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 - How do you choose your inputs?
 - May be very time-consuming.
 - Very reliable results if done correctly.
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 - Very reliable results if done correctly.
 - Results only hold for a specific machine and for a specific set of inputs.
- Theoretical analysis in a specific model of computation.
 - Gives asymptotic bounds like "this algorithm always runs in time $\mathcal{O}(n^2)$ ".
 - Typically focuses on the worst case.
 - Can give lower bounds like "any comparison-based sorting algorithm needs at least $\Omega(n \log n)$ comparisons in the worst case".

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Example 1

Suppose n numbers from the interval $\{1,\ldots,N\}$ have to be sorted. In this case we usually say that the input length is n instead of e.g. $n\log N$, which would be the number of bits required to encode the input.

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 Calculate running time and storage space etc. on a simplified, idealized model of computation, e.g. Random Access Machine (RAM), Turing Machine (TM), . . .

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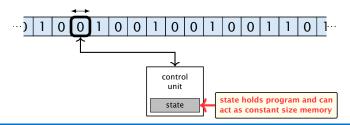
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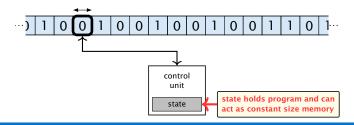
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Version 2. is often easier, but focusing on one type of operation makes it more difficult to obtain meaningful results.

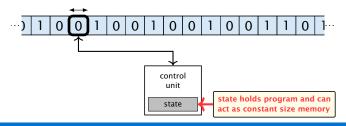
Very simple model of computation.



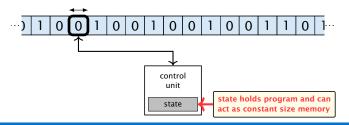
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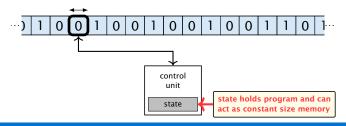


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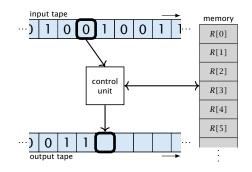


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- Very good model for discussing computability, or polynomial vs. exponential time.
- Some simple problems like recognizing whether input is of the form xx, where x is a string, have quadratic lower bound.
- \Rightarrow Not a good model for developing efficient algorithms.



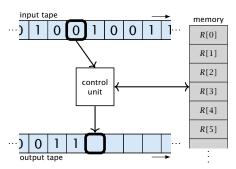
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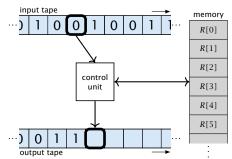


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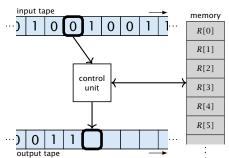
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 - ► R[i] := R[j] + R[k];R[i] := -R[k];

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Bounded word RAM model: cost is uniform but the largest value stored in a register may not exceed 2^w , where usually $w = \log_2 n$.

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more general: probability measure μ

$$C_{\text{avg}}(n) := \sum_{x \in I_n} \mu(x) \cdot C(x)$$

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- amortized complexity: The average cost of data structure operations over a worst case sequence of operations.
- randomized complexity:

 The algorithm may use random bits. Expected running time (over all possible choices of random bits) for a fixed input x. Then take the worst-case over all x with |x| = n.