

no lecture on Thursday this week
next and last lecture on Monday next week

Algorithms: Correctness & Efficiency – G52ACE

Randomised Algorithms: Quicksort (guest lecture, covering for Prof Logan)

Thomas Gärtner

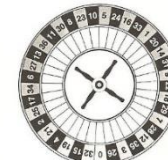
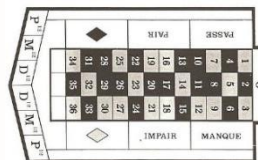
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**University of
Nottingham**

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Quicksort

QuickSort(S)

input: set S of numbers

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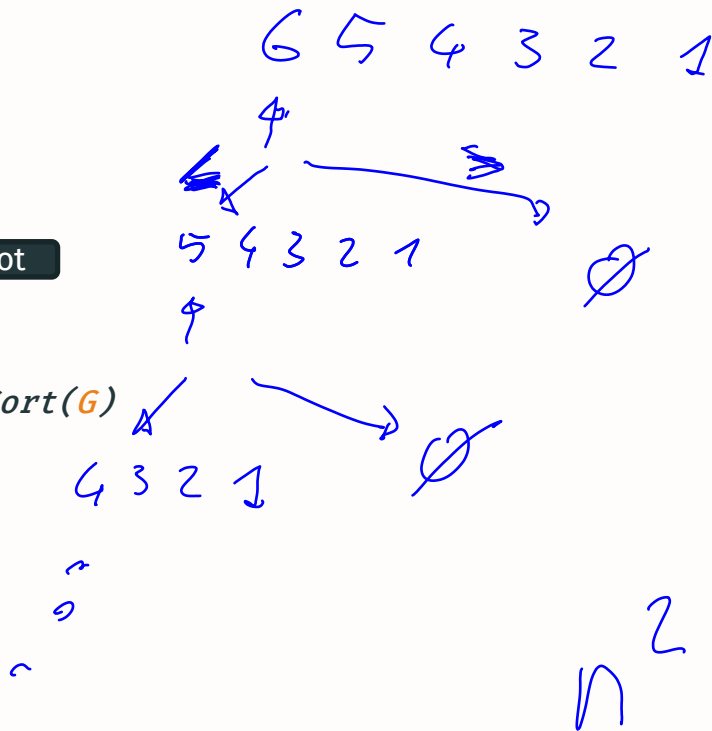
0: if $|S| \leq 1$ then return [e for e in **S**]

1: pick first element **e** from **S**

2: split **S** into subsets: splitter / pivot

- set **G** of elements greater than **e**
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3: return QuickSort(**L**) \oplus [**e**] \oplus QuickSort(**G**)



Lower bound: How many comparisons will we need?

How many comparisons will any comparison-based sorting algorithm need to make?

There are $n!$ different orderings for n elements

Each set of comparisons partitions the orderings into sets of “compatible” orderings

Example $\{a, b, c\}$, $3! = 6$

$a < b$
 $b < c$ |

1 ✓ ✓ $a b c$
X X ✓ $a < b$
X ✓ X $b a c$
X ✓ X $b c a$
X X ✓ $c a b$
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$a < b$ is compatible with 3 orders

$a > b$ is compatible with 3 orders

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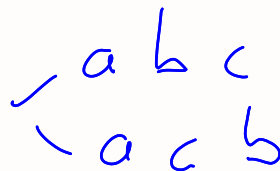
$a < b$ is compatible with 3 orders: $a < b < c, a < c < b, c < a < b$

$a > b$ is compatible with 3 orders: $b < a < c, b < c < a, c < b < a$

$a < b, b < c$ uniquely determine an order

$a < b, a < c$ do not uniquely determine an order

2^k “buckets”



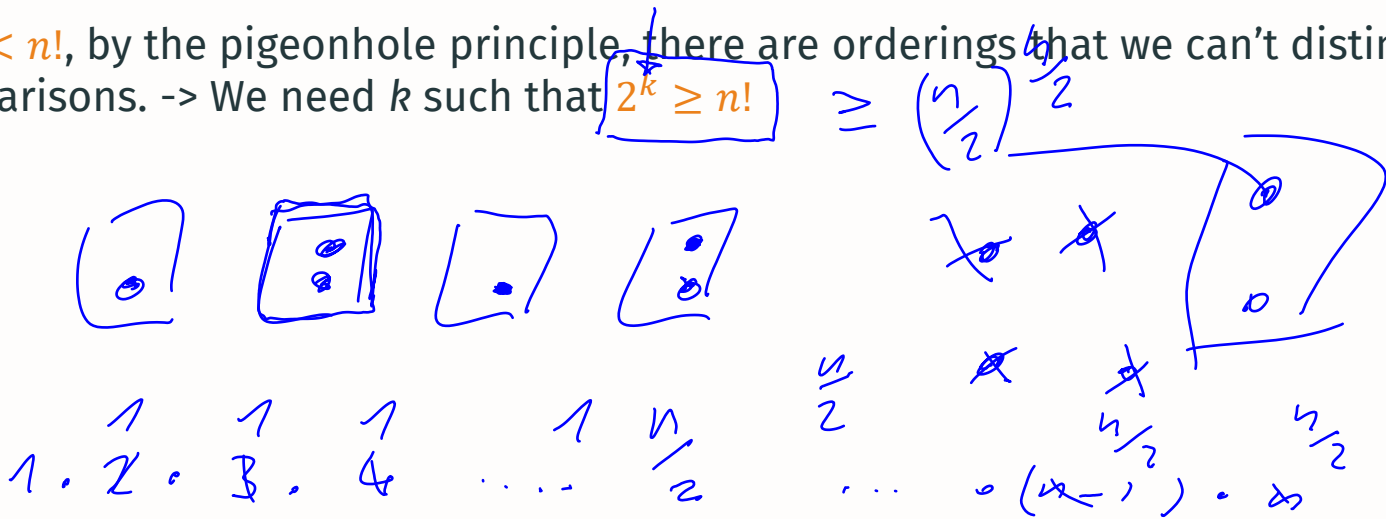
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Therefore $k \geq n/2 \log_2 n/2$

Handwritten derivation showing the steps from the inequality $2^k \geq (n/2)^{n/2}$ to the final complexity $O(n \log n)$. The derivation includes taking the logarithm of both sides and simplifying the resulting expression.

$$2^k \geq \left(\frac{n}{2}\right)^{n/2}$$
$$\log_2 2^k \geq \log_2 \left(\frac{n}{2}\right)^{n/2}$$
$$k \geq \frac{n}{2} \log_2 \frac{n}{2}$$
$$O(n \log n)$$

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Therefore $k \geq n/2 \log_2 n/2$

To obtain a better estimate we can use $n! \geq e (n/e)^n$
which gives $k \geq 1.4 (n - 1) \ln n$

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best case: $n \log n$



worst case: n^2



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average case?

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best case: $n \log n$

(~~average case~~)

expected runtime
for worst input

“Randomised Algorithm”

worst case: n^2

Probability

the **relative frequency** of a probabilistic event

how many times the “experiment” was “successful”

how many times you conducted an “experiment”

in expectation

Probability

set of “successful” outcomes

$$P(\omega \in A) = P(A)$$

random event

the **relative frequency** of a probabilistic event

how many times the “experiment” was “successful”

how many times you conducted an “experiment”

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Probability

$$P(\omega \in A) = P(A)$$

the **relative frequency** of a probabilistic event

how many times the “experiment” was “successful”

how many times you conducted an “experiment”

in expectation

if there is no reason that one event would happen more often than another

$$P(\omega) = \frac{1}{\text{how many different events can happen}} = \frac{1}{|\Omega|}$$

Probability

sample space Ω , possible events 2^Ω , probability $P: 2^\Omega \rightarrow \mathbb{R}^+$

$$P(\emptyset) = 0, P(\Omega) = 1, \forall A, B \subseteq \Omega: (A \cap B = \emptyset) \Rightarrow P(A \cup B) = P(A) + P(B)$$

the **relative frequency** of a probabilistic event

Probability—fair dice—sampling “uniformly at random”



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We call the **relative frequency** of a probabilistic event,

it's **probability** e.g. $P(6) = P(5) = \dots = \frac{1}{6}; P(\{5,6\}) = \frac{1}{3}$

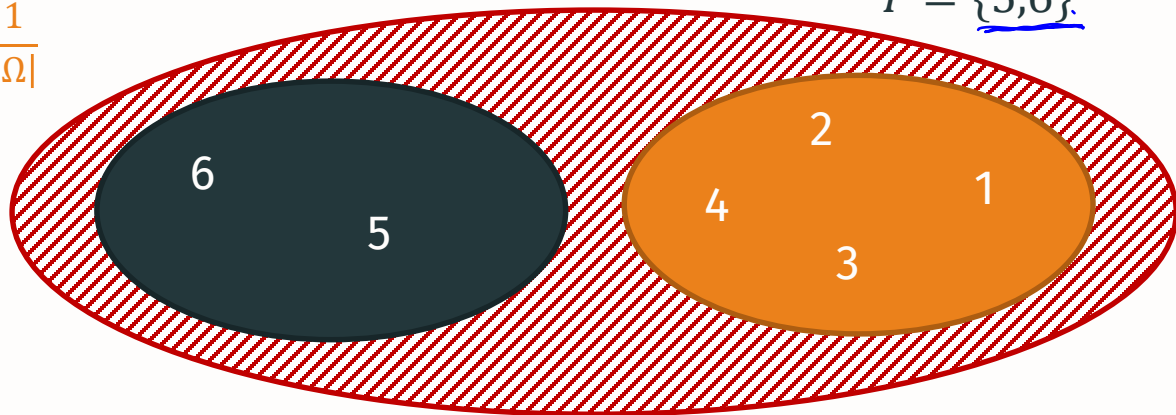
For fair dice (=uniform sampling)

$$\forall \omega \in \Omega: P(\omega) = \frac{1}{|\Omega|}$$

$$P(T) = \frac{|T|}{|\Omega|}$$

$$\Omega = \llbracket 6 \rrbracket$$

$$T = \{5,6\}$$



Random Variables & Expectation—linearity

Random variables assign values to outcomes of random experiments:

$$X: \Omega \rightarrow \mathbb{R}$$

The expected value of is a random variable's **probability-weighted** average:

$$\mathbb{E}(X) = \sum_{\omega \in \Omega} X(\omega) \cdot P(\omega)$$

Random Variables & Expectation

Random variables assign values to outcomes of random experiments:

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Examples

- number of eyes on a dice
- payoff after a game was won/lost
- number of times an event occurs in repeated experiments
- runtime of an algorithm

The **expected value** of is a random variable's **probability-weighted** average:

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Examples

- average number of eyes thrown with a dice
- average payoff
- average number of times an event occurs in repeated experiments
- average runtime of an algorithm
- worst-case expected runtime of an algorithm

over
input
and
algorithm

over input over alg

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$$X = \sum_{i=1}^n \sum_{j=i+1}^n X_{ij}$$

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$$\mathbb{E}[X] = \sum_{i=1}^n \sum_{j=i+1}^n \mathbb{E}[X_{ij}]$$

$$\mathbb{E} \left[\sum_i \sum_j X_{ij} \right] = \sum_i \sum_j \left(\mathbb{E} [X_{ij}] \right)$$

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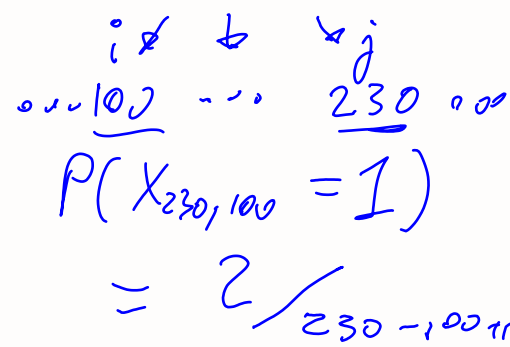
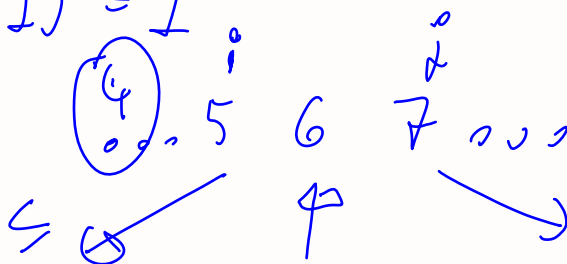
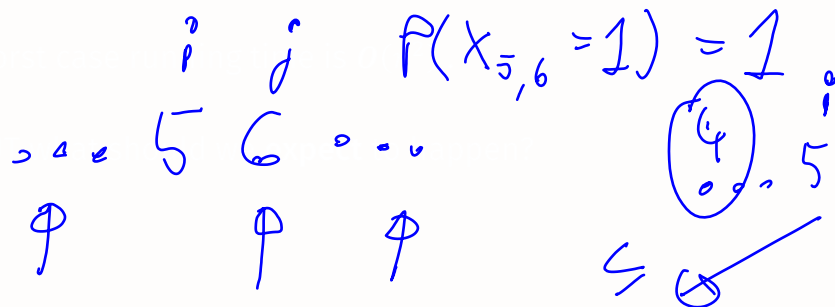
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now

$$P(X_{ij} = 1) = ?$$



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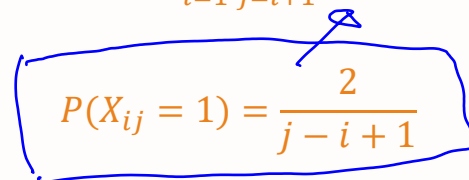
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$$P(X_{ij} = 1) = \frac{2}{j - i + 1}$$

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therefore

$$\mathbb{E}[X] \leq \sum_{i=1}^n 2 \cdot \left(\frac{1}{2} + \frac{1}{3} + \cdots + \frac{1}{n-i+1} \right)$$

$\approx \dots \frac{1}{n}$

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Excursion: Harmonic numbers

$$H_n = 1 + \frac{1}{2} + \frac{1}{3} + \cdots + \frac{1}{n} = \sum_{i=1}^n 1/i$$

$$n = 2^k$$

$$H_{2^k} = 1 + \frac{1}{2} + \frac{1}{3} + \cdots + \frac{1}{2^k} = \sum_{i=1}^{2^k} 1/i$$

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$$\begin{aligned} & \frac{1}{2} \\ & + \frac{\frac{1}{2} + \frac{1}{3}}{2} \\ & + \frac{\frac{1}{4} + \frac{1}{5} + \frac{1}{6} + \frac{1}{7}}{2} \\ & \vdots \\ & + \frac{1}{2^{k-1}} + \frac{1}{2^{k-1}+1} + \cdots + \frac{1}{2^k-1} \end{aligned}$$

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Handwritten notes and inequalities illustrating the growth of harmonic numbers:

Left side (rows):

- $\frac{1}{2} \approx \frac{1}{4} + \frac{1}{4}$
- $\frac{1}{2} = \frac{1}{8} + \frac{1}{8} + \frac{1}{8} + \frac{1}{8}$
- $\frac{1}{2} = \frac{1}{2^k} + \frac{1}{2^k} + \cdots + \frac{1}{2^k}$

Central boxed inequalities (rows):

- ≤ 1
- $\leq \frac{1}{2} + \frac{1}{3}$
- $\leq \frac{1}{4} + \frac{1}{5} + \frac{1}{6} + \frac{1}{7}$
- \dots
- $\leq \frac{1}{2^{k-1}} + \frac{1}{2^{k-1}+1} + \cdots + \frac{1}{2^k-1}$

Right side (rows):

- ≤ 1
- $\leq \frac{1}{2} + \frac{1}{2} = 1$
- $\leq \frac{1}{4} + \frac{1}{4} + \frac{1}{4} + \frac{1}{4} = 1$
- $\leq \frac{1}{2^{k-1}} + \frac{1}{2^{k-1}} + \cdots + \frac{1}{2^{k-1}} = 1$

Handwritten note: } 6 rows

Excursion: Harmonic numbers

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$$H_{2^k} = 1 + \frac{1}{2} + \frac{1}{3} + \dots + \boxed{\frac{1}{2^k}} = \sum_{i=1}^{2^k} 1/i$$

$$n = 2^k \Rightarrow \frac{1}{2} \log_2 n \leq H_n \leq 1 + \log_2 n$$

$$k = \log_2 n \quad \& \quad 2^k = n$$

$$\begin{aligned} & \frac{1}{2} \\ & \frac{1}{4} + \frac{1}{4} \\ & \frac{1}{8} + \frac{1}{8} + \frac{1}{8} + \frac{1}{8} \end{aligned}$$

$$\begin{aligned} & \leq 1 \\ & \leq \frac{1}{2} + \frac{1}{3} \\ & \leq \frac{1}{4} + \frac{1}{5} + \frac{1}{6} + \frac{1}{7} \\ & \dots \end{aligned}$$

$$\frac{1}{2^k} + \frac{1}{2^k} + \dots + \frac{1}{2^k}$$

$$\leq \frac{1}{2^{k-1}} + \frac{1}{2^{k-1}+1} + \dots + \frac{1}{2^k-1}$$

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$$\leq \frac{1}{2^{k-1}} + \frac{1}{2^{k-1}} + \dots + \frac{1}{2^{k-1}}$$

$$\frac{k}{2} \leq \underline{H_{2^k} - 2^{-k}} \leq k$$

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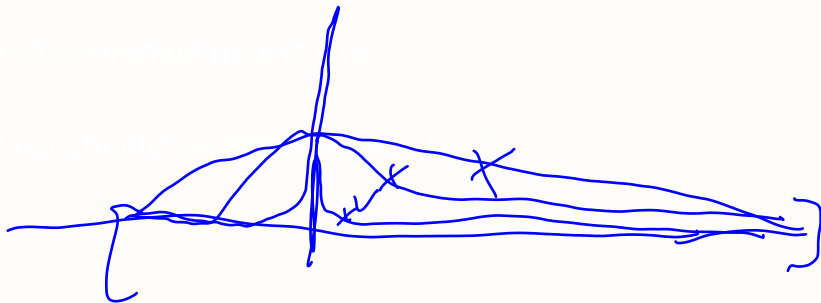
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Moore's Law

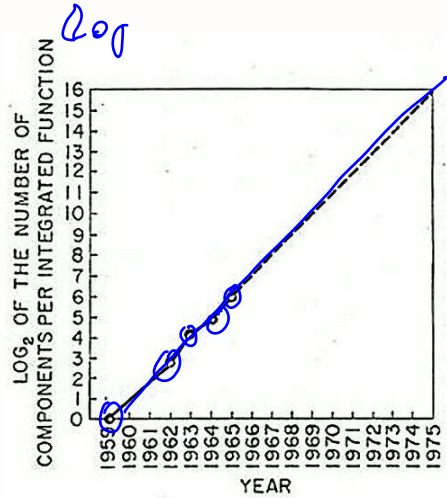
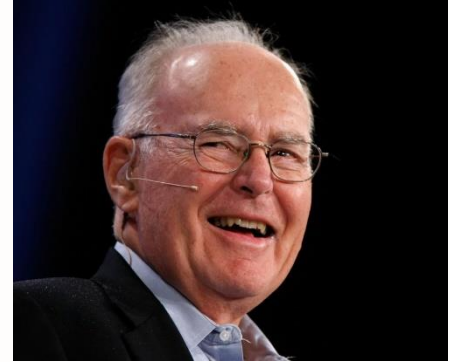


Fig. 2 Number of components per integrated function for minimum cost per component extrapolated vs time.



Gordon Moore, 1965

"Cramming more components onto integrated circuits"

Moore's Law

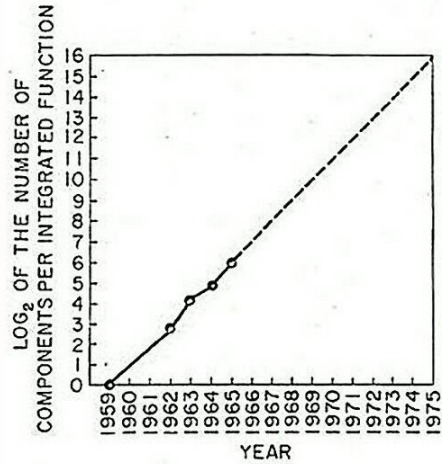
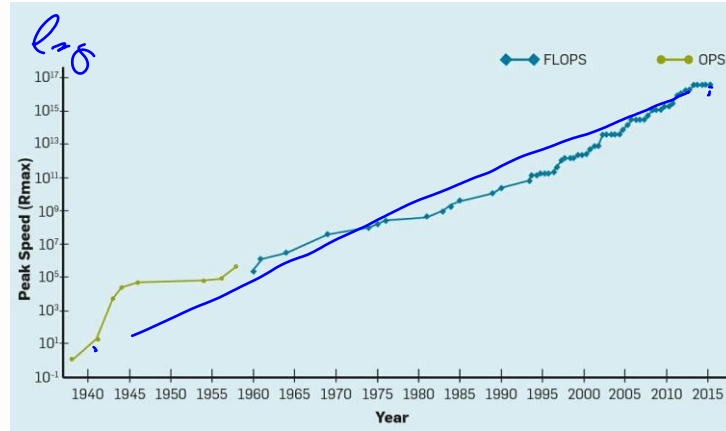


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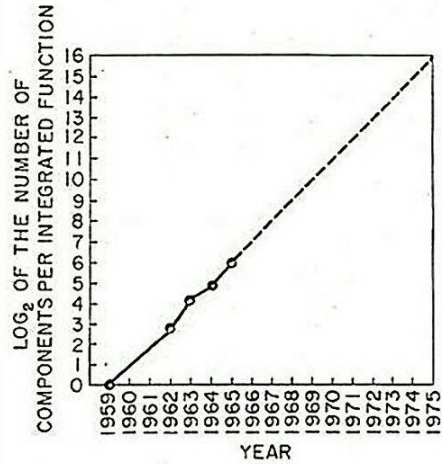
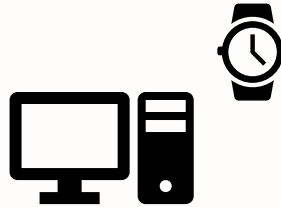
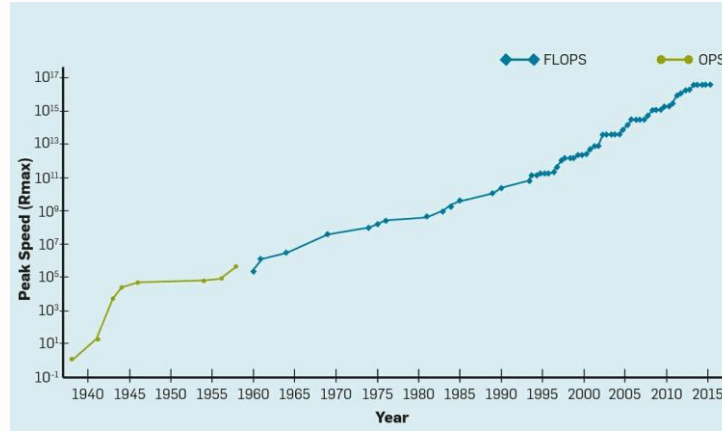
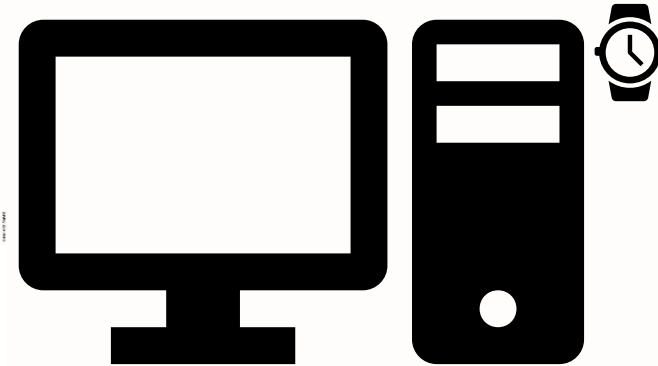


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THE SHEER SCALE OF GROWTH IN RECENT YEARS



Moore's Law

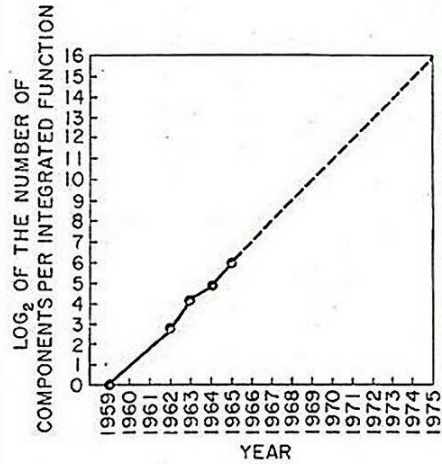
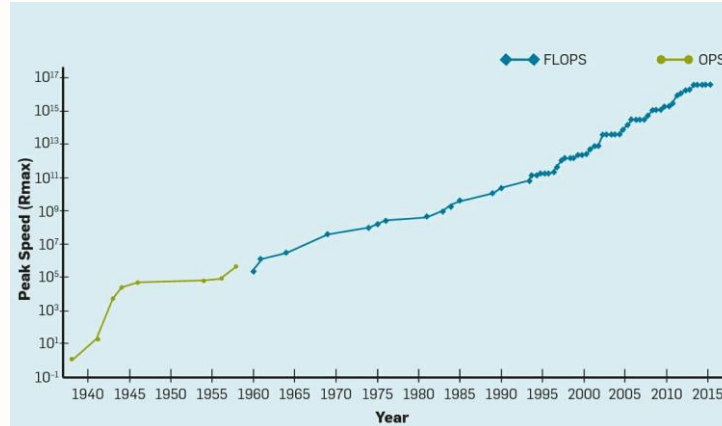
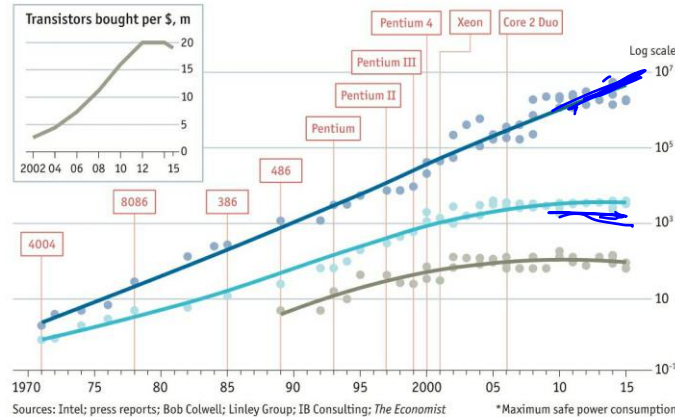


Fig. 2 Number of components per integrated function for minimum cost per component extrapolated vs time.



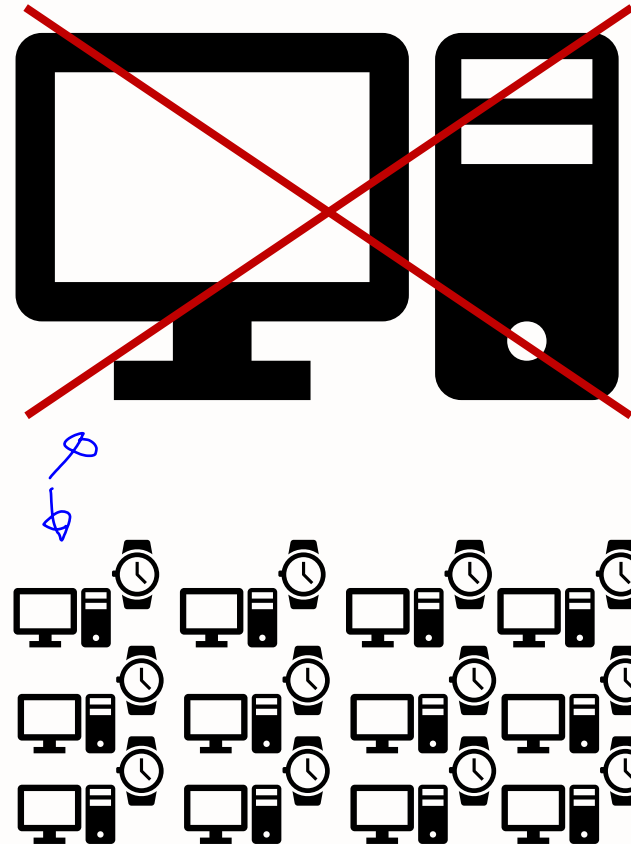
Stuttering

● Transistors per chip, '000 ● Clock speed (max), MHz ● Thermal design power*, w □ Chip introduction dates, selected



Sources: Intel; press reports; Bob Colwell; Linley Group; IB Consulting; The Economist

*Maximum safe power consumption



Moore's Law vs Nick's Class



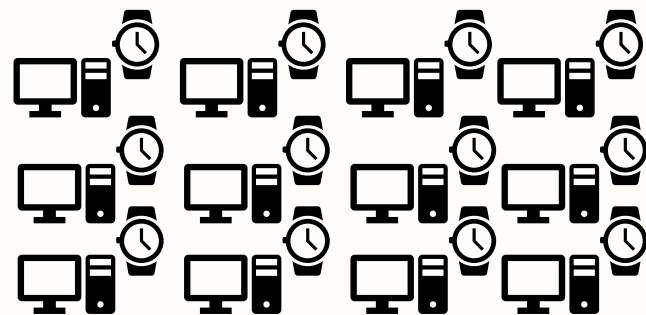
[Nicholas John Pippenger](#)



[Stephen Cook](#)

The complexity class NC contains algorithms that

- need a **polylogarithmic** number of steps (in the worst case)



Moore's Law vs Nick's Class



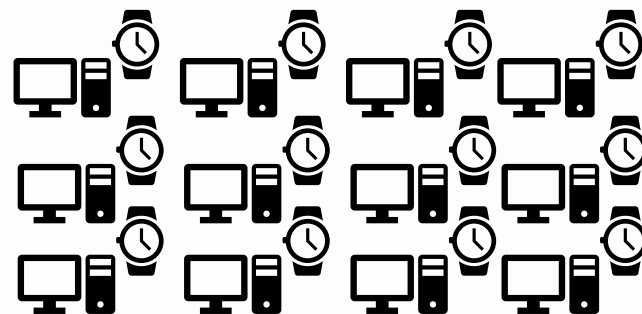
[Nicholas John Pippenger](#)



[Stephen Cook](#)

The complexity class **NC** contains algorithms that

- are correct
- use a polynomial number of processors
- need a **polylogarithmic** number of steps (in the worst case)



PRAM – Parallel Random Access Machine

In each clock tick, each processor can read or write one memory location, or do some basic operation.

PRAM machines can have concurrent/exclusive reads and writes (CRCW, CREW, ERCW!, EREW).

Expected Algorithm Runtime: Quicksort (->)

QuickSort(S)

input: set S of numbers

output: sorted list of elements from S

0: if $|S| \leq 1$ then return $[e \text{ for } e \text{ in } S]$

1: pick random element e from S (uniformly)

2: split S into subsets:

- set G of elements greater than e
- set L of elements less than e

3: return $\text{QuickSort}(L) \oplus [e] \oplus \text{QuickSort}(G)$

Parallel Quicksort (CRCW model)

QuickSort(S)

input: array **s** of numbers $s = (s_1, s_2, \dots, s_n)$ / processors P_1, P_2, \dots, P_n

ensure: array **s** is sorted

0: check in parallel if $s_1 \leq s_2 \leq \dots \leq s_n$ then return

1: pick random element **e** from **S** (uniformly)

2: in parallel for all **i**: if $s_i \leq e$ then $b_i \leftarrow 0$ else $b_i \leftarrow 1$

3: compute in parallel prefix and postfix sums

$$r_i \leftarrow \sum_{j \leq i} b_j, o_i \leftarrow \sum_{j \geq i} b_j \quad (\forall i \in \llbracket n \rrbracket)$$

4: if not $n/4 \leq r_n \leq 3n/4$ then goto 1

5: in parallel for all **i**: if $b_i = 0$ then $t_{r_i} \leftarrow s_i$ else $t_{o_i} \leftarrow s_i$

6: return $\text{QuickSort}(t_{1:r_n}) \oplus \text{QuickSort}(t_{r_n+1:n})$

Time
 $\log n$

n processor

Parallel Quicksort (CRCW model)

QuickSort(S)

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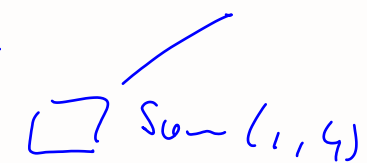
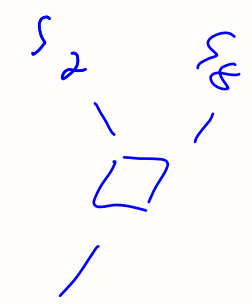
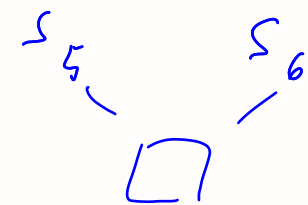
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each parallel instruction can be performed in time $\log n$

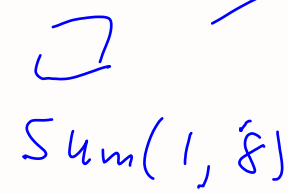
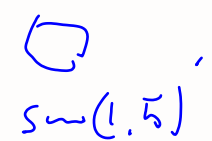
As before, we expect to need $2 \ln n$ recursions

So the overall runtime is $\ln^2 n$ using n processors



Proof

Summation



Questions?