

# Easy parallelization in R (using the foreach package)

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## Welcome to this R workshop about loops loops loops!

Before we can begin, please make sure that you can run the code chunk below.

Code chunks are areas of “R Markdown Notebooks” that allow us to execute our scripts. You can use the “Run” commands at the top of the code editor pane to run all or parts of chunks, or use the Cmd + Enter shortcut to Run any highlighted code, or the current line.

```
# Set the root to point to wherever /howtoforeach/ is on your computer
setwd( "~/Git/howtoforeach" ) # Change

## Load required packages

# For workshop related tasks
library( tictoc )      # for obtaining accurate code run time
require( lme4 )        # we will bootstrap a few models; only using a few functions so saving memory by
library( beeper )      # sometimes I like an auditory notification when long processes finish

# For making foreach run correctly
library( tidyverse )  # for data processing and better syntax
library( foreach )    # for parallel loop procedures across cores
library( parallel )   # for setting up your machine for multi-core processing
library( doParallel ) # for setting up your machine for multi-core processing (also)
```

We will do some additional setup specific to **foreach** when that package is introduced.

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## Part 1: Getting started

### To loop or not to loop

Beginning coders often start by writing looping scripts because they highlight the power of using computers to automate tasks. In general, “for” loops are used for tasks with a clear end point, and “while” loops are used for tasks with indefinite ends points.

For example, if you want to repeat a sequence for every row in a dataframe, you would use **for()**, but if you want to repeat a sequence until a condition is met, you would use a **while()** loop, which runs while the argument of **while()** remains TRUE.

You may have received advice to “not write loops”, but there is nothing inherently incorrect about writing loops; it’s legitimate code! This advice comes from the fact that some people believe that slow code is bad code, but I disagree.

Sometimes, writing a loop is easier because it follows a sequential flow that humans are used to reasoning through. When R starts breaking up calculations across cores, rewriting code into faster languages, and other performance-enhancing strategies, sometimes it’s difficult to understand what is really happening at the same level.

Also, sequential processing is inherent to many computational tasks, such as text processing, repeated sampling (bootstrapping), and web scraping. So, rather than preaching that nobody should write another loop, I will show you how to write loops that are more computationally efficient.

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## What’s the solution?

Let’s assume we have to write our code as a loop for some reason, but we also want the code to execute as quickly as possible. The main reason why loops tend to run slow is due to their sequential nature—the computer has to wait to process Item 100 until it can process Item 99—and every processing delay piles up.

Our solution will be to split up the single sequence into multiple chunks, so Items 1 through 33, 34 through 67, and 68 through 100 will be processed in parallel, cutting execution time up to 66%.

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## “I need the absolute fastest code!”

I highly recommend the `tidyverse` collection of R packages. If you can substitute a loop for a tidyverse statement, it will be orders of magnitude faster in most cases. It is also faster to write, and easier to understand, which can be just as important as having a working script.

If you don’t mind learning different syntax, you can also look at `data.table` which is generally faster than `tidyverse`, but is limited to rectangular (column-based) data.

“Vectorized” is a catch-all term for these super-fast packages that operate at multiple points of the data at the same time. Sometimes vectorized code is also run in parallel—it depends on the package and what the specific process is.

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## Part 2: The humble loop

### Loading data files with a loop

In the next code chunk, you’ll see the humble loop loading our text data files, and a vectorized version for comparison. We will time our operations using the package ‘tictoc’.

Notice in this example how much more work goes into writing a loop, but it’s clear and easy to modify. For example, I can write a progress message into the loop, but not into its vectorized equivalent.

Also, see how I have to create a container for the looped data; otherwise each loop would overwrite the previous one, and we would not be able to access the results of each iteration.

```
#####
# Sequential (looping) import #
#####

# List the files in the /data/ directory to build the filepaths
filepaths = paste0( getwd(), "/data/", list.files( path = "./data/", pattern = ".csv"))

# Let's take a peek at the first file, file.1
file.1 = read_csv( filepaths[[1]], col_types = cols()) # TIP: col_types = cols() suppresses warnings

# Empty version of file.1 (so I don't need to specify colnames later)
reviews.seq = file.1[FALSE,]

# Looping the list of files (slow and lots of lines)
tic("Looping version") # start timer
# I am demonstrating a 'for' loop for now, but there are also 'while' loops
for( file in seq_along( filepaths ) ) {
  print( paste("Loading file", file, "out of", paste0(length(filepaths)) ) ) # try this with vectori
  the.file = read_csv( filepaths[[ file ]], col_types = cols() ) # read the data into the.file
  reviews.seq = rbind( reviews.seq, the.file) # add it the.file to the bottom of reviews.seq
}
```

```
## [1] "Loading file 1 out of 3"
## [1] "Loading file 2 out of 3"
## [1] "Loading file 3 out of 3"
```

```
toc() # end timer
```

```
## Looping version: 0.28 sec elapsed
```

For comparison, all of the above is accomplished in 1 line here:

```
#####
# Vectorized import #
#####

tic("Vectorized version") # start timer

# With the plyr package, we can apply read_csv directly to the filepaths list
reviews = plyr::ldply( .data = filepaths, .fun = read_csv, col_types = cols() )

toc() # end coder timer
```

```
## Vectorized version: 0.258 sec elapsed
```

```
# Hopefully the two objects are of the same size.
ifelse( identical( dim(reviews), dim(reviews.seq) ),
        "PASS: Results match in size", "FAIL: Results do not match in size")
```

```
## [1] "PASS: Results match in size"
```

We will conclude our comparisons of looping versus vectorized code here. . . if you want speed, avoid sequential operations and instead use functions that automatically vectorize the computation, such as `tidyverse` or `data.table`.

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## Choosing a computational environment

Before we transition into executing parallel code, let's talk about another important factor in performance—how much processing power is available? If you need more capacity than a single computer can provide, then you may want to run the code on a purpose-made computational environment.

For example, Northwestern University maintains a research computing cluster called Quest with 11,800 cores, which are optimized for running code for its students and faculty. Basic Access is free. A non-free option is Amazon AWS, which will allow you to (glossing over details) create a virtual computer for you to run code on.

To state the obvious, if your code will be running for a few days, it's more convenient to have it run on a non-essential non-personal computer!

More info: <https://www.it.northwestern.edu/research/user-services/quest/overview.html>

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## Part 3: The parallel loop

### Text processing example

Text is typically slow to process, for a variety of reasons, and the way we work with the data during text analysis is often sequential. This makes text processing a good candidate for use of parallel loops.

For this example, we will use the data we loaded, which consists of ~50k job reviews from Amazon employees, scraped from indeed.com. Expect this chunk to take extra time to process.

```
# For each row, let's extract the state and count the length of the review details
{ # start timing
tic("Sequential text processing")

for ( row in 1:nrow( reviews ) ) {

  # Using a regular expression to extract pairs of capitalized letters
  reviews$state[row] = str_extract( reviews$Work_Location[row], "\\b[A-Z]{2}")

  # nchar() counts the characters
  reviews$nchar[row] = nchar( reviews$Review_Details[row] )

}

toc() } # end timing
```

```
## Sequential text processing: 39.031 sec elapsed
```

---

## Now let's split the loop over our cores, using foreach

First we have to set up the computer to run our loops in parallel. Out of the box, `foreach` does not run in parallel; it requires the `doParallel` package in order to use multiple cores. This only has to be done once per R session, and usually I do this immediately after loading libraries.

```
# Returns the number of cores that R can see
n.cores = detectCores()

# I like to keep one core in reserve. Otherwise, if the R session crashes, it can make your whole compu
cl = makeCluster( n.cores - 1 )

# Initiate an R process on the earmarked cores
registerDoParallel( cl )
```

Now that we have N-1 cores ready to receive commands from R, we will take the previous loop and divide it across the available cores. The following chunk calls `foreach()` to create a loop of row indexes which we will use to process each review in two ways: 1. We will attempt to extract the state abbreviation code (e.g. IL, CA, NY) from the review. 2. We will count the number of characters in the review. The results will be passed to `rbind` to create an extended version of the original data.

```
{ # begin timing
tic("Parallel text processing")

# Same functions as our sequential loop, but using foreach() and %dopar%
reviews.processed = foreach( row = 1:nrow( reviews ),
  .combine = "rbind",
  .packages = c("stringr")) %dopar% { # Packages need to be loaded for each core

  # Extract state code (2 capital letters)
  the.state = str_extract( reviews$Work_Location[row], "\\b[A-Z]{2}")

  # Extract review length
  the.nchar = nchar( reviews$Review_Details[row] )

  # I'm essentially appending 2 columns to the original data by extending each row
  return( c( reviews[row,], the.state, the.nchar ) )

} # end of foreach loop

toc() } # end timing chunk
```

```
## Parallel text processing: 38.529 sec elapsed
```

```
# Adding pretty column names to our extended data
reviews.processed = as.data.frame( reviews.processed )
colnames( reviews.processed ) = c(colnames( reviews ), # starting point
  "state.code", # I could also left join
  "review.len") # but this is fast and clean

reviews = reviews.processed
```

## Benchmarks

Dan's 2012 MacBook (no Zoom): 45s

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### Data chunking example

In the previous example, we looped each row by creating an iterator out of `reviews`. This is not a great use of parallel processing though, because there is little to gain when splitting 48k rows across a few cores. It's usually more efficient to break up your data into bigger chunks, for example, by state. You always want to parallelize the slowest part of the process, if possible.

For example, I usually use `foreach()` when web scraping because the slowest part of the process is making a connection with the remote server, so I have each core processing a different URL usually. I make sure I only ever have to call a URL once, and once I have made a connection, I used vectorized code to extract the target data, which minimizes the time per page.

```
# Let's get a list of all the (confirmed) states that appear in the $state field. We confirm states by
state.list = as.data.frame( table( unlist( reviews$state ) ) )
the.states = intersect( state.abb, reviews$state )

{ # beginning of timer
tic("Parallel loop of all states")

# Parallel loop of the states
reviews.bystate = foreach( state = seq_along( the.states ),
  .combine = "rbind",
  .packages = c("tictoc")) %dopar% { # let's do it in base R to avoid calling packages here

  # Collect the rows for this state
  the.state = unlist( the.states[state] )
  state.rows = reviews[ which( reviews$state.code == the.state ), ]

  # Let's time each state, out of curiosity
  tic(paste("State:", the.state))

  # Sum the nchar column
  state.sum = sum ( as.integer( state.rows$review.len ) )

  # Let's get the average rating too, while we're at it
  state.avg = round( mean( as.integer( state.rows$Rating ) ) )

  # Returning a vector on each loop, which I rbind into a matrix (fast!)
  return( c(the.state, state.sum, state.avg, toc()$toc) )

} # end reviews.bystate
toc() } # end of timer
```

```
## Parallel loop of all states: 2.278 sec elapsed
```

```
# Turn the matrix into a df w/ pretty colnames
reviews.bystate = as.data.frame( reviews.bystate )
colnames( reviews.bystate ) = c( "state.code", "review.len.sum", "rating.mean", "processing time")
```

It's much faster to process state-by-state versus row-by-row, but why?

### Why is it faster to loop state-by-state?

The 47 states represented in the data will be equally divided across multiple cores on my computer. For example, since I am allowing R to utilize 3 cores, each processor only has to handle about 15 states, rather than a single core handling all 47 one at a time.

Maybe Core #1 processed Michigan and Florida and Core #2 processed Illinois and California. We have a lot of control over this process, but by default `foreach()` is optimized to balance performance with being a general purpose tool.

Using `foreach()` is a little bit like dividing your data into thirds and having two colleagues process them for you while you run your share. At the end of the day, you'll have 3 sets of results which, when combined, are identical to if you ran the process sequentially.

---

## Coding Challenge A

The first challenge involves repairing the `$state` variable, which you can tell by looking at `state.list` contains many errors. Here's some meta-code that shows the basic process I will guide you through:

*Split Place\_Of\_Work\_1 to get state*

*Loop city,state strings to extract state abbreviation* `foreach(city_state_string)`

*Validate extracted data using built-in list, state.abb* `if(state in state.abb) str_extract(state)`

```
# write_rds( reviews, "./data/snapshot_a.rds") in coding_challenges you will load this file
```

---

## Part 4: Going deeper into `foreach()`

Now that we've covered the basic setup and syntax of `foreach()`, we will talk more about its parameters and use cases.

### Combining results

Besides being able to use multiple iterators to generate loops, `foreach` is also flexible in how it formats its output. Depending on how much memory a

*Concatenate* `.combine = 'c'`

I use this method when I want the result to be a list, or list of lists.

*Column/row bind* `.combine = 'cbind'/.combine = 'rbind'`

I use this method when I want the result to be a dataframe, tibble, or matrix.

*Basic operators* `.combine = '+'/.combine = '*'`

These methods make sense for bootstrapping and simulating data.

*Custom function* `.combine = 'cfun'`

```
# Adapted from the foreach documentation
comb.c      = foreach(i=icount(4), .combine='c') %dopar% exp(i)
comb.cbind  = foreach(icount(4), .combine='cbind') %dopar% rnorm(4)
comb.add    = foreach(icount(4), .combine='+') %dopar% rnorm(4)

# TIP: You can include conditional logic to foreach calls using `:%` and `when()` ... somewhat like Py

# As long as `i` is between 1-5, wrote the corresponding letter to the console
foreach( i = icount( 32 ) ) %:% when( i <= 5 ) %dopar% letters[i]
```

```
## [[1]]
## [1] "a"
##
## [[2]]
## [1] "b"
##
## [[3]]
## [1] "c"
##
## [[4]]
## [1] "d"
##
## [[5]]
## [1] "e"
```

---

## Nested foreach loops

A common text analysis problem is computing the similarity score between different strings. There are a lot of scores we could use, but to keep things simple, we simply take the proportion of words in common.

Loops can contain other loops, which can be convenient for checking all combinations of variables. In our toy problem, we are comparing vectors of words, so the first thing we need to do is create a dataset consisting of these vectors.

```
# Create a list of vectors, one for each review, containing a vector of words
grid_words = unlist( reviews$Place_Of_Work_1, reviews$Place_Of_Work_2 ) %>%
  tolower() %>%
  str_extract_all( pattern = '[a-z]+' )

# Print the head below
head( grid_words )
```

```
## [[1]]
## [1] "amazon"    "shopper"    "current"    "employee"    "dedham"    "ma"          "december"
```



```
##
## [[2]]
## [1] "warehouse" "team"      "member"    "former"    "employee"  "richmond"
## [7] "va"        "december"
##
## [[3]]
## [1] "warehouse" "order"      "picker"    "driver"    "both"
## [6] "inbound"   "and"        "outbound"  "capacities" "former"
## [11] "employee"  "charleston" "tn"        "december"
##
## [[4]]
## [1] "sortation" "associate" "at"        "delivery"  "station"   "former"
## [7] "employee"  "austin"    "tx"        "december"
##
## [[5]]
## [1] "warehouse" "worker"     "former"    "employee"  "pendergrass"
## [6] "ga"        "december"
##
## [[6]]
## [1] "amazon"    "warehouse" "associate" "current"   "employee"  "brooklyn"
## [7] "park"      "mn"        "december"
```

The easiest way to exhaustively compare two vectors is by using `expand.grid()` which finds all combinations. For example:

```
expand.grid( state.abb, state.abb )
```

```
##      Var1 Var2
## 1      AL  AL
## 2      AK  AL
## 3      AZ  AL
## 4      AR  AL
## 5      CA  AL
## 6      CO  AL
## 7      CT  AL
## 8      DE  AL
## 9      FL  AL
## 10     GA  AL
## 11     HI  AL
## 12     ID  AL
## 13     IL  AL
## 14     IN  AL
## 15     IA  AL
## 16     KS  AL
## 17     KY  AL
## 18     LA  AL
## 19     ME  AL
## 20     MD  AL
## 21     MA  AL
## 22     MI  AL
## 23     MN  AL
## 24     MS  AL
## 25     MO  AL
```

## 26	MT	AL
## 27	NE	AL
## 28	NV	AL
## 29	NH	AL
## 30	NJ	AL
## 31	NM	AL
## 32	NY	AL
## 33	NC	AL
## 34	ND	AL
## 35	OH	AL
## 36	OK	AL
## 37	OR	AL
## 38	PA	AL
## 39	RI	AL
## 40	SC	AL
## 41	SD	AL
## 42	TN	AL
## 43	TX	AL
## 44	UT	AL
## 45	VT	AL
## 46	VA	AL
## 47	WA	AL
## 48	WV	AL
## 49	WI	AL
## 50	WY	AL
## 51	AL	AK
## 52	AK	AK
## 53	AZ	AK
## 54	AR	AK
## 55	CA	AK
## 56	CO	AK
## 57	CT	AK
## 58	DE	AK
## 59	FL	AK
## 60	GA	AK
## 61	HI	AK
## 62	ID	AK
## 63	IL	AK
## 64	IN	AK
## 65	IA	AK
## 66	KS	AK
## 67	KY	AK
## 68	LA	AK
## 69	ME	AK
## 70	MD	AK
## 71	MA	AK
## 72	MI	AK
## 73	MN	AK
## 74	MS	AK
## 75	MO	AK
## 76	MT	AK
## 77	NE	AK
## 78	NV	AK
## 79	NH	AK

## 80	NJ	AK
## 81	NM	AK
## 82	NY	AK
## 83	NC	AK
## 84	ND	AK
## 85	OH	AK
## 86	OK	AK
## 87	OR	AK
## 88	PA	AK
## 89	RI	AK
## 90	SC	AK
## 91	SD	AK
## 92	TN	AK
## 93	TX	AK
## 94	UT	AK
## 95	VT	AK
## 96	VA	AK
## 97	WA	AK
## 98	WV	AK
## 99	WI	AK
## 100	WY	AK
## 101	AL	AZ
## 102	AK	AZ
## 103	AZ	AZ
## 104	AR	AZ
## 105	CA	AZ
## 106	CO	AZ
## 107	CT	AZ
## 108	DE	AZ
## 109	FL	AZ
## 110	GA	AZ
## 111	HI	AZ
## 112	ID	AZ
## 113	IL	AZ
## 114	IN	AZ
## 115	IA	AZ
## 116	KS	AZ
## 117	KY	AZ
## 118	LA	AZ
## 119	ME	AZ
## 120	MD	AZ
## 121	MA	AZ
## 122	MI	AZ
## 123	MN	AZ
## 124	MS	AZ
## 125	MO	AZ
## 126	MT	AZ
## 127	NE	AZ
## 128	NV	AZ
## 129	NH	AZ
## 130	NJ	AZ
## 131	NM	AZ
## 132	NY	AZ
## 133	NC	AZ

##	134	ND	AZ
##	135	OH	AZ
##	136	OK	AZ
##	137	OR	AZ
##	138	PA	AZ
##	139	RI	AZ
##	140	SC	AZ
##	141	SD	AZ
##	142	TN	AZ
##	143	TX	AZ
##	144	UT	AZ
##	145	VT	AZ
##	146	VA	AZ
##	147	WA	AZ
##	148	WV	AZ
##	149	WI	AZ
##	150	WY	AZ
##	151	AL	AR
##	152	AK	AR
##	153	AZ	AR
##	154	AR	AR
##	155	CA	AR
##	156	CO	AR
##	157	CT	AR
##	158	DE	AR
##	159	FL	AR
##	160	GA	AR
##	161	HI	AR
##	162	ID	AR
##	163	IL	AR
##	164	IN	AR
##	165	IA	AR
##	166	KS	AR
##	167	KY	AR
##	168	LA	AR
##	169	ME	AR
##	170	MD	AR
##	171	MA	AR
##	172	MI	AR
##	173	MN	AR
##	174	MS	AR
##	175	MO	AR
##	176	MT	AR
##	177	NE	AR
##	178	NV	AR
##	179	NH	AR
##	180	NJ	AR
##	181	NM	AR
##	182	NY	AR
##	183	NC	AR
##	184	ND	AR
##	185	OH	AR
##	186	OK	AR
##	187	OR	AR

##	188	PA	AR
##	189	RI	AR
##	190	SC	AR
##	191	SD	AR
##	192	TN	AR
##	193	TX	AR
##	194	UT	AR
##	195	VT	AR
##	196	VA	AR
##	197	WA	AR
##	198	WV	AR
##	199	WI	AR
##	200	WY	AR
##	201	AL	CA
##	202	AK	CA
##	203	AZ	CA
##	204	AR	CA
##	205	CA	CA
##	206	CO	CA
##	207	CT	CA
##	208	DE	CA
##	209	FL	CA
##	210	GA	CA
##	211	HI	CA
##	212	ID	CA
##	213	IL	CA
##	214	IN	CA
##	215	IA	CA
##	216	KS	CA
##	217	KY	CA
##	218	LA	CA
##	219	ME	CA
##	220	MD	CA
##	221	MA	CA
##	222	MI	CA
##	223	MN	CA
##	224	MS	CA
##	225	MO	CA
##	226	MT	CA
##	227	NE	CA
##	228	NV	CA
##	229	NH	CA
##	230	NJ	CA
##	231	NM	CA
##	232	NY	CA
##	233	NC	CA
##	234	ND	CA
##	235	OH	CA
##	236	OK	CA
##	237	OR	CA
##	238	PA	CA
##	239	RI	CA
##	240	SC	CA
##	241	SD	CA

##	242	TN	CA
##	243	TX	CA
##	244	UT	CA
##	245	VT	CA
##	246	VA	CA
##	247	WA	CA
##	248	WV	CA
##	249	WI	CA
##	250	WY	CA
##	251	AL	CO
##	252	AK	CO
##	253	AZ	CO
##	254	AR	CO
##	255	CA	CO
##	256	CO	CO
##	257	CT	CO
##	258	DE	CO
##	259	FL	CO
##	260	GA	CO
##	261	HI	CO
##	262	ID	CO
##	263	IL	CO
##	264	IN	CO
##	265	IA	CO
##	266	KS	CO
##	267	KY	CO
##	268	LA	CO
##	269	ME	CO
##	270	MD	CO
##	271	MA	CO
##	272	MI	CO
##	273	MN	CO
##	274	MS	CO
##	275	MO	CO
##	276	MT	CO
##	277	NE	CO
##	278	NV	CO
##	279	NH	CO
##	280	NJ	CO
##	281	NM	CO
##	282	NY	CO
##	283	NC	CO
##	284	ND	CO
##	285	OH	CO
##	286	OK	CO
##	287	OR	CO
##	288	PA	CO
##	289	RI	CO
##	290	SC	CO
##	291	SD	CO
##	292	TN	CO
##	293	TX	CO
##	294	UT	CO
##	295	VT	CO

##	296	VA	CO
##	297	WA	CO
##	298	WV	CO
##	299	WI	CO
##	300	WY	CO
##	301	AL	CT
##	302	AK	CT
##	303	AZ	CT
##	304	AR	CT
##	305	CA	CT
##	306	CO	CT
##	307	CT	CT
##	308	DE	CT
##	309	FL	CT
##	310	GA	CT
##	311	HI	CT
##	312	ID	CT
##	313	IL	CT
##	314	IN	CT
##	315	IA	CT
##	316	KS	CT
##	317	KY	CT
##	318	LA	CT
##	319	ME	CT
##	320	MD	CT
##	321	MA	CT
##	322	MI	CT
##	323	MN	CT
##	324	MS	CT
##	325	MO	CT
##	326	MT	CT
##	327	NE	CT
##	328	NV	CT
##	329	NH	CT
##	330	NJ	CT
##	331	NM	CT
##	332	NY	CT
##	333	NC	CT
##	334	ND	CT
##	335	OH	CT
##	336	OK	CT
##	337	OR	CT
##	338	PA	CT
##	339	RI	CT
##	340	SC	CT
##	341	SD	CT
##	342	TN	CT
##	343	TX	CT
##	344	UT	CT
##	345	VT	CT
##	346	VA	CT
##	347	WA	CT
##	348	WV	CT
##	349	WI	CT

##	350	WY	CT
##	351	AL	DE
##	352	AK	DE
##	353	AZ	DE
##	354	AR	DE
##	355	CA	DE
##	356	CO	DE
##	357	CT	DE
##	358	DE	DE
##	359	FL	DE
##	360	GA	DE
##	361	HI	DE
##	362	ID	DE
##	363	IL	DE
##	364	IN	DE
##	365	IA	DE
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##	2053	AZ	TN
##	2054	AR	TN
##	2055	CA	TN
##	2056	CO	TN
##	2057	CT	TN
##	2058	DE	TN
##	2059	FL	TN
##	2060	GA	TN
##	2061	HI	TN
##	2062	ID	TN
##	2063	IL	TN
##	2064	IN	TN
##	2065	IA	TN
##	2066	KS	TN
##	2067	KY	TN
##	2068	LA	TN
##	2069	ME	TN
##	2070	MD	TN
##	2071	MA	TN
##	2072	MI	TN
##	2073	MN	TN
##	2074	MS	TN
##	2075	MO	TN
##	2076	MT	TN
##	2077	NE	TN

##	2078	NV	TN
##	2079	NH	TN
##	2080	NJ	TN
##	2081	NM	TN
##	2082	NY	TN
##	2083	NC	TN
##	2084	ND	TN
##	2085	OH	TN
##	2086	OK	TN
##	2087	OR	TN
##	2088	PA	TN
##	2089	RI	TN
##	2090	SC	TN
##	2091	SD	TN
##	2092	TN	TN
##	2093	TX	TN
##	2094	UT	TN
##	2095	VT	TN
##	2096	VA	TN
##	2097	WA	TN
##	2098	WV	TN
##	2099	WI	TN
##	2100	WY	TN
##	2101	AL	TX
##	2102	AK	TX
##	2103	AZ	TX
##	2104	AR	TX
##	2105	CA	TX
##	2106	CO	TX
##	2107	CT	TX
##	2108	DE	TX
##	2109	FL	TX
##	2110	GA	TX
##	2111	HI	TX
##	2112	ID	TX
##	2113	IL	TX
##	2114	IN	TX
##	2115	IA	TX
##	2116	KS	TX
##	2117	KY	TX
##	2118	LA	TX
##	2119	ME	TX
##	2120	MD	TX
##	2121	MA	TX
##	2122	MI	TX
##	2123	MN	TX
##	2124	MS	TX
##	2125	MO	TX
##	2126	MT	TX
##	2127	NE	TX
##	2128	NV	TX
##	2129	NH	TX
##	2130	NJ	TX
##	2131	NM	TX



##	2132	NY	TX
##	2133	NC	TX
##	2134	ND	TX
##	2135	OH	TX
##	2136	OK	TX
##	2137	OR	TX
##	2138	PA	TX
##	2139	RI	TX
##	2140	SC	TX
##	2141	SD	TX
##	2142	TN	TX
##	2143	TX	TX
##	2144	UT	TX
##	2145	VT	TX
##	2146	VA	TX
##	2147	WA	TX
##	2148	WV	TX
##	2149	WI	TX
##	2150	WY	TX
##	2151	AL	UT
##	2152	AK	UT
##	2153	AZ	UT
##	2154	AR	UT
##	2155	CA	UT
##	2156	CO	UT
##	2157	CT	UT
##	2158	DE	UT
##	2159	FL	UT
##	2160	GA	UT
##	2161	HI	UT
##	2162	ID	UT
##	2163	IL	UT
##	2164	IN	UT
##	2165	IA	UT
##	2166	KS	UT
##	2167	KY	UT
##	2168	LA	UT
##	2169	ME	UT
##	2170	MD	UT
##	2171	MA	UT
##	2172	MI	UT
##	2173	MN	UT
##	2174	MS	UT
##	2175	MO	UT
##	2176	MT	UT
##	2177	NE	UT
##	2178	NV	UT
##	2179	NH	UT
##	2180	NJ	UT
##	2181	NM	UT
##	2182	NY	UT
##	2183	NC	UT
##	2184	ND	UT
##	2185	OH	UT

##	2186	OK	UT
##	2187	OR	UT
##	2188	PA	UT
##	2189	RI	UT
##	2190	SC	UT
##	2191	SD	UT
##	2192	TN	UT
##	2193	TX	UT
##	2194	UT	UT
##	2195	VT	UT
##	2196	VA	UT
##	2197	WA	UT
##	2198	WV	UT
##	2199	WI	UT
##	2200	WY	UT
##	2201	AL	VT
##	2202	AK	VT
##	2203	AZ	VT
##	2204	AR	VT
##	2205	CA	VT
##	2206	CO	VT
##	2207	CT	VT
##	2208	DE	VT
##	2209	FL	VT
##	2210	GA	VT
##	2211	HI	VT
##	2212	ID	VT
##	2213	IL	VT
##	2214	IN	VT
##	2215	IA	VT
##	2216	KS	VT
##	2217	KY	VT
##	2218	LA	VT
##	2219	ME	VT
##	2220	MD	VT
##	2221	MA	VT
##	2222	MI	VT
##	2223	MN	VT
##	2224	MS	VT
##	2225	MO	VT
##	2226	MT	VT
##	2227	NE	VT
##	2228	NV	VT
##	2229	NH	VT
##	2230	NJ	VT
##	2231	NM	VT
##	2232	NY	VT
##	2233	NC	VT
##	2234	ND	VT
##	2235	OH	VT
##	2236	OK	VT
##	2237	OR	VT
##	2238	PA	VT
##	2239	RI	VT

##	2240	SC	VT
##	2241	SD	VT
##	2242	TN	VT
##	2243	TX	VT
##	2244	UT	VT
##	2245	VT	VT
##	2246	VA	VT
##	2247	WA	VT
##	2248	WV	VT
##	2249	WI	VT
##	2250	WY	VT
##	2251	AL	VA
##	2252	AK	VA
##	2253	AZ	VA
##	2254	AR	VA
##	2255	CA	VA
##	2256	CO	VA
##	2257	CT	VA
##	2258	DE	VA
##	2259	FL	VA
##	2260	GA	VA
##	2261	HI	VA
##	2262	ID	VA
##	2263	IL	VA
##	2264	IN	VA
##	2265	IA	VA
##	2266	KS	VA
##	2267	KY	VA
##	2268	LA	VA
##	2269	ME	VA
##	2270	MD	VA
##	2271	MA	VA
##	2272	MI	VA
##	2273	MN	VA
##	2274	MS	VA
##	2275	MO	VA
##	2276	MT	VA
##	2277	NE	VA
##	2278	NV	VA
##	2279	NH	VA
##	2280	NJ	VA
##	2281	NM	VA
##	2282	NY	VA
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##	2284	ND	VA
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##	2286	OK	VA
##	2287	OR	VA
##	2288	PA	VA
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##	2290	SC	VA
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##	2295	VT	VA
##	2296	VA	VA
##	2297	WA	VA
##	2298	WV	VA
##	2299	WI	VA
##	2300	WY	VA
##	2301	AL	WA
##	2302	AK	WA
##	2303	AZ	WA
##	2304	AR	WA
##	2305	CA	WA
##	2306	CO	WA
##	2307	CT	WA
##	2308	DE	WA
##	2309	FL	WA
##	2310	GA	WA
##	2311	HI	WA
##	2312	ID	WA
##	2313	IL	WA
##	2314	IN	WA
##	2315	IA	WA
##	2316	KS	WA
##	2317	KY	WA
##	2318	LA	WA
##	2319	ME	WA
##	2320	MD	WA
##	2321	MA	WA
##	2322	MI	WA
##	2323	MN	WA
##	2324	MS	WA
##	2325	MO	WA
##	2326	MT	WA
##	2327	NE	WA
##	2328	NV	WA
##	2329	NH	WA
##	2330	NJ	WA
##	2331	NM	WA
##	2332	NY	WA
##	2333	NC	WA
##	2334	ND	WA
##	2335	OH	WA
##	2336	OK	WA
##	2337	OR	WA
##	2338	PA	WA
##	2339	RI	WA
##	2340	SC	WA
##	2341	SD	WA
##	2342	TN	WA
##	2343	TX	WA
##	2344	UT	WA
##	2345	VT	WA
##	2346	VA	WA
##	2347	WA	WA

##	2348	WV	WA
##	2349	WI	WA
##	2350	WY	WA
##	2351	AL	WV
##	2352	AK	WV
##	2353	AZ	WV
##	2354	AR	WV
##	2355	CA	WV
##	2356	CO	WV
##	2357	CT	WV
##	2358	DE	WV
##	2359	FL	WV
##	2360	GA	WV
##	2361	HI	WV
##	2362	ID	WV
##	2363	IL	WV
##	2364	IN	WV
##	2365	IA	WV
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##	2367	KY	WV
##	2368	LA	WV
##	2369	ME	WV
##	2370	MD	WV
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##	2374	MS	WV
##	2375	MO	WV
##	2376	MT	WV
##	2377	NE	WV
##	2378	NV	WV
##	2379	NH	WV
##	2380	NJ	WV
##	2381	NM	WV
##	2382	NY	WV
##	2383	NC	WV
##	2384	ND	WV
##	2385	OH	WV
##	2386	OK	WV
##	2387	OR	WV
##	2388	PA	WV
##	2389	RI	WV
##	2390	SC	WV
##	2391	SD	WV
##	2392	TN	WV
##	2393	TX	WV
##	2394	UT	WV
##	2395	VT	WV
##	2396	VA	WV
##	2397	WA	WV
##	2398	WV	WV
##	2399	WI	WV
##	2400	WY	WV
##	2401	AL	WI

##	2402	AK	WI
##	2403	AZ	WI
##	2404	AR	WI
##	2405	CA	WI
##	2406	CO	WI
##	2407	CT	WI
##	2408	DE	WI
##	2409	FL	WI
##	2410	GA	WI
##	2411	HI	WI
##	2412	ID	WI
##	2413	IL	WI
##	2414	IN	WI
##	2415	IA	WI
##	2416	KS	WI
##	2417	KY	WI
##	2418	LA	WI
##	2419	ME	WI
##	2420	MD	WI
##	2421	MA	WI
##	2422	MI	WI
##	2423	MN	WI
##	2424	MS	WI
##	2425	MO	WI
##	2426	MT	WI
##	2427	NE	WI
##	2428	NV	WI
##	2429	NH	WI
##	2430	NJ	WI
##	2431	NM	WI
##	2432	NY	WI
##	2433	NC	WI
##	2434	ND	WI
##	2435	OH	WI
##	2436	OK	WI
##	2437	OR	WI
##	2438	PA	WI
##	2439	RI	WI
##	2440	SC	WI
##	2441	SD	WI
##	2442	TN	WI
##	2443	TX	WI
##	2444	UT	WI
##	2445	VT	WI
##	2446	VA	WI
##	2447	WA	WI
##	2448	WV	WI
##	2449	WI	WI
##	2450	WY	WI
##	2451	AL	WY
##	2452	AK	WY
##	2453	AZ	WY
##	2454	AR	WY
##	2455	CA	WY

```
## 2456 CO WY
## 2457 CT WY
## 2458 DE WY
## 2459 FL WY
## 2460 GA WY
## 2461 HI WY
## 2462 ID WY
## 2463 IL WY
## 2464 IN WY
## 2465 IA WY
## 2466 KS WY
## 2467 KY WY
## 2468 LA WY
## 2469 ME WY
## 2470 MD WY
## 2471 MA WY
## 2472 MI WY
## 2473 MN WY
## 2474 MS WY
## 2475 MO WY
## 2476 MT WY
## 2477 NE WY
## 2478 NV WY
## 2479 NH WY
## 2480 NJ WY
## 2481 NM WY
## 2482 NY WY
## 2483 NC WY
## 2484 ND WY
## 2485 OH WY
## 2486 OK WY
## 2487 OR WY
## 2488 PA WY
## 2489 RI WY
## 2490 SC WY
## 2491 SD WY
## 2492 TN WY
## 2493 TX WY
## 2494 UT WY
## 2495 VT WY
## 2496 VA WY
## 2497 WA WY
## 2498 WV WY
## 2499 WI WY
## 2500 WY WY
```

But this will not work for `grid_words`, because it's too large

```
## NULL
```

This seems to be an example of an operation better accomplished through loops, since it's computationally intensive and we don't have the resources to run a vectorized version of this script.

Since loops break up the computation, it's possible to run long processes (with interruptions, although I'm not showing that).

```

# For the purposes of this workshop, you may want to down-sample before this operation (start small)
loop_words = sample( grid_words, 100 )

# Parallel loop of loops to compare
{tic("Exhaustive similarity comparisons") # timing starts
reviews.compared =
  foreach( loop.outer = seq_along( loop_words ), .combine='rbind' ) %:%
    foreach( loop.inner = seq_along( loop_words ), .combine='c', .packages = 'dplyr' ) %do% {

      # Count words total and words in common, using set theory (fast)
      words.inner      = length( loop_words[ loop.inner ][[1]] )
      words.total      = length( union( loop_words[ loop.outer ][[1]], loop_words[ loop.inner ][[1]] ) )
      words.in.common  = length( intersect( loop_words[ loop.outer ][[1]], loop_words[ loop.inner ][[1]] ) )

      # Calculate the score
      # the.score = words.in.common / words.total words.inner
      the.score = words.in.common / words.inner

      # Return the output
      return( the.score )
    }
  }
toc()} # timing ends

```

```
## Exhaustive similarity comparisons: 6.158 sec elapsed
```

```

# DING
# beep()

```

## Benchmarks

```
10...0.1s 100...3.3s 500...73.7s 1k...321s 2k...1252s
```

---

## Coding Challenge B

The second challenge involves finding the mean `$Rating` of each city/state combination. Create an outer loop that lists the cities for which there are reviews, then create an inner loop that calculates the mean `$Rating` on each iteration. Here's some meta-code to get you started thinking about this problem:

```
Outer Loop foreach(state) %:% city_list = unique cities in state
```

```
Inner Loop foreach(city) %dopar% calculate the mean review$Rating
```

Good luck!

```
write_rds( reviews, "./data/snapshot_b.rds" ) # in coding_challenges you will load this file
```

---



## Bootstrapping

So far the examples have focused on `for` loops iterating over lists and dataframes, but this is not the whole story. For many purposes, you might want to use a `while` loop, which terminates when/while certain conditions are being met.

Another common use for loops is bootstrapping, which involves running a process for a specified number of times. For example, we can run the same model 1,000 times over different subsets of the same data and compare the results...

```
# Let's time our bootstrap loop
{
  tic("Bootstrapping a linear model n times")

  bootstrap = foreach( icount( 1000 ), # how many loops?
    .combine = rbind, # each run of the model will generate 1 row
    .packages = c ("dplyr", "lme4")) %dopar% { # parallel mode
    #.packages = c ("dplyr", "lme4")) %do% { # non-parallel mode

      # Sample some random rows to make things run faster
      sample = reviews %>% sample_n(nrow(reviews) * 0.33, replace = FALSE)

      # Fit the model predicting sentiment based on length and rating
      model = lm( data = sample,
                  formula = unlist(Rating) ~ unlist(nchar) )

      # Extract model coefficients
      coefs = coefficients( model )

      # Output
      return( coefs )
    }

  toc()
}
```

```
## Bootstrapping a linear model n times: 42.816 sec elapsed
```

```
# We can get the mean of each coefficient as follows
colMeans( bootstrap )
```

```
## (Intercept) unlist(nchar)
## 3.477841008 -0.002841132
```

The parallel version typically runs 33% faster for me.

---

## Coding Challenge C

For this challenge, you will be bootstrapping a model of review sentiment, based on a quick text analysis that results in each review being assigned a score based on how positive or negative, according to an unsupervised

algorithm. We will be trying to predict the sentiment based on the rating and review length. Here's some meta-code to get you thinking about this problem:

```
Sentiment data generation sentimentr( review_text )  
Loop a linear model of sentiment lm( sentiment ~ rating + length )  
Abstract over the results colMeans(bootstrapped_results)
```

---

## Part 6: Final thoughts

I don't often write loops, but when I do, they run in parallel.

To get the most out of parallel loops, you will need to have access to as many cores (processors) as possible. Consider using remote computing resources, such as Quest (Northwestern). For the best performance, use vectorized code whenever possible, such as the functions provided in `tidyverse` and `data.table`, rather than loops.

That said, many processes are easy to write as loops because the code runs sequentially, which is easy to conceptualize. Rewriting loops using vectorized functions essentially gives you more power as a researcher, because it unlocks the computing power you already have. More importantly, *it saves your time!*

Happy looping looping looping!

### One last step

Best practice would be to close the script with stopping the cluster...

```
# This severs the connection between R and the 'extra' cores we were leveraging.  
stopCluster( cl )
```