Lecture 9: Cluster computing

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BIOST 561: Computational Skills For Biostatistics I

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In Lecture 8, you learned how to compute without R.

Today, you'll learn how to transfer those skills to computing on a cluster.

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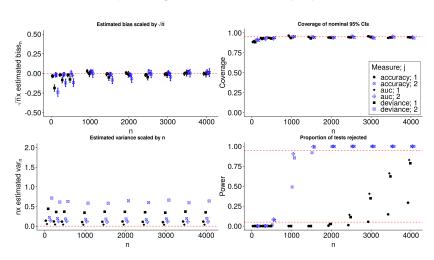
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Data analysis:

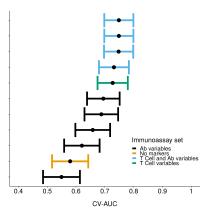
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All markers	0.748 [0.697, 0.799]
T Cells + Fx Ab	0.732 [0.680, 0.784]
T Cells	0.727 [0.675, 0.780]
lgG3	0.695 [0.638, 0.751]
IgG + IgA + IgG3	0.688 [0.629, 0.746]
IgG + IgA + IgG3 + Fx Ab	0.658 [0.598, 0.718]
IgG + IgA	0.620 [0.559, 0.682]
No markers	0.580 [0.517, 0.643]
Fx Ab	0.550 [0.485, 0.614]



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analyses without keeping an R session open (for the duration!) on my personal machine?

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*: we will discuss this further!

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We will use the cluster to do this!

Part I: coding for the cluster

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Today, I'll provide an alternative method (this should not always replace the simulator!).

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Modular code: each file has a single task

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Modular code: each file has a single task

Your turn!

Go to the code subdirectory. Then answer these questions with a partner:

- 1. what does do_one do? What are its arguments?
- 2. what does generate_data do? What are its arguments?
- 3. Write effective comments in each file so that **future you** understands each!

Coding for the cluster: modular code

Answers and Brian's comments (to be posted after class)

Coding for the cluster: setting the seed

Coding for the cluster: #! and executables

Coding for the cluster: saving output

Coding for the cluster: debugging

Coding for the cluster: compiling output

Part II: using the cluster

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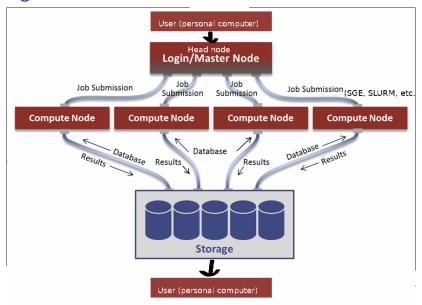
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- a head node (where you are)
- compute nodes (where your code gets run)
- submission system (how your code gets run)
- *other groups have similar resources, e.g.,
 - hyak (managed by UW-IT)
 - pearson (statistical genetics group only)
 - gizmo (Fred Hutchinson Cancer Research Center only)
 - Microsoft Azure, Amazon Web Services (AWS)

Using the cluster



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We'll practice good habits for being nice as we go along.

Using the cluster: logging in (Windows, incl. box)

- 1. Open TeraTerm [or your favorite secure shell (SSH) client]
- 2. Enter the address of your favorite cluster, e.g., bayes.biostat.washington.edu
- 3. Make sure that the "New connection" window is filled out as in the figure, except for "Host" (screenshot from box)
- 4. Click OK, enter your UW BIOST username and password when prompted (user and pass you use for box)



Using the cluster: logging in (Mac/Linux)

- 1. Open a Terminal window
- 2. Type ssh mynetid@cluster.washington.edu
 - replace mynetid with Your UW NetID, and
 - replace cluster with Your cluster, e.g., bayes
- Enter password (same password as box) when prompted (the field will remain blank but your password will be received)

Using the cluster: logging in

Your turn! Take 2 minutes to complete this activity **by yourself** (if you don't have a computer, write down how you would do this).

- 1. Log into the cluster
- 2. What is the name of the directory that you are when you log in?
- Create a directory called robust_ses in your home directory
- Create a directory called robust_ses on your computer, under biost561/lecture9

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- shell scripts

are all used in exactly the same way!

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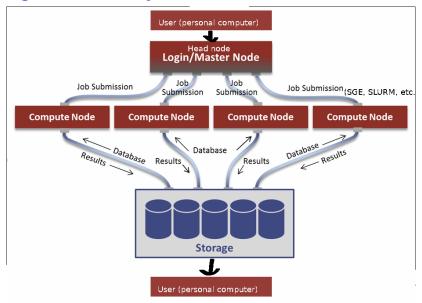
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Example job:



Using the cluster: submitting jobs

Workhorse command on SGE: qsub

Many options, including:

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Many options, including:

Options are specified with a single -, as you saw in Lecture 8.

Using the cluster: checking jobs

Using the cluster: other helpful SGE commands

Appendix: other useful cluster things

Useful things: Windows file compatability