



COMP110: Principles of Computing  
**5: Complexity**



# Research Journal



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- ▶ Use any presentation software to prepare slides (e.g. Beamer, PowerPoint, Google Slides)

# Research journal presentations — Why?

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- ▶ A chance for informal feedback

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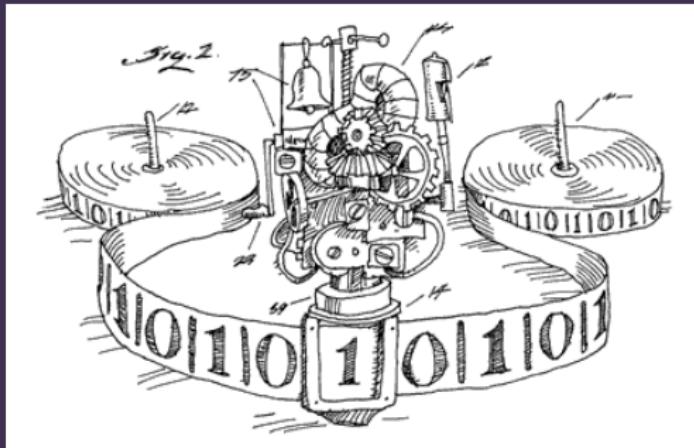
# Research journal

- ▶ You should also be working on the journal itself!
- ▶ Peer review (draft needed) in **week 8**
- ▶ This week's worksheet is the last one for a while, to give you time to work on your research journal

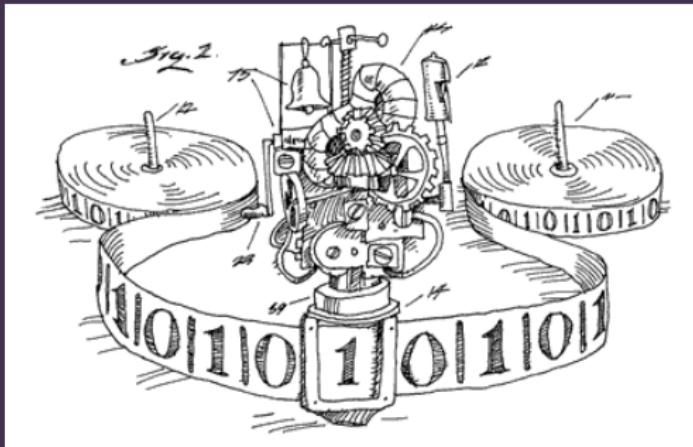
# Turing machines



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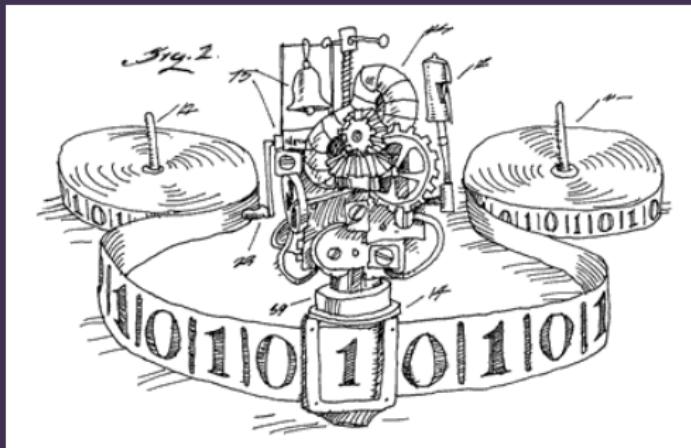


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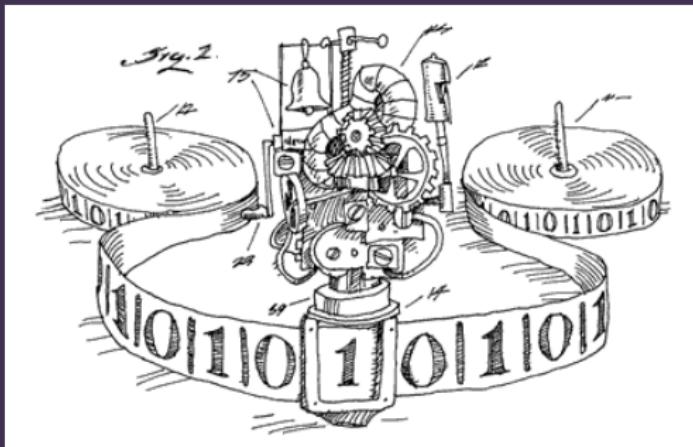
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- ▶ Introduced in 1936 by Alan Turing
- ▶ Theoretical model of a “computer”
  - ▶ I.e. a machine that carries out computations (calculations)

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  - ▶ Where to move the tape head: one space to the left, or one space to the right

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- ▶ A machine, language or system is **Turing complete** if it can simulate a Turing machine
- ▶ (In practice, nothing can simulate an infinite tape, so we just assume a large enough tape)

# Examples of Turing complete systems

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  - ▶ NB:  $A$  may be **infinite**
- ▶ A function  $f : A \rightarrow B$  is **computable** if there exists a Turing machine which computes  $f$ 
  - ▶ I.e. given an encoding of  $a \in A$  as input, the Turing machine outputs an encoding of  $f(a)$

# An uncomputable function

The **halting problem**

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- ▶  $A =$  the set of all Turing machines (encoded as transition tables)

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- ▶ There is **no** Turing machine that computes  $f$
- ▶  $f$  is **uncomputable**

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- ▶ Church-Turing tells us that Turing machines are as powerful as any other computer
- ▶ Therefore if a function is uncomputable, there is **no conceivable machine** that can compute it

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- ▶ Write a software tool that, given a C# program, predicts whether that program can go into an infinite loop
- ▶ Your tool must work for **all** C# programs, considering **all** possible inputs to the program
- ▶ This task is impossible!

# Search



# Search

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- ▶ We have a list of names, each with some data associated
- ▶ We want to find one of them

# Linear search

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procedure FIND(name, list)
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procedure FIND(name, list)
    for each item in list do
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procedure FIND(name, list)
    for each item in list do
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procedure FIND(name, list)
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# How long does it take?

Socrative room code: FALCOMPED

- ▶ Suppose there are 25 items in the list

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- ▶ Suppose there are 25 items in the list
- ▶ In the **best case**, how many items do we need to visit before finding the one we want?
- ▶ How about in the **worst case**?

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- ▶ How about if there are 50 items?

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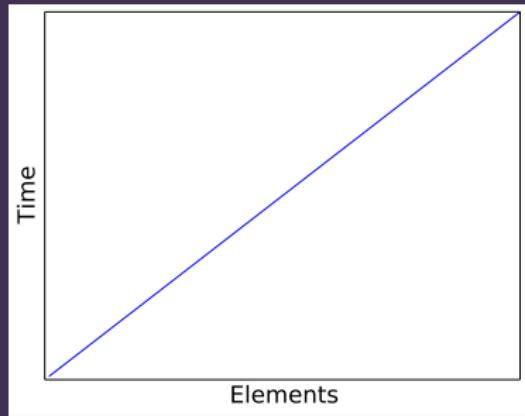
- ▶ If there are 25 items in the list, the **worst case** number of items visited is 25
- ▶ How about if there are 50 items?
- ▶ How about 100 items?

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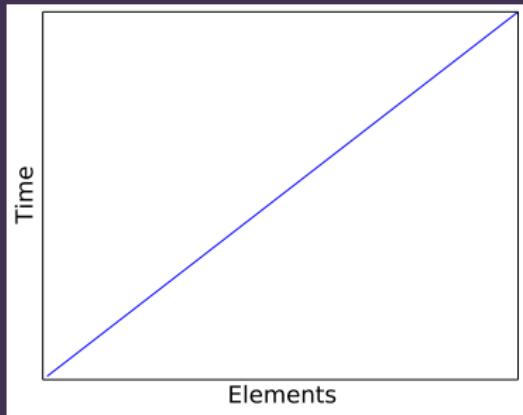
- ▶ If there are 25 items in the list, the **worst case** number of items visited is 25
- ▶ How about if there are 50 items?
- ▶ How about 100 items?
- ▶ If the number of items **doubles**, what happens to the amount of time the search takes?

# Linear time



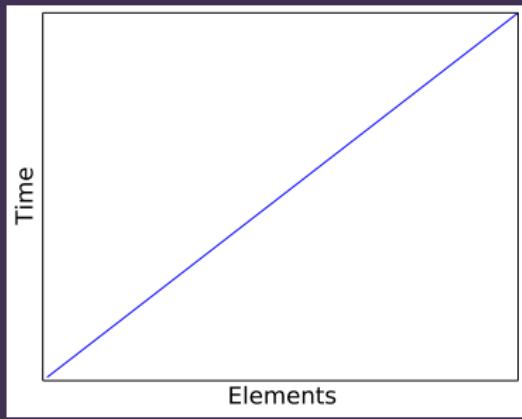
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- ▶ Linear search is said to have **linear time complexity**
- ▶ Also written as  $O(n)$  **time complexity**

# Searching a sorted list

- ▶ If the list is **sorted** in alphabetical order, we can do better than linear...

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# Find “Lopez, Jeffrey”

Anderson, Martha  
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Brown, Janet  
Clark, Stephanie  
Collins, Jane  
Cox, Shirley  
Davis, Marilyn  
Diaz, Harold  
Gonzalez, Adam  
Henderson, Lawrence  
Hughes, Aaron  
Kelly, Philip  
→ Lewis, Rose  
Lopez, Jeffrey  
Miller, Jeremy  
Parker, Debra  
Perez, Diana  
Russell, Mildred  
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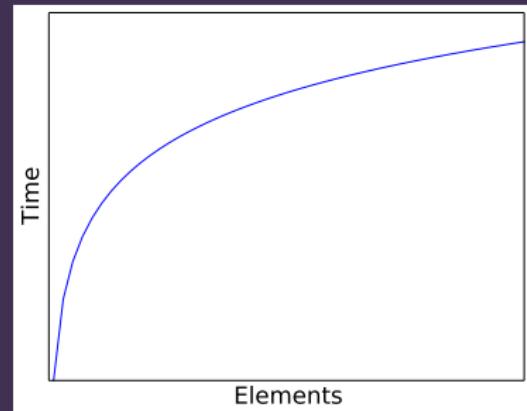
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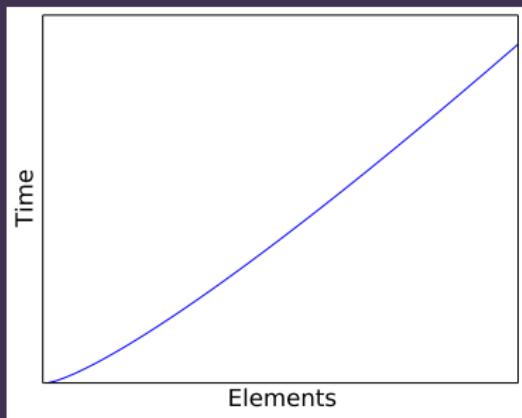
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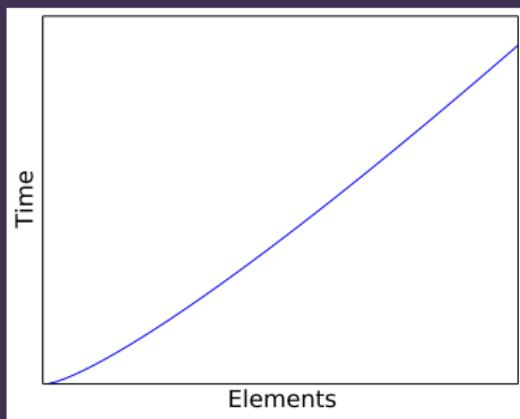
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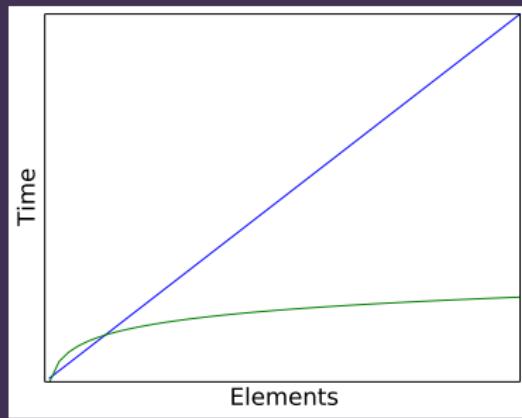
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- ▶ **Copying** (half of) a list is **linear**  $O(n)$
- ▶ The actual running time would be  $O(n \log n)$
- ▶ Use **pointers** into the list instead of copying

# Binary search vs linear search



- ▶ So binary search is better than linear search... right?

# Hashing

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- ▶ Use these numbers to assign each element to a “bin” where it can be found

:	:
112	Ward, Jessica
113	Baker, Theresa
114	Collins, Jane
115	—
116	—
117	Hughes, Aaron
118	—
119	—
120	—
121	—
122	Brown, Janet
123	—
124	—
125	Gonzalez, Adam Lewis, Rose
126	—
127	—
128	—
129	—
130	—
131	—
132	Young, Frank
:	:

# Hash look-up

98	Diaz, Harold
99	Parker, Debra
	Perez, Diana
	White, Amanda
112	Ward, Jessica
113	Baker, Theresa
114	Collins, Jane
117	Hughes, Aaron
122	Brown, Janet
125	Gonzalez, Adam
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132	Young, Frank
135	Kelly, Philip
138	Cox, Shirley
142	Clark, Stephanie
144	Scott, Michelle
145	Miller, Jeremy
147	Davis, Marilyn
149	Lopez, Jeffrey
151	Anderson, Martha
158	Williams, Billy
162	Sanders, Phillip
171	Russell, Mildred
175	Stewart, Howard
183	Henderson, Lawrence

“Lopez, Jeffrey”

# Hash look-up

98	Diaz, Harold
99	Parker, Debra
	Perez, Diana
	White, Amanda
112	Ward, Jessica
113	Baker, Theresa
114	Collins, Jane
117	Hughes, Aaron
122	Brown, Janet
125	Gonzalez, Adam
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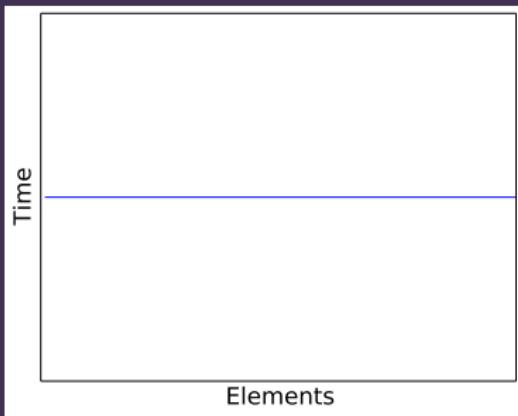
$$12 + 15 + 16 + 5 + 26 + 10 + 5 + 6 + 6 + \\ 18 + 5 + 25 = 149$$

# How long does it take?



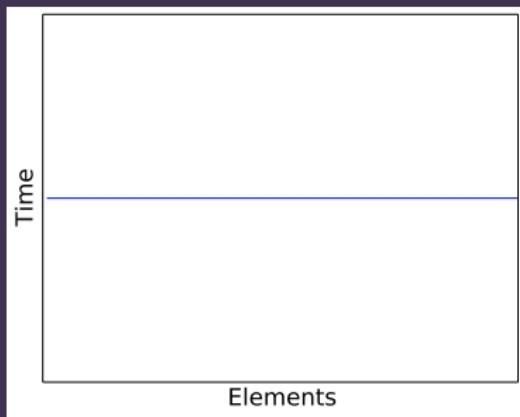
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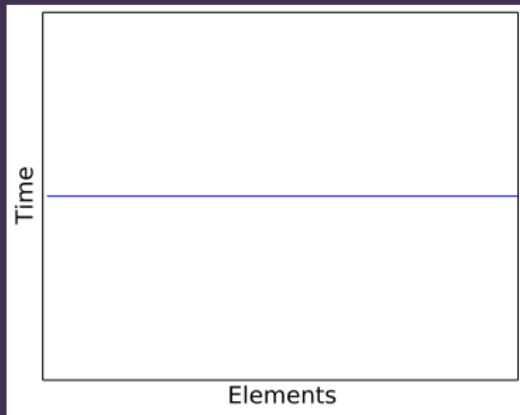
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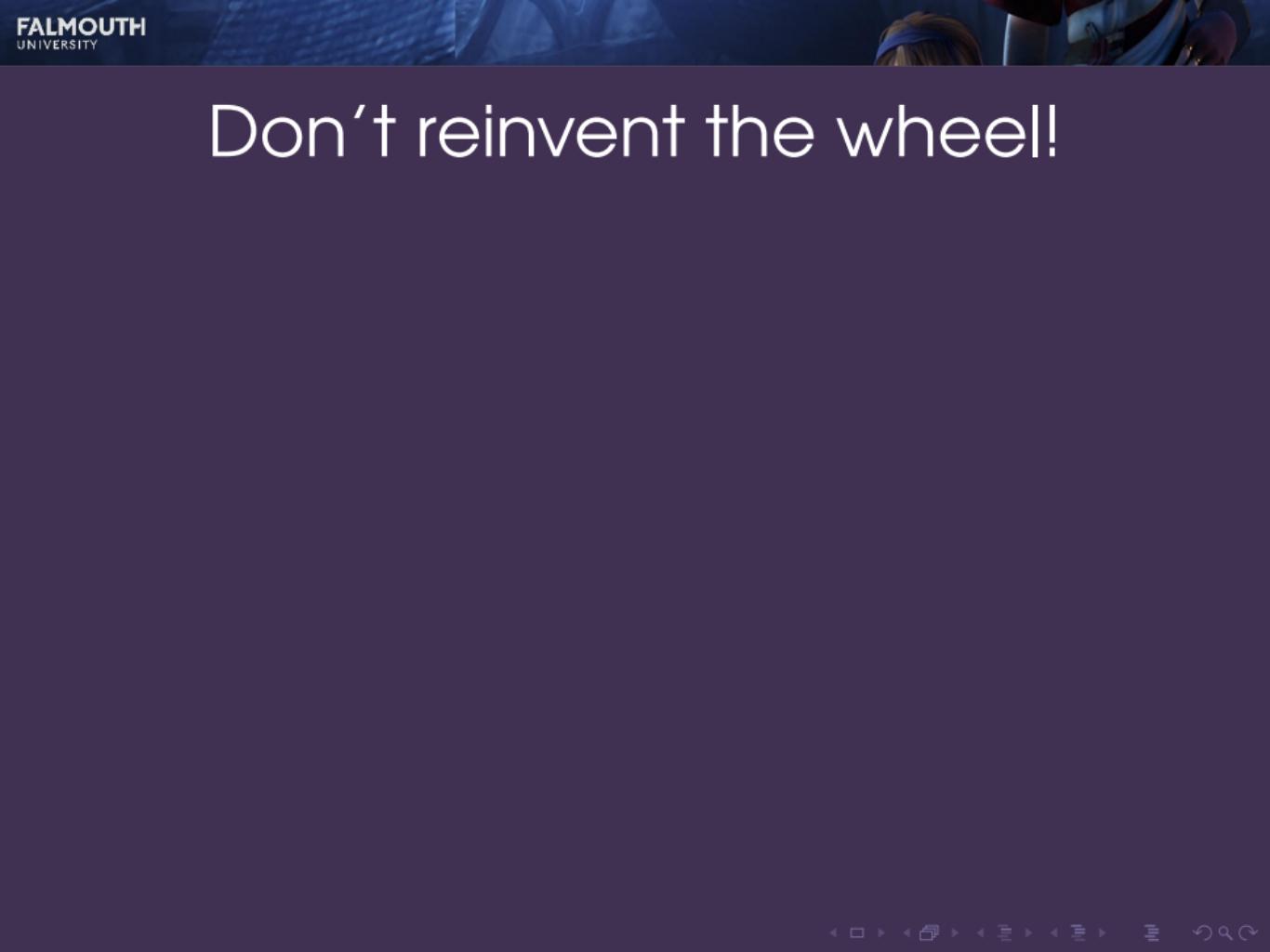


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- ▶ When there are collisions, need to fall back on something like linear or binary search within each bin



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# More on complexity



# Common complexity classes

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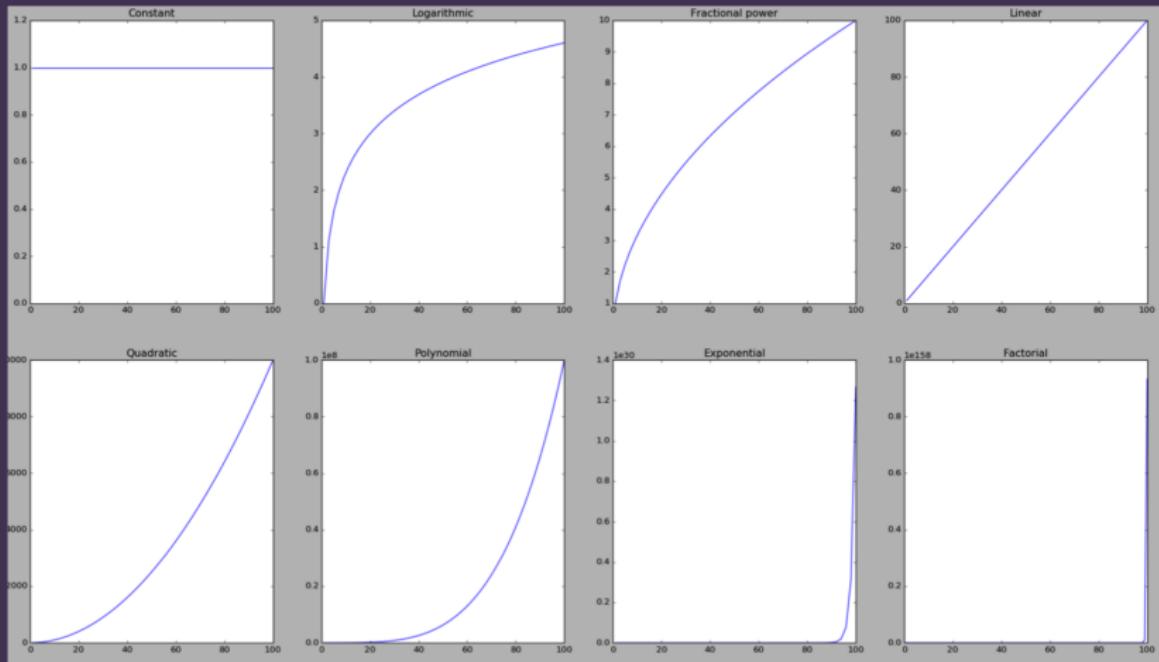
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- ▶ Multiply **compound** algorithms
  - ▶ If an algorithm does  $n$  “things” and each “thing” is  $O(n)$ , then the overall algorithm is  $O(n^2)$

# Quadratic complexity

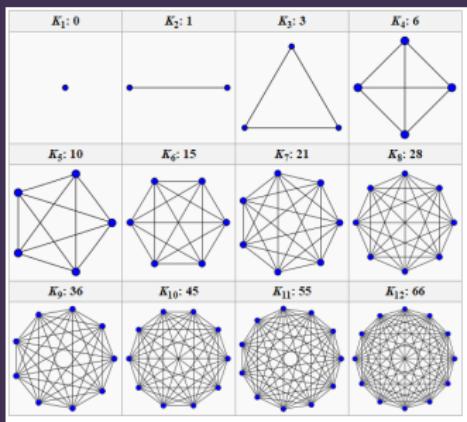
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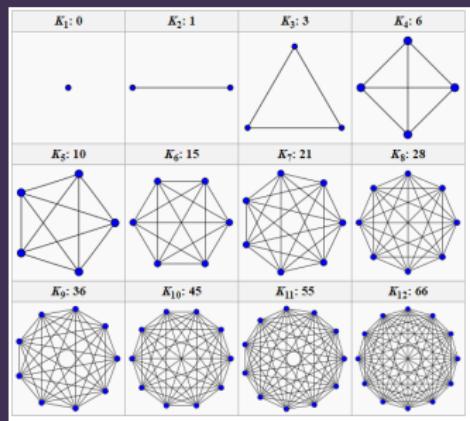
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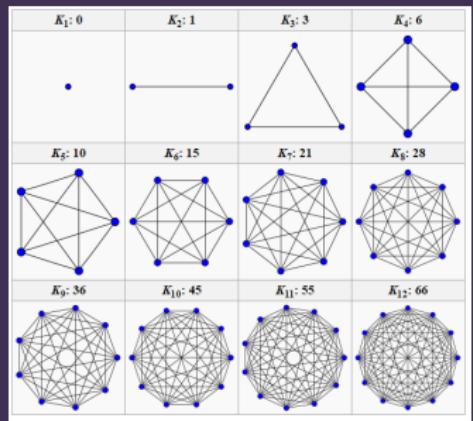


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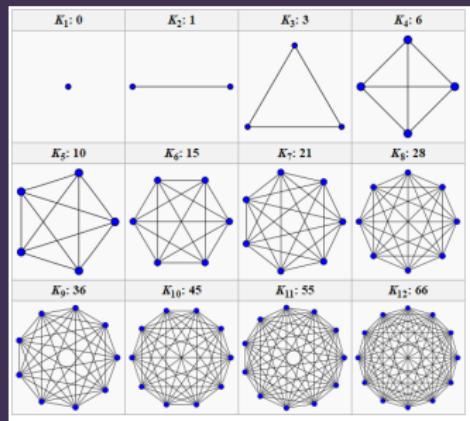
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  - ▶ Further reading: spatial hashing, quadtrees, octrees, Verlet lists

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- ▶ Are there any problems in  $NP$  but not in  $P$ ?

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- ▶ Choice of data structures and algorithms can have a large impact on the efficiency of your software
- ▶ ... but only if scalability is actually a factor

# Open workshop



# Open workshop

- ▶ Use the rest of this session to get on with your **research journal**
- ▶ Post in the chat if you have questions or need help!