

Evening session: Introduction to Cloud and GPU computing

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Cloud computing

Why?

1. Access more (CPUs, memory, storage) or different (e.g., GPUs) compute resources than easily available locally
2. Run workflows that apply the same steps (e.g., bulk RNAseq (pseudo)-alignment; scRNAseq preprocessing) across many different samples
3. Access ‘consortium’ data already stored in the cloud, perhaps requiring authentication, e.g., GTEx
4. Exploit novel cloud-based services, e.g., Google Bigtable (highly scalable database)

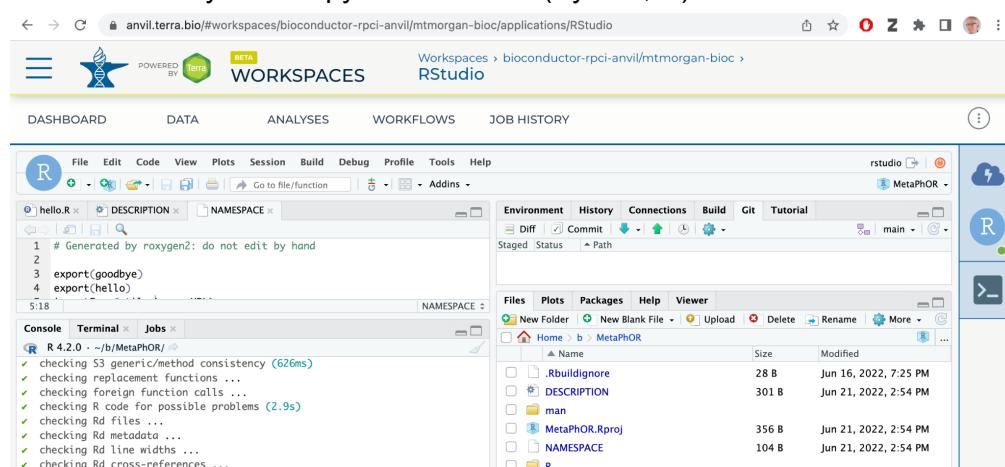
What?

Do-it-yourself

- Google, Amazon, or Azure (Microsoft) cloud resources
- (Virtually) unlimited access to compute power – CPUs, memory, storage
- Highly flexible, but requires familiarity with cloud provider tools for managing resources

Pre-configured

- E.g., In the US, [Seven Bridges](#) or NIH NHGRI [AnVIL](#)
- A ‘higher level’ interface requiring less technical knowledge of cloud provider tools for resource management
- Emphasizing particular use cases
 - Interactive analysis in Jupyter notebooks (Python, R) or RStudio



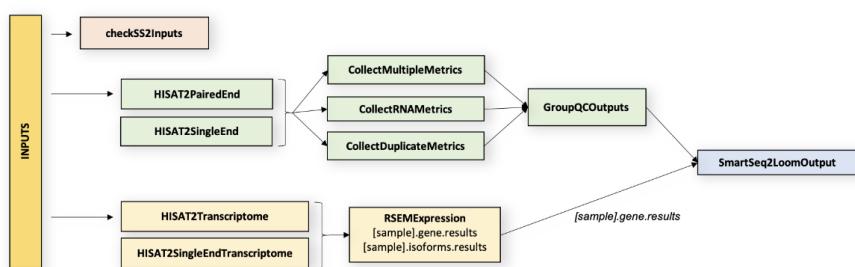
- Large-scale workflows described using Workflow Description Language
 - RNAseq pseudo-alignment (Kayla Interdonato)

```

19 task salmon_quant {
20   File fastq1
21   File fastq2
22   File transcriptome_index
23   String transcriptome_index_name
24
25   String quant_out = basename(fastq1, ".fastq.gz")
26
27   command {
28     tar -xvzf ${transcriptome_index}
29     salmon quant -i ${transcriptome_index_name} -l A \
30       -1 ${fastq1} \
31       -2 ${fastq2} \
32       -p 8 --validateMappings -o ${quant_out}
33     tar -cvzf ${quant_out}.tar.gz ${quant_out}
34   }
35
36   runtime {
37     docker: "combinelab/salmon:1.3.0"
38   }

```

- Human Cell Atlas [Smart-seq2 single-cell pipeline](#)



- ‘Single sign on’, including access to restricted data resources (if appropriate!)

How? – AnVIL

Google account + credit card

- Yes, cloud computing costs money, and a payment method has to be established!
- Not likely your credit card, but perhaps tied to an account associated with your institution & grant

Use of unrestricted data

- ‘Workspace’ defining a project
- ‘Cloud environment’ describing resources for interactive computation – CPU,

The screenshot shows the 'RStudio Cloud Environment' configuration interface. At the top, it displays cost information: Running cloud compute cost (\$0.25 per hr), Paused cloud compute cost (\$0.01 per hr), and Persistent disk cost (\$2.00 per month). Below this, the 'Application configuration' section lists installed software (RStudio (R 4.2.0, Bioconductor 3.15, Python 3.8.10)) and provides update and version details (Updated: May 24, 2022, Version: 3.15.1). The 'Cloud compute profile' section allows setting CPU (4) and Memory (26 GB) configurations, with an option to enable GPUs (Beta). A note at the bottom indicates memory, disk space, and workflow support.

- memory, disk space
- ‘Workflows’ for large-scale computation
- ‘Bucket’ for storing static (e.g., FASTQ) data

Use of restricted data

- E.g., suppose dbGaP has provided me with access to GTEx data
- Link my AnVIL account with dbGaP
- My AnVIL account grants access to cloud-based resources associated GTEx
- Restrictions on use may be imposed, e.g., data must remain in AnVIL environment

Benefits

Flexibility

- Today I need a very large computer for my interactive analysis, but tomorrow I do not

- For this project, one step involves summarizing 100's of FASTQ files from a single cell experiment to count matrices; I want to do this summary quickly so that I can get to the more interesting biological questions in an interactive analysis
- For some data sets, restricted access is only available through AnVIL or other clouds

Independence

- To some extent, we are free from the constraints of our local system administrators and IT departments – to get a bigger computer or submit a large job, we just start a larger instance or launch a workflow with appropriate resources.
- ‘Someone else’ (usually an expert) has taken care of technical things, e.g., the ability to install any Bioconductor package; configuration of salmon for pseudo-alignment.

Costs

- General purpose compute nodes (e.g., 4 cores, 32 MB memory) are not expensive, and are only billed while in use. Useful in workflows and interactive analysis
- Data movement (e.g., from the bucket where the data is stored to the disk of the compute node) is inexpensive and very fast.

Challenges

New concepts

- Buckets for data storage
- Formal workflow description languages
- Understanding cloud provider resources, e.g., billing, security, ...

Costs

- Cloud providers are making a lot of money; as a corollary, we’re paying a lot for cloud services!
- Cost control – unintentional consumption of a large amount of resources, either accidentally (oops, I meant to submit 100 FASTQ files to a workflow, not the same file 100 times!) or because the resources are too easily accessible (yes, I want to do 500,000 simulations; no, I didn’t realize that would cost \$60,000).
- Data ingress (uploading data to the cloud) is usually ‘free’, but data egress is costly – vendor lock-in

Bioconductor resources (examples)

- [AnVIL](#) package for working with AnVIL & cloud resources
- [hca](#) and [cellxgenedp](#) packages for Human Cell Atlas data access (not really restricted to the cloud, but data access is particularly fast on the cloud)
- [Rcwl](#) for workflow management within R
- Bioconductor [docker images](#) for easy deployment to cloud environments

Intro to GPUs

What is a GPU?

- The graphics processing unit, or GPU, has become one of the most important types of computing technology to dramatically accelerate workloads in high-performance computing (HPC), deep learning, and more.
- Designed for parallel processing
- GPUs were originally designed to accelerate the rendering of 3D graphics
- While CPUs have continued to deliver performance increases through architectural innovations, faster clock speeds, and the addition of cores, GPUs are specifically designed to accelerate computer graphics workloads.
- GPUs are more programmable than ever before, affording them the flexibility to accelerate a broad range of applications that go well beyond traditional graphics rendering.

GPU for Machine learning

- Some of the most exciting applications for GPU technology involve AI and machine learning.
- GPUs incorporate an extraordinary amount of computational capability, they can deliver incredible acceleration in workloads that take advantage of the highly parallel nature of GPUs, such as image recognition.
- Many of today's deep learning technologies rely on GPUs working in conjunction with CPUs.

GPGPU - General Purpose GPU

Applications in Bioinformatics -

https://en.wikipedia.org/wiki/General-purpose_computing_on_graphics_processing_units#Bioinformatics

Packages in Bioconductor using deep learning

- These packages can be sped up using GPUs

ttgsea - Tokenizing Text of Gene Set Enrichment Analysis

DeepPINCS - Protein Interactions and Networks with Compounds based on Sequences using Deep Learning

VAExprs - Generating Samples of Gene Expression Data with Variational Autoencoders

GenProSeq - Generating Protein Sequences with Deep Generative Models

Using GPU on AnVIL

- Choose the cloud environment with 'Enable GPU' option
- Choose the GPGPU type you'd like to use
 - Choose the number of GPUs
- Start the environment

Cloud Environment

X

A cloud environment consists of application configuration, cloud compute and persistent disk(s).

Running cloud compute cost

\$2.18 per hr

Paused cloud compute cost

\$0.01 per hr

Persistent disk cost

\$2.00 per month

Application configuration i

R / Bioconductor: (Python 3.7.12, R 4.1.3, Bioconductor 3.14, tidyverse 1...

What's installed on this environment?

Updated: May 31, 2022

Version: 2.1.3 

Cloud compute profile

CPUs

16 

Memory (GB)

60 

Enable GPUs BETA

[Learn more about GPU cost and restrictions.](#) 

GPU type

NVIDIA Tesla T4 

GPUs

4 

Startup script

URI

Compute type

Standard VM 

Enable autopause

[Learn more about autopause.](#) 

30

minutes of inactivity

Location BETA



Select... 