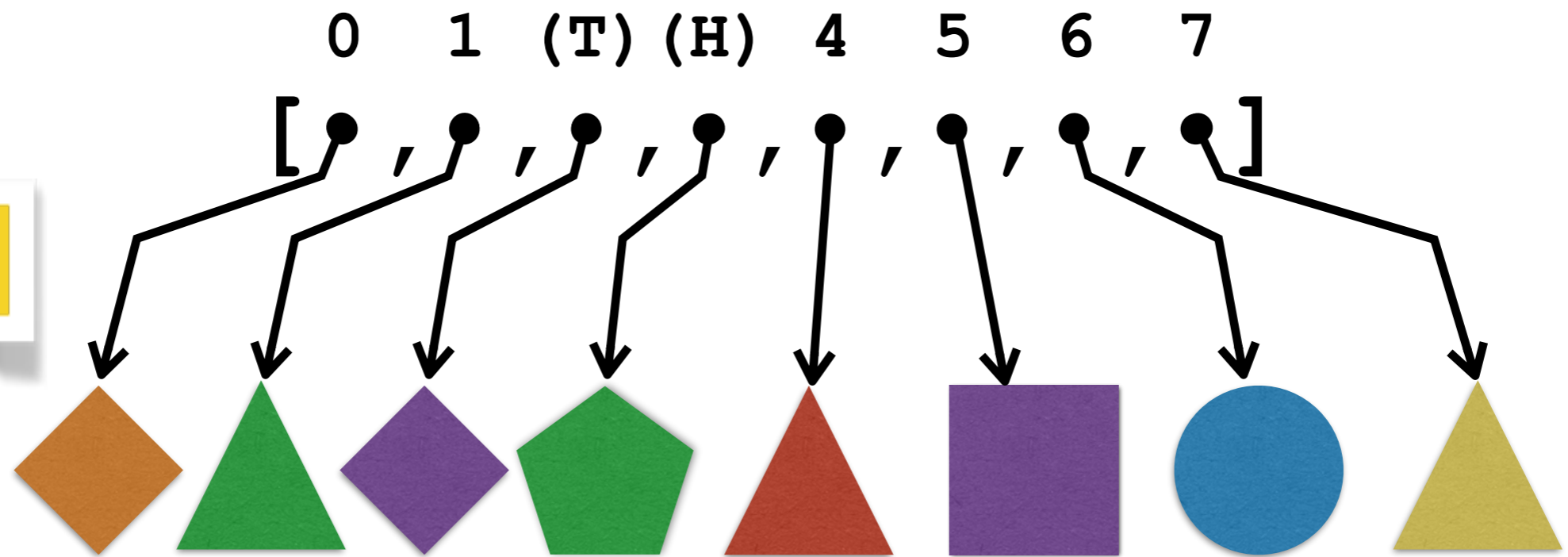
head=3, tail=2, (size=8)



# Queue as Array



FULL !!



head=3, tail=2, (size=8)

# Queue as Array

```
enqueue( element ){    // array implementation
    if ( size == length)
        increase length of array // *** SEE BELOW **

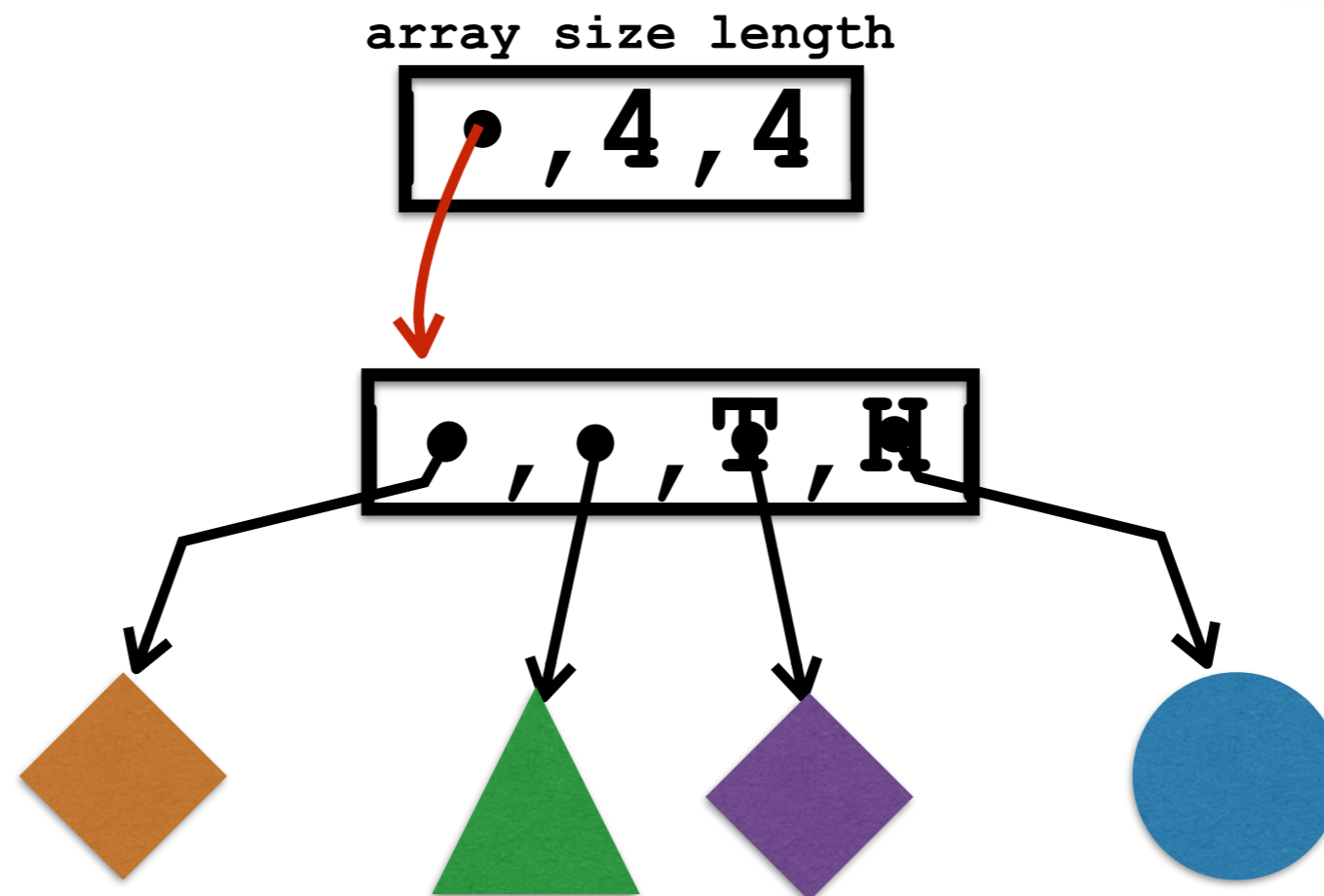
    a[ (head + size) % length ] = element
    size = size + 1
}
```

```
dequeue(){
    out = a[head]
    head = (head + 1) % length
    size = size - 1
    return out
}
```

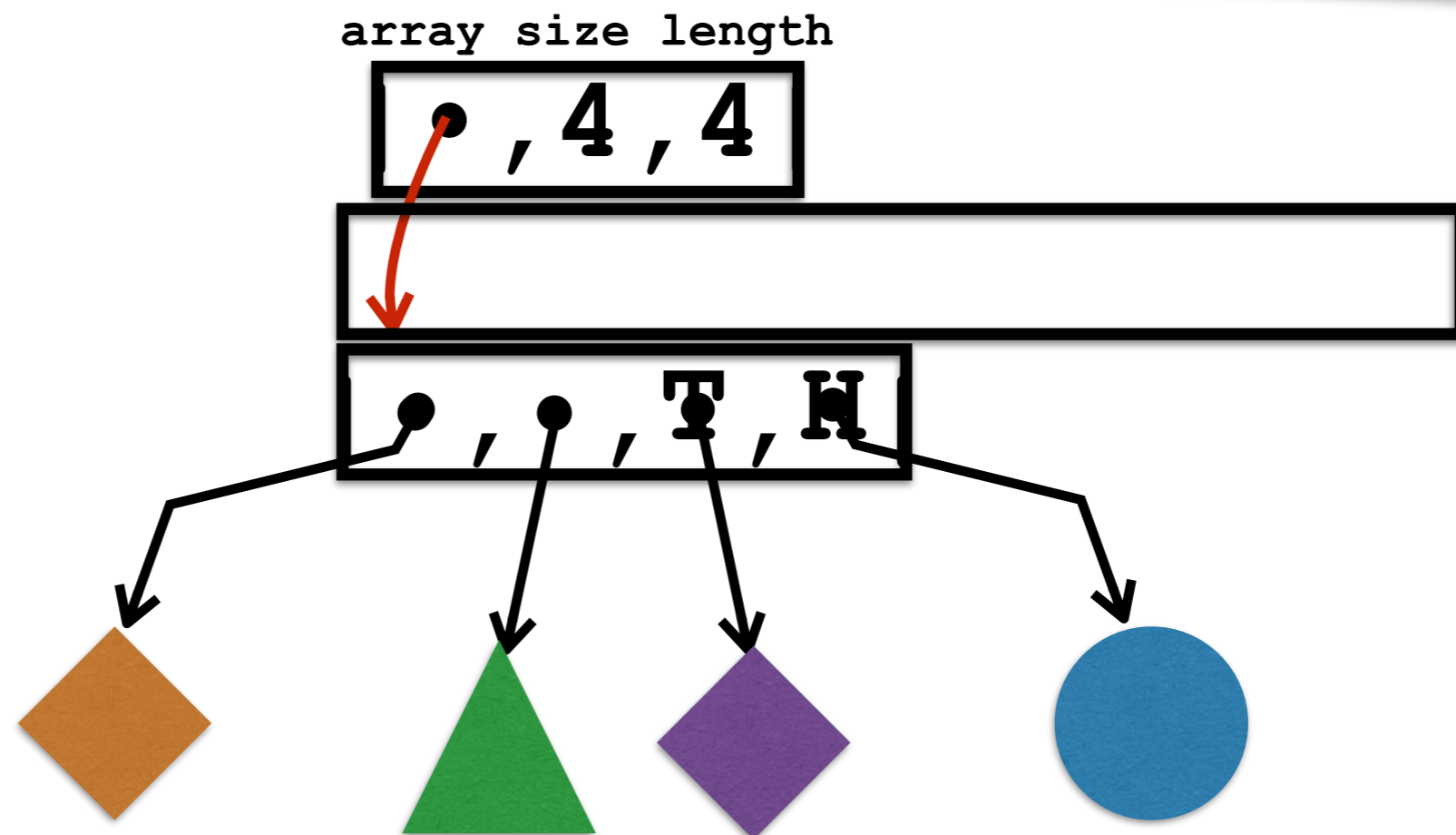
# Queue as Array

```
// copy the length elements to a new bigger array
create a bigger array
for i = 0 to small.length-1
    big[i] = small[ (head + i) % small.length ]
head = 0
tail = small.length-1
size = small.length
```

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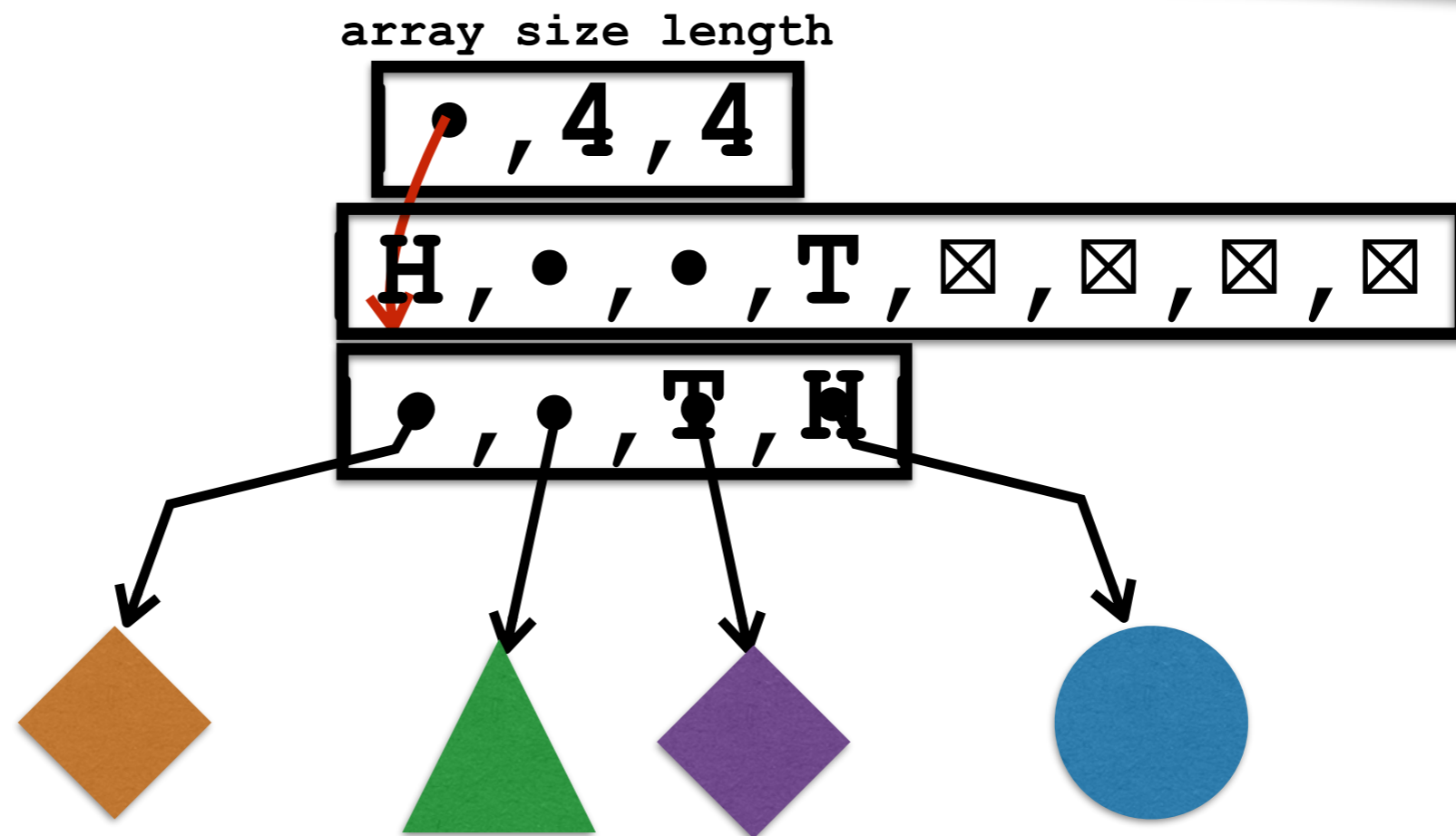
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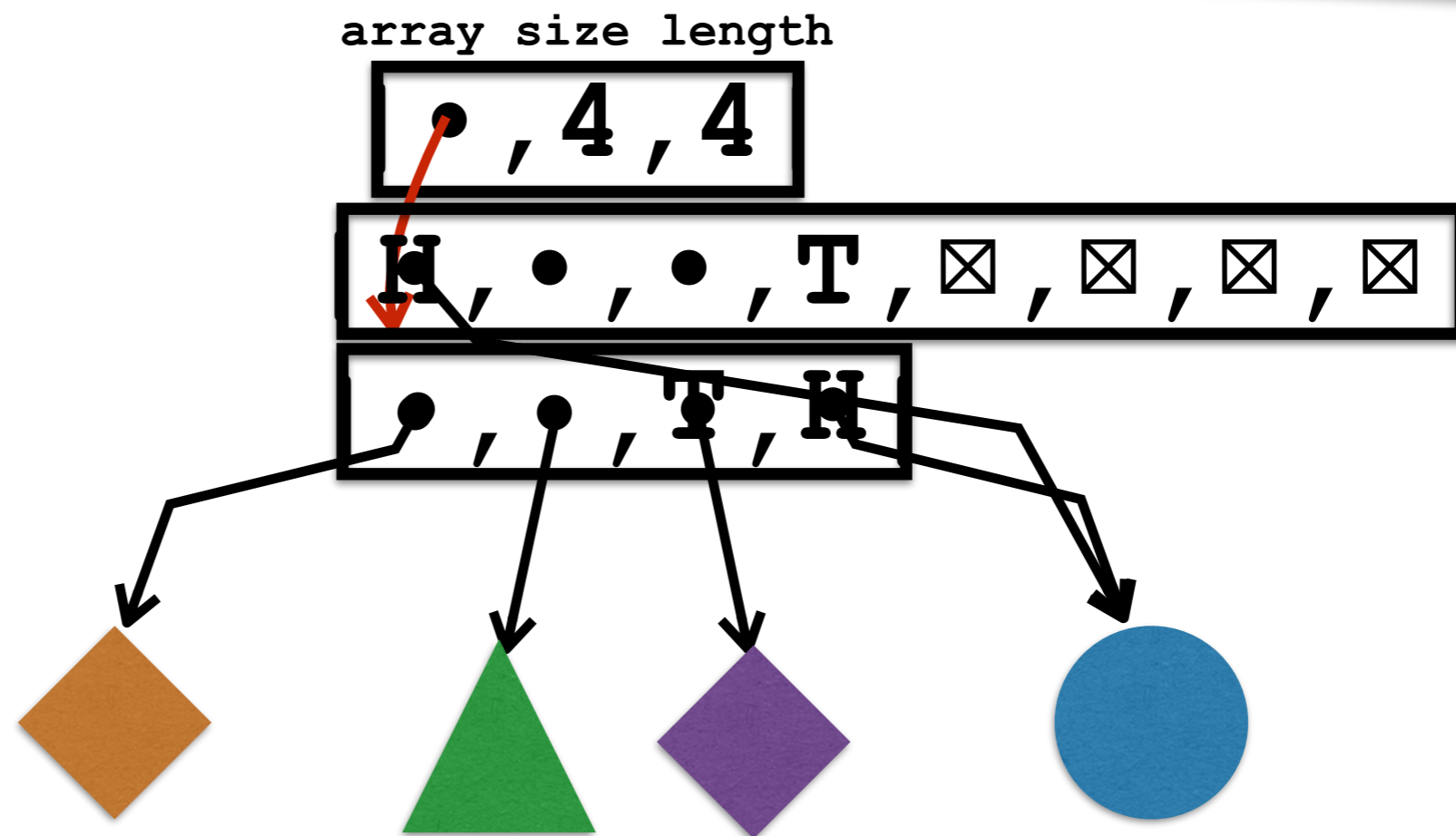
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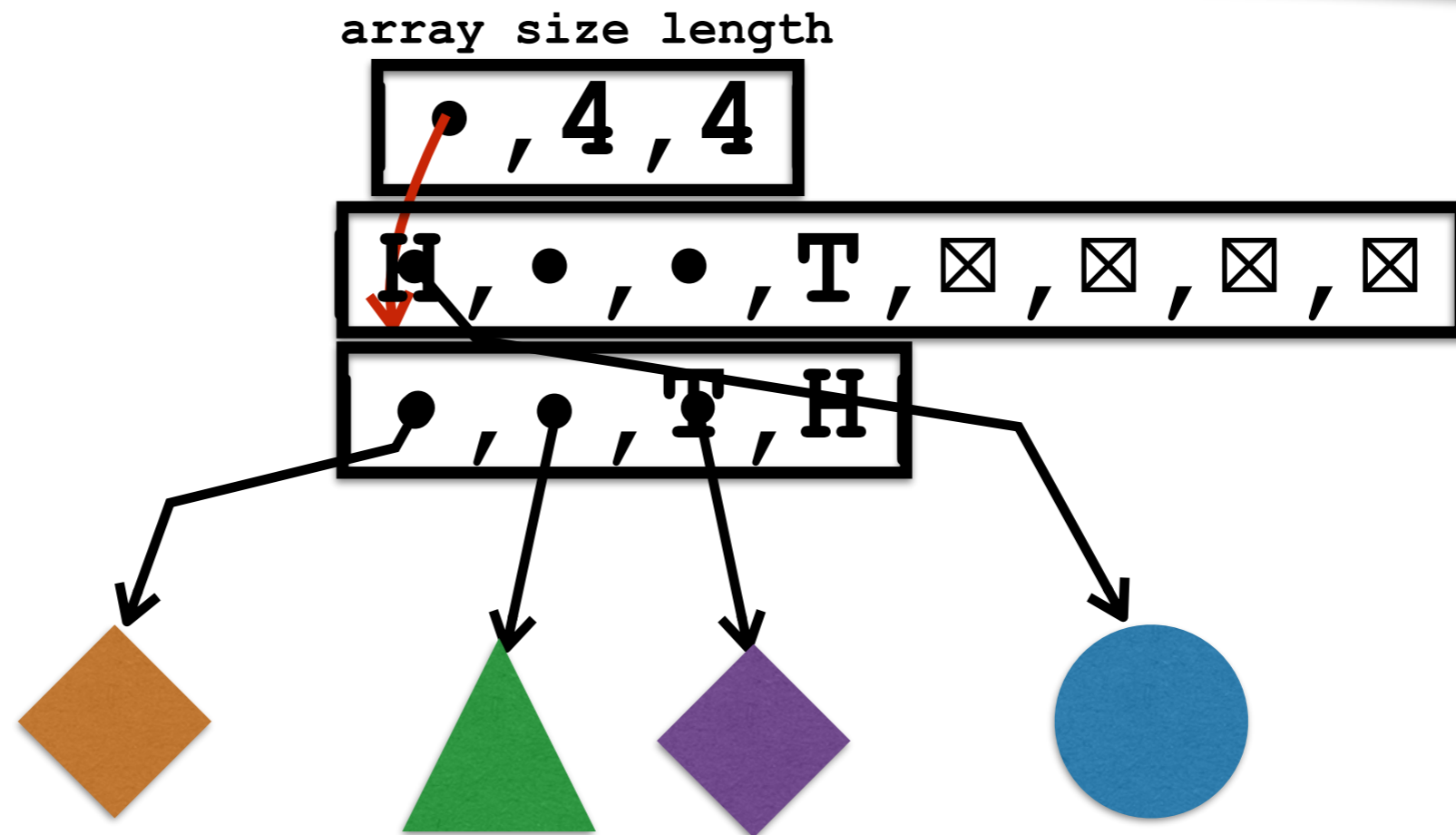
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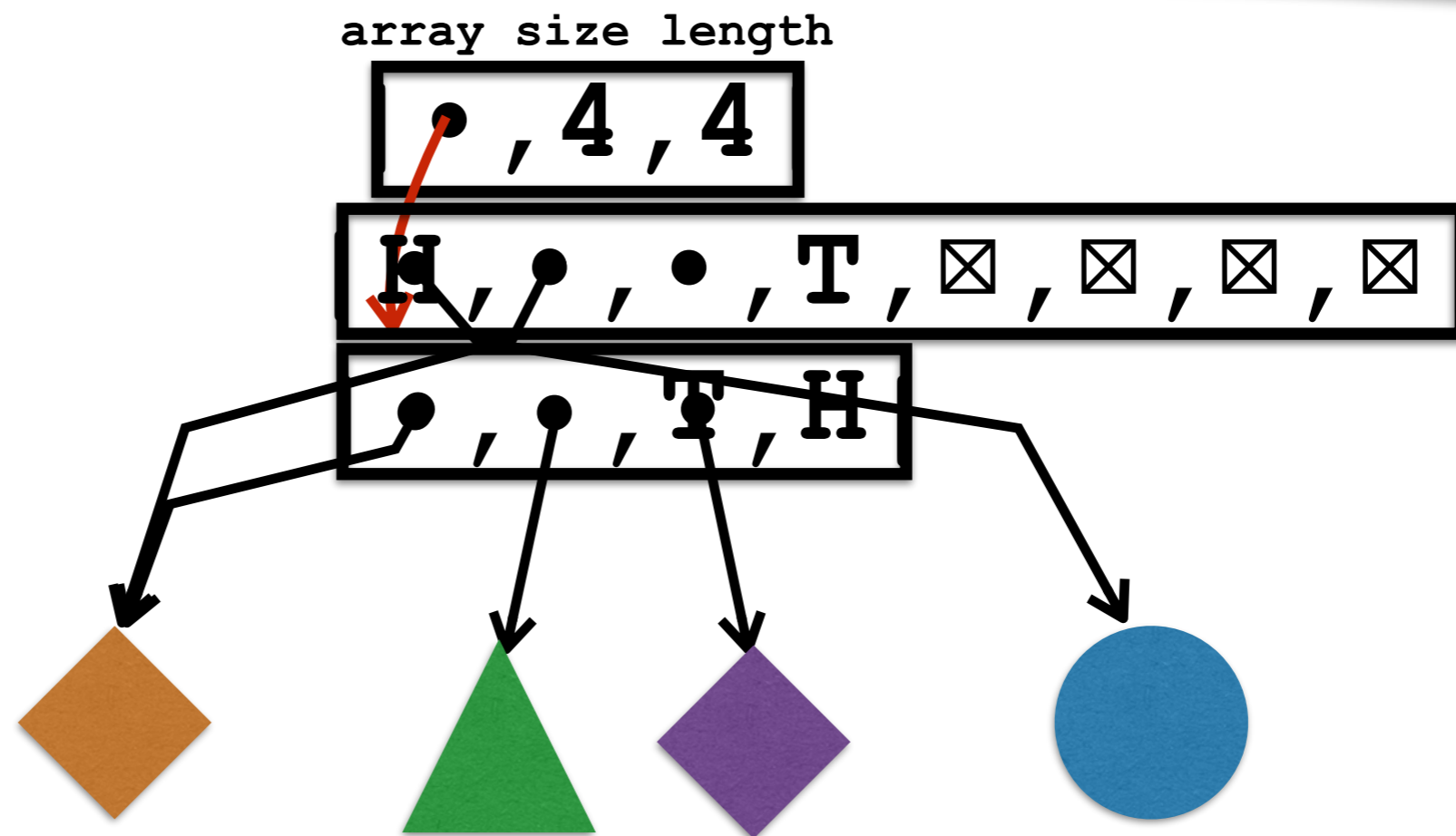




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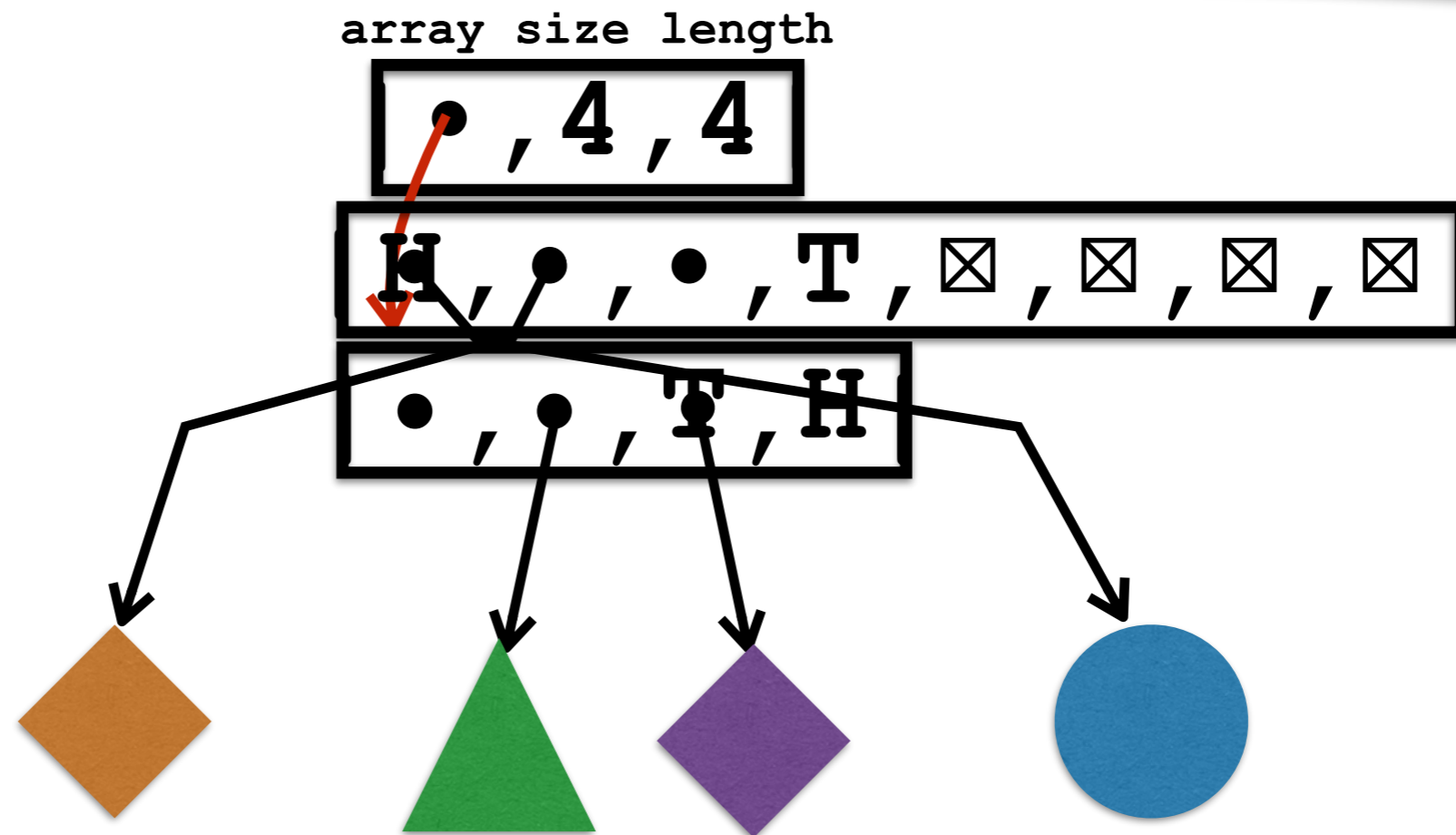
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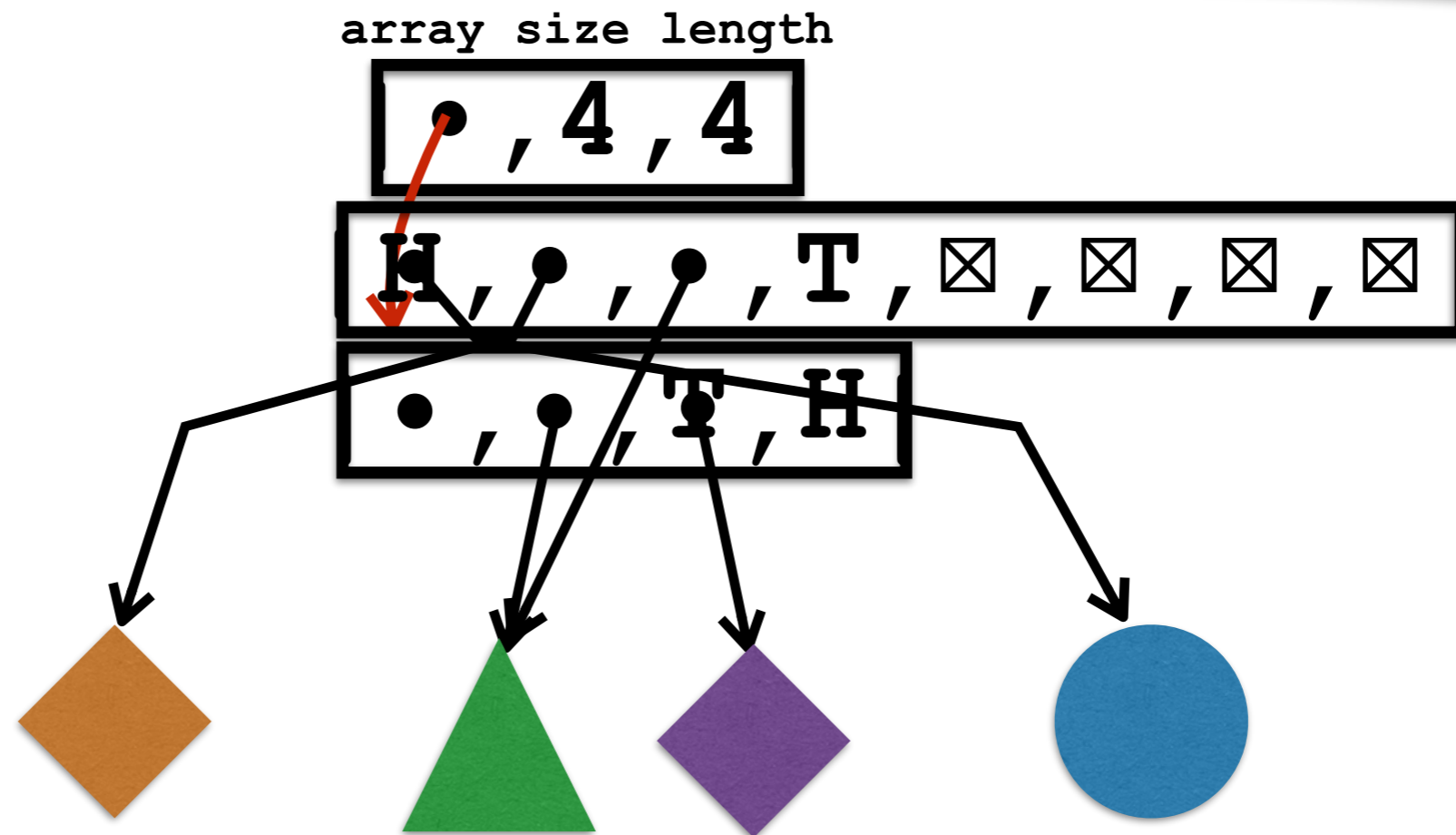
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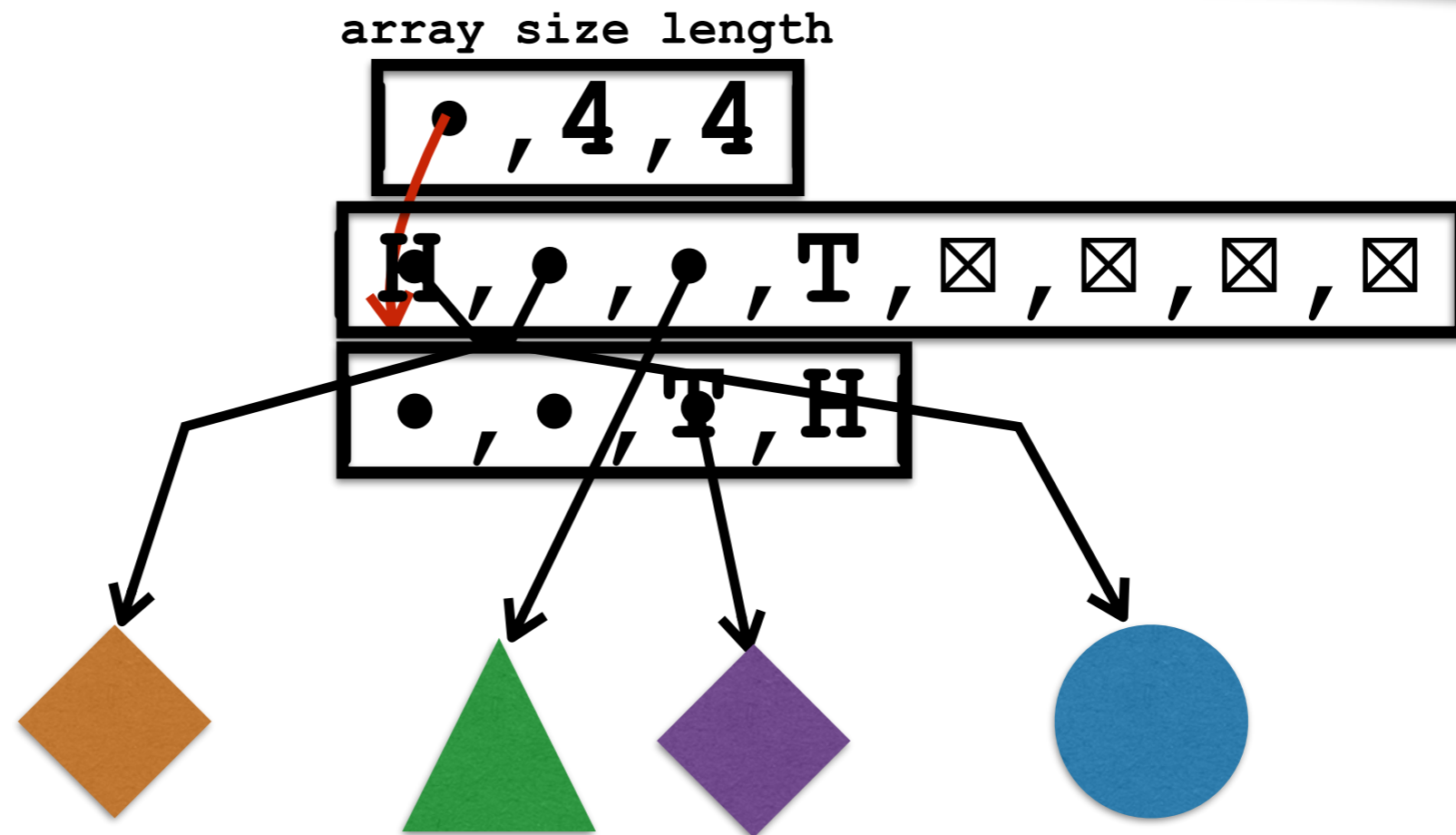
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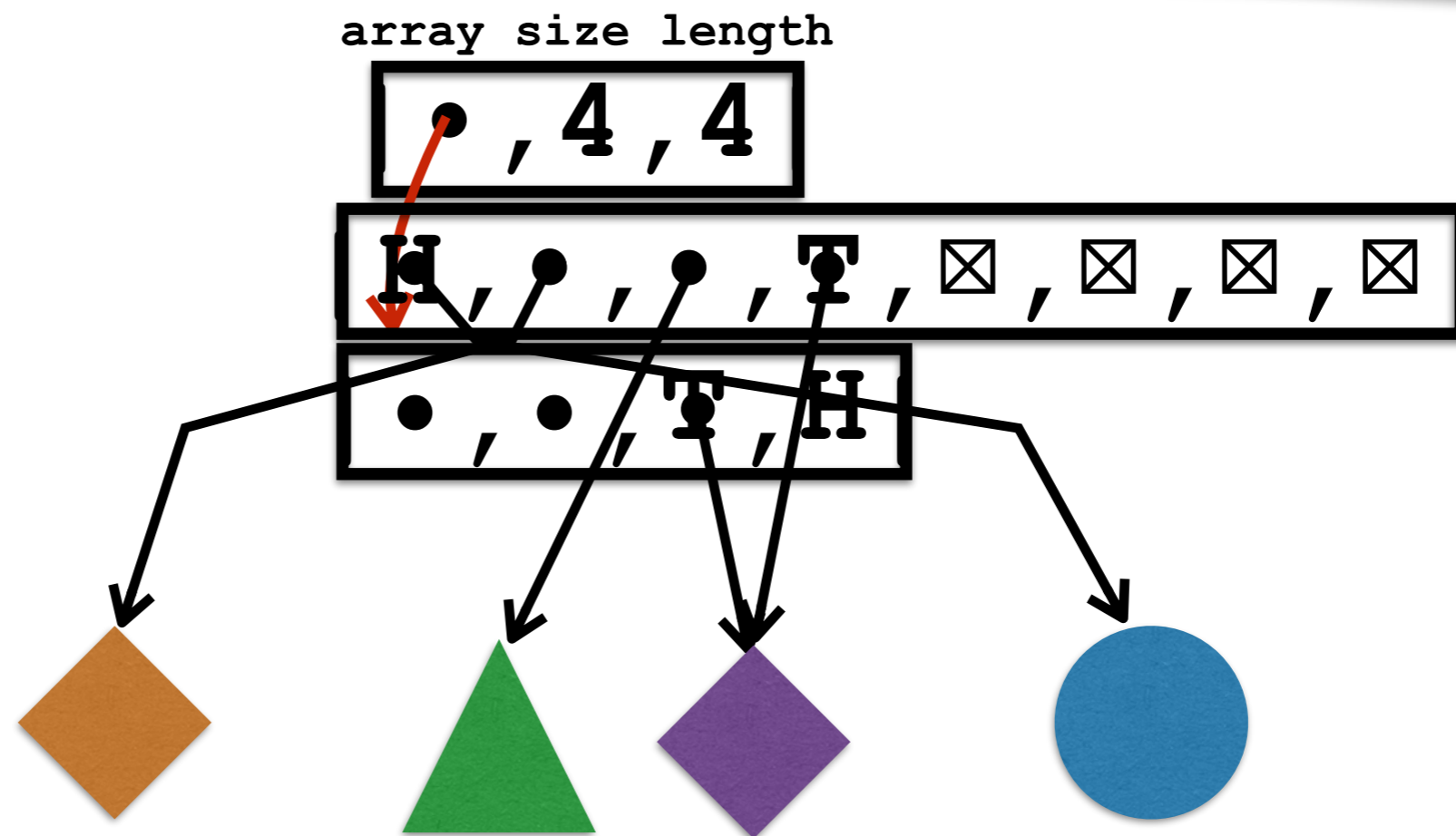
```



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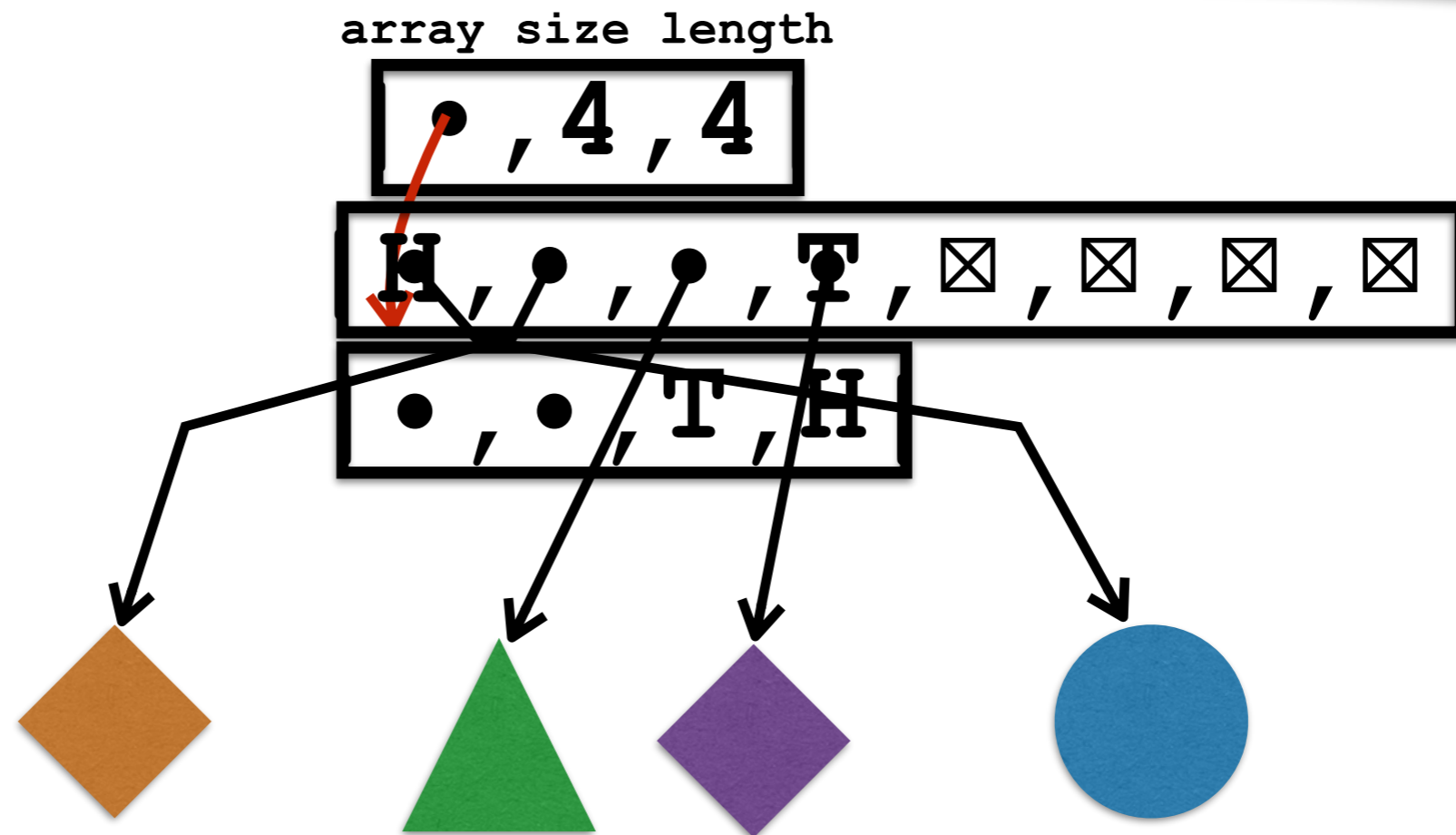
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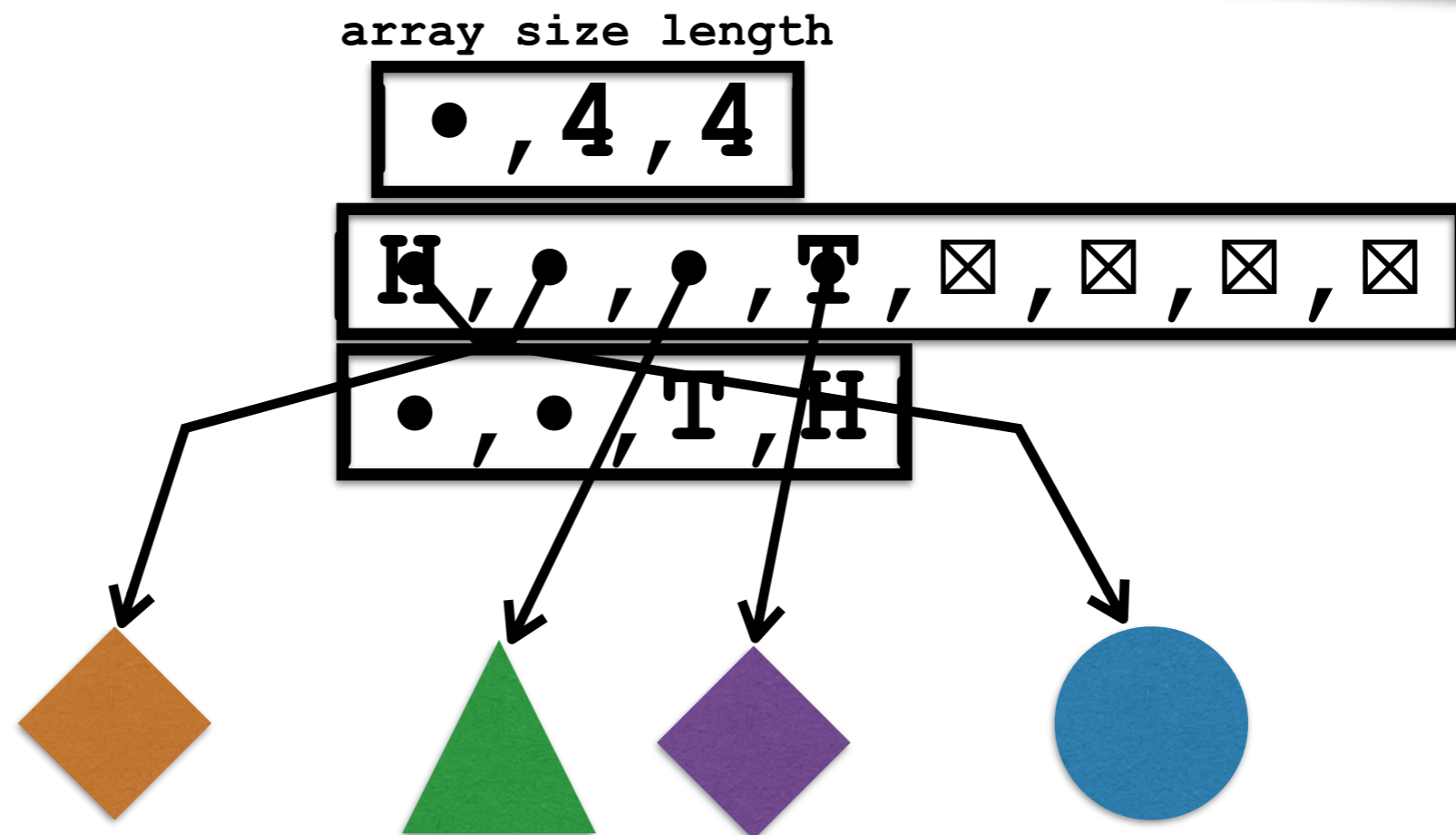
```



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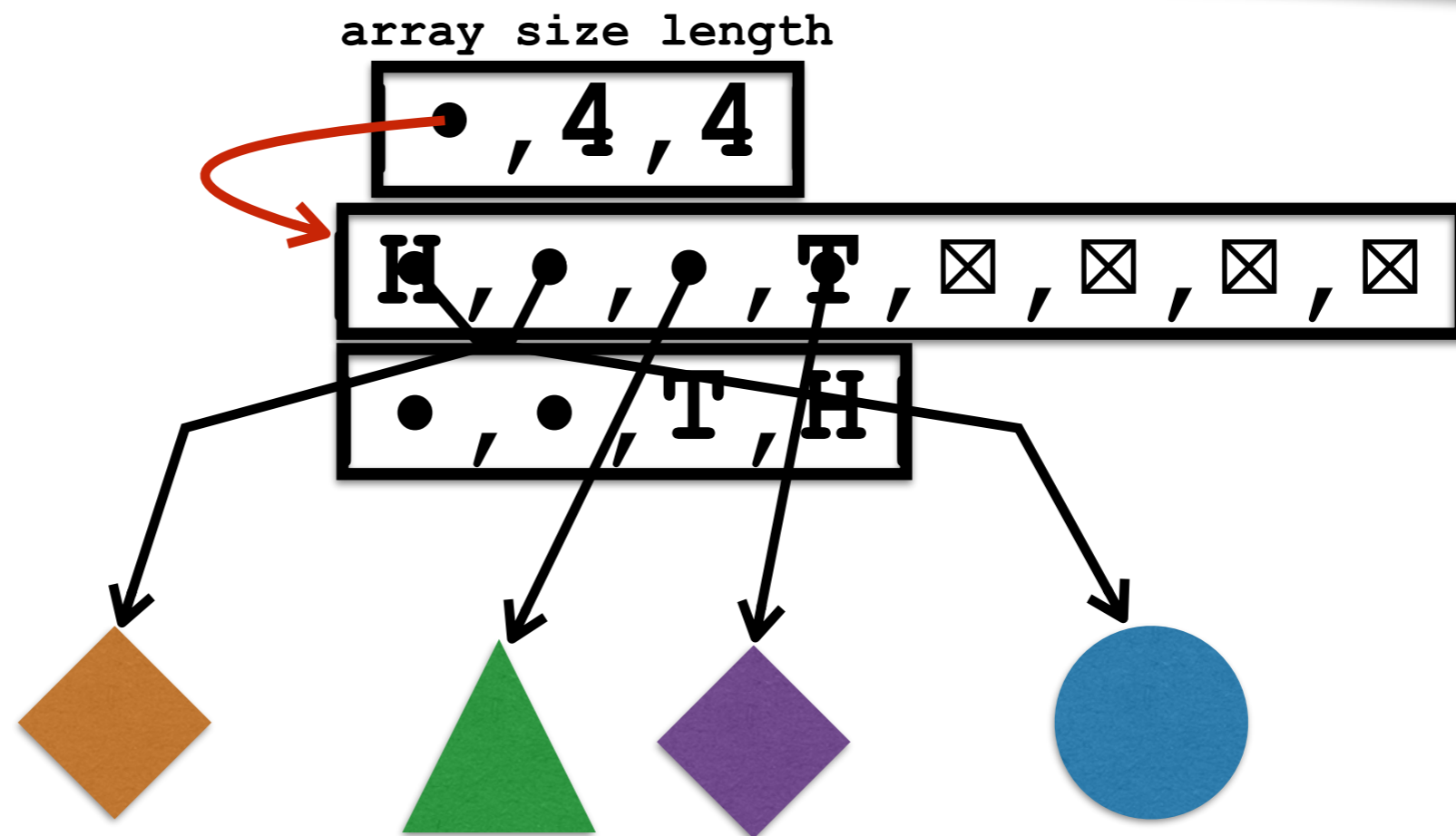
```



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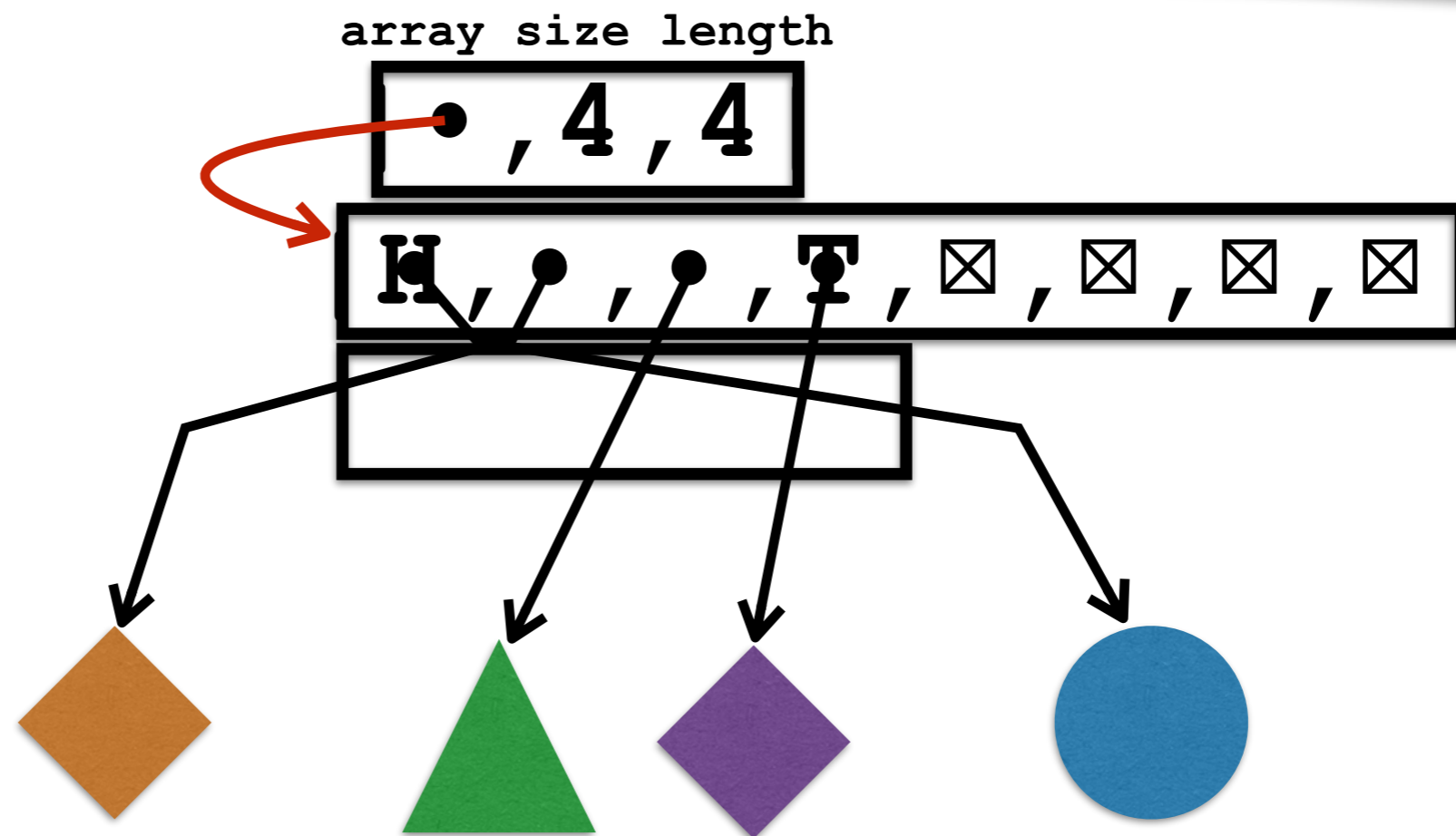




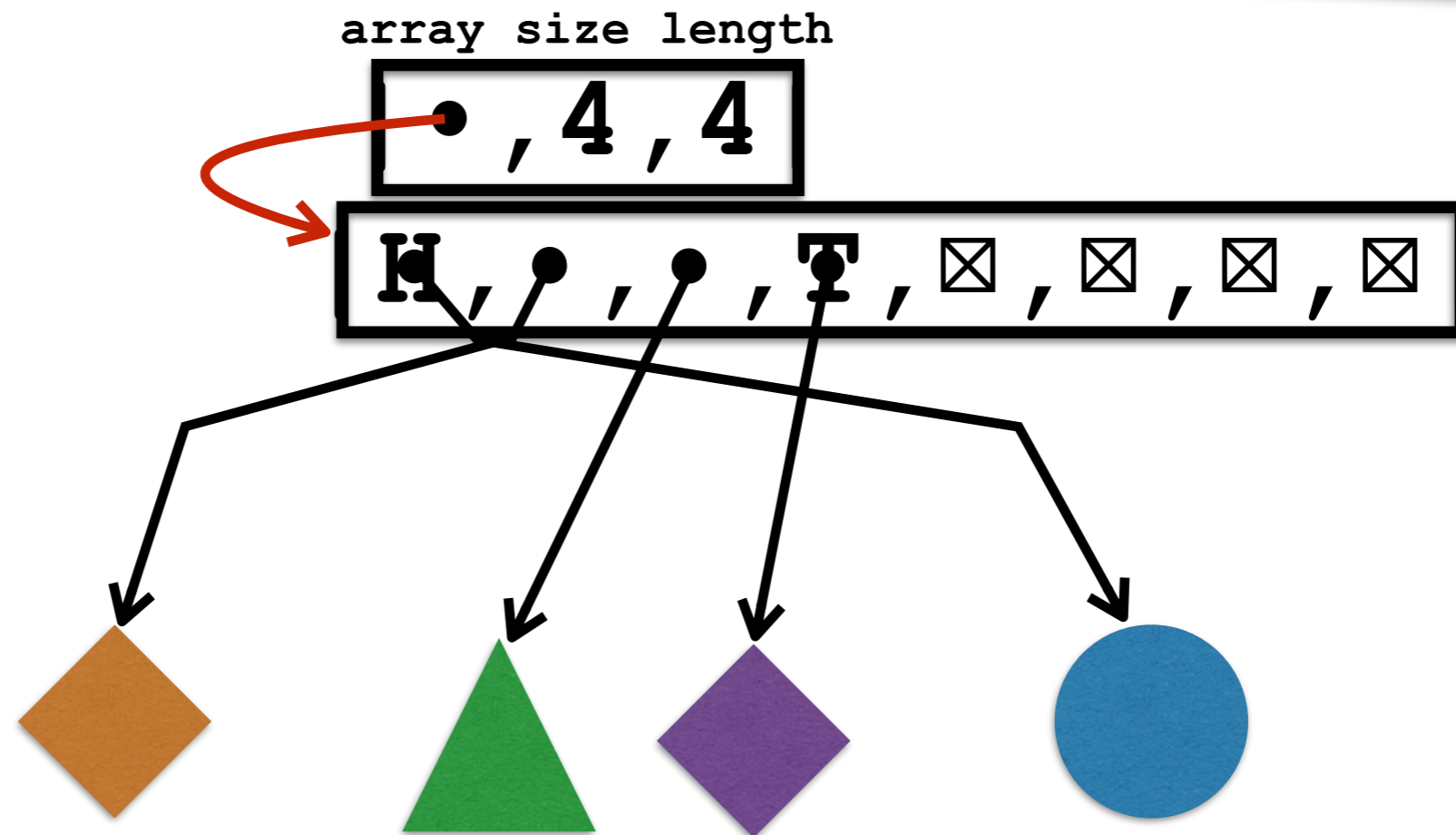
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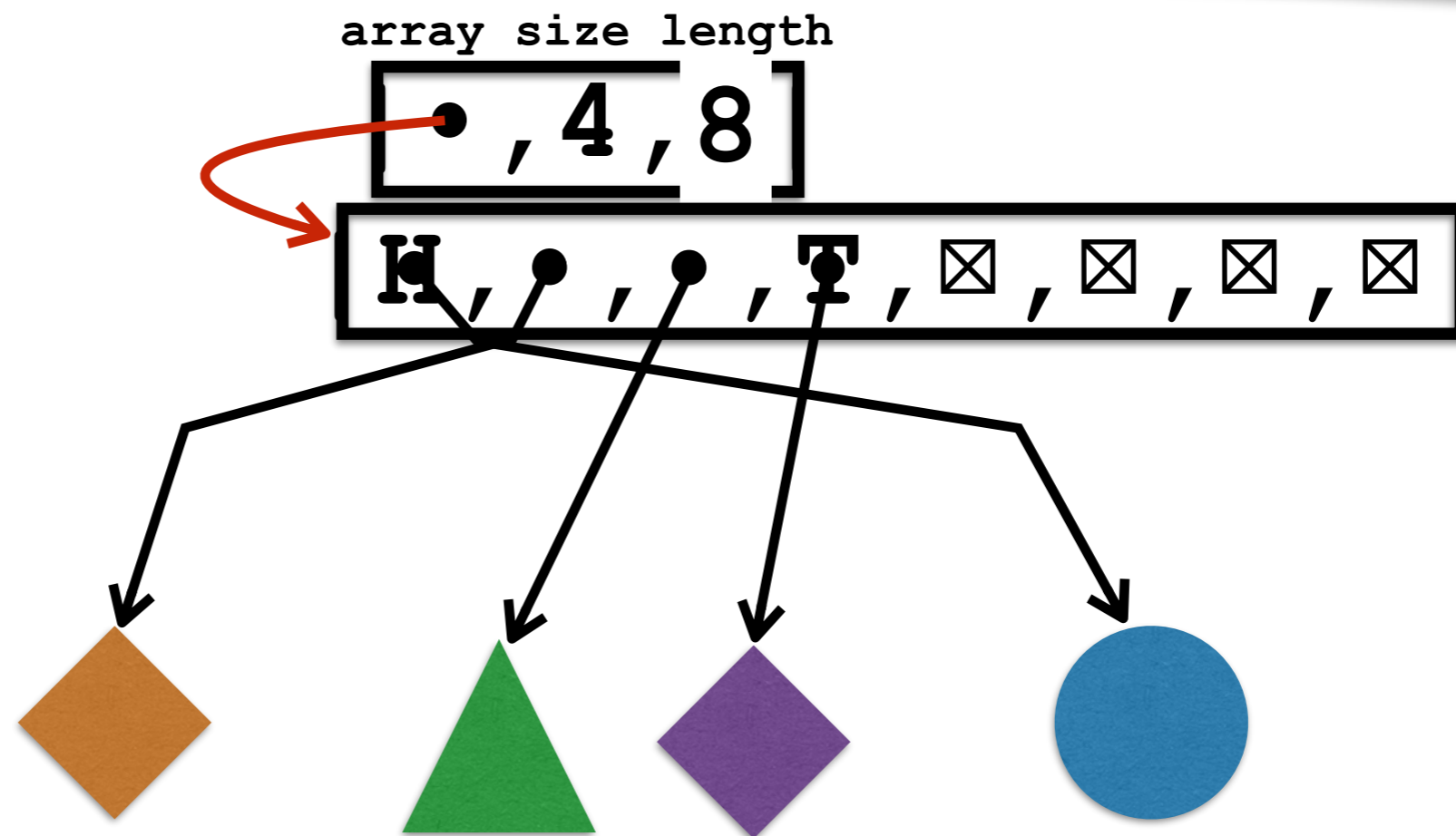
```



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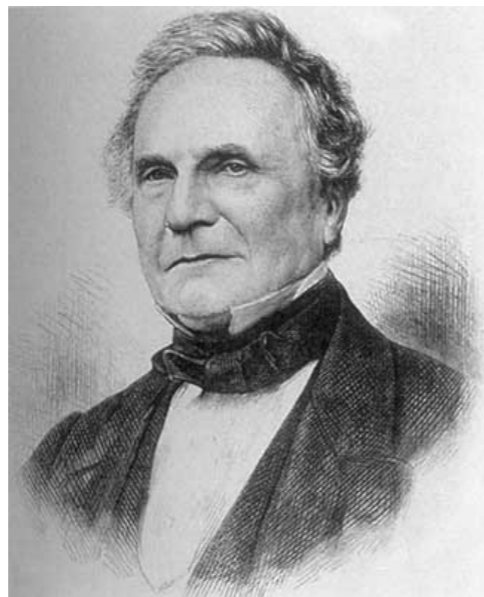
# Running Times and Asymptotic Notation



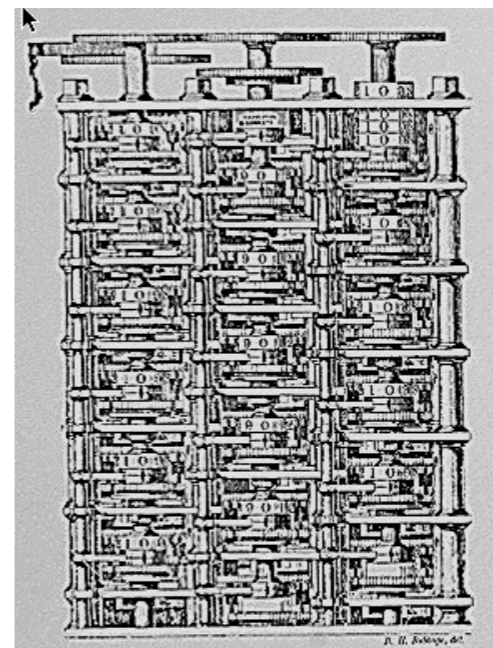
# Computational Tractability

As soon as an Analytic Engine exists, it will necessarily guide the future course of the science. Whenever any result is sought by its aid, the question will arise - By what course of calculation can these results be arrived at by the machine in the shortest time?

- Charles Babbage



Charles Babbage (1864)



Analytic Engine (schematic)

# Computational Tractability

**Brute force.** For many non-trivial problems, there is a natural brute force search algorithm that tries every possible solution.

- Typically takes  $2^N$  time or worse for inputs of size  $N$ .
- Unacceptable in practice.



even worse :  $N!$  for some problems

**Desirable scaling property.** When the input size doubles, the algorithm should only slow down by some constant factor  $C$ .

There exists constants  $a > 0$  and  $d > 0$  such that on every input of size  $N$ , its running time is bounded by  $aN^d$  steps.

**Def.** An algorithm is **poly-time** if the above scaling property holds.



choose  $C = 2^d$

# Worst Case Analysis

# Worst Case Analysis

**Worst case running time.** Obtain bound on **largest possible** running time of algorithm on any input of a given size  $N$ .

- Generally captures efficiency in practice.
- Draconian view, but hard to find effective alternative.

**Average case running time.** Obtain bound on running time of algorithm on **random** input as a function of input size  $N$ .

- Hard (or impossible) to accurately model real instances by random distributions.
- Algorithm tuned for a certain distribution may perform poorly on other inputs.



Winter 2016

**COMP-250: Introduction  
to Computer Science**

Lecture 8, February 4, 2016