# Bitvectors with runs and the successor/predecessor problem



Data Compression Conference

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Wednesday, 25th March 2020, Snowbird, Utah



This work has received funding from the European Union's Horizon 2020 research and innovation programme under the Marie Skłodowska-Curie grant agreement No 690941





# Outline

#### Introduction

Background

**D**zombit-vector

Experimental evaluation

Conclusions

#### **D** Future work



Search engines: looking for "data compression"

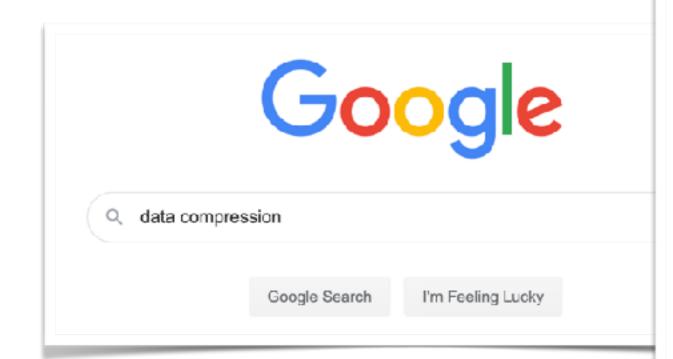


Search engines: looking for "data compression"

Google						
Q data compres	ssion		Ŷ			
	Google Search	I'm Feeling Lucky				



Search engines: looking for "data compression"



About 243,000,000 results (0.55 seconds)

en.wikipedia.org > wiki > Data\_compression 💌

#### Data compression - Wikipedia

In signal processing, data compression, source coding, or bit-rate reduction is the process of encoding information using fewer bits than the original ... Lossless · Lossy · Theory · Uses

en.wikipedia.org > wiki > Lossless\_compression 💌

#### Lossless compression - Wikipedia

Lossless compression is a class of **data compression** algorithms that allows the original data to be perfectly reconstructed from the compressed data. By contrast ...

www.techopedia.com > definition > data-compression - -

#### What is Data Compression? - Definition from Techopedia

Data compression is the process of modifying, encoding or converting the bits structure of data in such a way that it consumes less space on disk. It enables reducing the storage size of one or more data instances or elements. Data compression is also known as source coding or bit-rate reduction.

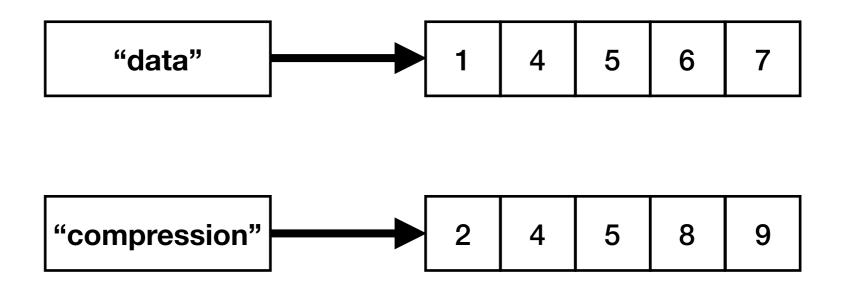
www.britannica.com > technology > data-compression \*

#### Data compression | computing | Britannica

Data compression, also called compaction, the process of reducing the amount of data needed for the storage or transmission of a given piece of information, ...

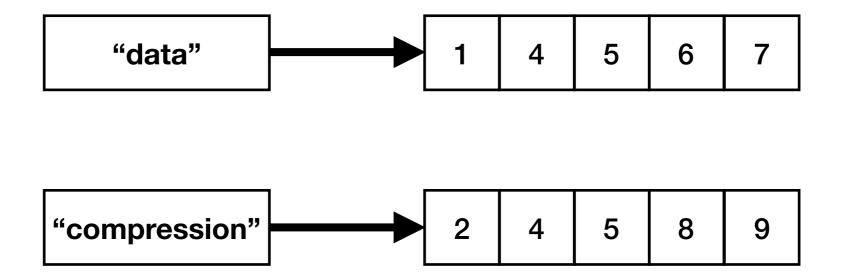


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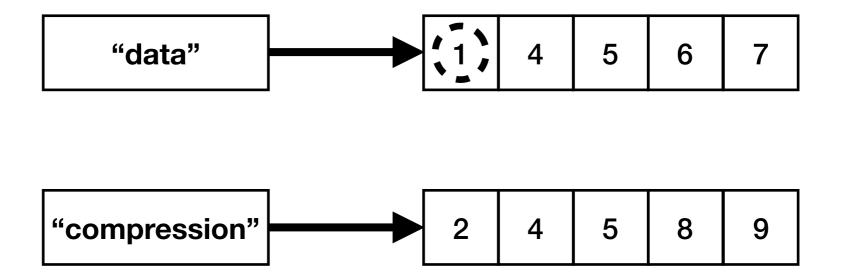


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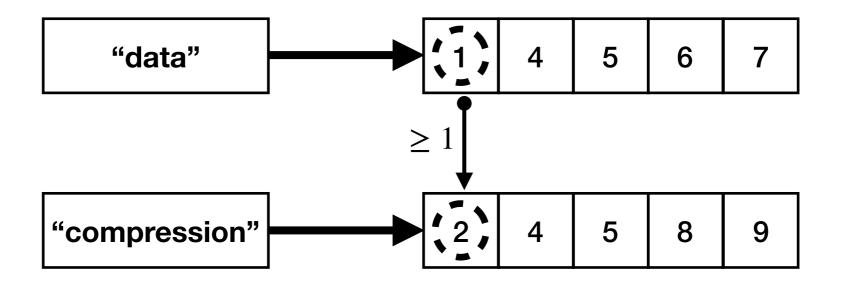


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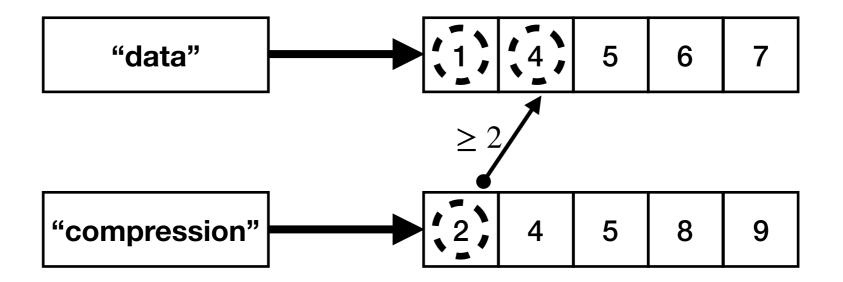


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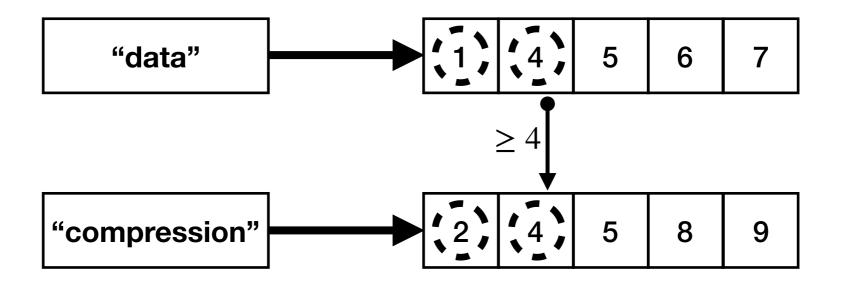


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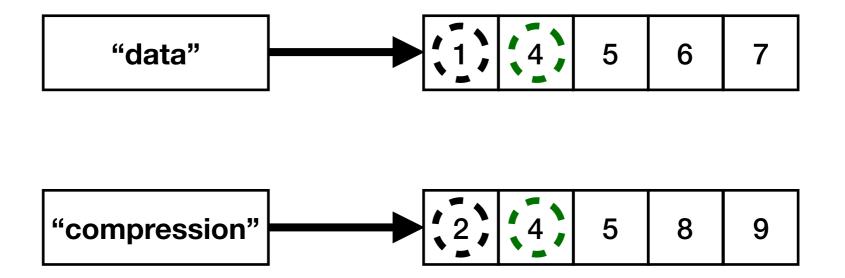


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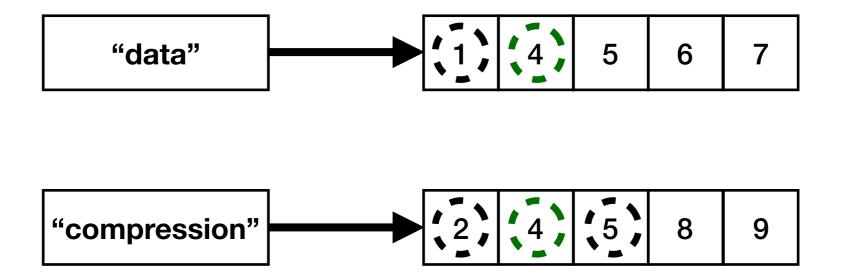


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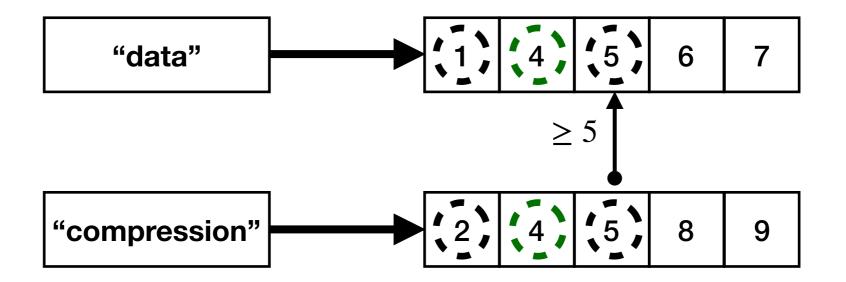


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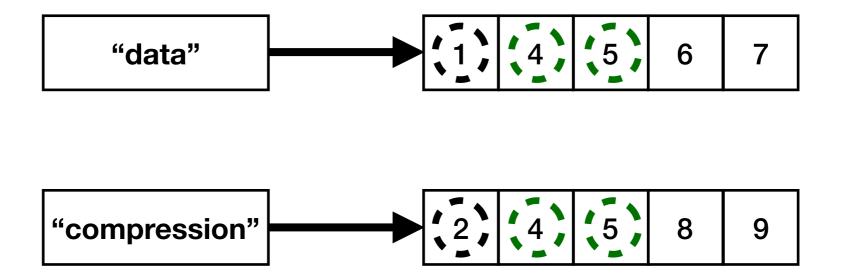


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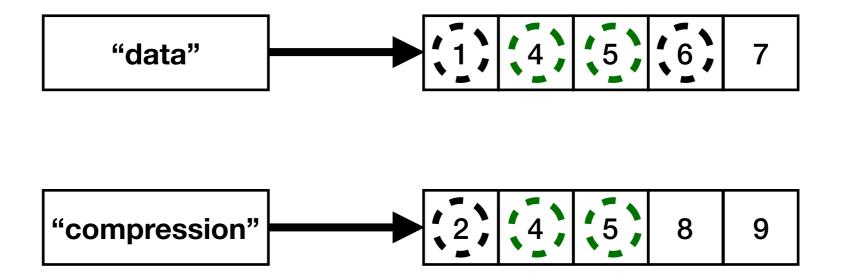


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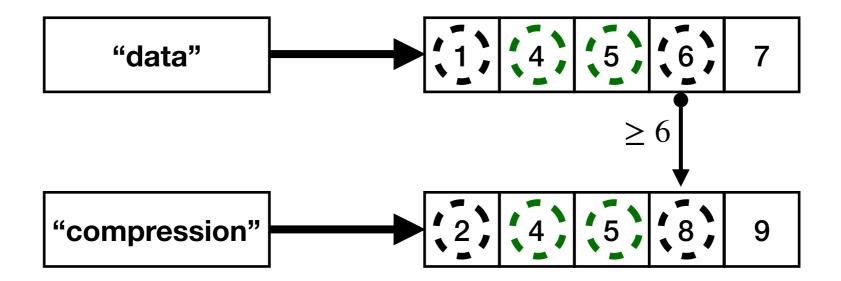


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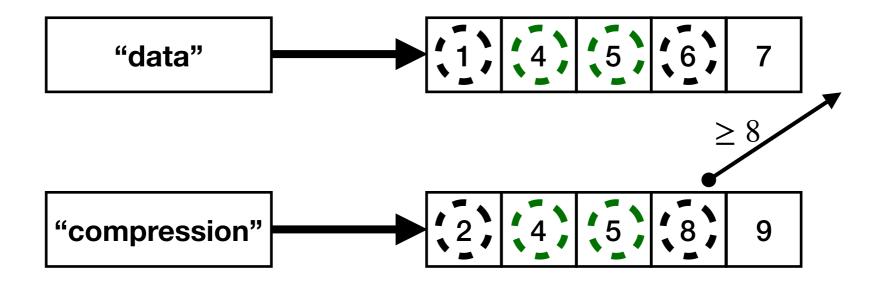


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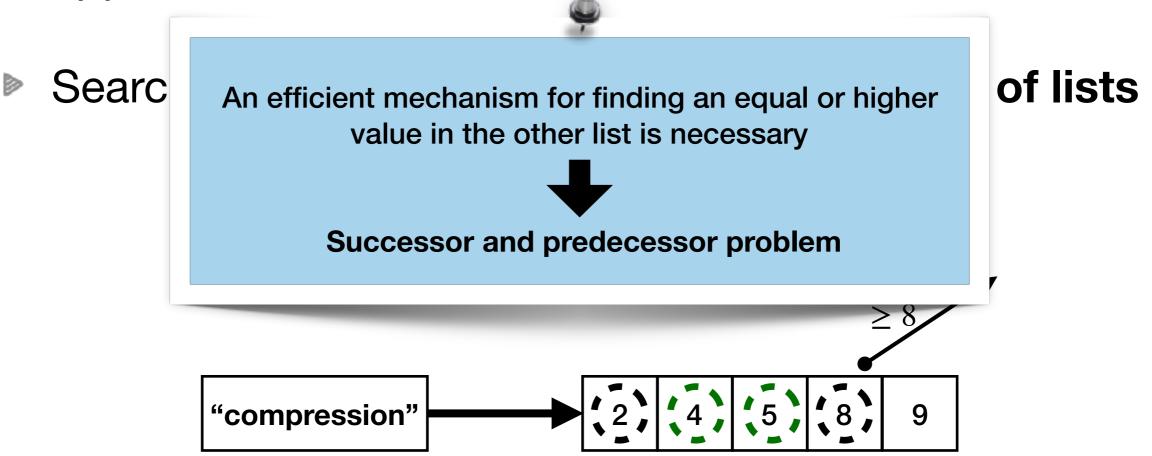


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#### Successor and predecessor

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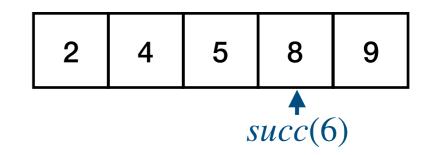
2 4	5	8	9
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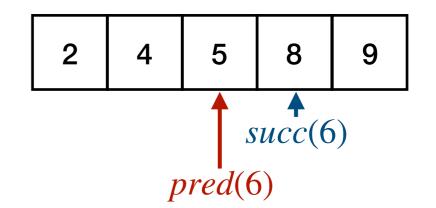




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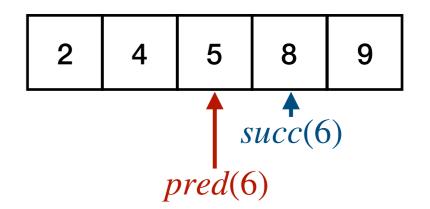


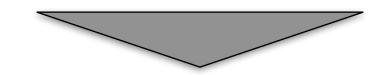
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1	2	3	4	5	6	7	8	9
0	1	0	1	1	0	0	1	1

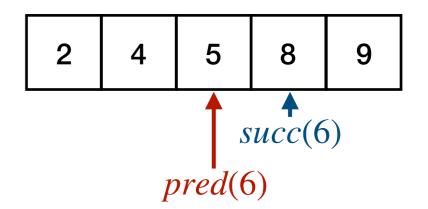


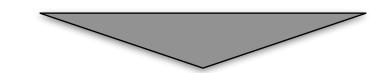
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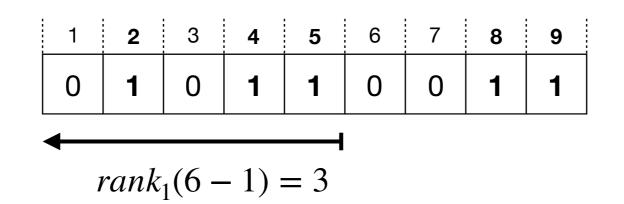
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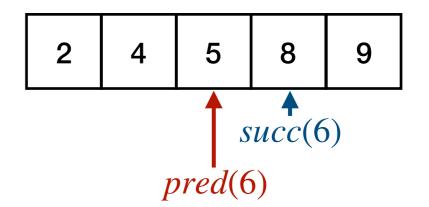


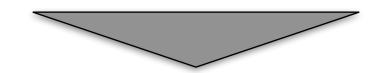


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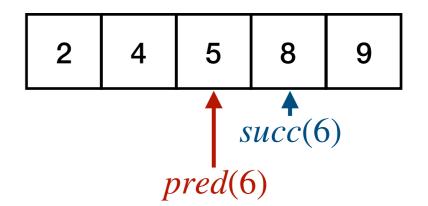


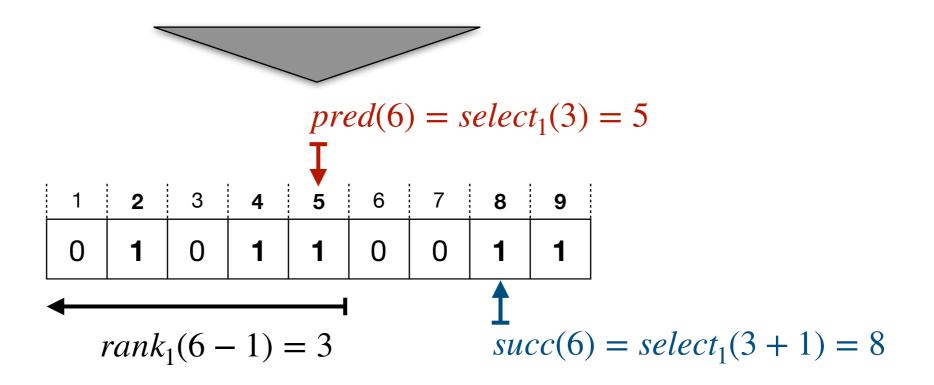




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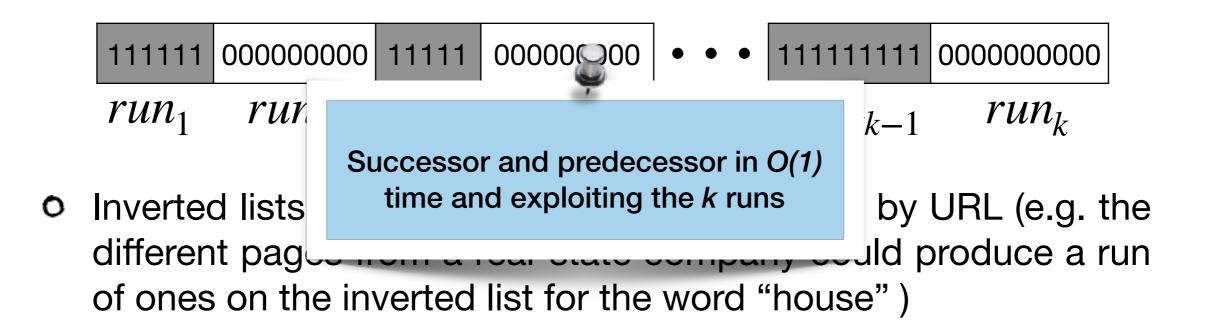
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- Hybrid bitvector: splits the bitvector into different parts and choose the best technique (sparse, runs, plain) for each individual division. [hybrid]



:

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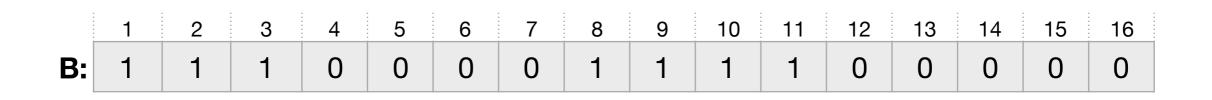


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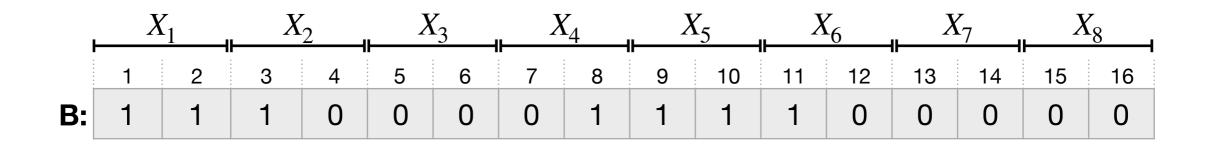


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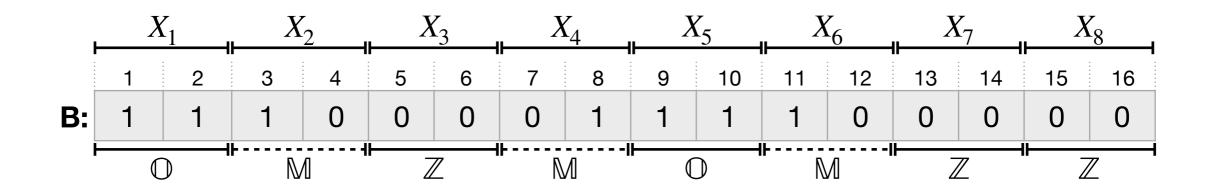




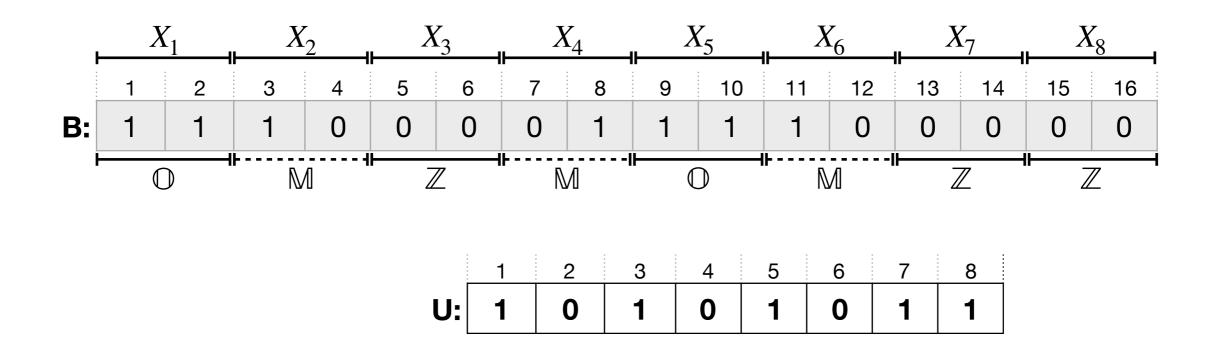




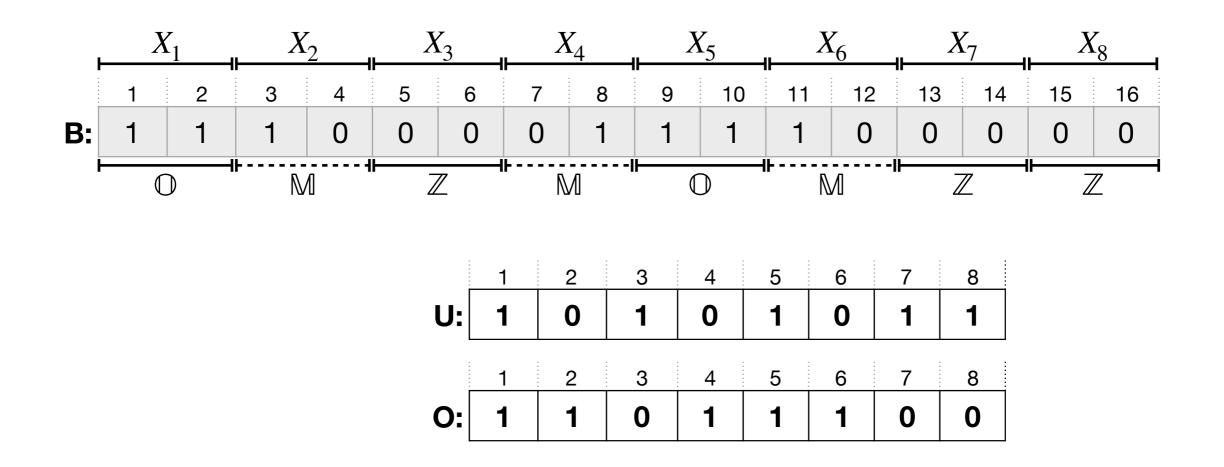




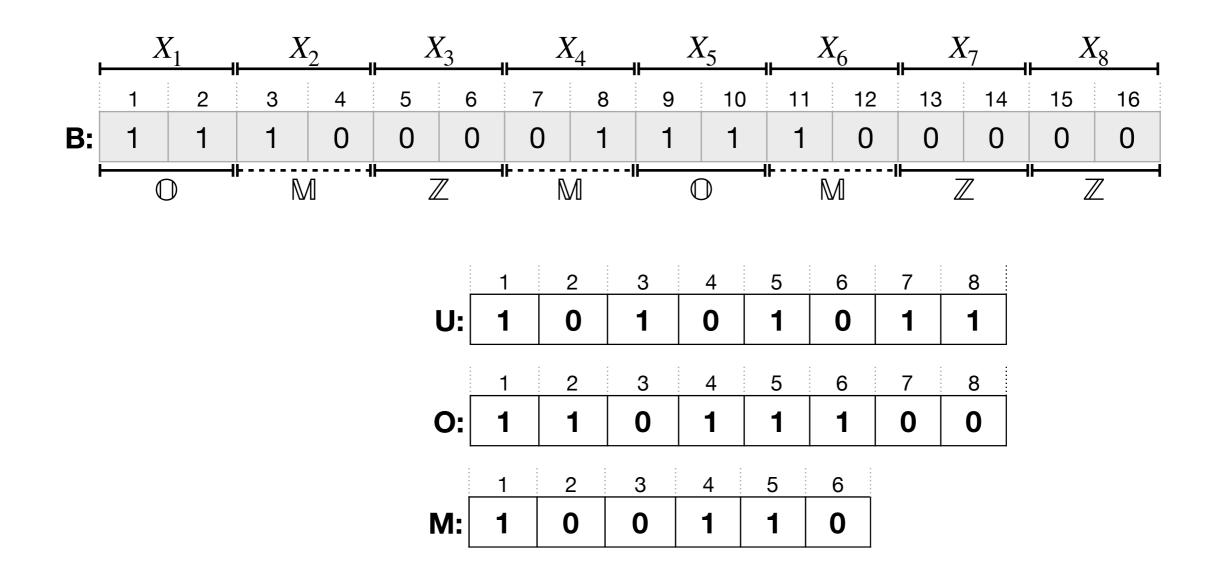




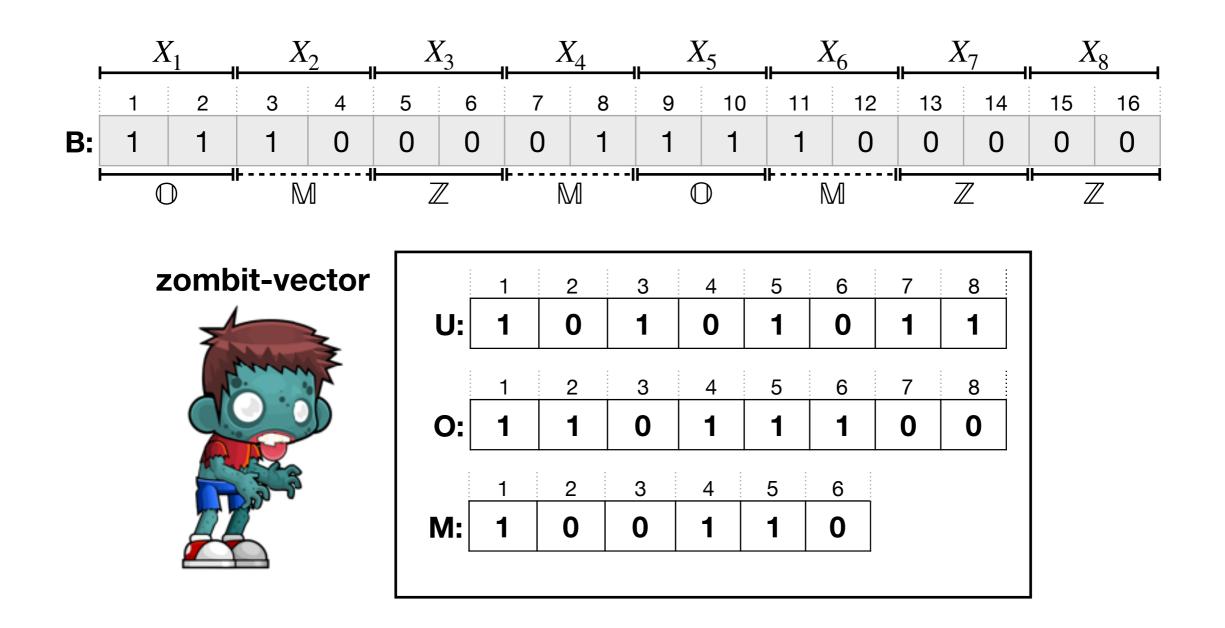




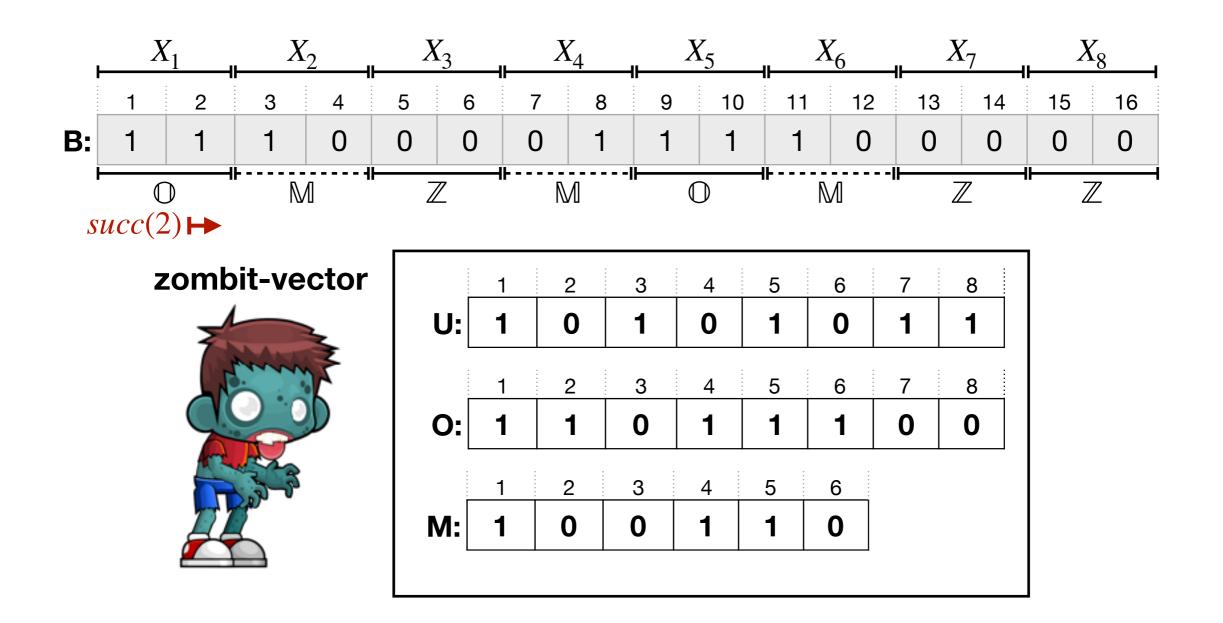




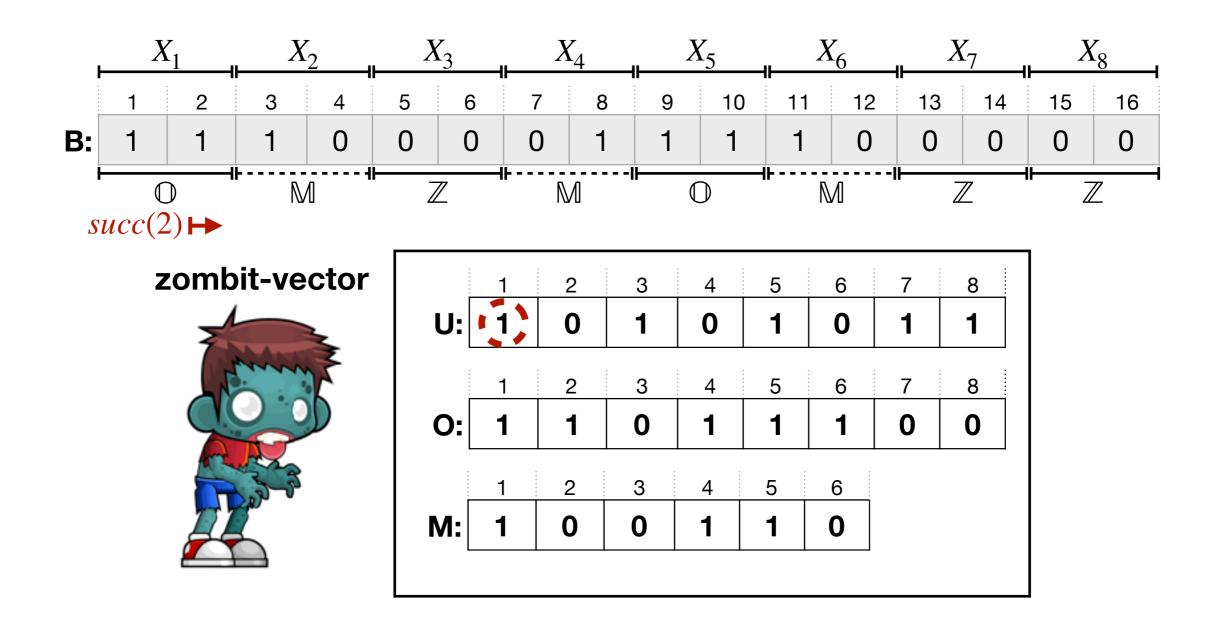




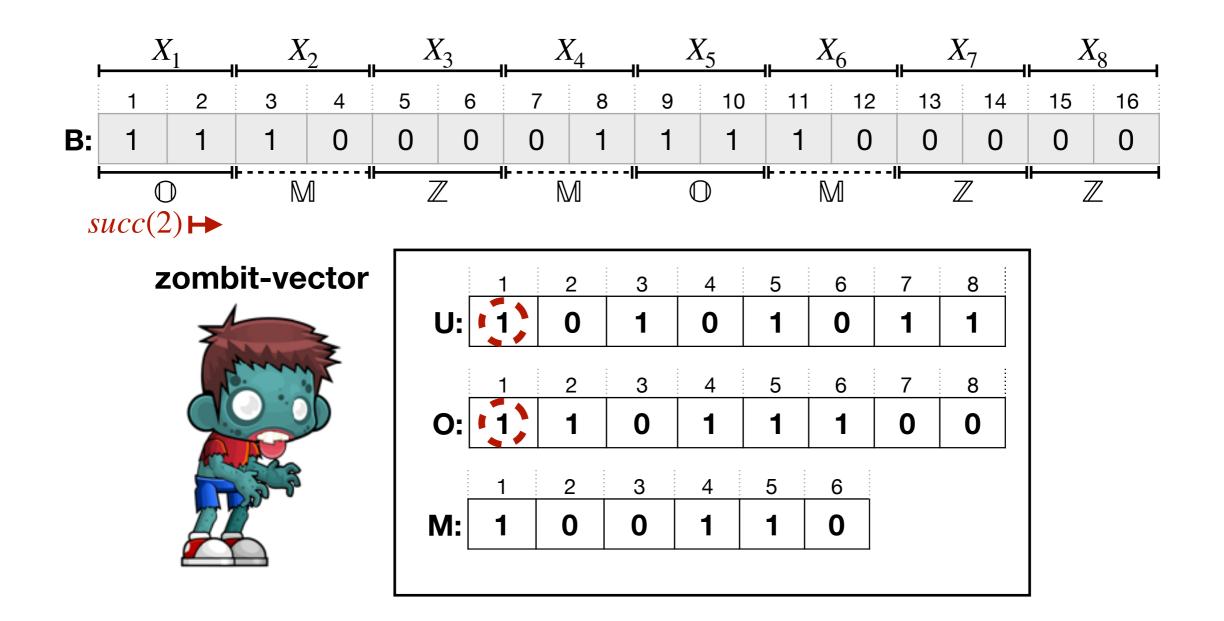




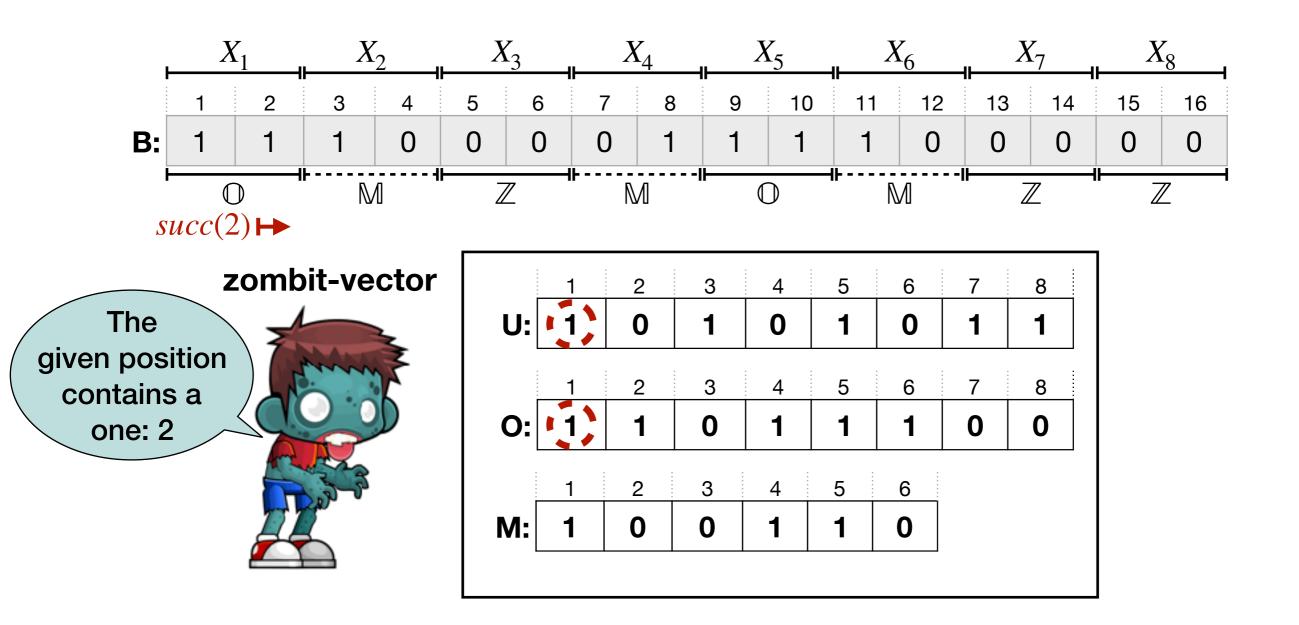




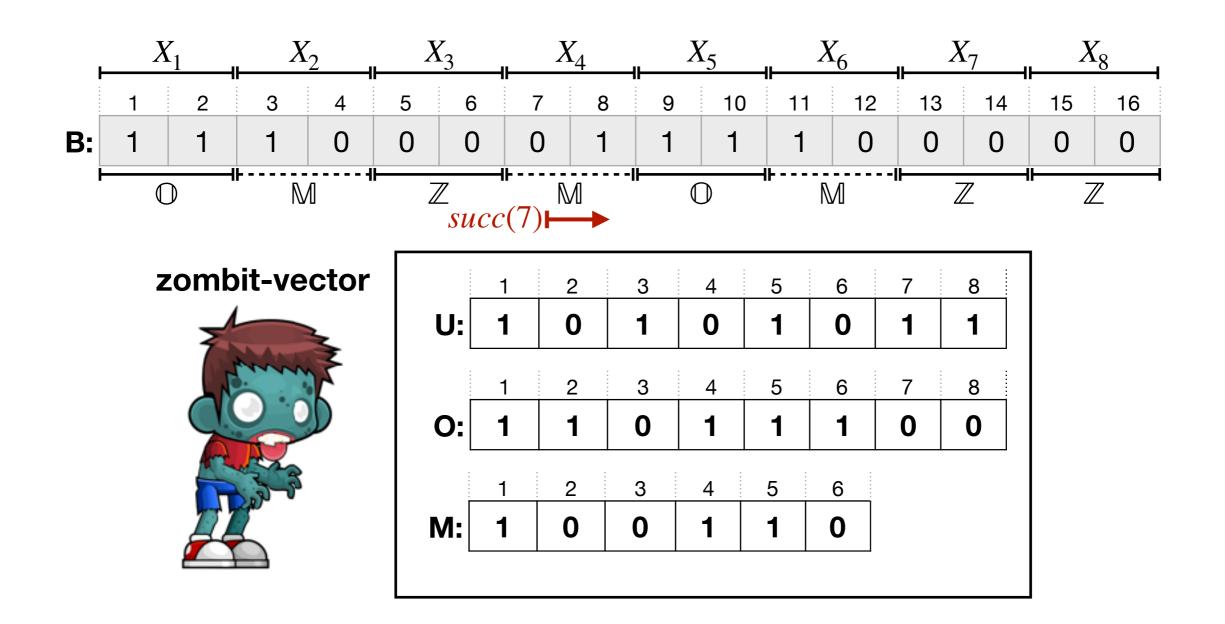




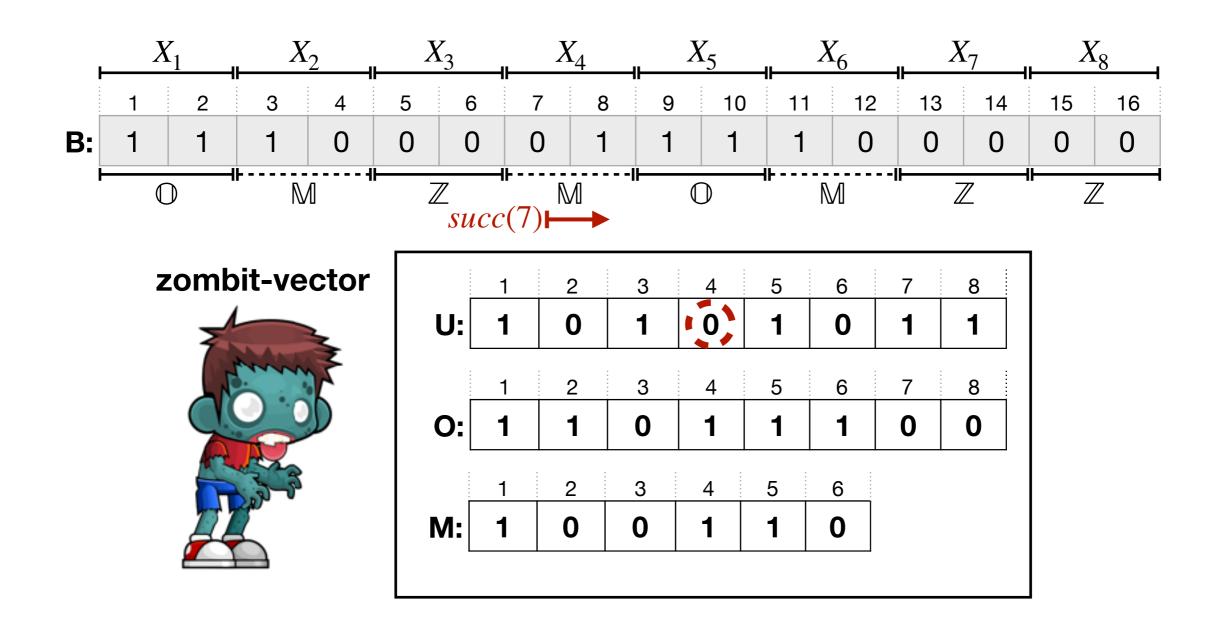




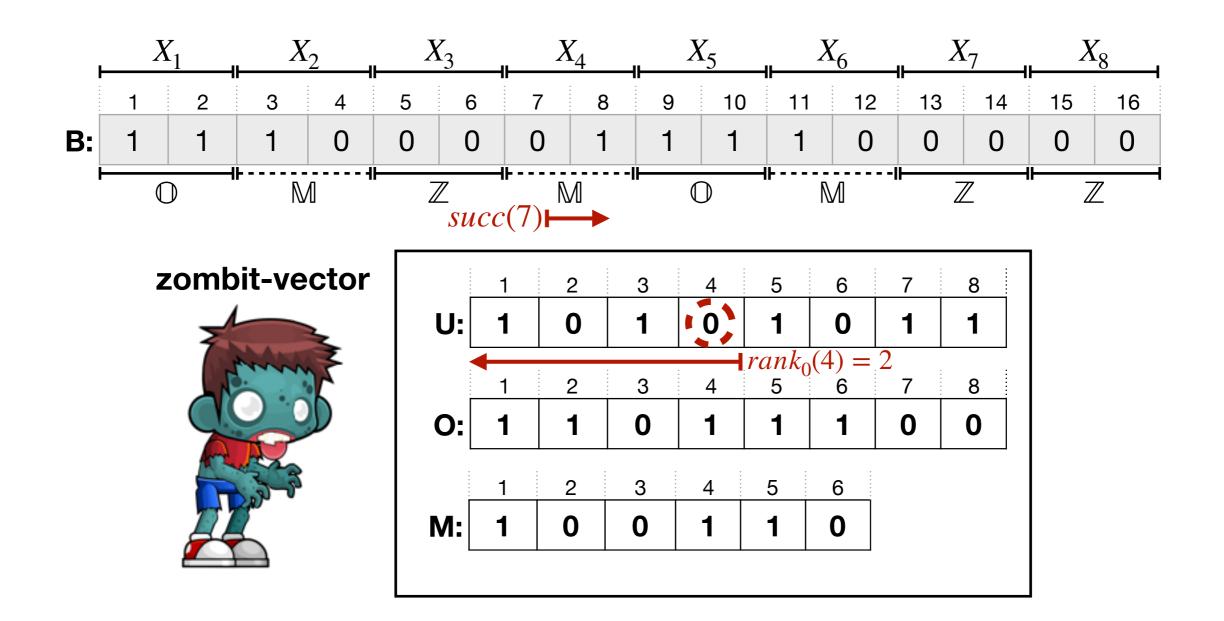




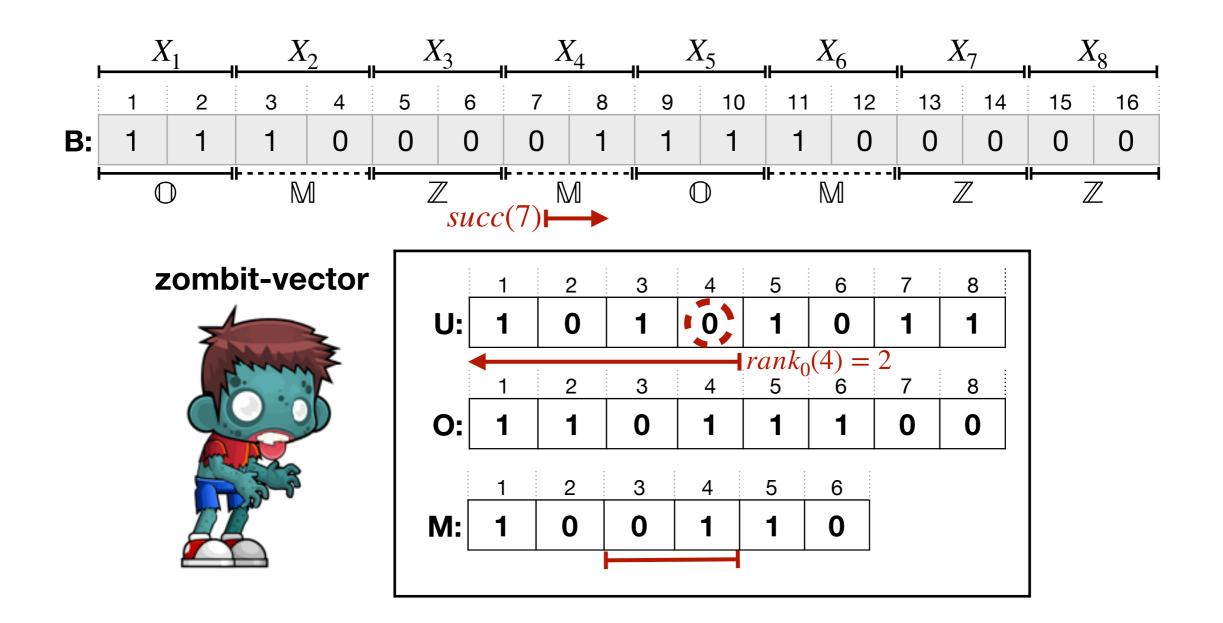




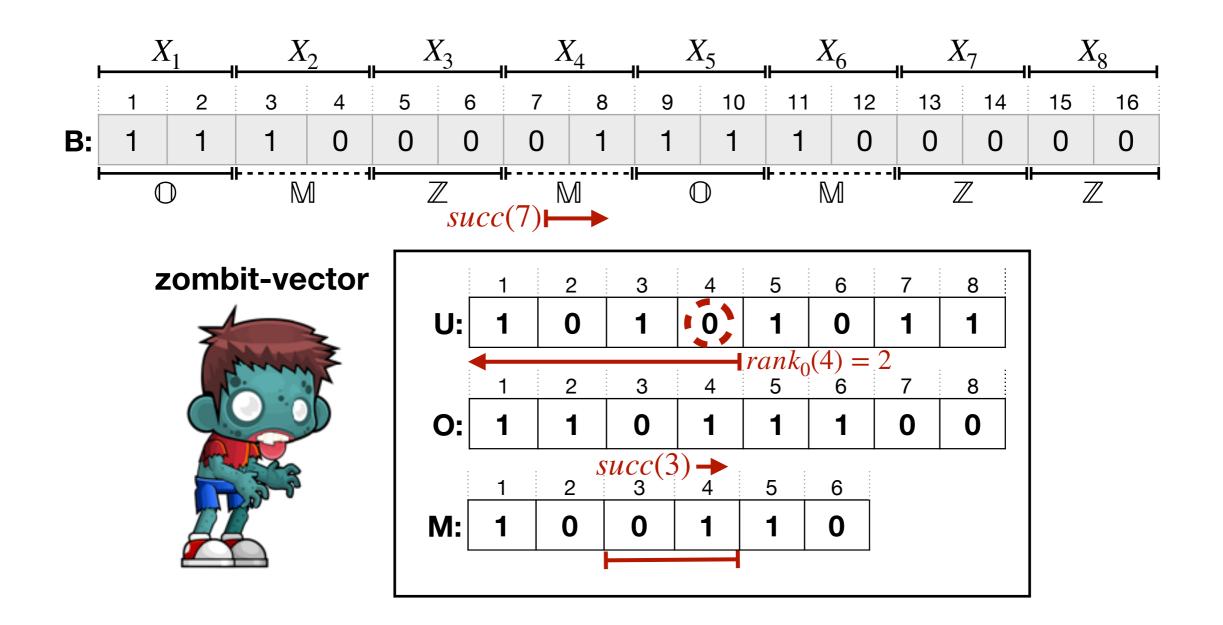




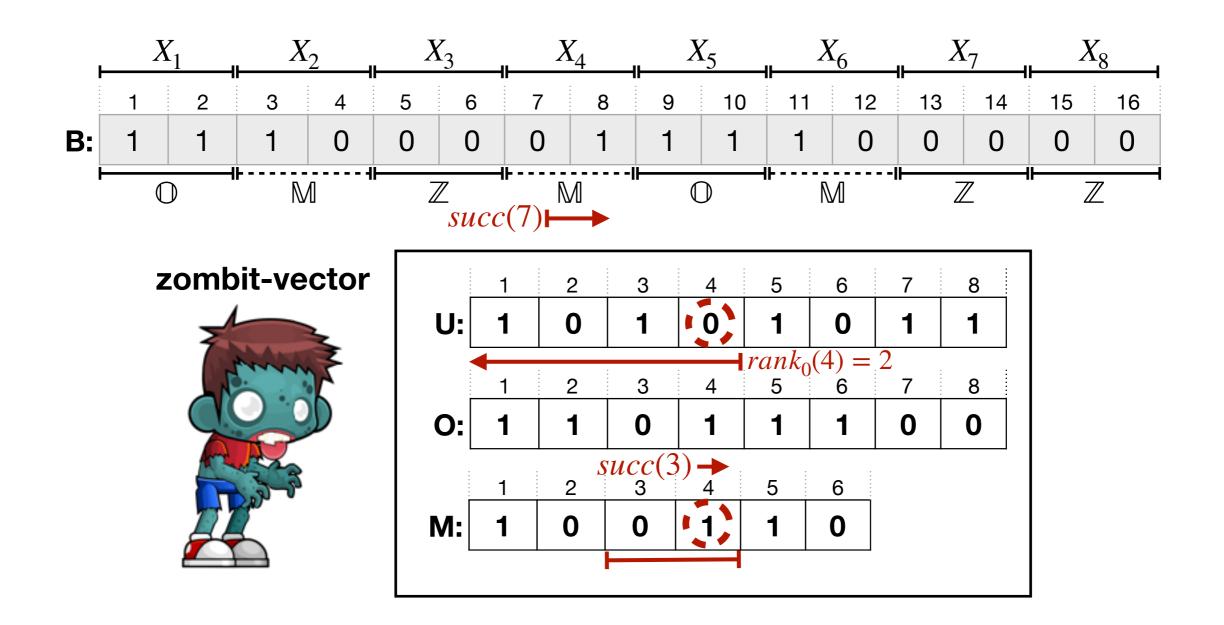




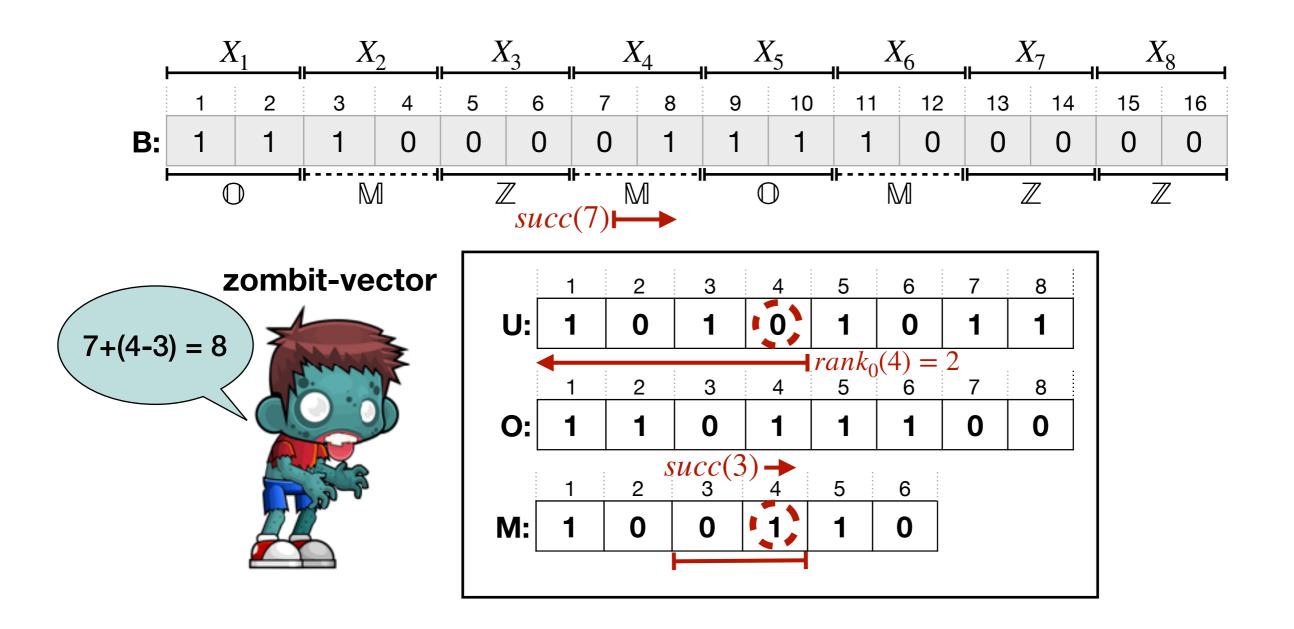




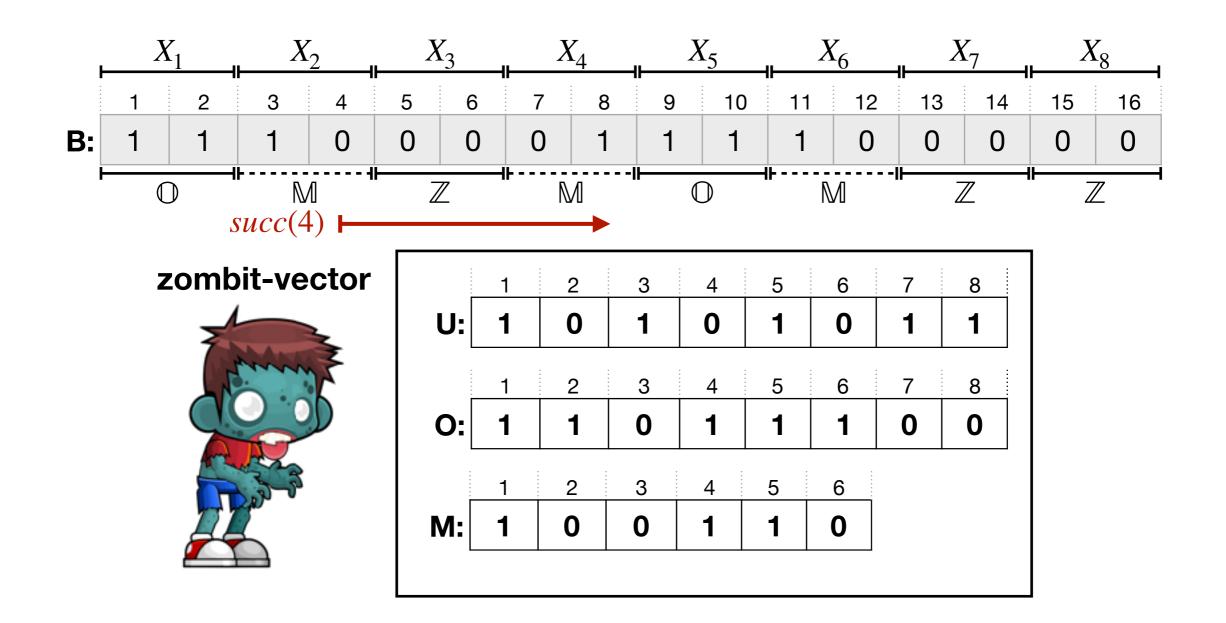




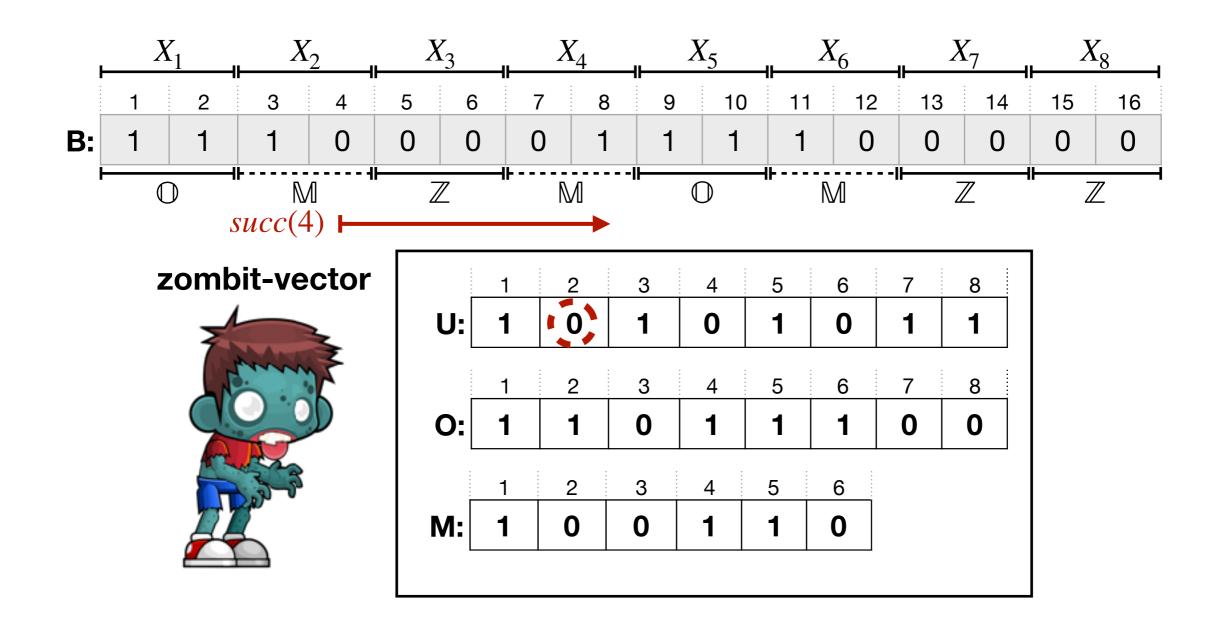




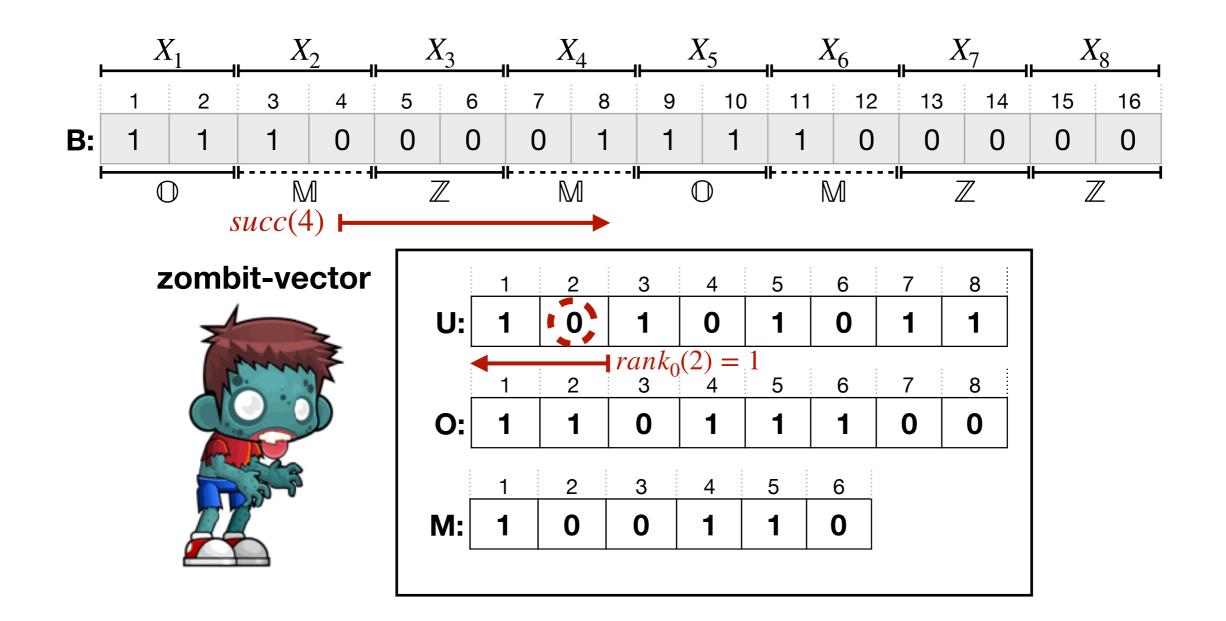




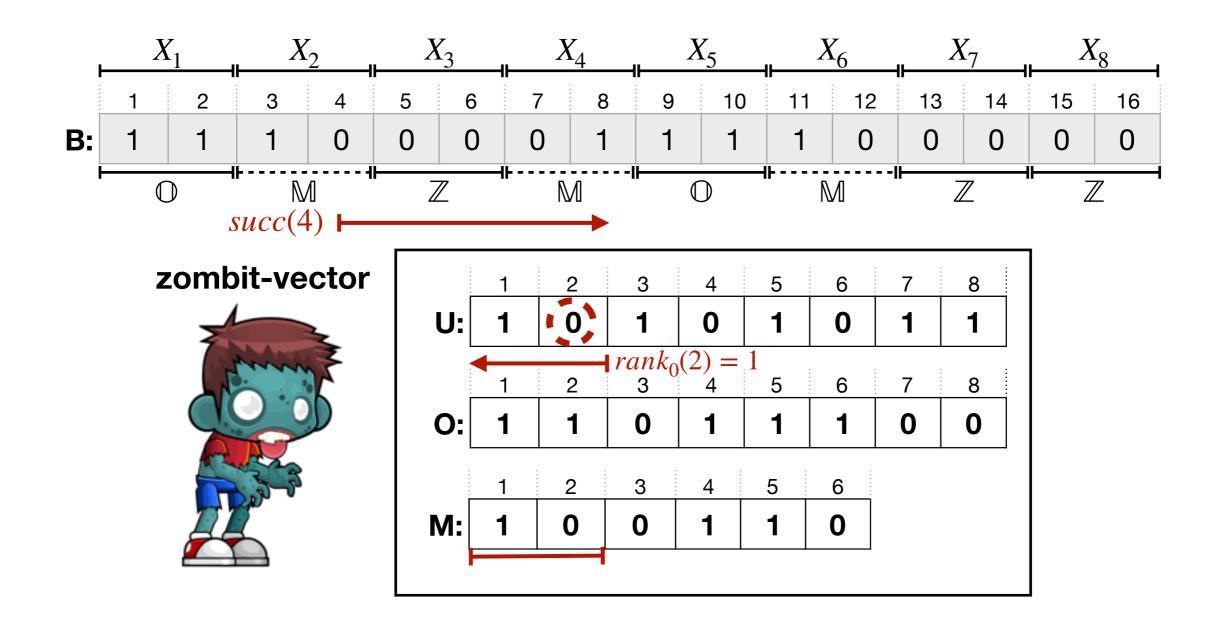




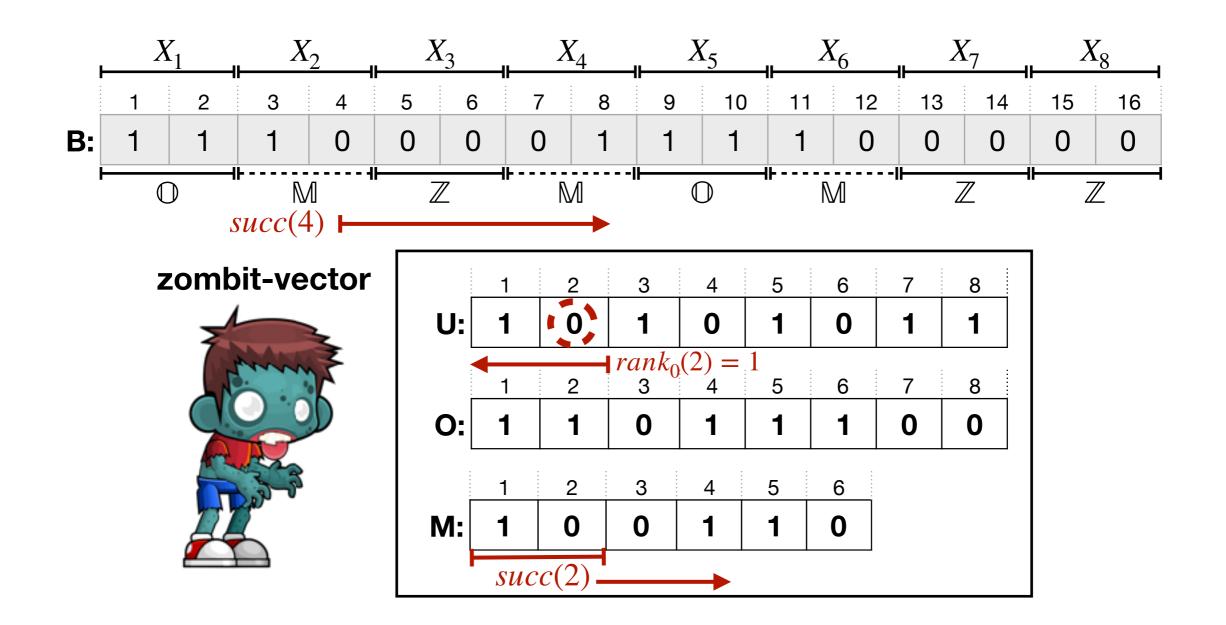




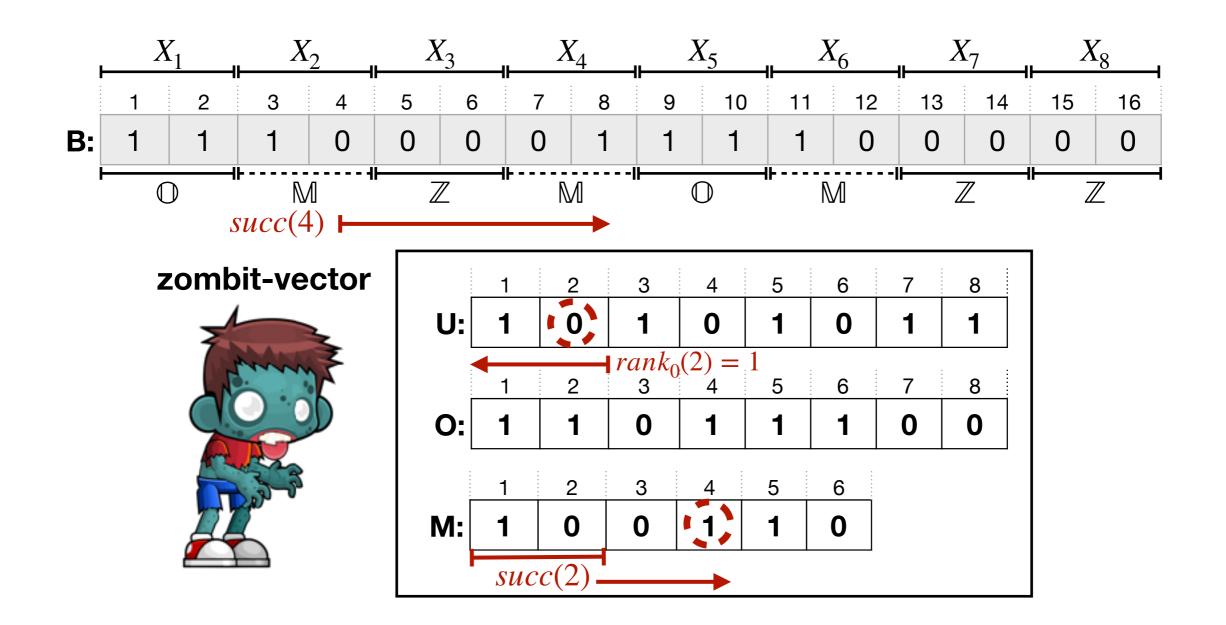




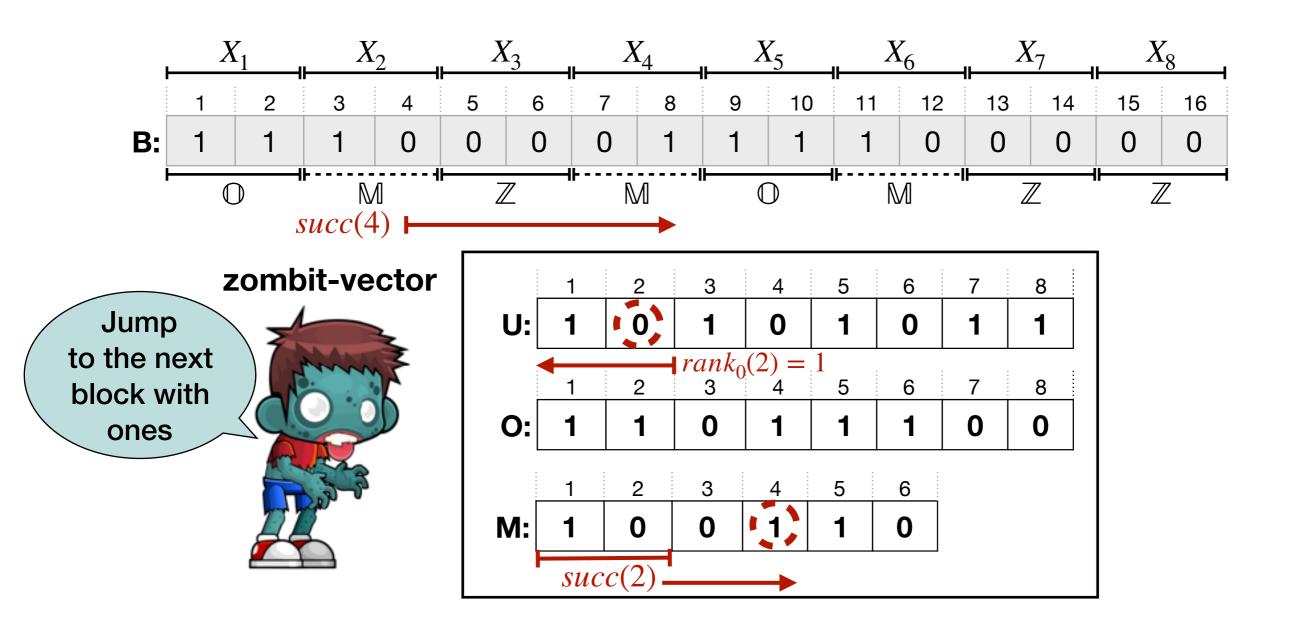




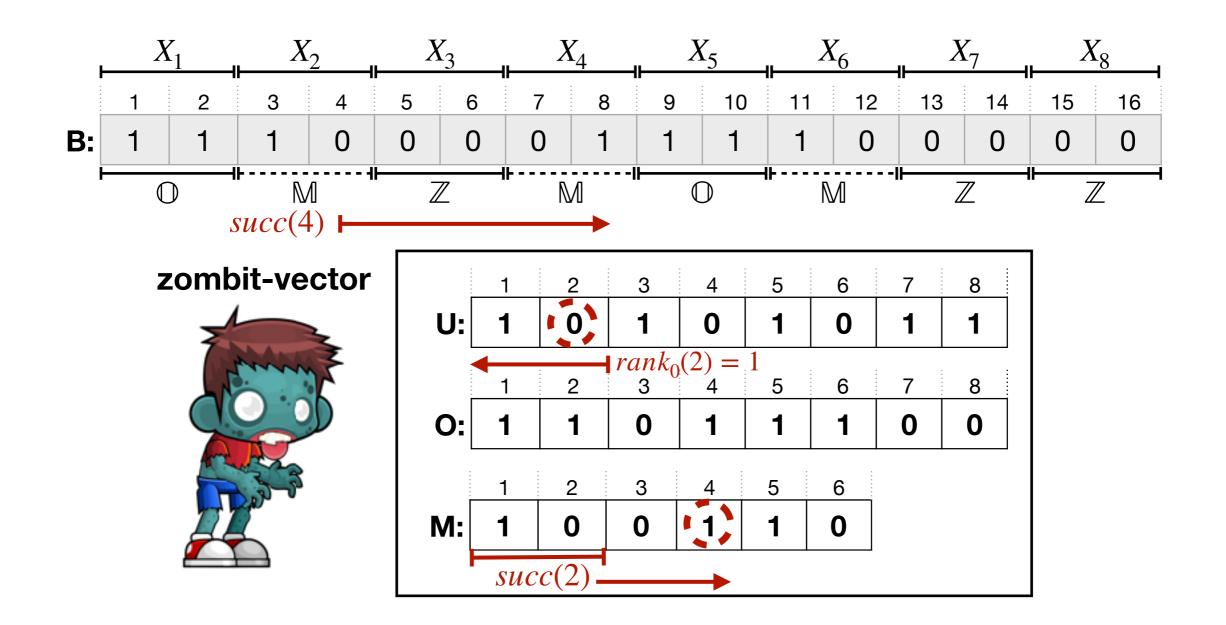




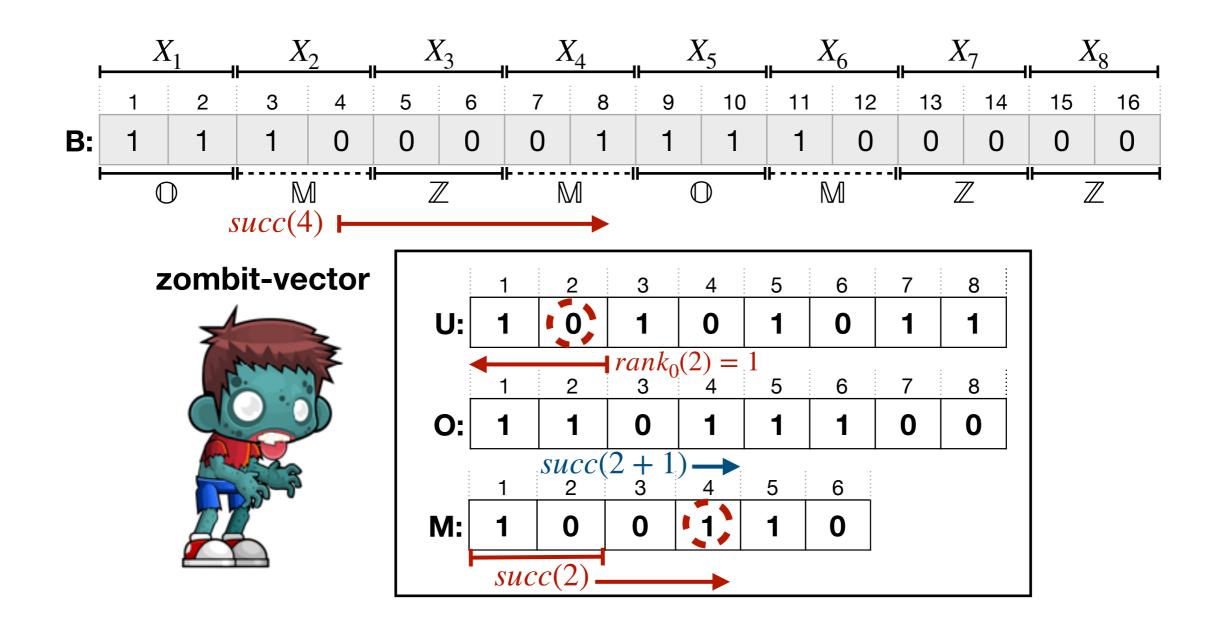




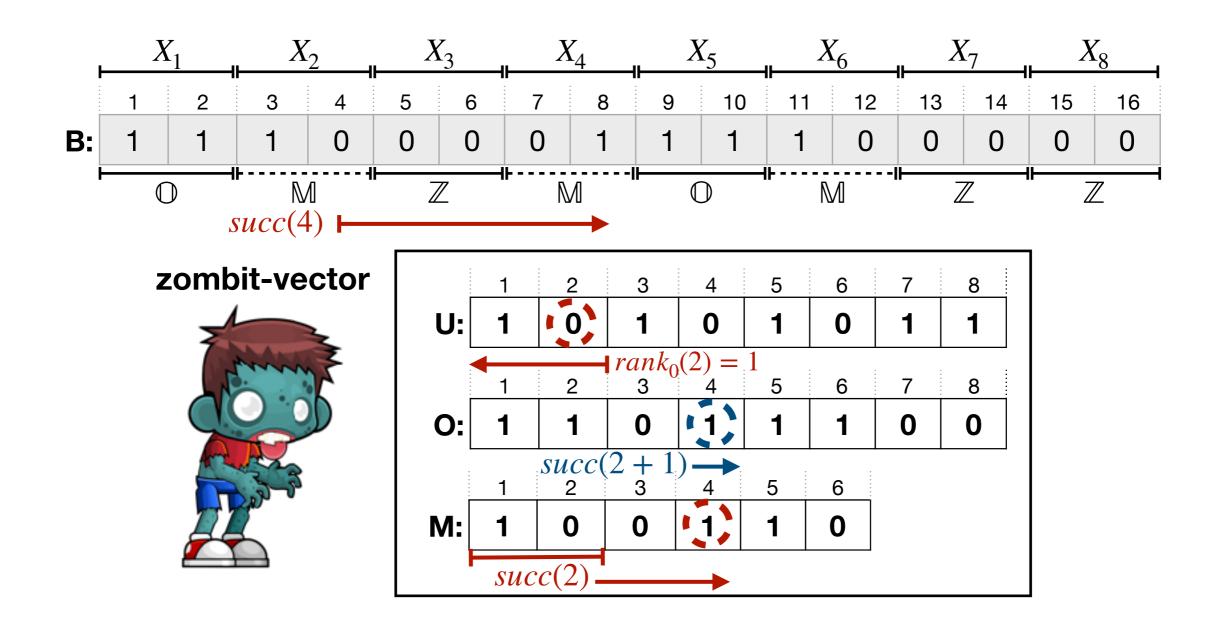




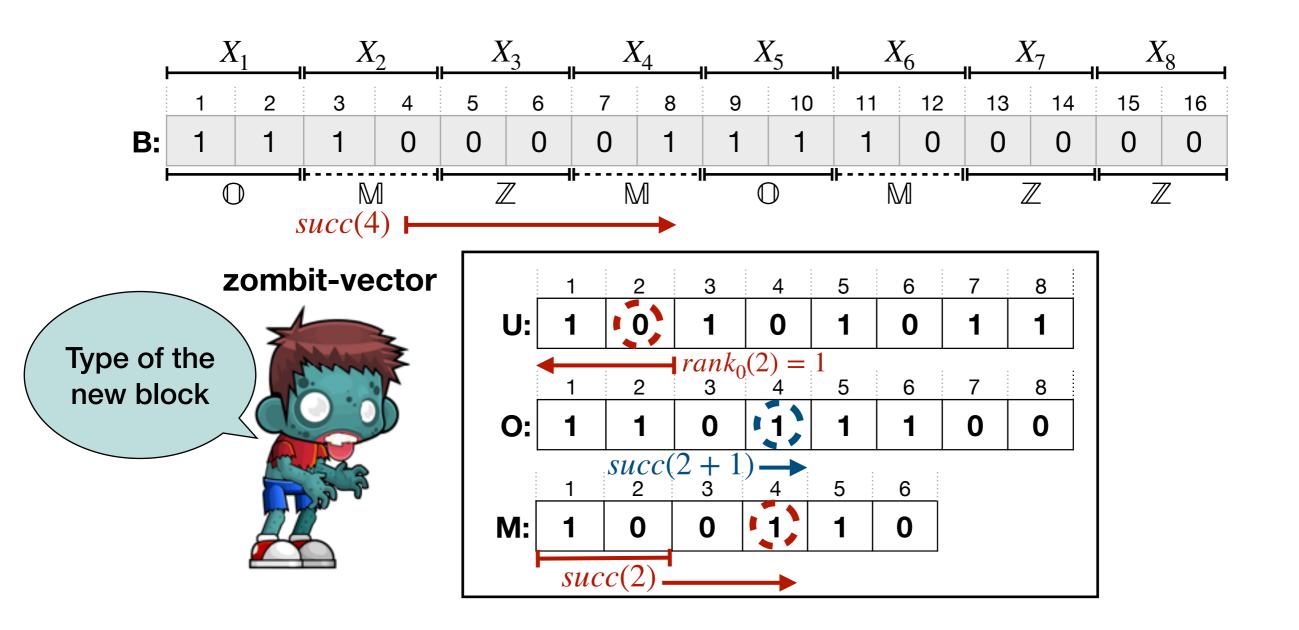




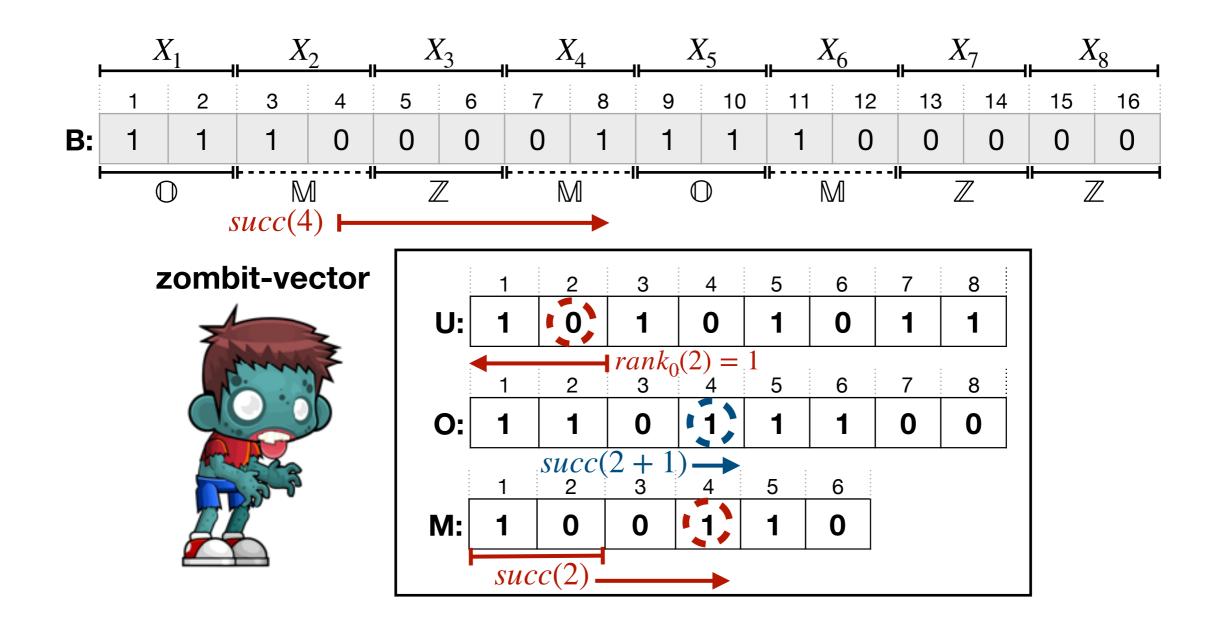




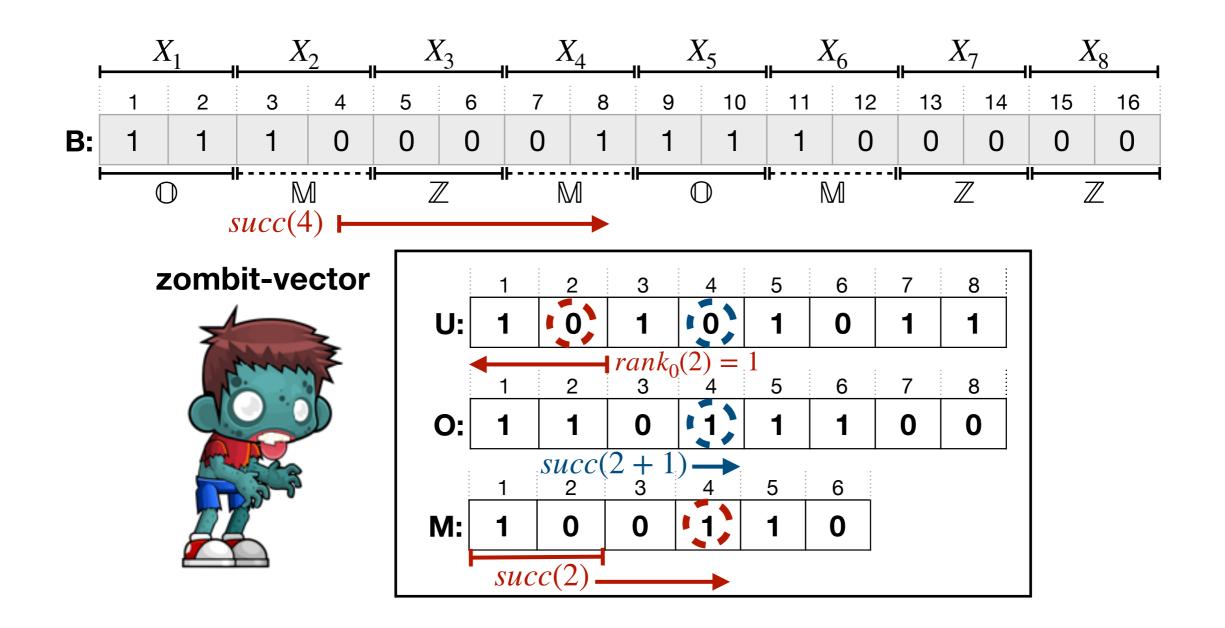




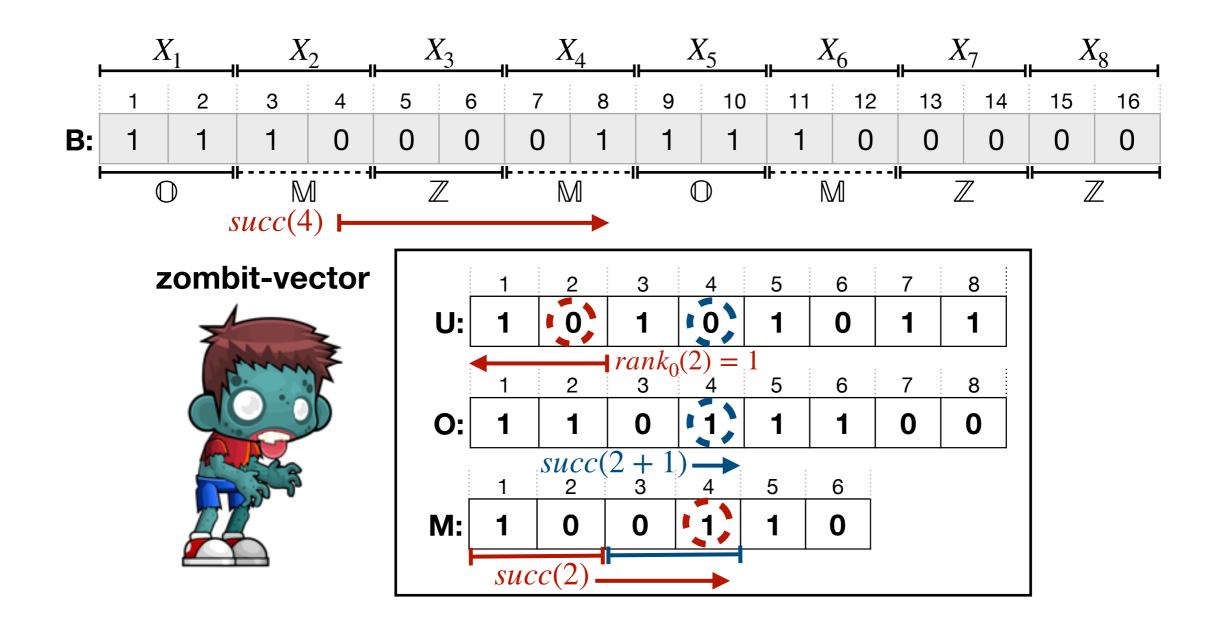




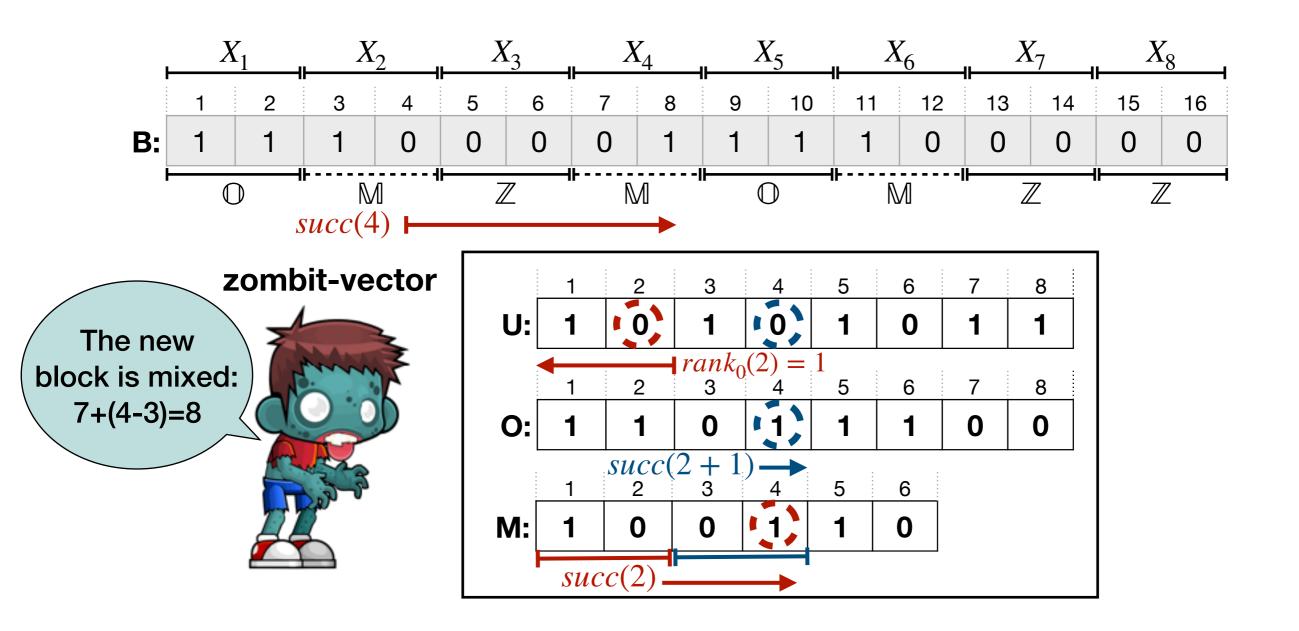




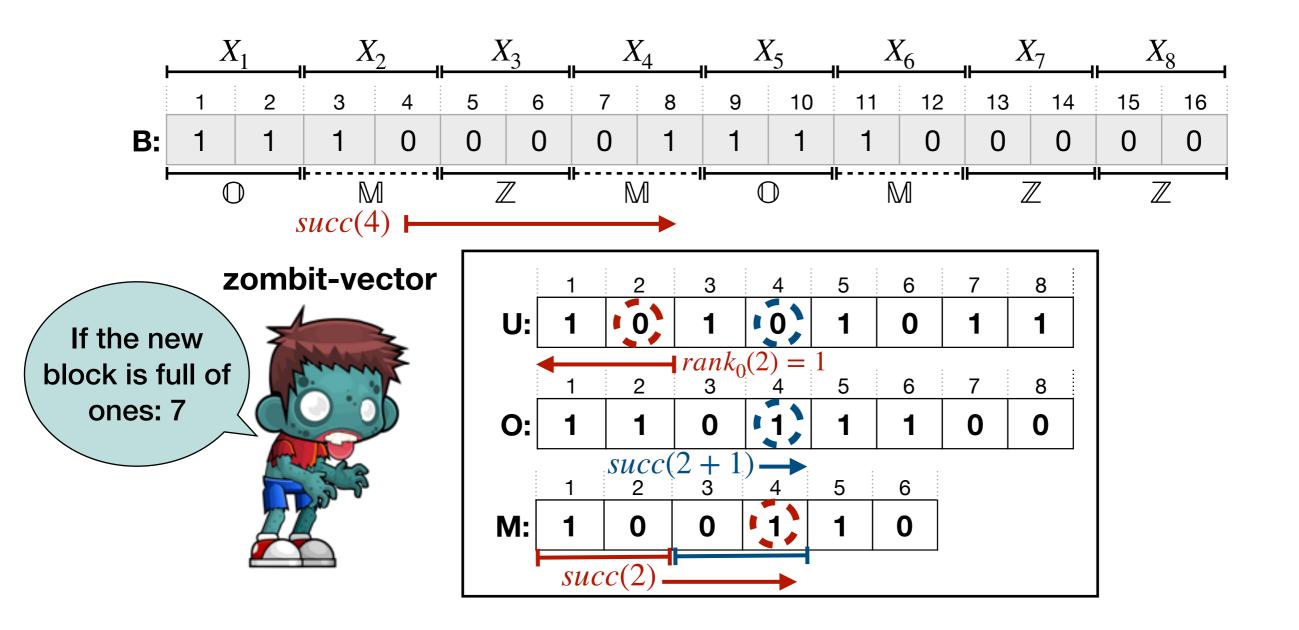














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- *zombit-rec* converges to O(k) bits and solves access, rank, succ and pred in O(log(n/k)) time.



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oz-vector	$O(\log k)$	$k \log \frac{2n}{k} + O(k)$
hybrid	$O(\log n)$	$min(k,m)\lceil \log b \rceil + o(n)$
zombit	<b>O</b> (1)	$O(\sqrt{kn})$
zombit-rec	$O(log \frac{n}{k})$	O(k)



# Outline

### **Introduction**

### **Mackground**

**Zombit-vector** 

### **C** Experimental evaluation

Conclusions

### **D** Future work



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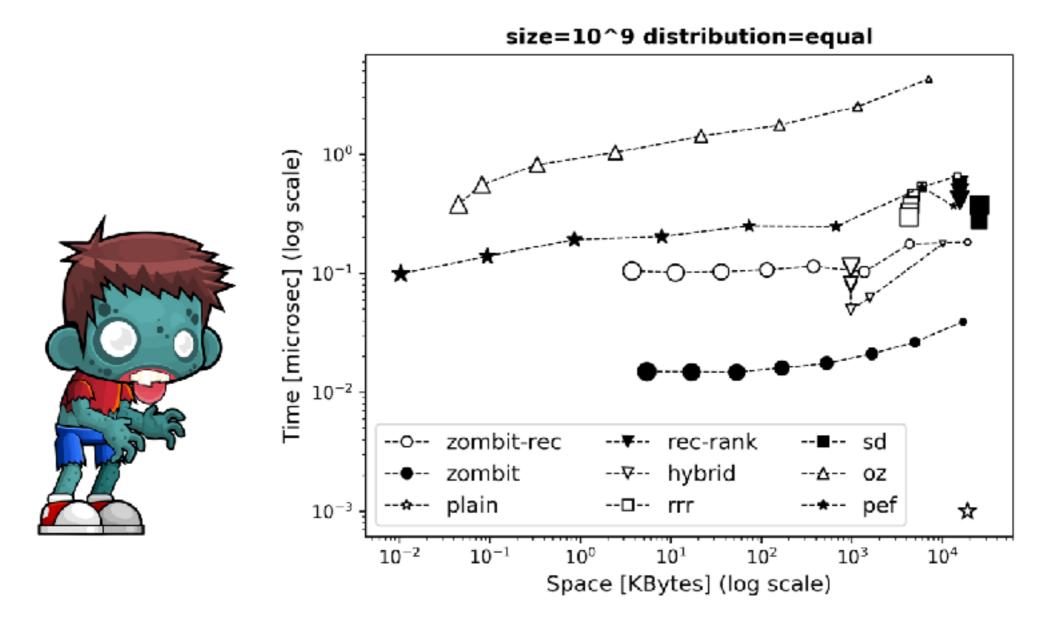
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- Synthetic bitvectors with different sizes 10<sup>9</sup>, 10<sup>8</sup> and 10<sup>7</sup> with different lengths of runs.

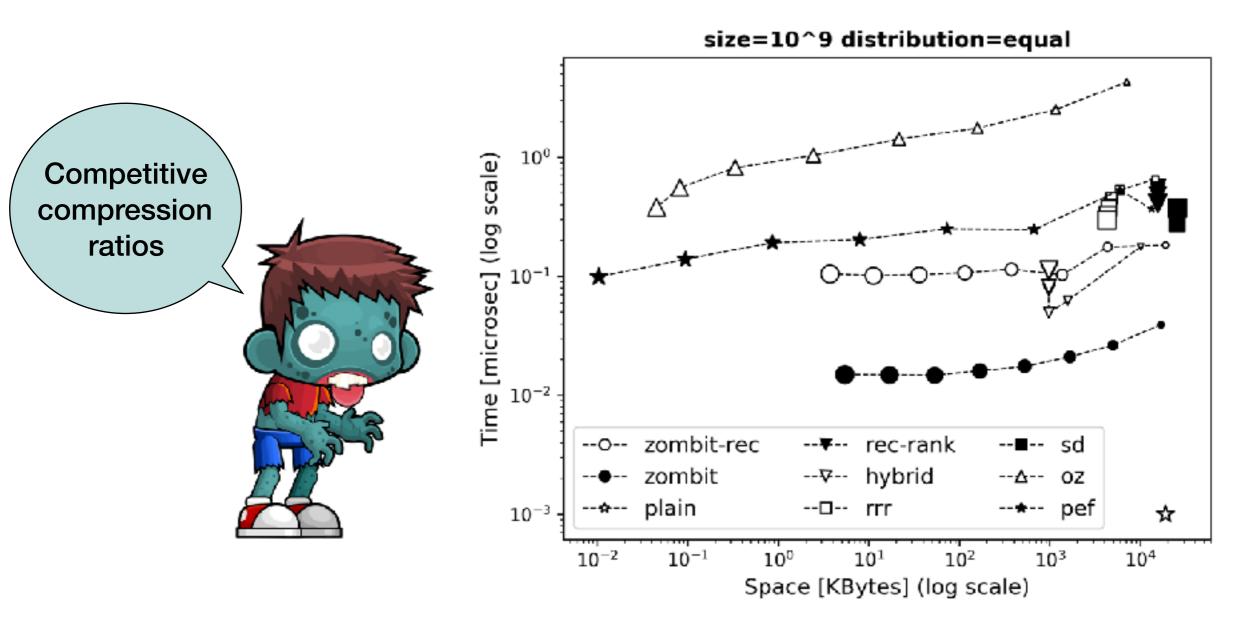


#### Successor operation



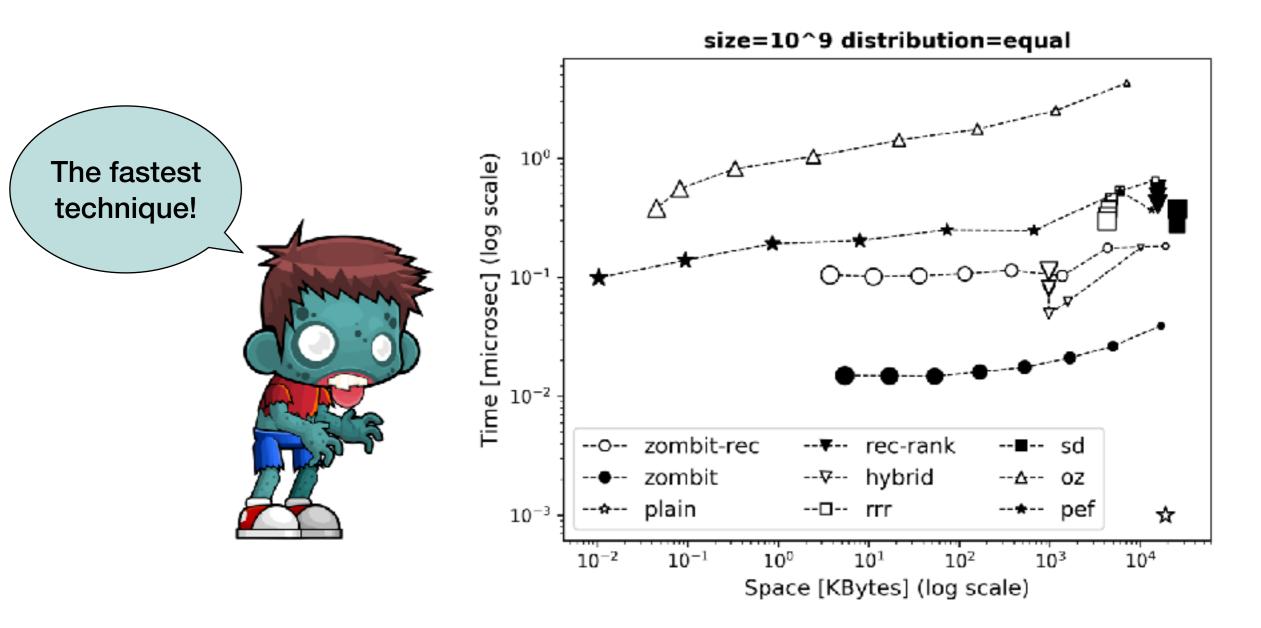


#### Successor operation



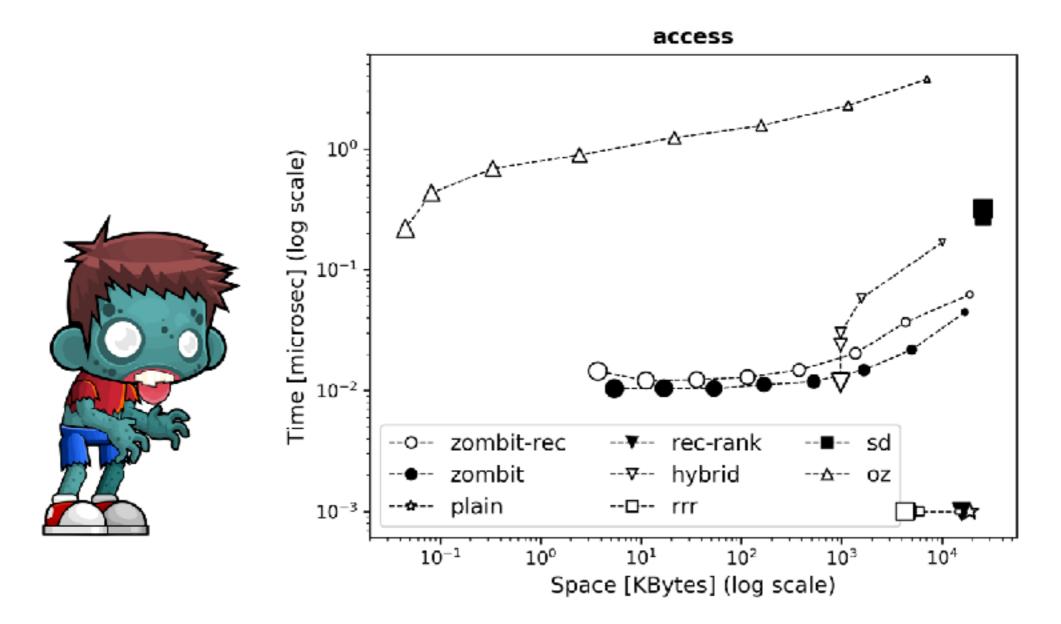


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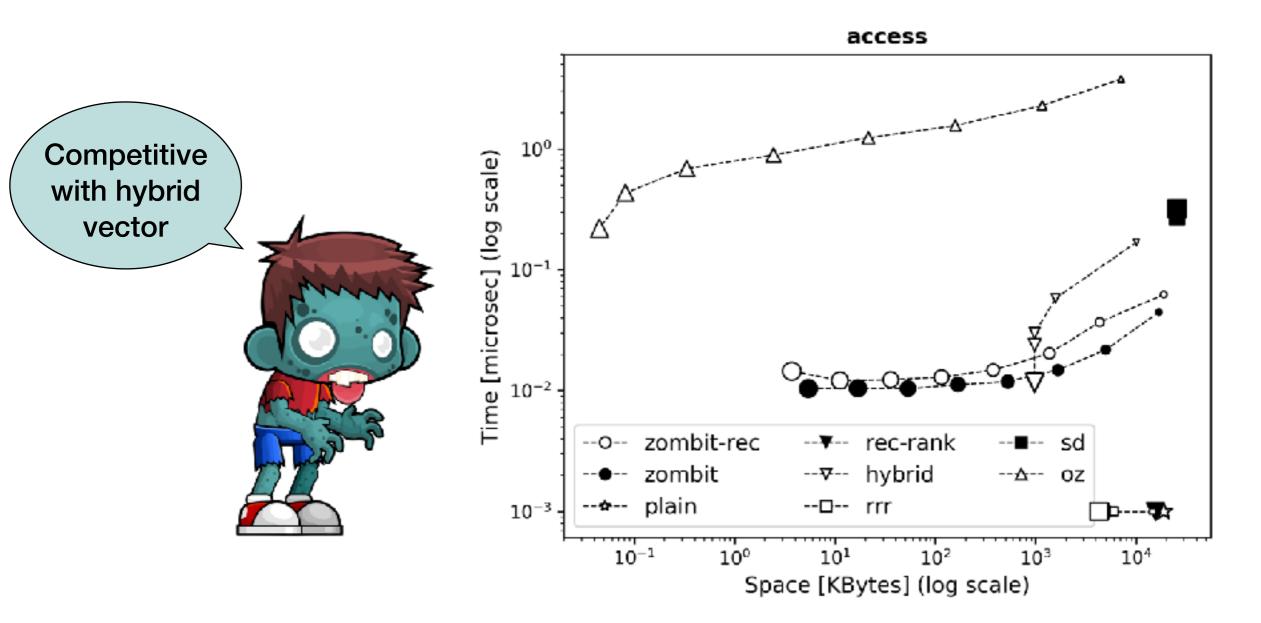


#### Access operation



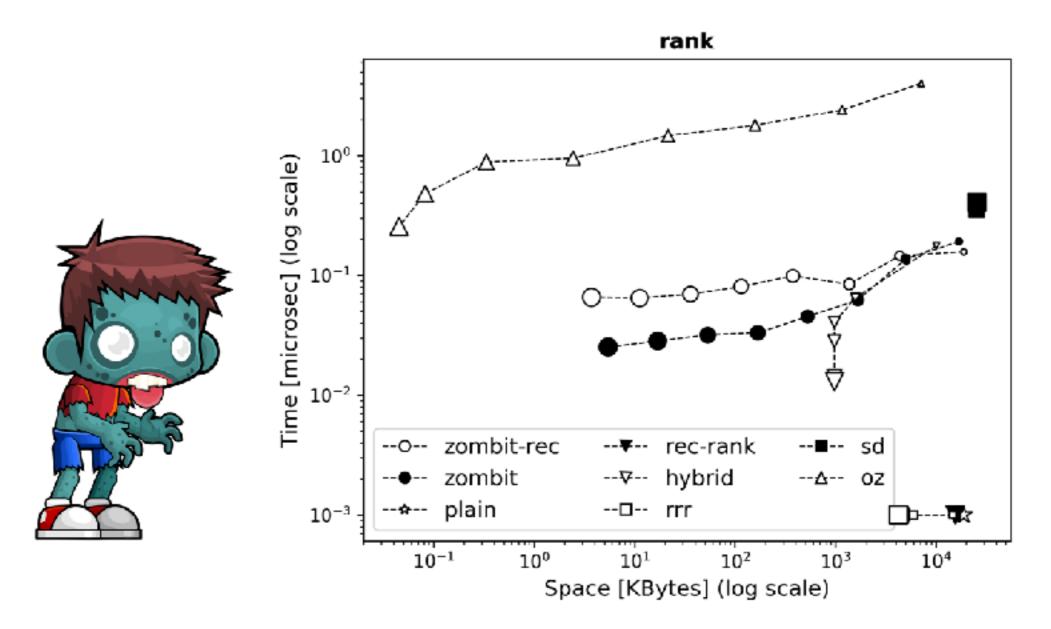






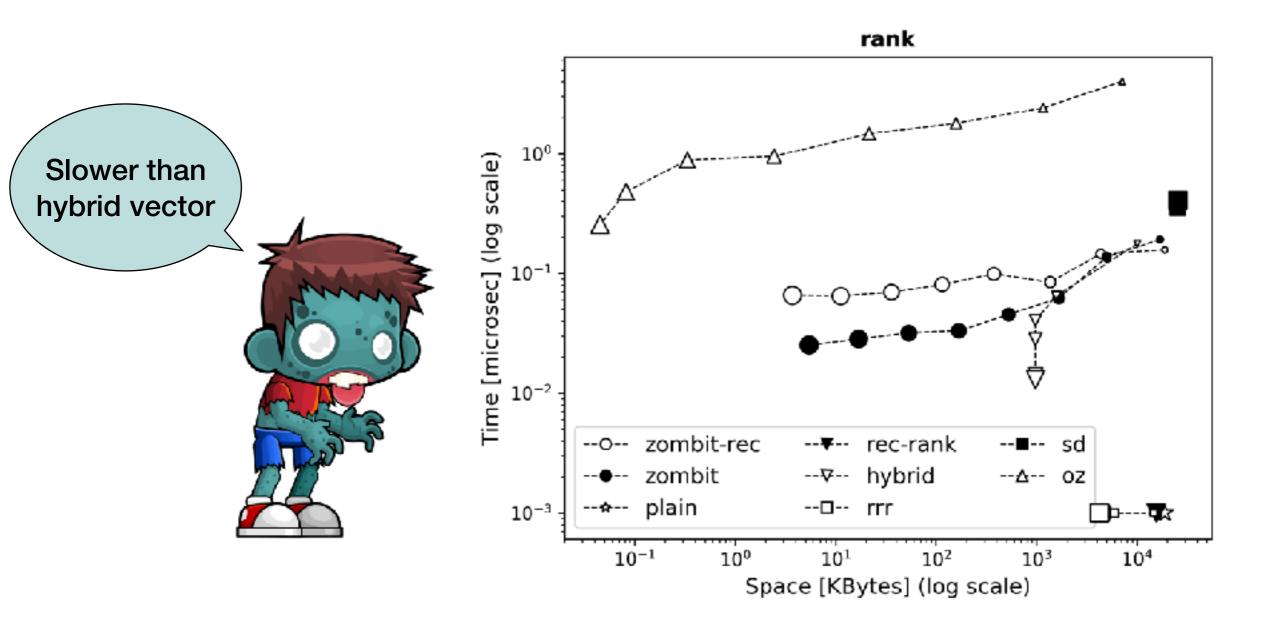


#### Rank operation





#### Rank operation





# Outline

### **Introduction**

- **Mackground**
- **Zombit-vector**
- **Mathematic** Experimental evaluation

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#### Experimental evaluation with **real data**.





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### Future work

- Experimental evaluation with real data.
- Improving the compression ratios of bitvectors with short runs.
- Solving select operation with o(n) bits of extra-space.







# Thanks for your attention!!



# Bitvectors with runs and the successor/predecessor problem



Data Compression Conference

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Wednesday, 25th March 2020, Snowbird, Utah



This work has received funding from the European Union's Horizon 2020 research and innovation programme under the Marie Skłodowska-Curie grant agreement No 690941



