



*Translational Data Deep Dive*

# **Building Computational Workflows**

March 11th, 2026, DaSL Lounge and Teams

# We are recording...



# What is a Deep Dive?

A hands-on, interactive quarterly series to help build practical data science skills at Fred Hutch

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## Learn Concepts

*Understand the **what**, **why**, and **when** behind data science approaches.*



## See the Tools in Action

*Live demonstrations using Hutch-supported tools and infrastructure.*



## Ask & Engage

*These sessions are interactive ask questions in person or online using Slido.*



## Get your hands dirty!

*Try things yourself during the co-working time*

# Who we are?

*And why should we be teaching you about computational workflows?*



**Sita:** A geneticist by training, a computational biologist by choice, I have used WDL's for bulk genomics and single-cell analysis.

**Taylor Firman:** A bioinformatics software engineer that has used Bash/SLURM/Python in biophysics & genomics, but is much happier having discovered WDL and Docker.

**Emma Bishop:** A bioinformatician who learned workflows to run single-cell pipelines in the cloud and now uses WDL instead of bash scripts to analyze T cells.

# Who else is here to help today

## Scientific Computing

- Dan Tenenbaum
- Michael Gutteridge

## **PROOF Workbench** *(A website to run WDL workflows on the Hutch cluster)*

- Scott Chamberlain
- Sean Kross

## Training Team

- Chris Lo
- Ted Laderas

## Cirro

- Sam Minot

# Before we begin...

*Some housekeeping to make this session work for everyone*

- You can find the materials associated with this Deep Dive
  - [Deep Dive Page](#)
  - [GitHub](#)
- We will be using [Slido](#)
  - Just go to slido.com and enter this code **#8758973**
  - More details on how to use Slido here
- We follow the Training Teams [DaSL Participation guidelines](#)



# Learning Objectives



## Recognize When Workflows Are Useful

Learn *when* a computational workflow is the right solution for your analysis



## Understand the Structure of a WDL

Learn how to read and interpret the key components of **WDL**, a language to write workflows in



## Don't Build Workflows from Scratch

Know about how you can leverage resources like the **WILDS WDL Library** to assemble your own workflow



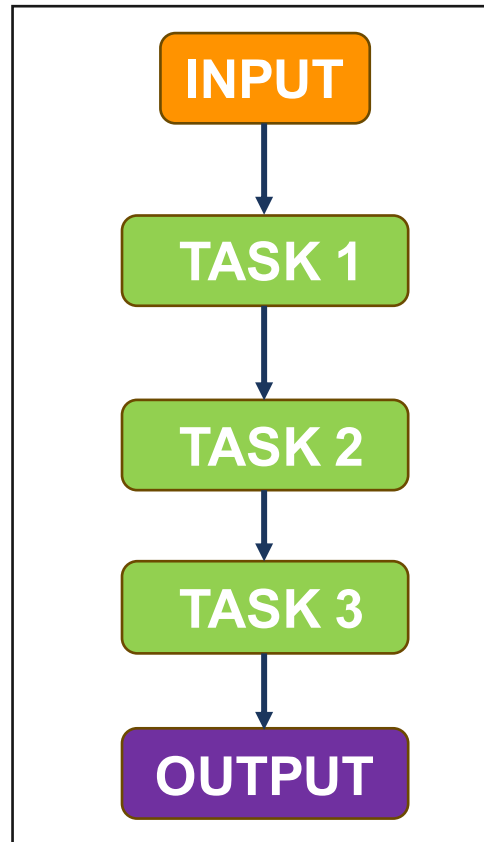
## Learn how to run workflows at Fred Hutch

See how workflows can be run using **PROOF Workbench** and Hutch infrastructure

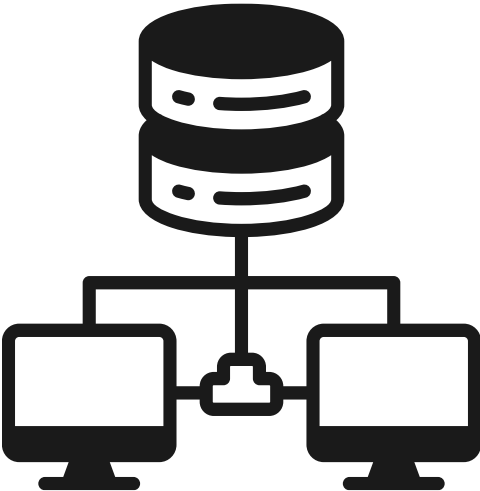
# What is a Workflow?

A workflow describes all the steps needed to turn input data into results.

## WORKFLOW

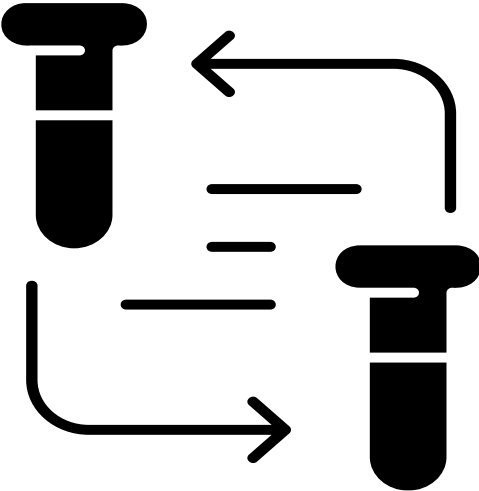


# What are the advantages of writing and running a workflow?



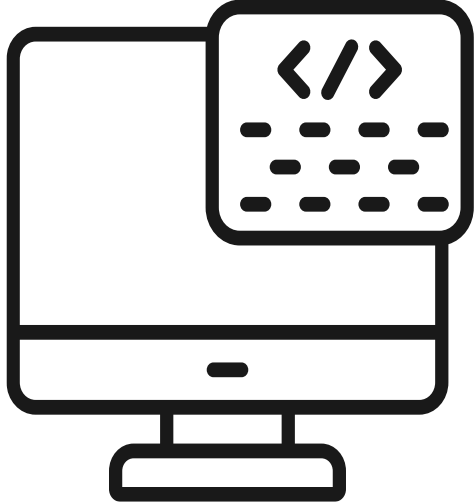
**Scalable**

Many samples



**Reproducible**

Defined steps  
↓  
Same results



**Shareable**

Reusable across teams

# When to use a workflow?

<b>Make a Workflow When...</b>	<b>DON'T Make One When...</b>
<b>It's repetitive</b>	<b>You're still figuring things out</b>
<b>It's multi-step</b>	<b>It's exploratory</b>
<b>It's multi-sample</b>	<b>It's tiny</b>
<b>It needs to be reproducible</b>	<b>It's temporary</b>

# So how do you write a workflow?

There are special workflow languages (code) in which you can give instructions... we like



## **Readable**

*workflows look like step-by-step instructions*

## **Modular**

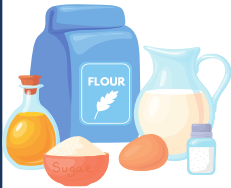
*tools are separated into reusable tasks*


## **Built for genomics pipelines**


*designed for large bioinformatics analyses*

# How do I get started with WDL workflows?

You will usually need 3 files

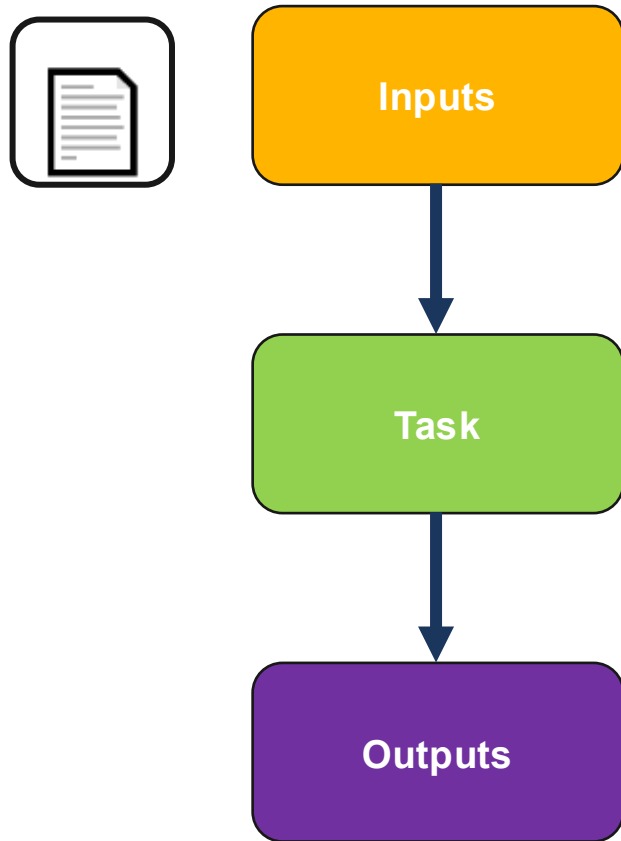
 <h2>Ingredients</h2> <p><i>What data to use</i></p>
<code>inputs.json</code>
Specifies <b>inputs</b>
File paths (FASTQs, BAMs, etc.)
Defines any input values for a task

 <h2>Equipment</h2> <p><i>What resources to use</i></p>
<code>options.json</code>
How many times a task should retry if it fails
Should you reuse previous results
Where outputs are saved

 <h2>The Recipe</h2> <p><i>What tasks to run</i></p>
<code>workflow.wdl</code>
Defines the tasks
Connects inputs → tasks → outputs
Declares dependencies between steps

# hello\_world Workflow

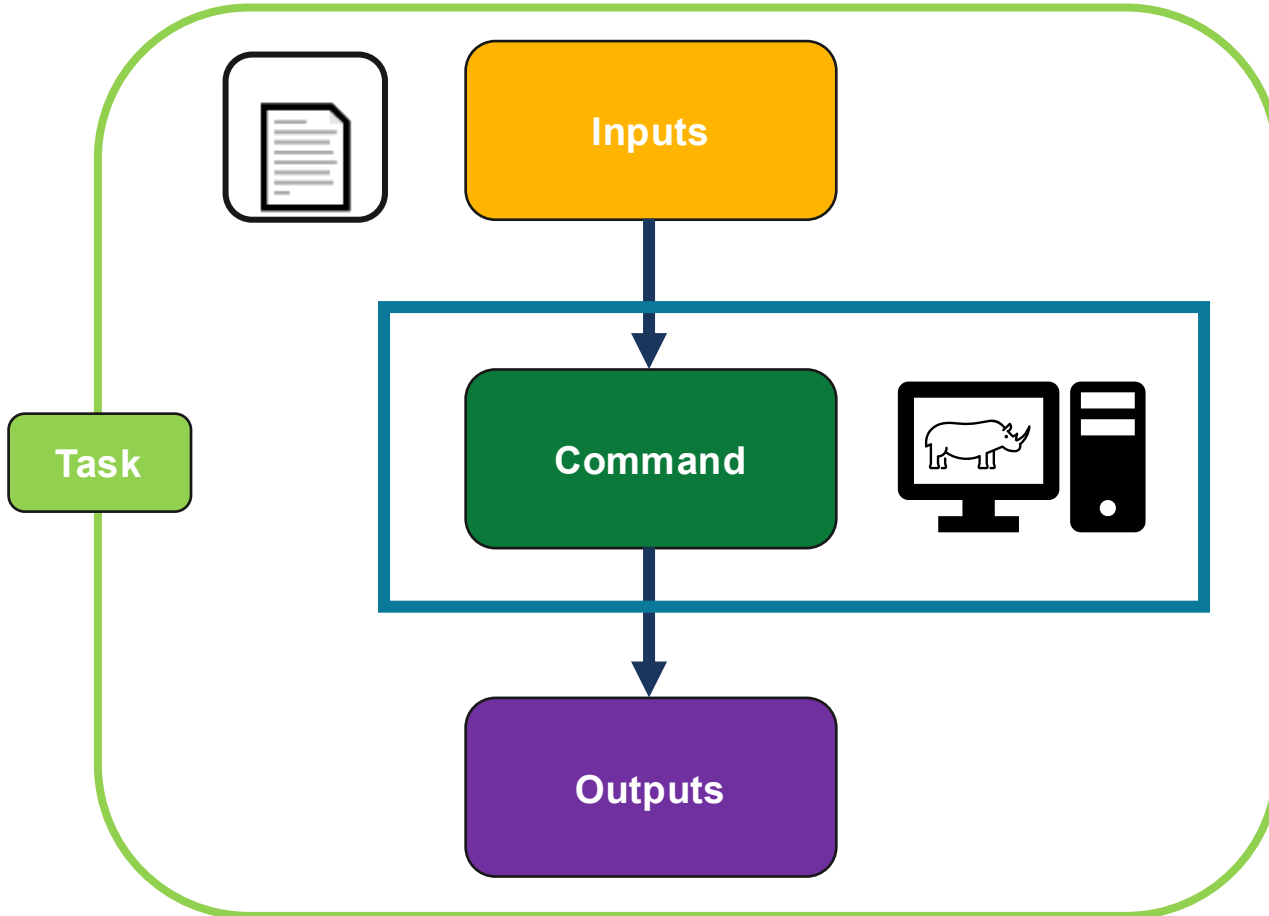
Human-readable overview of what's happening



```
workflow hello world {  
  meta {  
    description: "Writes a personalized greeting"  
    outputs: {  
      out_file: "A file containing a greeting"  
    }  
  }  
  
  parameter_meta {  
    your_name: "Name to greet"  
    your_number: "Your favorite number"  
  }  
  
  input {  
    String your_name = "Fred"  
    Int your_number = 29  
  }  
  
  call greeting {  
    input:  
      name = your_name,  
      number = your_number  
  }  
  
  output {  
    File out_file = greeting.out  
  }  
}
```

# “Hello World” Task: greeting

Very similar vibes, but at the “building-block” task level:



Fred Hutch Cancer Center

```
task greeting {
  meta { ...
  }

  parameter_meta { ...
  }

  input {
    String name
    Int number
  }

  command <<<
    echo "Hello, ~{name}!" > "~{name}.txt"
    echo "Your favorite number is ~{number}" \
    >> "~{name}.txt"
  >>>

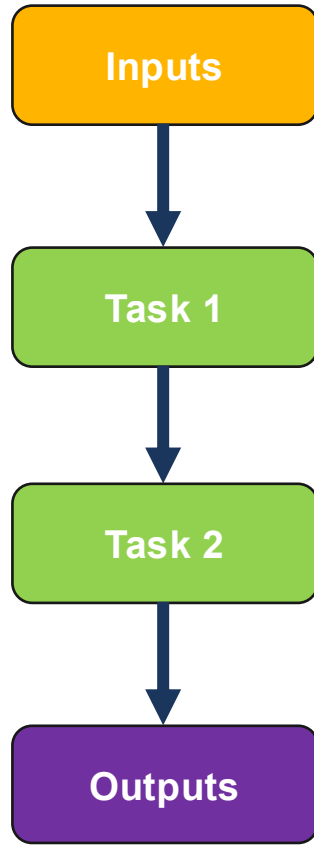
  output {
    File out = "~{name}.txt"
  }

  runtime {
    docker: "getwilds/awscli:2.27.49"
    memory: "2 GB"
    cpu: 1
  }
}
```

# A real-world example: ww-sra-salmon

Benefits of WDL become clear at scale...

- 1 Specify what data I'm interested in
- 2 Download data from source
- 3 Post-process as necessary
- 4 Save final results locally



```
import modules/ww-sra/ww-sra.wdl as sra_tasks
import modules/ww-salmon/ww-salmon.wdl as salmon_tasks

workflow sra_salmon {
  input {
    Array[String] sra_id_list
    File salmon_index_zip
  }

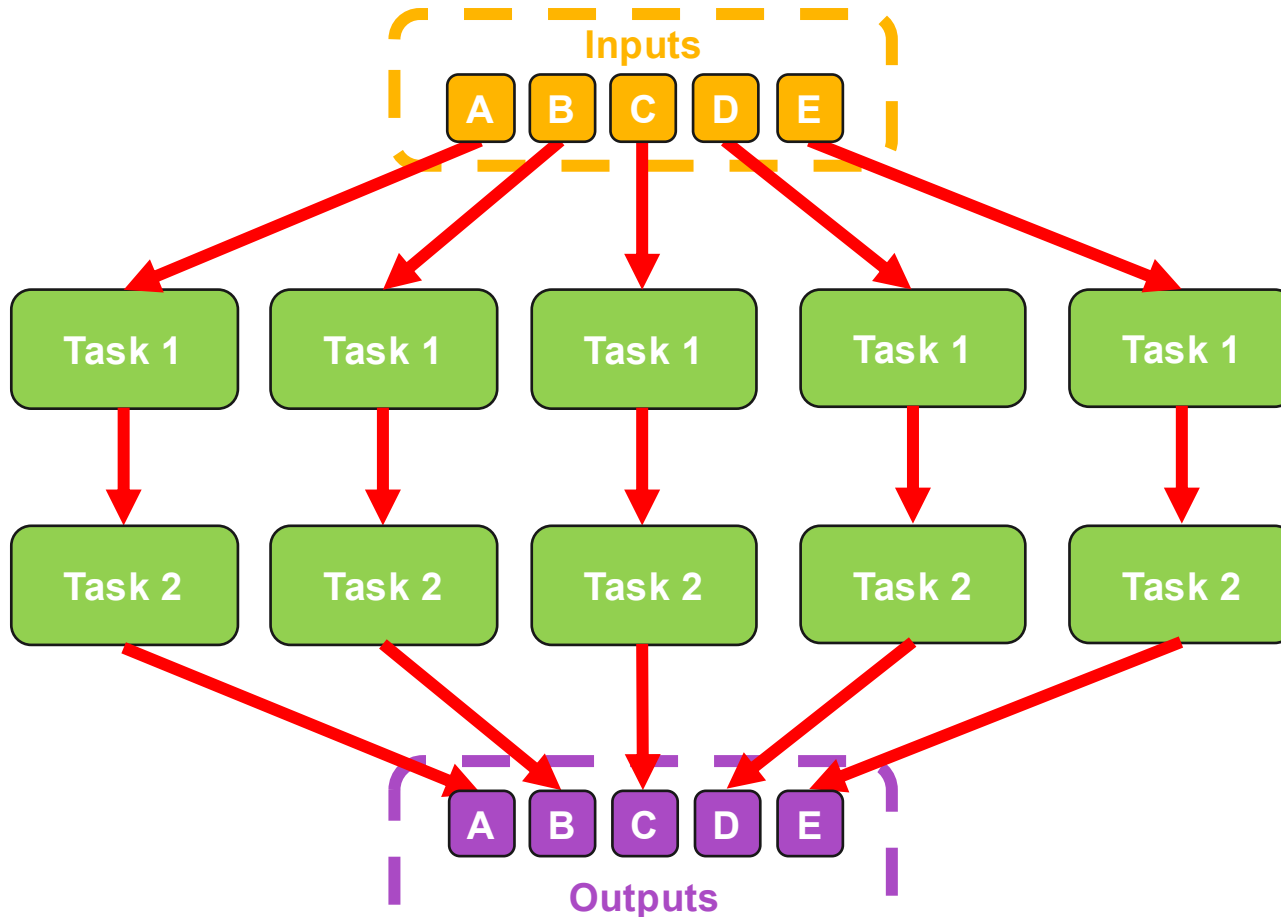
  # Download FASTQ files from SRA
  call sra_tasks.fastqdump { input: sra_id = id }

  # Quantify each sample with Salmon
  call salmon_tasks.quantify { input:
    salmon_index_dir = salmon_index_zip,
    fastq_r1 = fastqdump.r1_end,
    fastq_r2 = fastqdump.r2_end,
  }

  output {
    Array[File] quant_dirs = quantify.salmon_quant_dir
  }
}
```

# A real-world example: ww-sra-salmon

Benefits of WDL become clear at scale...



Fred Hutch Cancer Center

```
import modules/ww-sra/ww-sra.wdl as sra_tasks
import modules/ww-salmon/ww-salmon.wdl as salmon_tasks

workflow sra_salmon {
  input {
    Array[String] sra_id_list
    File salmon_index_zip
  }

  # Scatter across nodes for max parallelization
  scatter ( id in sra_id_list ){

    # Download FASTQ files from SRA
    call sra_tasks.fastqdump { input: sra_id = id }



    # Quantify each sample with Salmon
    call salmon_tasks.quantify { input:
      salmon_index_dir = salmon_index_zip,
      fastq_r1 = fastqdump.r1_end,
      fastq_r2 = fastqdump.r2_end,
    }

  }

  output {
    Array[File] quant_dirs = quantify.salmon_quant_dir
  }
}
```

# Customize the workflows as needed

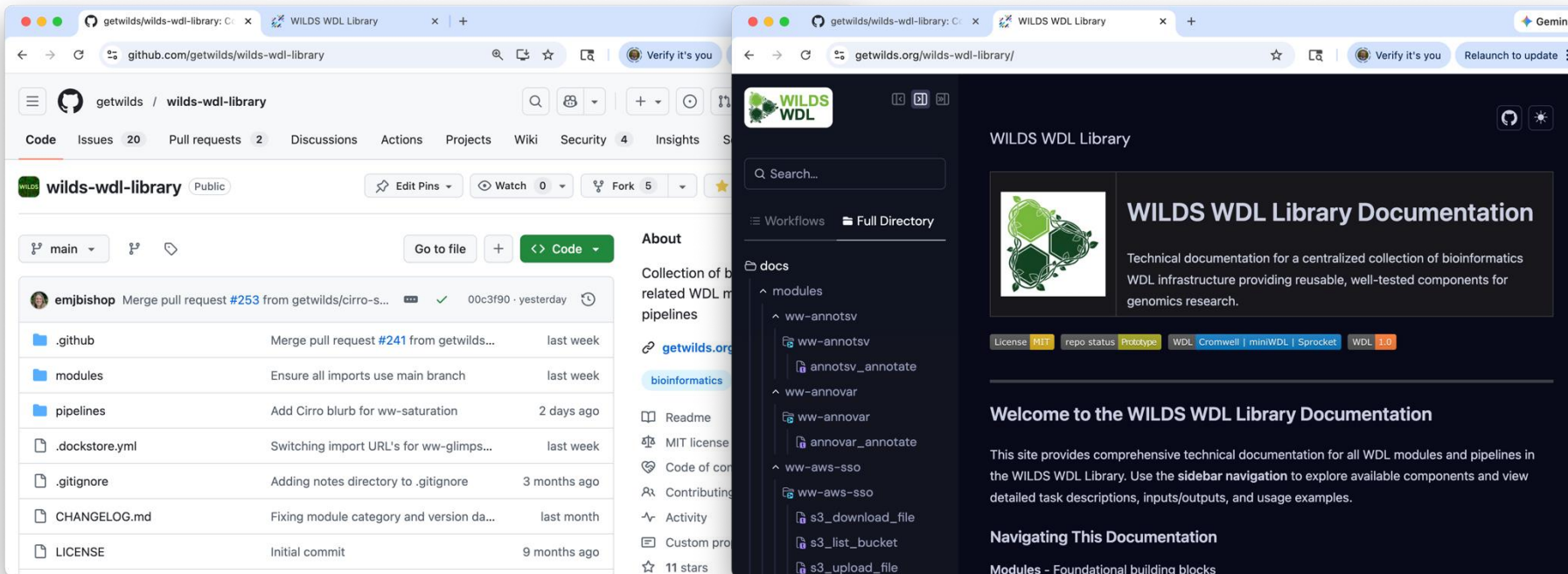
Straightforward to customize a workflow to suit your research needs

I want to...	Solution	Bioinformatics Example
...provide my own datasets	Update the inputs!	Array[File] → [file1.txt, ...]
...use a different tool	Swap out the module import!	ww-salmon → ww-star
...add an extra tool	Import another module!	RNAseq → ww-deseq2
...save my results to S3	There's a module for that!!	ww-aws-sso (Upload & Download)
...make a new module	We'll help you build one!!!	 WILDS WDL Library WILDS Docker Library 

# Easiest way to write a WDL: use someone else's!

Plenty of existing WDL workflows out there... including right here at Fred Hutch!

- WILDS WDL Library: curated collection of pre-built, well-tested, well-documented workflows
- We can help you build one via the WILDS WDL Development Program!!



More details here!!!



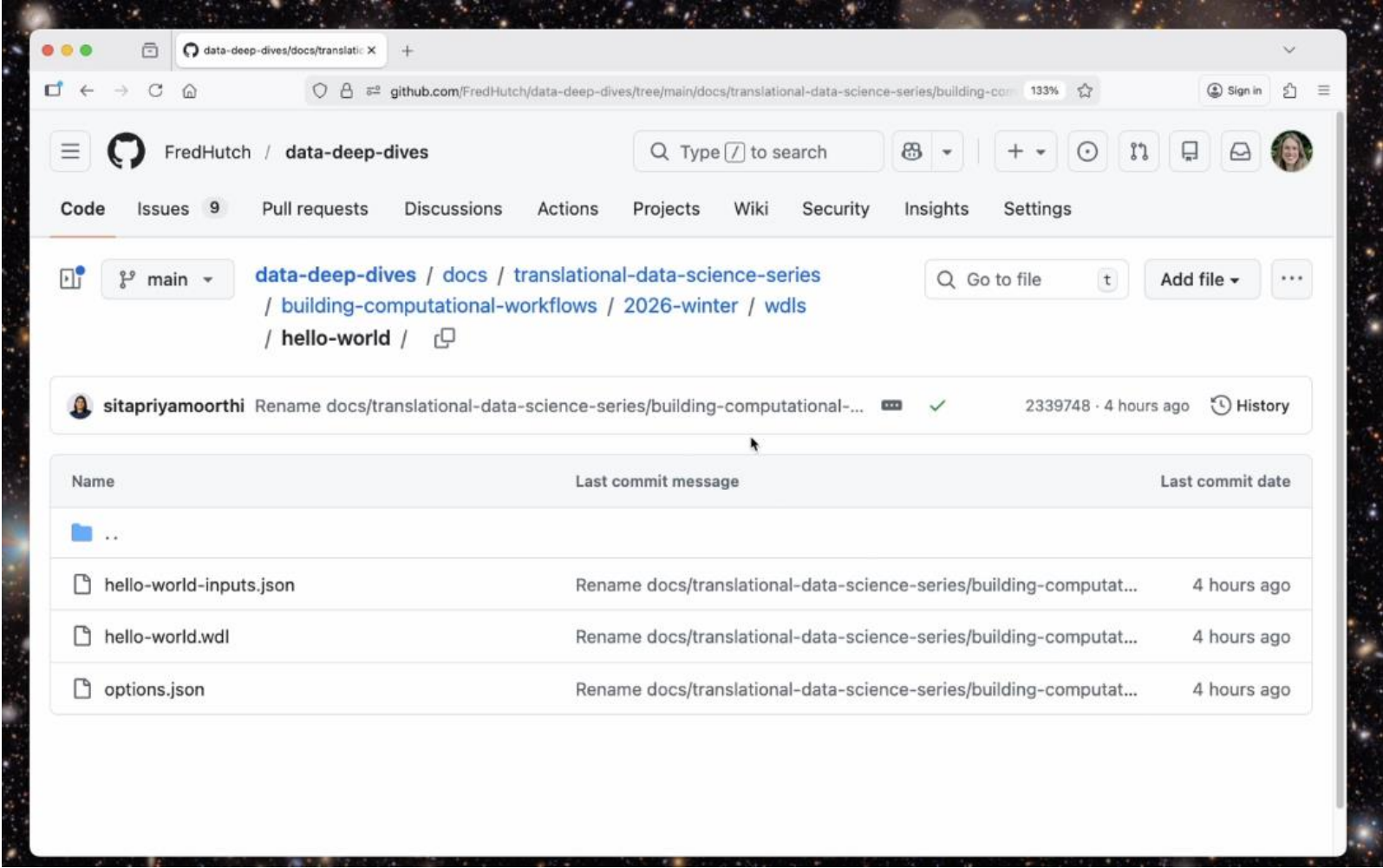
## Other Resources

[Broad's WARP](#)  
[GATK Workflows](#)  
[BioWDL](#)  
[Dockstore](#)

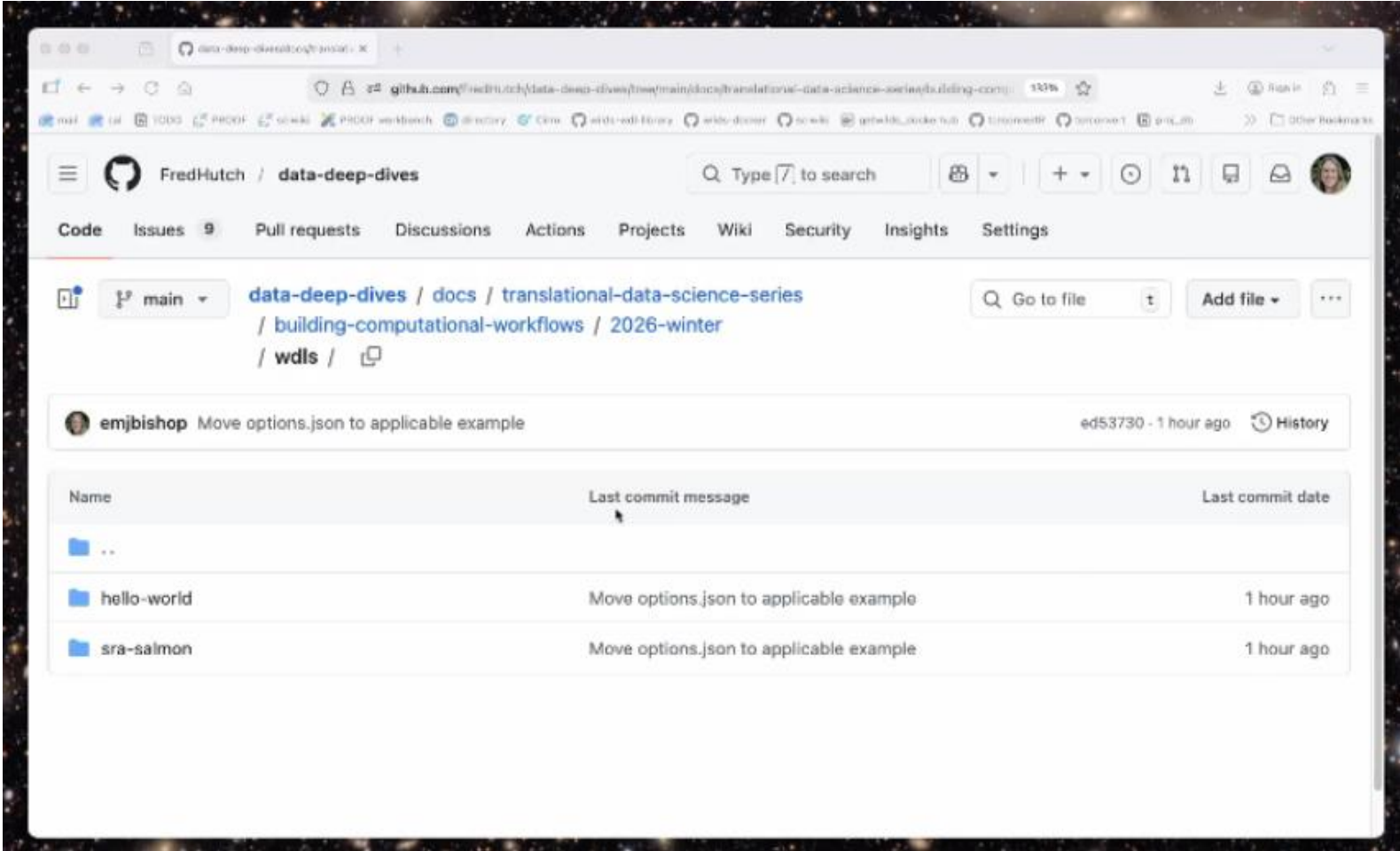


**Let's see workflows in action!**

# Demo: hello-world.wdl



# Demo: sra-salmon.wdl

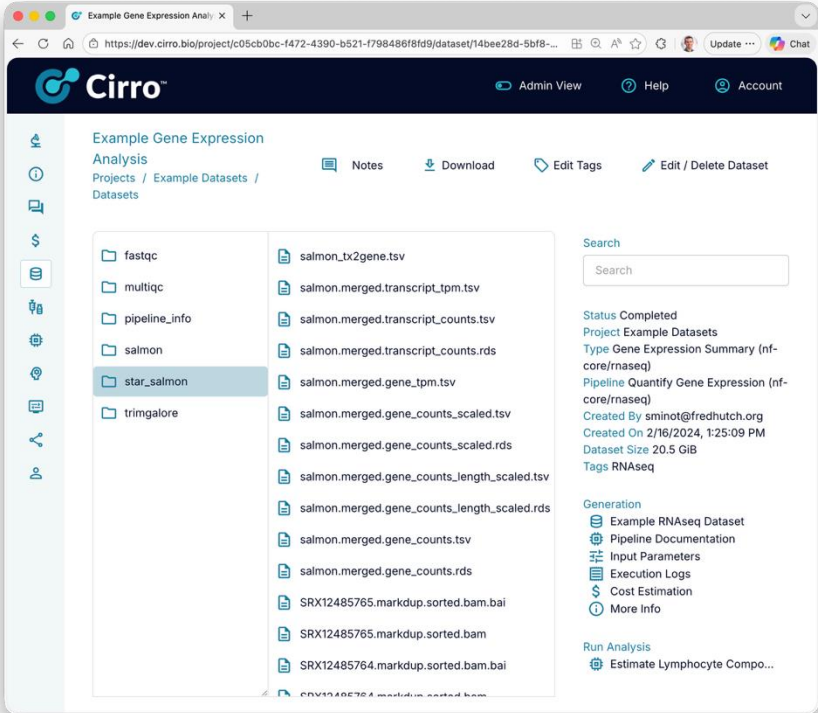
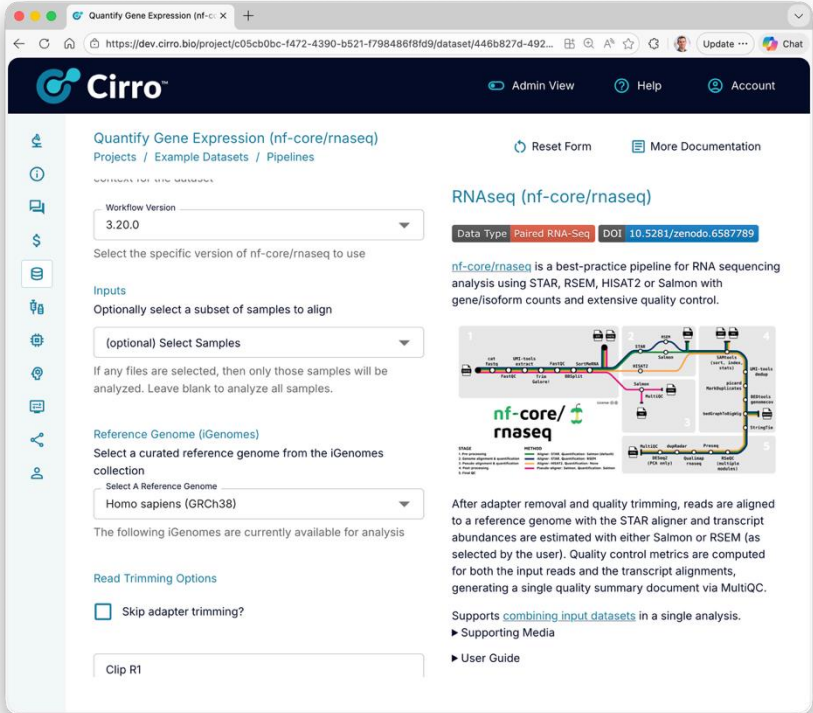
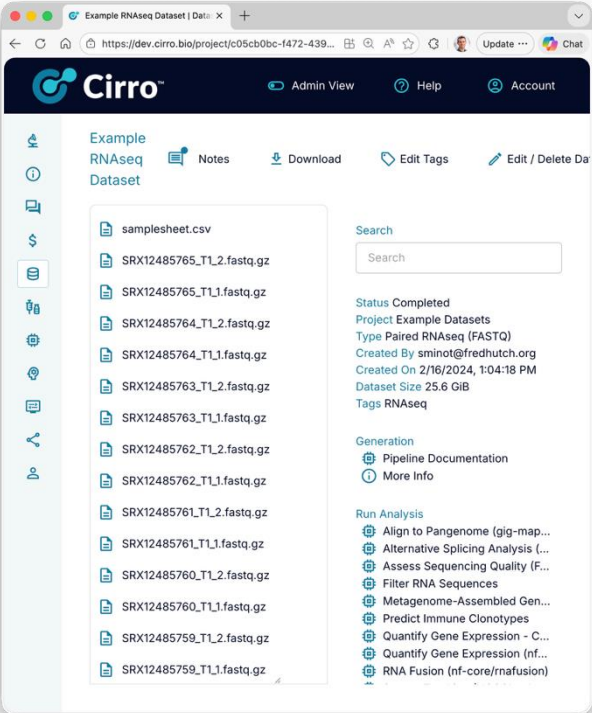


# Running WDLs at Fred Hutch

Tool	Interface	Can run	Good for
<b>Sprocket</b>	Command line	Any WDL	Development, running workflows manually
<b>PROOF Workbench</b>	Website	Any WDL	Running at scale, validating before moving to the cloud
<b>Cirro</b>	Website	Cirro WDLs	Using established pipelines

**Also popular: miniWDL, Terra, AnVIL**

# Deploying Workflows in the Cirro Data Platform



1. Ingest Raw Data

2. User Friendly Pipeline Execution

3. Pipeline Outputs + Provenance

# Tips for Success

- **Test then scale:** Start with a few samples first
- **Spot-check outputs:** Ensure they look reasonable
- **Check workflow status:** Confirm it started and didn't fail
- **Read logs to troubleshoot:** Scan for error messages to get context

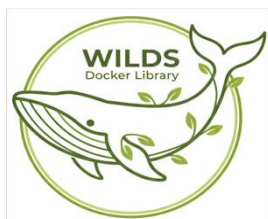
 Reach out on Slack, we've probably seen your issue before

## Fred Hutch WDL Resources



### [WILDS WDL Library](#)

Validated WDLs



### [WILDS Docker Library](#)

Validated Docker images



### [PROOF Workbench](#)

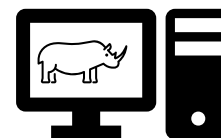
Run WDLs on the cluster through a website

## Fred Hutch Data Science Help



### [FH-Data Slack](#)

Crowdsourcing, Q&A



### [SciComp](#)

Get help with the cluster, the cloud, data storage



### [SciWiki](#)

Useful knowledgebase



### [Data House Calls](#)

Short meetings for personalized help

# Data Science Lab Training Spring Quarter, 2026

## Courses

- Bash for Bioinformatics\*
- Machine Learning for Python\*(new!)
- Bioconductor for Genomics\*
- Intro to R

*\*has prerequisites*

## Workshops

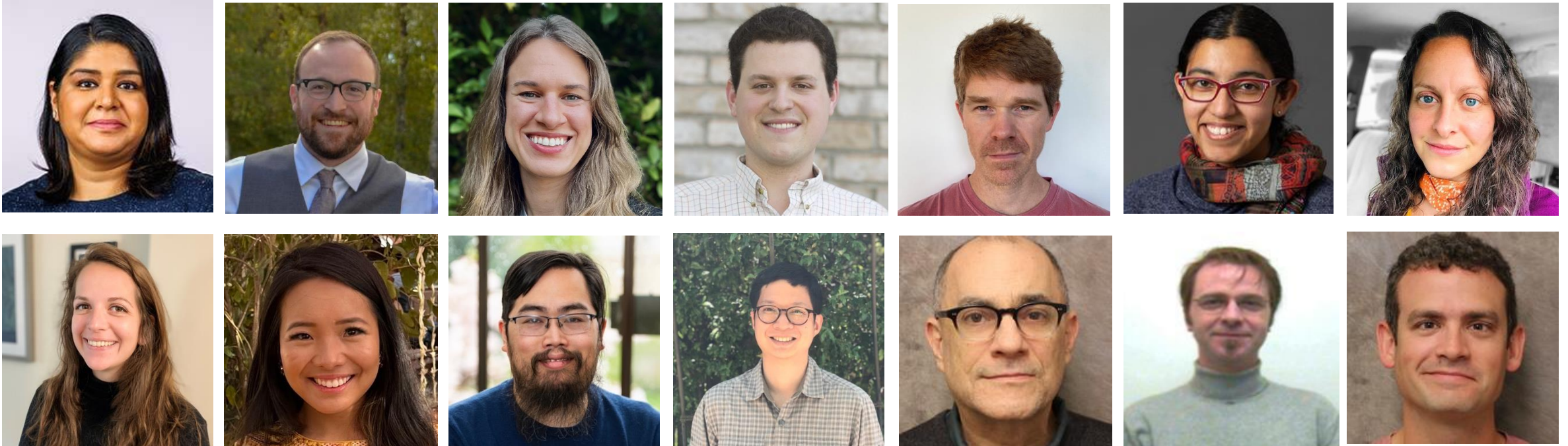
- Better Tables
- Intro to Command Line
- Intro to Cluster Computing

Registration opens soon!  
**Quarter begins April 28**

For more info: *send a blank email to* [dasltraining-join@lists.fhcrc.org](mailto:dasltraining-join@lists.fhcrc.org)

# Thank you to everyone who made this Deep Dive possible!

And thank you all for participating!!



Fred Hutch Office of the Chief Data Officer (OCDO)  
Fred Hutch Scientific Computing (SciComp)



**Thank you**