

Measuring and Predicting Running Time

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Outline

Running times of different implementations

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- ▶ We have two implementations of PhoneDirectory: ArrayBasedPD and SortedPD.



Running times of different implementations

- ▶ We have two implementations of PhoneDirectory: ArrayBasedPD and SortedPD.
- ▶ Each has implementations of find, add, and remove.



Running times of different implementations

- ▶ We have two implementations of PhoneDirectory: ArrayBasedPD and SortedPD.
- ▶ Each has implementations of find, add, and remove.
- ▶ Can we compare their speeds?



find



find

- ▶ ArrayBasedPD.find



find

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 - ▶ Jay, Bob, Zoe, Ian, Ann, Eve



find

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 - ▶ Jay, Bob, Zoe, Ian, Ann, Eve
 - ▶ Look for Vic?



- ▶ `ArrayBasedPD.find`
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 - ▶ Have to compare Vic with n entries, where $n = \text{size}$, which is 6.

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- ▶ `SortedPD.find`



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- ▶ `SortedPD.find`
 - ▶ Only really helpful when n (size) is large.



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 - ▶ Jay, Bob, Zoe, Ian, Ann, Eve
 - ▶ Look for Vic?
 - ▶ Have to compare Vic with n entries, where $n = \text{size}$, which is 6.
- ▶ `SortedPD.find`
 - ▶ Only really helpful when n (size) is large.
 - ▶ Requires $\log_2 n$ comparisons



addOrChangeEntry

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- ▶ `ArrayBasedPD.addOrChangeEntry`



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addOrChangeEntry

- ▶ ArrayBasedPD.addOrChangeEntry
 - ▶ Jay, Bob, Zoe, Ian, Ann, Eve
 - ▶ Add Vic?



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- ▶ `SortedPD.addOrChangeEntry`
 - ▶ Also has to call find and wait for find to finish.



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 - ▶ add uses n array accesses. Actually $n - 1$ reads and n writes, where n is 7. So $2n - 1$.



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 - ▶ Abe, Ann, Bob, Eve, Ian, Jay, Zoe
 - ▶ add uses n array accesses. Actually $n - 1$ reads and n writes, where n is 7. So $2n - 1$.
 - ▶ Total time is $\log_2 n$ comparisons (find) plus $2n - 1$ array accesses (add).



removeEntry

removeEntry

- ▶ `ArrayBasedPD.removeEntry`



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removeEntry

- ▶ ArrayBasedPD.removeEntry
 - ▶ Jay, Bob, Zoe, Ian, Ann, Eve
 - ▶ Who takes longest to remove? Jay?



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 - ▶ Jay, Bob, Zoe, Ian, Ann, Eve
 - ▶ Who takes longest to remove? Jay?
 - ▶ removeEntry calls find.



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 - ▶ removeEntry calls find.
 - ▶ find takes 1 comparison to find Jay.



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 - ▶ remove takes 2 array accesses to remove Jay.



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 - ▶ remove takes 2 array accesses to remove Jay.
 - ▶ Total time for 1 comparison and 2 array accesses.



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 - ▶ removeEntry calls remove.
 - ▶ Eve, Bob, Zoe, Ian, Ann
 - ▶ remove takes 2 array accesses to remove Jay.
 - ▶ Total time for 1 comparison and 2 array accesses.
 - ▶ What about Eve? (Last entry)



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 - ▶ Jay, Bob, Zoe, Ian, Ann, Eve
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 - ▶ `remove` takes 2 array accesses to remove Jay.
 - ▶ Total time for 1 comparison and 2 array accesses.
 - ▶ What about Eve? (Last entry)
 - ▶ Call to `find` takes n comparisons.



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 - ▶ Eve, Bob, Zoe, Ian, Ann
 - ▶ `remove` takes 2 array accesses to remove Jay.
 - ▶ Total time for 1 comparison and 2 array accesses.
 - ▶ What about Eve? (Last entry)
 - ▶ Call to `find` takes n comparisons.
 - ▶ `add` still uses 2 array accesses to “remove” Eve (but it could be smarter).



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 - ▶ Jay, Bob, Zoe, Ian, Ann, Eve
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 - ▶ What about Eve? (Last entry)
 - ▶ Call to find takes n comparisons.
 - ▶ add still uses 2 array accesses to “remove” Eve (but it could be smarter).
 - ▶ So Eve is worst case, requiring time for n comparisons (find) and 2 array accesses (remove).



removeEntry

- ▶ `ArrayBasedPD.removeEntry`
 - ▶ Jay, Bob, Zoe, Ian, Ann, Eve
 - ▶ Who takes longest to remove? Jay?
 - ▶ `removeEntry` calls `find`.
 - ▶ `find` takes 1 comparison to find Jay.
 - ▶ `removeEntry` calls `remove`.
 - ▶ Eve, Bob, Zoe, Ian, Ann
 - ▶ `remove` takes 2 array accesses to remove Jay.
 - ▶ Total time for 1 comparison and 2 array accesses.
 - ▶ What about Eve? (Last entry)
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 - ▶ So Eve is worst case, requiring time for n comparisons (`find`) and 2 array accesses (`remove`).
- ▶ `SortedPD.removeEntry`



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 - ▶ Jay, Bob, Zoe, Ian, Ann, Eve
 - ▶ Who takes longest to remove? Jay?
 - ▶ `removeEntry` calls `find`.
 - ▶ `find` takes 1 comparison to find Jay.
 - ▶ `removeEntry` calls `remove`.
 - ▶ Eve, Bob, Zoe, Ian, Ann
 - ▶ `remove` takes 2 array accesses to remove Jay.
 - ▶ Total time for 1 comparison and 2 array accesses.
 - ▶ What about Eve? (Last entry)
 - ▶ Call to `find` takes n comparisons.
 - ▶ `add` still uses 2 array accesses to “remove” Eve (but it could be smarter).
 - ▶ So Eve is worst case, requiring time for n comparisons (`find`) and 2 array accesses (`remove`).
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 - ▶ Ann, Bob, Eve, Ian, Jay, Zoe

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 - ▶ Who takes longest to remove? Jay?
 - ▶ `removeEntry` calls `find`.
 - ▶ `find` takes 1 comparison to find Jay.
 - ▶ `removeEntry` calls `remove`.
 - ▶ Eve, Bob, Zoe, Ian, Ann
 - ▶ `remove` takes 2 array accesses to remove Jay.
 - ▶ Total time for 1 comparison and 2 array accesses.
 - ▶ What about Eve? (Last entry)
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- ▶ `SortedPD.removeEntry`
 - ▶ Ann, Bob, Eve, Ian, Jay, Zoe
 - ▶ Who is the worst to remove?



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 - ▶ Who takes longest to remove? Jay?
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 - ▶ `remove` takes 2 array accesses to remove Jay.
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 - ▶ What about Eve? (Last entry)
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- ▶ `SortedPD.removeEntry`
 - ▶ Ann, Bob, Eve, Ian, Jay, Zoe
 - ▶ Who is the worst to remove?
 - ▶ Did you figure out it was Ann?



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 - ▶ Who takes longest to remove? Jay?
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 - ▶ Eve, Bob, Zoe, Ian, Ann
 - ▶ `remove` takes 2 array accesses to remove Jay.
 - ▶ Total time for 1 comparison and 2 array accesses.
 - ▶ What about Eve? (Last entry)
 - ▶ Call to `find` takes n comparisons.
 - ▶ `add` still uses 2 array accesses to “remove” Eve (but it could be smarter).
 - ▶ So Eve is worst case, requiring time for n comparisons (`find`) and 2 array accesses (`remove`).
- ▶ `SortedPD.removeEntry`
 - ▶ Ann, Bob, Eve, Ian, Jay, Zoe
 - ▶ Who is the worst to remove?
 - ▶ Did you figure out it was Ann?
 - ▶ `find` takes $\log_2 n$ comparisons to locate Ann.



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 - ▶ Jay, Bob, Zoe, Ian, Ann, Eve
 - ▶ Who takes longest to remove? Jay?
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 - ▶ `removeEntry` calls `remove`.
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 - ▶ `remove` takes 2 array accesses to remove Jay.
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 - ▶ What about Eve? (Last entry)
 - ▶ Call to `find` takes n comparisons.
 - ▶ `add` still uses 2 array accesses to “remove” Eve (but it could be smarter).
 - ▶ So Eve is worst case, requiring time for n comparisons (`find`) and 2 array accesses (`remove`).
- ▶ `SortedPD.removeEntry`
 - ▶ Ann, Bob, Eve, Ian, Jay, Zoe
 - ▶ Who is the worst to remove?
 - ▶ Did you figure out it was Ann?
 - ▶ `find` takes $\log_2 n$ comparisons to locate Ann.
 - ▶ `add` takes n array reads and writes to move everyone else back.



removeEntry

- ▶ `ArrayBasedPD.removeEntry`
 - ▶ Jay, Bob, Zoe, Ian, Ann, Eve
 - ▶ Who takes longest to remove? Jay?
 - ▶ `removeEntry` calls `find`.
 - ▶ `find` takes 1 comparison to find Jay.
 - ▶ `removeEntry` calls `remove`.
 - ▶ Eve, Bob, Zoe, Ian, Ann
 - ▶ `remove` takes 2 array accesses to remove Jay.
 - ▶ Total time for 1 comparison and 2 array accesses.
 - ▶ What about Eve? (Last entry)
 - ▶ Call to `find` takes n comparisons.
 - ▶ `add` still uses 2 array accesses to “remove” Eve (but it could be smarter).
 - ▶ So Eve is worst case, requiring time for n comparisons (`find`) and 2 array accesses (`remove`).
- ▶ `SortedPD.removeEntry`
 - ▶ Ann, Bob, Eve, Ian, Jay, Zoe
 - ▶ Who is the worst to remove?
 - ▶ Did you figure out it was Ann?
 - ▶ `find` takes $\log_2 n$ comparisons to locate Ann.
 - ▶ `add` takes n array reads and writes to move everyone else back.
 - ▶ Bob, Eve, Ian, Jay, Zoe



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 - ▶ Who takes longest to remove? Jay?
 - ▶ `removeEntry` calls `find`.
 - ▶ `find` takes 1 comparison to find Jay.
 - ▶ `removeEntry` calls `remove`.
 - ▶ Eve, Bob, Zoe, Ian, Ann
 - ▶ `remove` takes 2 array accesses to remove Jay.
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- ▶ `SortedPD.removeEntry`
 - ▶ Ann, Bob, Eve, Ian, Jay, Zoe
 - ▶ Who is the worst to remove?
 - ▶ Did you figure out it was Ann?
 - ▶ `find` takes $\log_2 n$ comparisons to locate Ann.
 - ▶ `add` takes n array reads and writes to move everyone else back.
 - ▶ Bob, Eve, Ian, Jay, Zoe
 - ▶ Total is $\log_2 n$ comparisons (`find`) and $2n$ array accesses (`remove`).



Summary



Summary

- ▶ ArrayBasedPD



Summary

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 - ▶ find: n comparisons



Summary

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 - ▶ find: n comparisons
 - ▶ add: 1 array access (usually)



Summary

- ▶ ArrayBasedPD
 - ▶ find: n comparisons
 - ▶ add: 1 array access (usually)
 - ▶ remove: 2 array accesses



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 - ▶ find: n comparisons
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 - ▶ find: $\log_2 n$ comparisons



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- ▶ Constants don't matter.
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- ▶ Only the dominant term matters.
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- ▶ By the time you have opened the garage door, got in, started it up, etc., you will spend more time than just walking there.
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- ▶ So 75 microseconds.



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- ▶ So 75 microseconds.
- ▶ Notice that I used the same log base 10. You can't switch log bases in the middle, or you will get a different (and wrong) answer.



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 - ▶ $t = 74.997$
- ▶ Different log. Same answer!



I'M JUST OUTSIDE TOWN, SO I SHOULD
BE THERE IN FIFTEEN MINUTES.

ACTUALLY, IT'S LOOKING
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NO, WAIT, THIRTY SECONDS.



THE AUTHOR OF THE WINDOWS FILE
COPY DIALOG VISITS SOME FRIENDS.

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- ▶ Answer: repeat the experiment many times and take the average.



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- ▶ Accurate predictions can make or break a business and save millions of dollars.
- ▶ To improve the accuracy of a measurement, repeat it many times and take an average.
- ▶ For example, run it once to get an approximate time. Figure out how many times you can run it in one second. Run it that many times and take the average running time.

