ISB Cancer Genomics Cloud

NCI CBIIT Speaker Series December 9th 2015





ISB-CGC Team Members



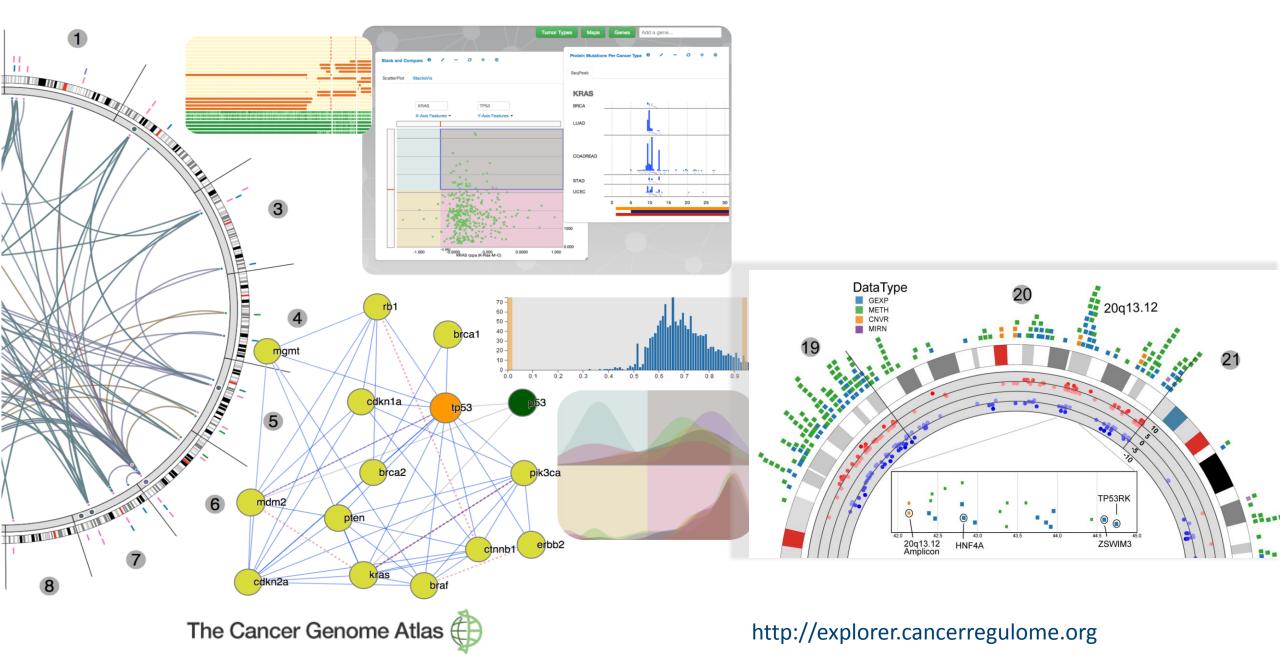
Ilya Shmulevich Sheila Reynolds Michael Miller Phyliss Lee Kelly Iverson Zack Rodebaugh Kalle Leinonen Abigail Hahn Eric Downes Roger Kramer



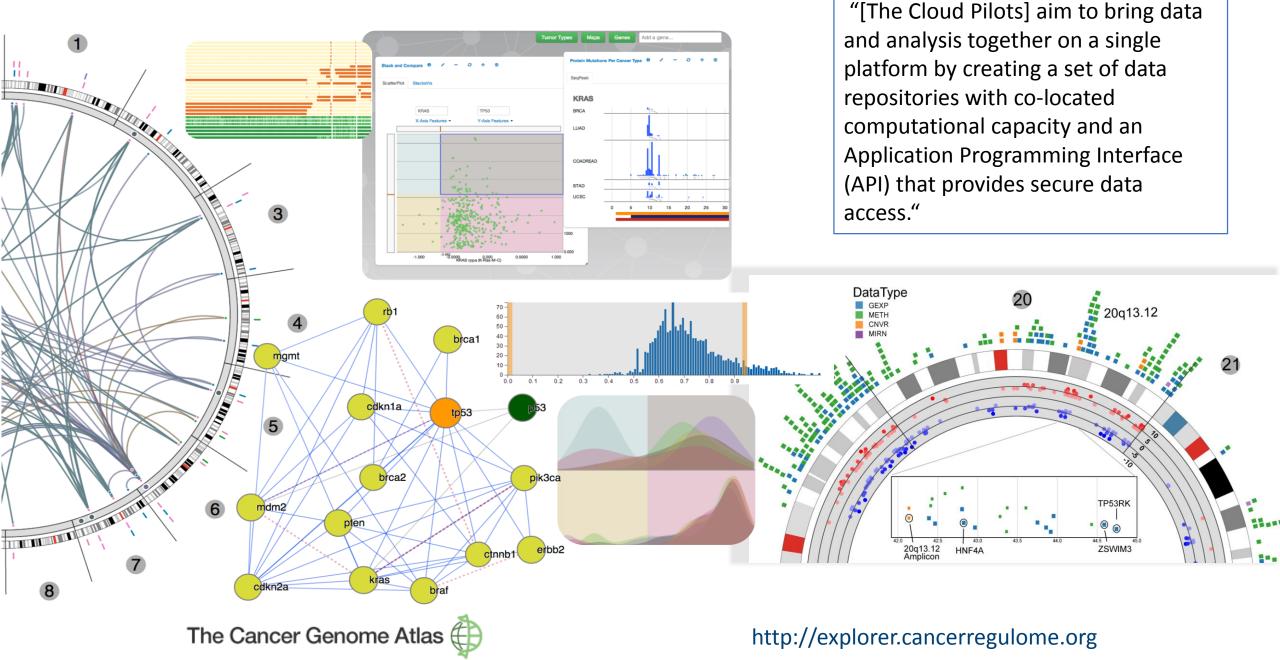
Jonathan Bingham Nicole Deflaux Matt Bookman Jaclyn Koller SRA Enduring Values. Inspired Performance."

David Pot Ross Casanova Sandeep Namburi Yan Zhang Brian Conn

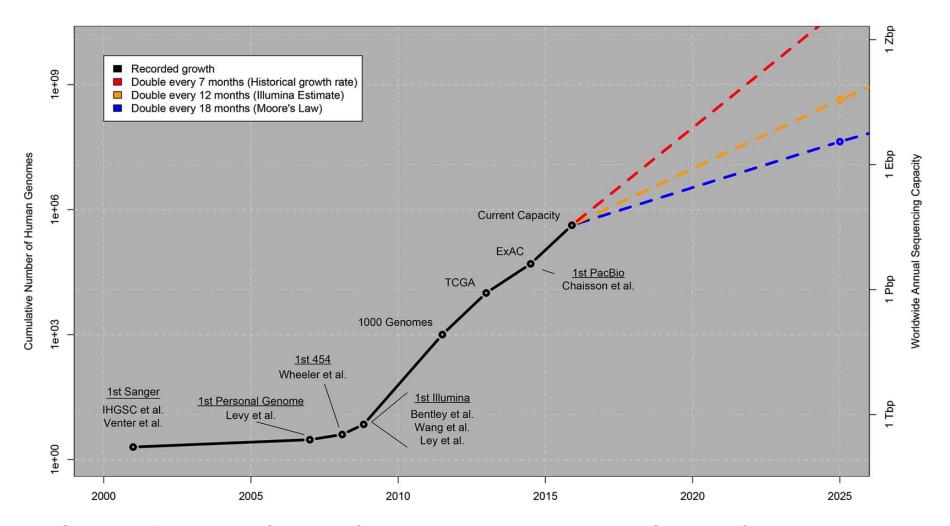
ISB GDAC in TCGA



ISB GDAC in TCGA

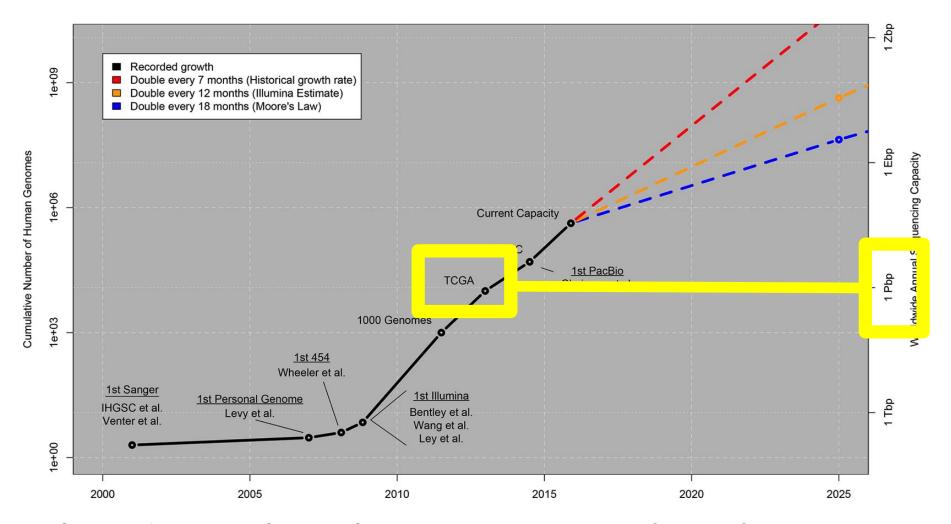


The Challenge of Big Data



Big Data: Astronomical or Genomical? Zachary D. Stephens, Skylar Y. Lee, Faraz Faghri, Roy H. Campbell, Chengxiang Zhai, Miles J. Efron, Ravishankar Iyer, Michael C. Schatz, Saurabh Sinha, Gene E. Robinson

The Challenge of Big Data



Big Data: Astronomical or Genomical? Zachary D. Stephens, Skylar Y. Lee, Faraz Faghri, Roy H. Campbell, Chengxiang Zhai, Miles J. Efron, Ravishankar Iyer, Michael C. Schatz, Saurabh Sinha, Gene E. Robinson

Cloud Paradigm Shift(s)

- Shift #1: Move data and existing pipelines to the cloud
 - all researchers access a single copy of the data
 - everyone saves time, money, and bandwidth
 - compute-power is "near" the data
 - pay only for minutes used
- Shift #2: Cloud-aware computing
 - rethink/redevelop approaches to fully leverage the power of the cloud
 - massively parallel, bursty, opportunistic computing

Cloud Paradigm Shift(s)

- Shift #1: Move data and existing pipelines to the cloud
 - all researchers access a single copy of the data
 - everyone saves time, money, and bandwidth
 - compute-power is "near" the data
 - pay only for minutes used
- Shift #2: Cloud-aware computing
 - rethink/redevelop approaches to fully leverage the power of the cloud
 - massively parallel, bursty, opportunistic computing
 - eg: use BigQuery to calculate expression association with mutation status for one gene takes 7s, doing it for all 20k genes takes less than 9s!

The ISB Cancer Genomics Cloud

- Goals
- Approach

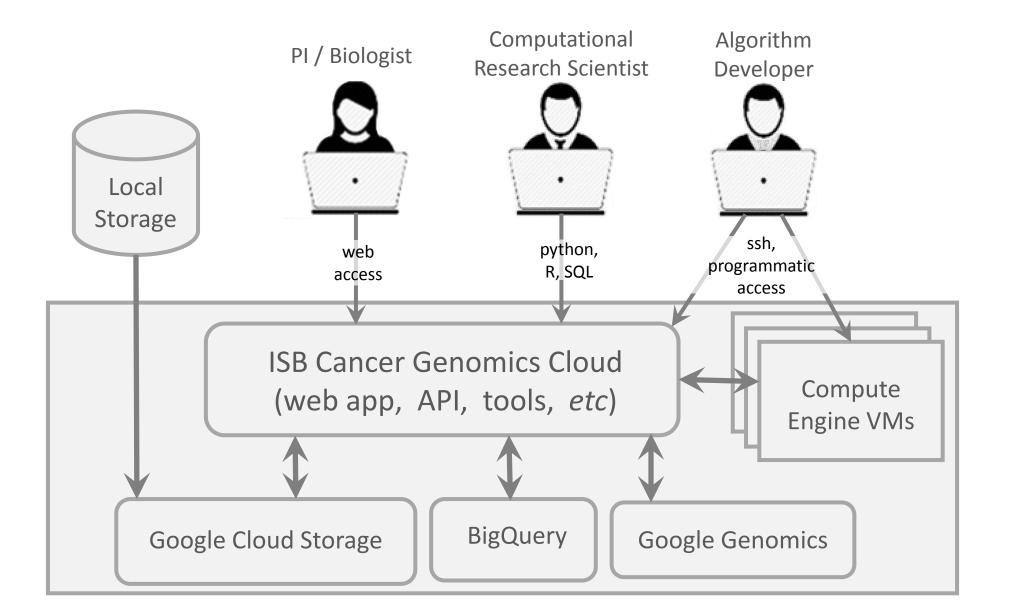
Primary Goals of the ISB-CGC

to make TCGA data, together with tools and computepower available and accessible to a broad range of users

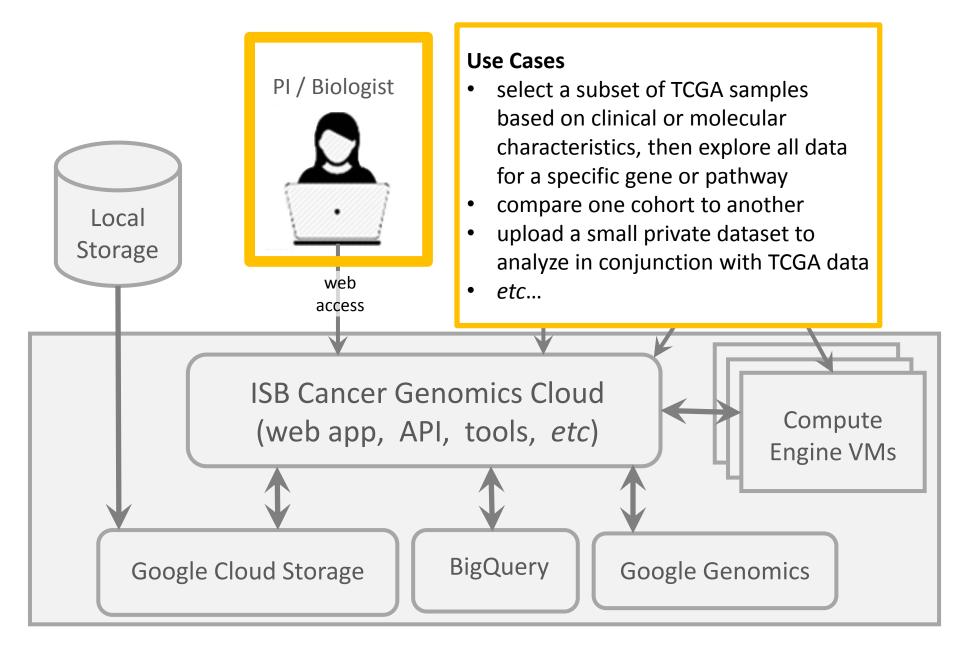
using multiple access modes:

- interactive web application
- scripting languages: R, Python, SQL
- direct programmatic access

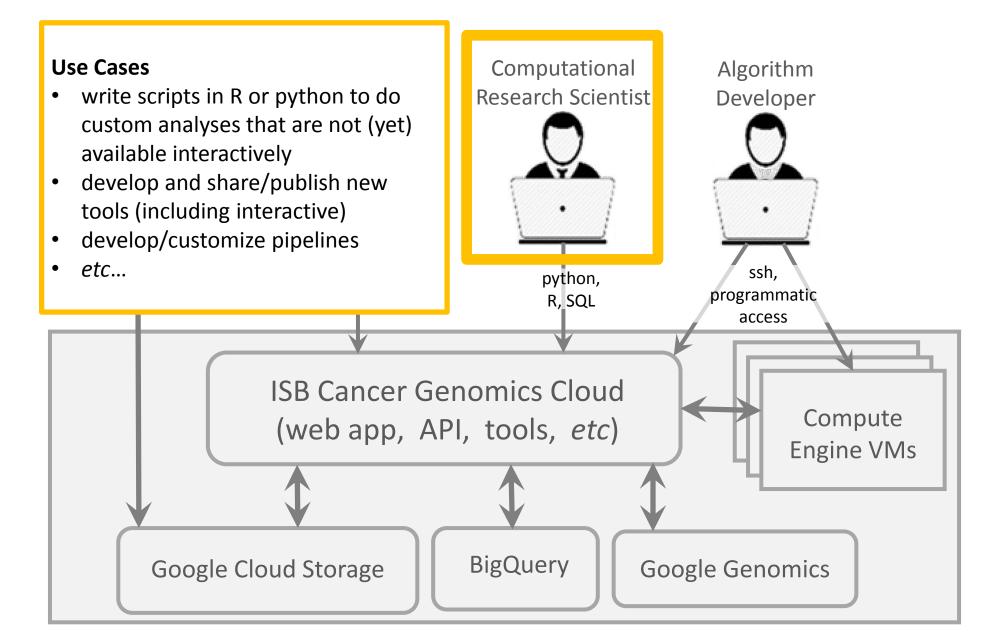
Platform & Tools targeted to a range of users:



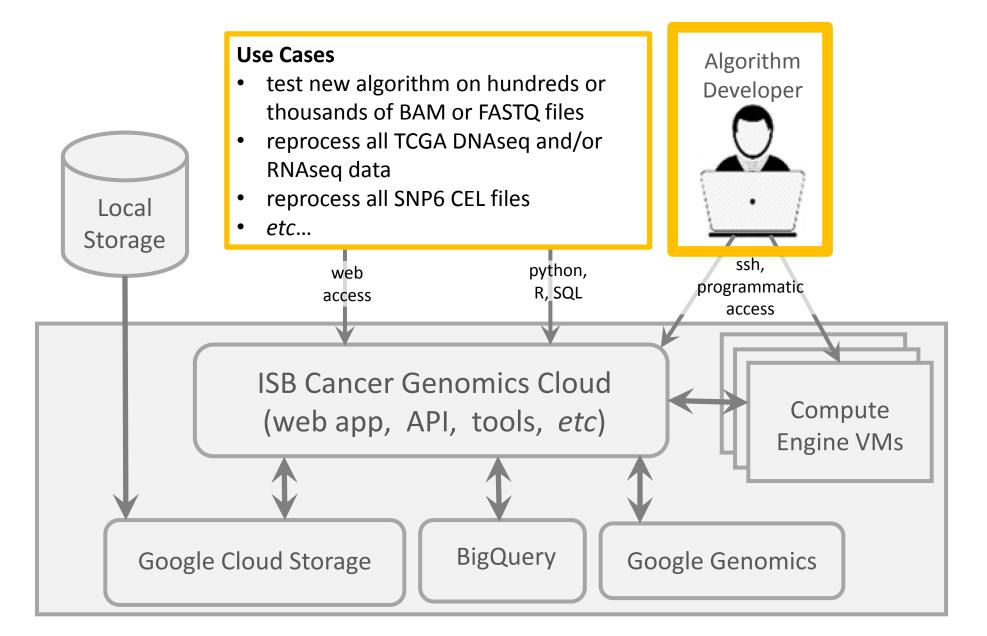
web access for the PI / Biologist:



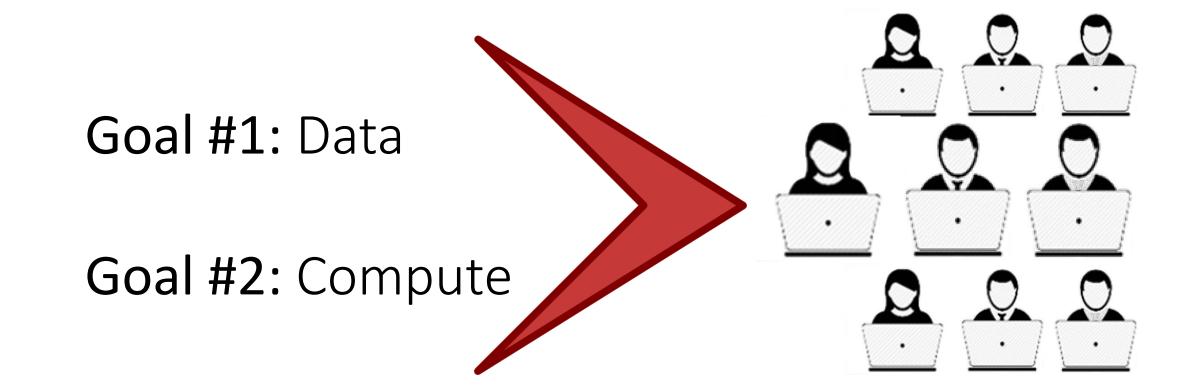
Python, R, and SQL for the Computational Scientist



programmatic access for the Algorithm Developer:



Primary Goals of the ISB-CGC



Goal #1: Data

1 PB Google Cloud Storage

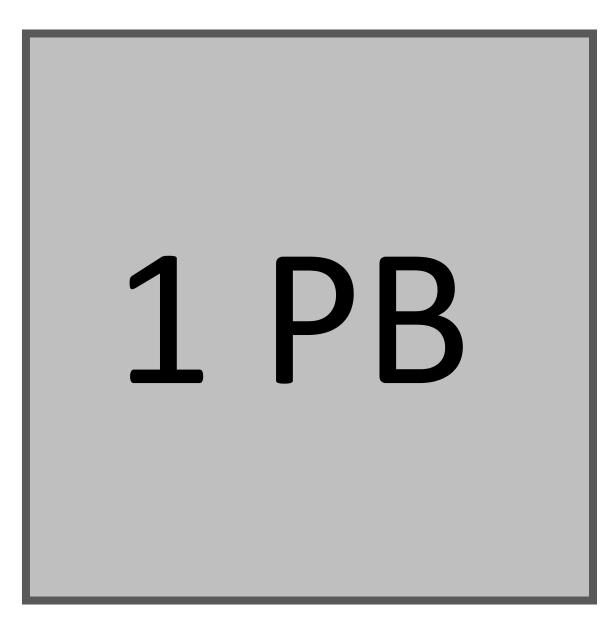
•



THE CANCER GENOME ATLAS

National Cancer Institute National Human Genome Research Institute





Total *size* of TCGA data hosted by ISB-CGC: **1 PB**

What is in there?

Low-level Sequence Data

Total *size* of TCGA data hosted by ISB-CGC: **1 PB**

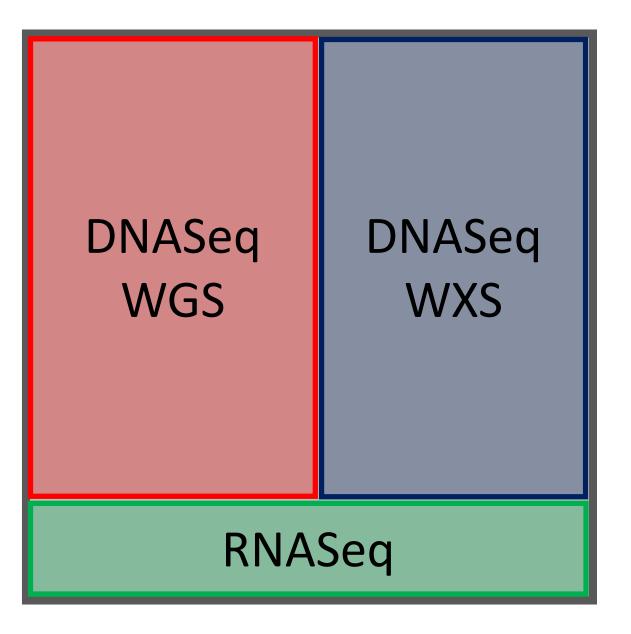
• 99.8% is low-level sequence data (Level-1)

DNASeq

RNASeq

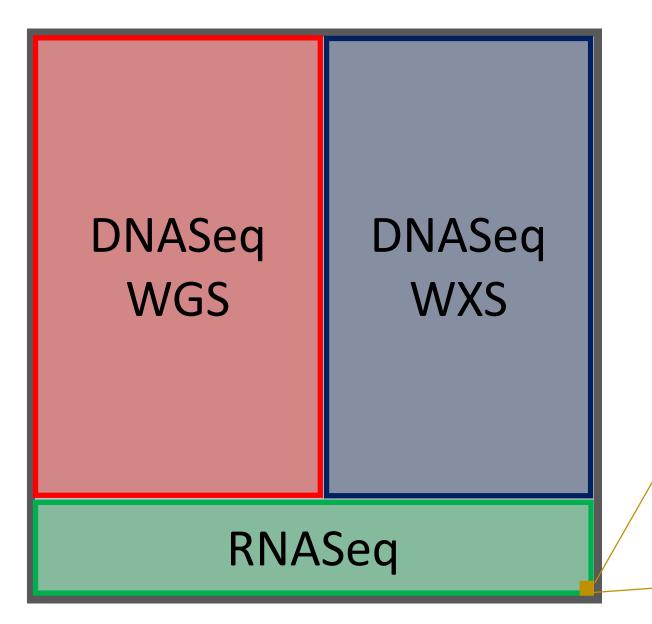
Total *size* of TCGA data hosted by ISB-CGC: **1 PB**

- 99.8% is low-level sequence data (Level-1)
 - 85% is DNASeq data
 - 15% is RNASeq data (including miRNAseq)



Total size of TCGA data hosted by ISB-CGC: 1 PB

- 99.8% is low-level sequence data (Level-1)
 - 85% is DNASeq data
 - 52% is whole genome sequence
 - 48% is exome sequence
 - 15% is RNASeq data (including miRNAseq)

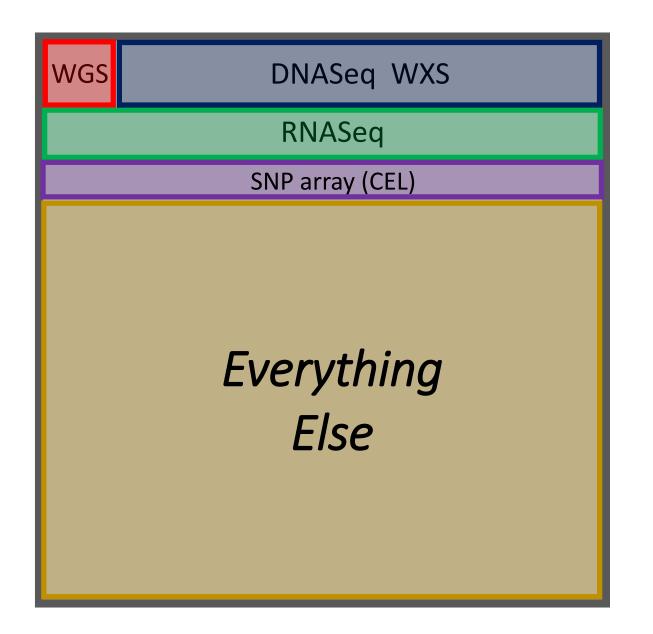


Total *size* of TCGA data hosted by ISB-CGC: **1 PB**

• 99.8% is low-level sequence data (Level-1)

- 85% is DNASeq data
 - 52% is whole genome sequence
 - 48% is exome sequence
- 15% is RNASeq data (including miRNAseq)

- 0.15% is low-level SNP array data (CEL files)
- 0.05% is *all* other data (Level-3, clinical, *etc*)



Total *number* of TCGA files hosted by ISB-CGC: **340K**

- 22% is low-level sequence data (Level-1)
 - 53% is DNASeq data
 - 10% is whole genome sequence
 - 90% is exome sequence
 - 47% is RNASeq data (including miRNAseq)
- 7% is low-level SNP array data (CEL files)
- 71% is *all* other data (Level-3, clinical, *etc*)

RNASeq						
SNP array (CEL)						
DNASeq (MAF, VCF)	clinical & biospecimen					
DNA methylation	miDNAcog					
Protein (RPPA)	miRNAseq					
RNASeq (gene, isoform, exon, junction, <i>etc</i>)	SNP array (genotype calls, allele- and segment- copy-number values)					

Total *number* of TCGA files hosted by ISB-CGC: **340K**

- 22% is low-level sequence data (Level-1)
 - 53% is DNASeq data
 - 10% is whole genome sequence
 - 90% is exome sequence
 - 47% is RNASeq data (including miRNAseq)
- 7% is low-level SNP array data (CEL files)
- 71% is *all* other data (Level-3, clinical, *etc*)

Goal #1: Data

ISB-CGC Phase 1

- Low-level sequence and SNP array data as *files* in Cloud Storage
- High-level data and annotations as *tables* in **BigQuery**

ISB-CGC Phase 2

- Low-level sequence data in **Google Genomics** (backed by Bigtable)
- Variant calls in **Google Genomics** and **BigQuery**

Goal #1: Data

ISB-CGC Phase 1

- Low-level sequence and SNP array data as *files* in Cloud Storage
- High-level data and annotations as *tables* in **BigQuery**

ISB-CGC Phase 2

- Low-level sequence data in **Google Genomics**
- Variant calls in **Google Genomics** and **BigQuery**
 - **BigQuery:** massively parallel analytics engine pushes queries out to thousands of machines and aggregates results in seconds
 - Google Genomics: read- and variant-optimized platform, supports the industry standard GA4GH API and can handle petabytes of data

Table Details: Clinical_data

Schema

ParticipantBarcode	STRING	NULLABLE	Describe this field
Study	STRING	NULLABLE	Describe this field
Project	STRING	NULLABLE	Describe this field
ParticipantUUID	STRING	NULLABLE	Describe this field
TSSCode	STRING	NULLABLE	Describe this field
age_at_initial_pathologic_diagnosis	INTEGER	NULLABLE	Describe this field
anatomic_neoplasm_subdivision	STRING	NULLABLE	Describe this field
batch_number	INTEGER	NULLABLE	Describe this field
bcr	STRING	NULLABLE	Describe this field
clinical_M	STRING	NULLABLE	Describe this field
clinical_N	STRING	NULLABLE	Describe this field
clinical_T	STRING	NULLABLE	Describe this field
clinical_stage	STRING	NULLABLE	Describe this field
colorectal_cancer	STRING	NULLABLE	Describe this field
country	STRING	NULLABLE	Describe this field
vital_status	STRING	NULLABLE	Describe this field
days_to_birth	INTEGER	NULLABLE	Describe this field
days_to_death	INTEGER	NULLABLE	Describe this field
days_to_last_known_alive	INTEGER	NULLABLE	Describe this field
days_to_last_followup	INTEGER	NULLABLE	Describe this field
days_to_initial_pathologic_diagnosis	INTEGER	NULLABLE	Describe this field
days_to_submitted_specimen_dx	INTEGER	NULLABLE	Describe this field
ethnicity	STRING	NULLABLE	Describe this field
frozen_specimen_anatomic_site	STRING	NULLABLE	Describe this field
gender	STRING	NULLABLE	Describe this field
gleason_score_combined	FLOAT	NULLABLE	Describe this field
histological_type	STRING	NULLABLE	Describe this field
history_of_colon_polyps	STRING	NULLABLE	Describe this field

Table Details: Biospecimen_data

Schema

ParticipantBarcode	STRING	NULLABLE	Describe this field
SampleBarcode	STRING	NULLABLE	Describe this field
SampleTypeLetterCode	STRING	NULLABLE	Describe this field
SampleType	STRING	NULLABLE	Describe this field
Study	STRING	NULLABLE	Describe this field
Project	STRING	NULLABLE	Describe this field
SampleTypeCode	STRING	NULLABLE	Describe this field
avg_percent_lymphocyte_infiltration	FLOAT	NULLABLE	Describe this field
avg_percent_monocyte_infiltration	FLOAT	NULLABLE	Describe this field
avg_percent_necrosis	FLOAT	NULLABLE	Describe this field
avg_percent_neutrophil_infiltration	FLOAT	NULLABLE	Describe this field
avg_percent_normal_cells	FLOAT	NULLABLE	Describe this field
avg_percent_stromal_cells	FLOAT	NULLABLE	Describe this field
avg_percent_tumor_cells	FLOAT	NULLABLE	Describe this field
avg_percent_tumor_nuclei	FLOAT	NULLABLE	Describe this field
batch_number	INTEGER	NULLABLE	Describe this field
bcr	STRING	NULLABLE	Describe this field
days_to_collection	FLOAT	NULLABLE	Describe this field
days_to_sample_procurement	FLOAT	NULLABLE	Describe this field
is_ffpe	STRING	NULLABLE	Describe this field
max_percent_lymphocyte_infiltration	FLOAT	NULLABLE	Describe this field
max_percent_monocyte_infiltration	FLOAT	NULLABLE	Describe this field
max_percent_necrosis	FLOAT	NULLABLE	Describe this field
max_percent_neutrophil_infiltration	FLOAT	NULLABLE	Describe this field
max_percent_normal_cells	FLOAT	NULLABLE	Describe this field
max_percent_stromal_cells	FLOAT	NULLABLE	Describe this field
max_percent_tumor_cells	FLOAT	NULLABLE	Describe this field
max_percent_tumor_nuclei	FLOAT	NULLABLE	Describe this field

Table Details: Annotations

Schema

annotationId	INTEGER	NULLABLE	Describe this field
annotationCategoryId	INTEGER	NULLABLE	Describe this field
annotationCategoryName	STRING	NULLABLE	Describe this field
annotationClassification	STRING	NULLABLE	Describe this field
annotationNoteText	STRING	NULLABLE	Describe this field
Study	STRING	NULLABLE	Describe this field
itemTypeName	STRING	NULLABLE	Describe this field
itemBarcode	STRING	NULLABLE	Describe this field
AliquotBarcode	STRING	NULLABLE	Describe this field
ParticipantBarcode	STRING	NULLABLE	Describe this field
SampleBarcode	STRING	NULLABLE	Describe this field
dateAdded	STRING	NULLABLE	Describe this field
dateCreated	STRING	NULLABLE	Describe this field
dateEdited	STRING	NULLABLE	Describe this field



Table Details: Clinical_data

Schema

Schema				
ParticipantBarcode		STRING	NULLABLE	Describe this field
Study		STRING	NULLABLE	Describe this field
Project		STRING	NULLABLE	Describe this field
ParticipantUUID		CTDING	NULLABLE	n
TSSCode	Table Details: Somatic_Mutatio	n_calls		
age_at_initial_path	Schema			
anatomic_neoplasr	ParticipantBarcode	STRING	NULLABLE	Describe this field
batch_number	Tumor_SampleBarcode	STRING	NULLABLE	Describe this field
bcr	Tumor_AliquotBarcode	STRING	NULLABLE	Describe this field
clinical_M	Tumor_SampleTypeLetterCode	STRING	NULLABLE	Describe this field
clinical_N	Normal_SampleBarcode	STRING	NULLABLE	Describe this field
clinical_T	Normal_AliquotBarcode	STRING	NULLABLE	Describe this field
clinical_stage	Normal_SampleTypeLetterCode	STRING	NULLABLE	Describe this field
colorectal_cancer	Study	STRING	NULLABLE	Describe this field
country	Annotation_Transcript	STRING	NULLABLE	Describe this field
vital_status	CCLE_ONCOMAP_Total_Mutations_In_Gene	INTEGER	NULLABLE	Describe this field
days_to_birth	COSMIC_Total_Alterations_In_Gene	INTEGER	NULLABLE	Describe this field
days_to_death	Center	STRING	NULLABLE	Describe this field
days_to_last_know	Chromosome	STRING	NULLABLE	Describe this field
days_to_last_follov	DNARepairGenes_Role	STRING	NULLABLE	Describe this field
days_to_initial_pat	DbSNP_RS	STRING	NULLABLE	Describe this field
days_to_submitted	Db SNP_Val_Status	STRING	NULLABLE	Describe this field
ethnicity	DrugBank	STRING	NULLABLE	Describe this field
frozen_specimen_a	End_Position	INTEGER	NULLABLE	Describe this field
gender	Entrez_Gene_Id	INTEGER	NULLABLE	Describe this field
gleason_score_con	GC_Content	FLOAT	NULLABLE	Describe this field
histological_type	GENCODE_Transcript_Name	STRING	NULLABLE	Describe this field
history_of_colon_p	GENCODE_Transcript_Status	STRING	NULLABLE	Describe this field
	GENCODE_Transcript_Type	STRING	NULLABLE	Describe this field
	GO_Biological_Process	STRING	NULLABLE	Describe this field
	GO_Cellular_Component	STRING	NULLABLE	Describe this field
	GO_Molecular_Function	STRING	NULLABLE	Describe this field
	Gene_Type	STRING	NULLABLE	Describe this field
	Genome_Change	STRING	NULLABLE	Describe this field

Table Details: Biospecimen_data

Schema

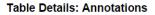
ParticipantBarcode	STRIN	G NULLABLE	Describe th	is field	a	nnotationId
SampleBarcode	STRIN	G NULLABLE	Describe th	is field	a	nnotationCategoryId
SampleTypeLetterCode	STRIN	RING NULLABLE Describe this field		is field	a	nnotationCategoryName
SampleType	STRIN	G NULLABLE	Describe th	is field	a	nnotationClassification
Study	STRIN	G NULLABLE	Describe th	is field	a	nnotationNoteText
Project	STRI	Table Deta	ils: Cop	y_Nur	nber_seg	ments
SampleTypeCode	STRI	Schema				
avg_percent_lymphocyte_infiltration	FLO/	Schema				
avg_percent_monocyte_infiltration	FLO A	ParticipantBar	code	STRING	NULLABLE	Describe this field
avg_percent_necrosis	FLOA	SampleBarcod	e	STRING	NULLABLE	Describe this field
avg_percent_neutrophil_infiltration	FLO A	SampleTypeLe	tterCode	STRING	NULLABLE	Describe this field
avg_percent_normal_cells	FLOA	AliquotBarcod	e	STRING	NULLABLE	Describe this field
avg_percent_stromal_cells	FLOA	Study		STRING	NULLABLE	Describe this field
avg_percent_tumor_cells	FLOA	Platform		STRING	NULLABLE	Describe this field
avg_percent_tumor_nuclei	FLOA	Chromosome		STRING	NULLABLE	Describe this field
batch_number	INTE	Start		INTEGER	R NULLABLE	Describe this field
bcr	STRI	End		INTEGER	R NULLABLE	Describe this field
days_to_collection	FLOA	Num Probes		INTEGER	R NULLABLE	Describe this field
days_to_sample_procurement	FLOA	-				
is_ffpe	STRI	Segment_Mear	1	FLOAT	NULLABLE	Describe this field
man managed househands indition	EL 0.47					

max nercent lymnhocyte infiltration FLOAT NULLABLE Describe this field

Table Details: DNA_Methylation_betas

Schema

ParticipantBarcode	STRING	NULLABLE	Describe this field
SampleBarcode	STRING	NULLABLE	Describe this field
SampleTypeLetterCode	STRING	NULLABLE	Refer: https://tcga-data.nci.nih.gov/datareports/codeTablesReport.htm
AliquotBarcode	STRING	NULLABLE	The Aliquot ID is an identifier/barcode of TCGA data. Refer: https://wiki.nci.nih.gov/display/TCGA/TCGA+barcode
Platform	STRING	NULLABLE	Refer: https://tcga-data.nci.nih.gov/datareports/codeTablesReport.htm
Study	STRING	NULLABLE	TCGA disease type
Probe_Id	STRING	NULLABLE	Illumina's CpG loci IDs. Refer: http://www.illumina.com/content/dam/illumina- marketing/documents/products/technotes/technote_cpg_loci_identification.pdf
Beta_Value	FLOAT	NULLABLE	The beta value (β) is used to estimate the methylation level of the CpG locus using the ratio of intensities between methylated and unmethylated alleles



Schema

annotationId	INTEGER	NULLABLE	Describe this field
annotationCategoryId	INTEGER	NULLABLE	Describe this field
annotationCategoryName	STRING	NULLABLE	Describe this field
annotationClassification	STRING	NULLABLE	Describe this field
annotationNoteText	STRING	NULLABLE	Describe this field
gments	STRING	NULLABLE	Describe this field



STRING Table Details: Protein_RPPA_data

STRING STRING

STRING

STRING

STRING

STRING

STRING

ParticipantBarcode	STRING	NULLABLE	Describe this field
SampleBarcode	STRING	NULLABLE	Describe this field
SampleTypeLetterCode	STRING	NULLABLE	Describe this field
AliquotBarcode	STRING	NULLABLE	Describe this field
Study	STRING	NULLABLE	Describe this field
Gene_Name	STRING	NULLABLE	Describe this field
Protein_Expression	FLOAT	NULLABLE	Describe this field
Protein_Name	STRING	NULLABLE	Describe this field
Protein_Basename	STRING	NULLABLE	Describe this field
Phospho	STRING	NULLABLE	Describe this field
antibody Source	STRING	NULLABLE	Describe this field
validation Status	STRING	NULLABLE	Describe this field

Table Details: Clinical_data

Schema

Jenema				
ParticipantBarcode		STRING	NULLABLE	Describe this field
Study		STRING	NULLABLE	Describe this field
Project		STRING	NULLABLE	Describe this field
ParticipantUUID	Table Detaile: Semetia Mutatia		NULLABLE	
TSSCode	Table Details: Somatic_Mutatio	on_cans		
age_at_initial_path	Schema			
anatomic_neoplasr	ParticipantBarcode	STRING	NULLABLE	Describe this field
batch_number	Tumor_SampleBarcode	STRING	NULLABLE	Describe this field
bcr	Tumor_AliquotBarcode	STRING	NULLABLE	Describe this field
clinical_M	Tumor_SampleTypeLetterCode	STRING	NULLABLE	Describe this field
clinical_N	Normal_SampleBarcode	STRING	NULLABLE	Describe this field
clinical_T	Normal_AliquotBarcode	STRING	NULLABLE	Describe this field
clinical_stage	Normal_SampleTypeLetterCode	STRING	NULLABLE	Describe this field
colorectal_cancer	Study	STRING	NULLABLE	Describe this field
country	Annotation_Transcript	STRING	NULLABLE	Describe this field
vital_status	CCLE_ONCOMAP_Total_Mutations_In_Gene	INTEGER	NULLABLE	Describe this field
days_to_birth	COSMIC_Total_Alterations_In_Gene	INTEGER	NULLABLE	Describe this field
days_to_death	Center	STRING	NULLABLE	Describe this field
days_to_last_know	Chromosome	STRING	NULLABLE	Describe this field
days_to_last_follov	DNARepairGenes_Role	STRING	NULLABLE	Describe this field
days_to_initial_pat	DbSNP_RS	STRING	NULLABLE	Describe this field
days_to_submitted	Db SNP_Val_Status	STRING	NULLABLE	Describe this field
ethnicity	DrugBank	STRING	NULLABLE	Describe this field
frozen_specimen_a	End_Position	INTEGER	NULLABLE	Describe this field
gender	Entrez_Gene_Id	INTEGER	NULLABLE	Describe this field
gleason_score_con	GC_Content	FLOAT	NULLABLE	Describe this field
histological_type	GENCODE_Transcript_Name	STRING	NULLABLE	Describe this field
history_of_colon_p	GENCODE_Transcript_Status	STRING	NULLABLE	Describe this field
	GENCODE_Transcript_Type	STRING	NULLABLE	Describe this field
	GO_Biological_Process	STRING	NULLABLE	Describe this field
	GO_Cellular_Component	STRING	NULLABLE	Describe this field
	GO_Molecular_Function	STRING	NULLABLE	Describe this field
	Gene_Type	STRING	NULLABLE	Describe this field
	Genome_Change	STRING	NULLABLE	Describe this field

Table Details: Biospecimen_data

Schema

ield..

field...

field...

field...

ield...

ield... field... ield...

ield...

ield...

	ParticipantBarcode		STRIN	G	NULLABLE	Describe	this field	ann	otationId	
	SampleBarcode		STRIN	G	NULLABLE	Describe	this field	ann	otationCatego	oryld
	SampleTypeLetterCode		STRIN	G	NULLABLE	Describe	this field	ann	otationCatego	ryName
	SampleType		STRIN	G	NULLABLE	Describe	he this field		annotationClassification annotationNoteText	
	Study		STRIN	G	NULLABLE	Describe				
	Project		STRI	STRI Table Details: Copy_Number_segments						
	SampleTypeCode		STRI Schema							
	avg_percent_lymphocyte	_infiltration	FLOA				_			
	avg_percent_monocyte_i	nfiltration	FLOA	Parti	icipantBar	code	STRING	NULLABLE	Describe this	field
-	avg_percent_necrosis		FLOA	Sam	pleBarcod	e	STRING	NULLABLE	Describe this	field
· .	avg_percent_neutrophil_i	infiltration	FLOA	Sam	pleTypeLe	tterCode	STRING	NULLABLE	Describe this	field
•	avg_percent_normal_cells	8	FLOA	Aliq	uotBarcod	e	STRING	NULLABLE	Describe this	field
	avg_percent_stromal_cell	s	FLOA	Stud	iy		STRING	NULLABLE	Describe this	field
	avg_percent_tumor_cells		FLOA	Platf	form		STRING	NULLABLE	Describe this	field
	avg_percent_tumor_nucle	ei	FLOA	Chro	omosome	mosome STF		NULLABLE	Describe this	field
	batch_number bcr		INTE	Fad		Table Details: mRNA_BCG			GSC_H	
			STRI			Schema				
	days_to_collection		FLOA	Num Probes		Schema				
	days_to_sample_procure	ment	STRI		Segment_Mean		ParticipantBarcode SampleBarcode		STRING	NULLAE
. –	is_ffpe								STRING	NULLAE
	max nercent lymnhocyte				NHI LARI F					
lab	ole Details: DN	A Metr				Deerik	SampleTyp	oeLetterCode	STRING	Table
		_	iylati	on_	betas	l Doorik	SampleTyp AliquotBar		STRING	
Sch	ema		iylati	on_	betas	, Deorrit				Schema
		_	-		_	·	AliquotBar		STRING	
Par	ticipantBarcode	STRING	NULL	ABLE	Describ	e this fie	AliquotBar Study Platform		STRING STRING STRING	Schema
Par		_	-	ABLE	Describ	·	AliquotBar Study Platform gene_id	rcode	STRING STRING STRING INTEGER	Schema Participa
Par San	ticipantBarcode	STRING	NULL	ABLE	Describ Describ	e this fie e this fie	AliquotBar Study Platform gene_id original_ge	rcode ene_symbol	STRING STRING STRING INTEGEF STRING	Schema Participa Samplet
Par San San	ticipantBarcode npleBarcode	STRING	NULL	ABLE ABLE ABLE	Describ Describ Refer: h	e this fie e this fie https://tc	AliquotBar Study Platform gene_id	rcode ene_symbol	STRING STRING STRING INTEGER	Schema Participa Samplel Aliquoti
Par San San Alic	ticipantBarcode npleBarcode npleTypeLetterCode	STRING STRING STRING	NULL	ABLE ABLE ABLE ABLE	Describ Describ Refer: h The Alic	e this fie e this fie https://to quot ID i	AliquotBar Study Platform gene_id original_ge	rcode ene_symbol	STRING STRING STRING INTEGEF STRING	Schema Particip Sample Aliquot Sample Study
Par San San Alic	ticipantBarcode npleBarcode npleTypeLetterCode quotBarcode tform	STRING STRING STRING STRING	NULL NULL NULL	ABLE ABLE ABLE ABLE	Describ Describ Refer: h The Alic Refer: h	e this fie e this fie https://tc quot ID i https://tc	AliquotBar Study Platform gene_id original_gen HGNC_gen	rcode ene_symbol ne_symbol	STRING STRING STRING INTEGER STRING	Schema Particip Samplet Aliquott Samplet Study Platform
Pari San San Alic Plat	ticipantBarcode npleBarcode npleTypeLetterCode quotBarcode tform dy	STRING STRING STRING STRING STRING	NULL NULL NULL NULL	ABLE ABLE ABLE ABLE ABLE	Describ Describ Refer: h The Alic Refer: h TCGA c	e this fie e this fie https://to uuot ID i https://to fisease t	AliquotBar Study Platform gene_id original_ge HGNC_gen RPKM gene_adde	rcode ene_symbol ne_symbol	STRING STRING STRING INTEGER STRING STRING FLOAT STRING	Schema Participi Samplel Aliquott Sample Study Platform mirna_id
Part San San Alic Plat Stu	ticipantBarcode npleBarcode npleTypeLetterCode quotBarcode tform	STRING STRING STRING STRING	NULL NULL NULL NULL	ABLE ABLE ABLE ABLE ABLE	Describ Describ Refer: h The Alic Refer: h TCGA c	e this fie e this fie tttps://tc tttps://tc fisease 1	AliquotBar Study Platform gene_id original_ge HGNC_gen RPKM gene_adde	ene_symbol eneasymbol enda	STRING STRING STRING INTEGEF STRING STRING FLOAT STRING	Schema Particip Samplet Aliquott Samplet Study Platform

Table Details: Annotations

Schema

annotationId	INTEGER	NULLABLE	Describe this field
annotationCategoryId	INTEGER	NULLABLE	Describe this field
annotationCategoryName	STRING	NULLABLE	Describe this field
annotationClassification	STRING	NULLABLE	Describe this field
annotationNoteText	STRING	NULLABLE	Describe this field
gments	STRING	NULLABLE	Describe this field

STRING Table Details: miRNA_expression

STRING Table Details: Protein_RPPA_data

s field	STRI	NG	Schema					
s field	STRI	NG	Dentiain ant Dan		STRING			
	STRING		Participantban	ParticipantBarcode		NULLABLE	Describe this field	
s field	STRI	NG	SampleBarcod	e	STRING	NULLABLE	Describe this field	
s field	STRI	NG	SampleTypeLe	tterCode	STRING	NULLABLE	Describe this field	
s field	STRI	NG	AliquotBarcod	e	STRING	NULLABLE	Describe this field	
s field	STRI	NG	Study		STRING	NULLABLE	Describe this field	
s field		ea	RPKM		STRING	NULLABLE	Describe this field	
		<u>с</u> ч.		ion	FLOAT	NULLABLE	Describe this field	
					STRING	NULLABLE	Describe this field	
N	ULLABLE	ABLE Describe this field		me	STRING	NULLABLE	Describe this field	
N	JLLABLE	De	scribe this field		STRING	NULLABLE	Describe this field	

Query Table

ParticipantBarcode	STRING	NULLABLE	Describe this field	
SampleBarcode	STRING	NULLABLE	Describe this field	
AliquotBarcode	STRING	NULLABLE	Describe this field	
SampleTypeLetterCode	STRING	NULLABLE	Describe this field	
Study	STRING	NULLABLE	Describe this field	
Platform	STRING	NULLABLE	Describe this field	
mirna_id	STRING	NULLABLE	Describe this field	
mirna_accession	STRING	NULLABLE	Describe this field	
normalized_count	FLOAT	NULLABLE	Describe this field	hylated alleles



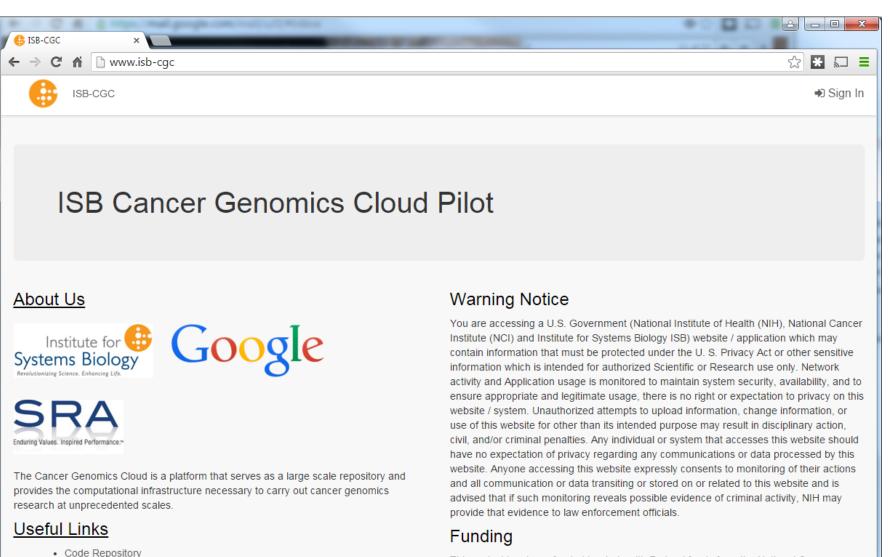
Bring your data to BigQuery!

- easily integrate with other BigQuery datasets ... if other people put their data and annotations into BigQuery tables
- eg Tute Genomics
- Let's put out a call to researchers to make data, annotations, etc available for all to use in BigQuery!
 - TCGA Level-3 data (500 GB) -- \$10 per month
 - Tute Genomics (649 GB and 8.6 billion rows) -- \$13 per month
 - GENCODE (593 MB table with 2.6 million rows) -- only 14 cents per year

Goal #2: Compute

- 1. PI / Biologist: web-based interaction
- 2. Computational Research Scientist: R, Python, SQL
- 3. Algorithm Developer: VMs, Container Engine, Dataproc, Dataflow

web access for the PI / Biologist

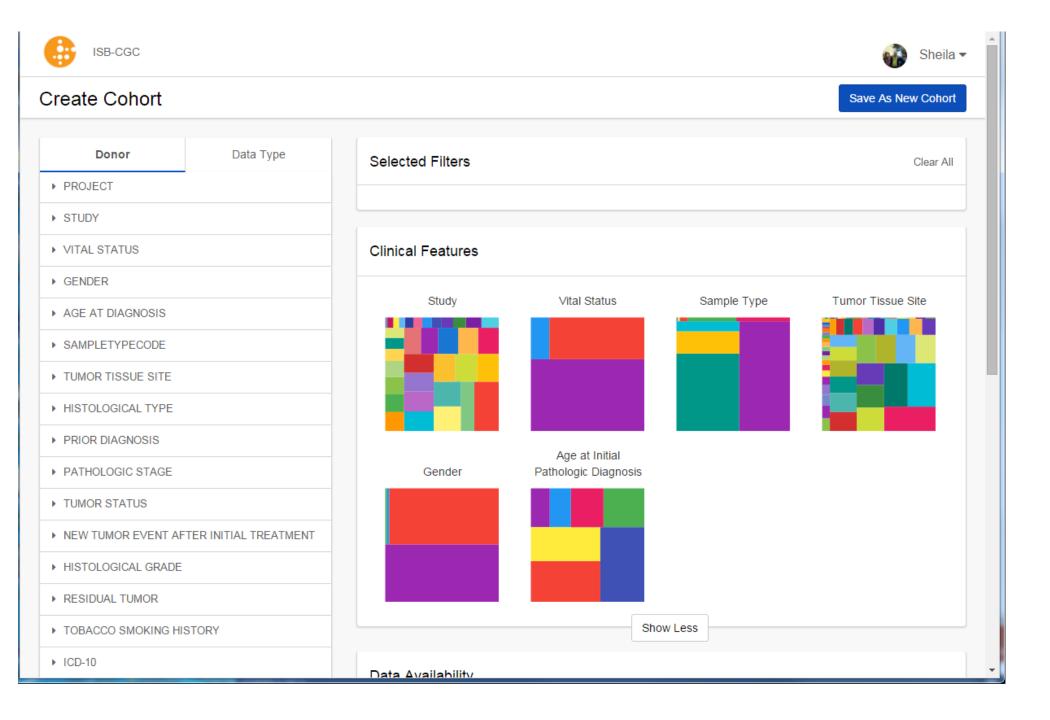


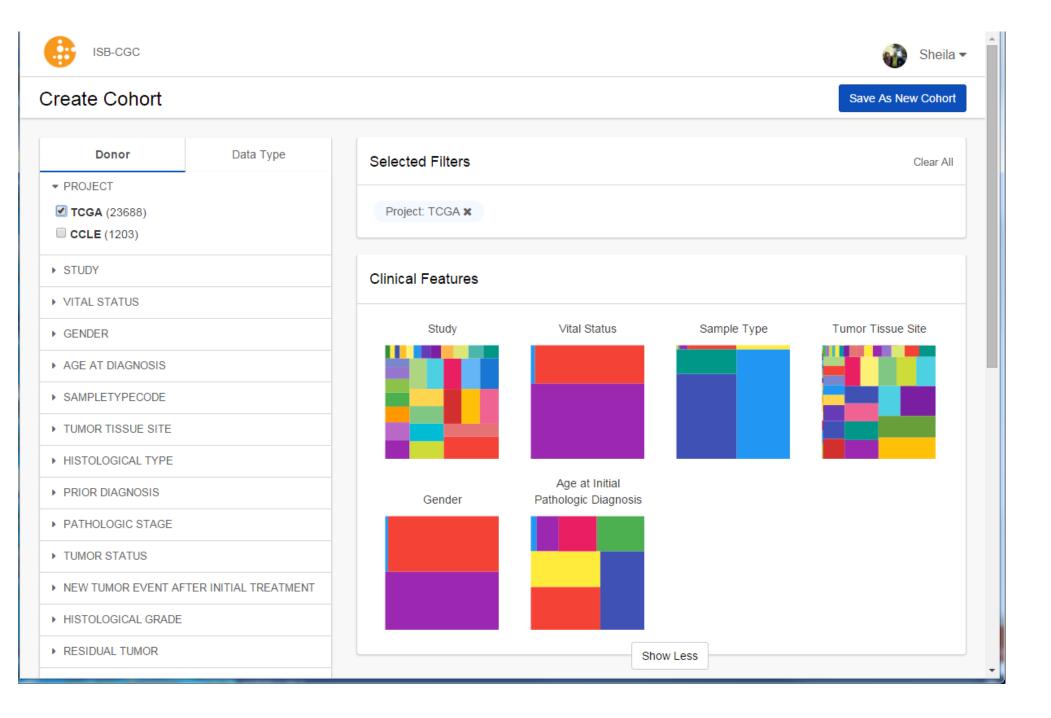
This project has been funded in whole with Federal funds from the National Cancer

Contract No. HHSN261201400007C.

Institute, National Institutes of Health, Department of Health and Human Services, under

- Code Repository
- Documentation







🚯 Sheila 🗸

Save As New Cohort

Create C Create Cohort

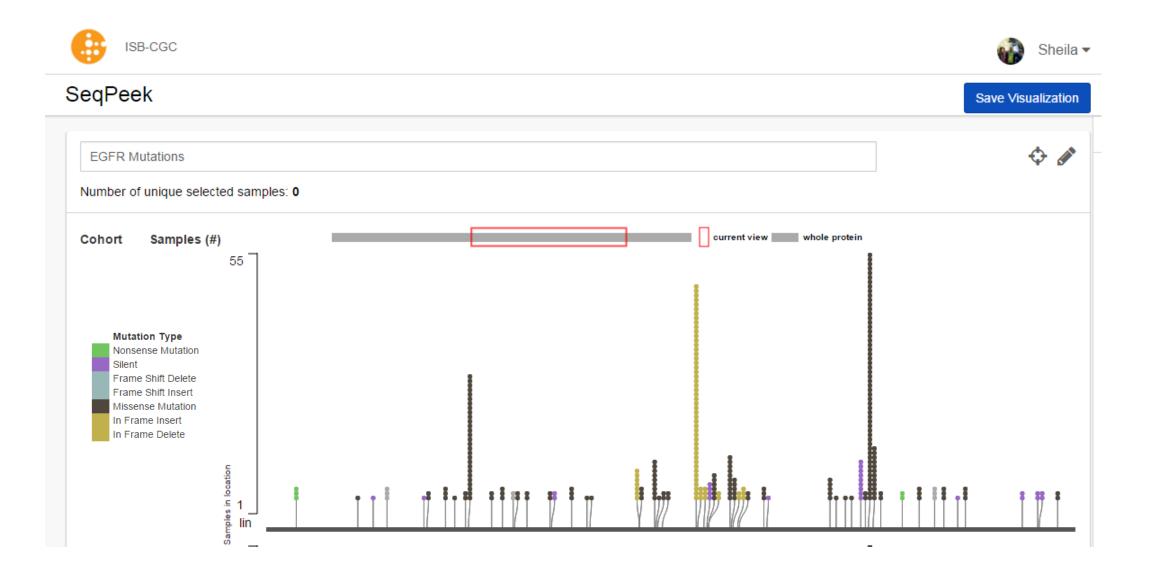
DonData Type• PROJECT• PROJECT• TCGA (2 • STUDY• STUDY• CCLE (12 • VITAL STATUS• VITAL STATUS• STUDY• Alive (493)• Dead (625)• Dead (625)• VITAL STA• Dead (625)• VITAL STA• GENDER• AGE AT DI • SAMPLETY• AGE AT DIAGNOSIS• TUMOR TIS• SAMPLETYPECODE• TUMOR TIS• TUMOR TISSUE SITE• PRIOR DIA • PRIOR DIAGNOSIS• PRIOR DIAGNOSIS• NEW TUMOR ST• TUMOR STATUS• NEW TUMO• NEW TUMOR EVENT AFTER INITIAL TREATMENT• HISTOLOG• NEW TUMOR EVENT AFTER INITIAL TREATMENT						
 PROJECT TCGA (2 STUDY CCLE (12 VITAL STATUS STUDY Alive (493) Dead (625) VITAL STA None (10) GENDER AGE AT DI AGE AT DI AGE AT DIA PAGE AT DIA PRIOR DIA PATHOLOG TUMOR ST NEW TUMC NEW TUMC NEW TUMC NEW TUMC 	Don	Donor	Data Type			
CCLE (12· STUDY· STUDY· VITAL STA· VITAL STA· VITAL STA· VITAL STA· VITAL STA· Oead (625)· None (10)· GENDER· AGE AT DI· AGE AT DI· AGE AT DIA· AGE AT DIAGNOSIS· TUMOR TIS· HISTOLOG· PRIOR DIA· PATHOLOC· NEW TUMO· NEW TUMO· NEW TUMO· NEW TUMO· NEW TUMO· NEW TUMO	▼ PROJECT	▶ PROJECT				
 VITAL STATUS STUDY Alive (493) Dead (625) None (10) GENDER AGE AT DI AGE AT DIAGNOSIS SAMPLET SAMPLET SAMPLETYPECODE TUMOR TIS TUMOR TISSUE SITE HISTOLOG PRIOR DIA PRIOR DIAGNOSIS PATHOLOC PATHOLOC PATHOLOG NEW TUMC NEW TUMC NEW TUMC NEW TUMOR EVENT AFTER INITIAL TREATMENT 		STUDY				
 VITAL STA Dead (625) None (10) GENDER AGE AT DI AGE AT DIAGNOSIS SAMPLET AGE AT DIAGNOSIS SAMPLETYPECODE TUMOR TIS TUMOR TISSUE SITE HISTOLOG HISTOLOGICAL TYPE PRIOR DIA PRIOR DIAGNOSIS PATHOLOC PATHOLOGIC STAGE TUMOR STATUS NEW TUMC NEW TUMC NEW TUMOR EVENT AFTER INITIAL TREATMENT 	CCLE (12	▼ VITAL STATUS				
 VITAL STA None (10) GENDER AGE AT DI AGE AT DIAGNOSIS SAMPLET AGE AT DIAGNOSIS SAMPLET SAMPLETYPECODE TUMOR TIS TUMOR TIS TUMOR TISSUE SITE HISTOLOG HISTOLOGICAL TYPE PRIOR DIA PRIOR DIAGNOSIS PATHOLOC PATHOLOGIC STAGE TUMOR ST NEW TUMC NEW TUMOR EVENT AFTER INITIAL TREATMENT 	STUDY	Alive (493)				
• GENDER • AGE AT DI • AGE AT DI • SAMPLET • SAMPLET • TUMOR TIS • TUMOR TIS • HISTOLOG • PRIOR DIA • PATHOLOC • TUMOR ST • NEW TUMO • NEW TUMO • NEW TUMO	VITAL STA					
 AGE AT DI AGE AT DIAGNOSIS SAMPLETY AGE AT DIAGNOSIS SAMPLETYPECODE TUMOR TIS TUMOR TISSUE SITE HISTOLOG HISTOLOGICAL TYPE PRIOR DIA PRIOR DIAGNOSIS PATHOLOG PATHOLOGIC STAGE TUMOR ST NEW TUMO NEW TUMO EVENT AFTER INITIAL TREATMENT 	GENDER					
 SAMPLET TUMOR TIS TUMOR TIS TUMOR TISSUE SITE HISTOLOG HISTOLOGICAL TYPE PRIOR DIA PRIOR DIAGNOSIS PATHOLOG PATHOLOGIC STAGE TUMOR ST NEW TUMO NEW TUMO NEW TUMOR EVENT AFTER INITIAL TREATMENT 	AGE AT DI	▶ GENDER				
 TUMOR TIS TUMOR TIS TUMOR TISSUE SITE TUMOR DIA PRIOR DIA PRIOR DIAGNOSIS PATHOLOC TUMOR ST NEW TUMC NEW TUMC NEW TUMOR EVENT AFTER INITIAL TREATMENT 	SAMPLET)	AGE AT DIAGNOSIS				
 HISTOLOG HISTOLOGICAL TYPE PRIOR DIA PATHOLOC PATHOLOGIC STAGE TUMOR ST NEW TUMC NEW TUMC NEW TUMOR EVENT AFTER INITIAL TREATMENT 		► SAMPLETYPECODE				
 PRIOR DIA PATHOLOG PATHOLOG PATHOLOGIC STAGE TUMOR ST NEW TUMO NEW TUMO NEW TUMOR EVENT AFTER INITIAL TREATMENT 		▶ TUMOR TISSUE SITE				
PATHOLOG PATHOLOGIC STAGE TUMOR ST NEW TUMO HISTOLOG PATHOLOGIC STATUS NEW TUMOR EVENT AFTER INITIAL TREATMENT	► HISTOLOG	HISTOLOGICAL TYPE				
FORMARY ST FORMARY ST FORMARY STATUS FORMARY STATUS	PRIOR DIA	PRIOR DIAGNOSIS				
TUMOR ST TUMOR STATUS NEW TUMO NEW TUMOR EVENT AFTER INITIAL TREATMENT	PATHOLO(▶ PATHOLOGIC STAGE				
NEW TUM NEW TUMOR EVENT AFTER INITIAL TREATMENT	TUMOR ST					
▶ HISTOLOG	► NEW TUM	► TUMOR STATUS				
	► HISTOLOG	NEW TUMOR EVENT AF	TER INITIAL TREATMENT			
RESIDUAL	▶ RESIDUAL	HISTOLOGICAL GRADE				
RESIDUAL TUMOR		▶ RESIDUAL TUMOR				

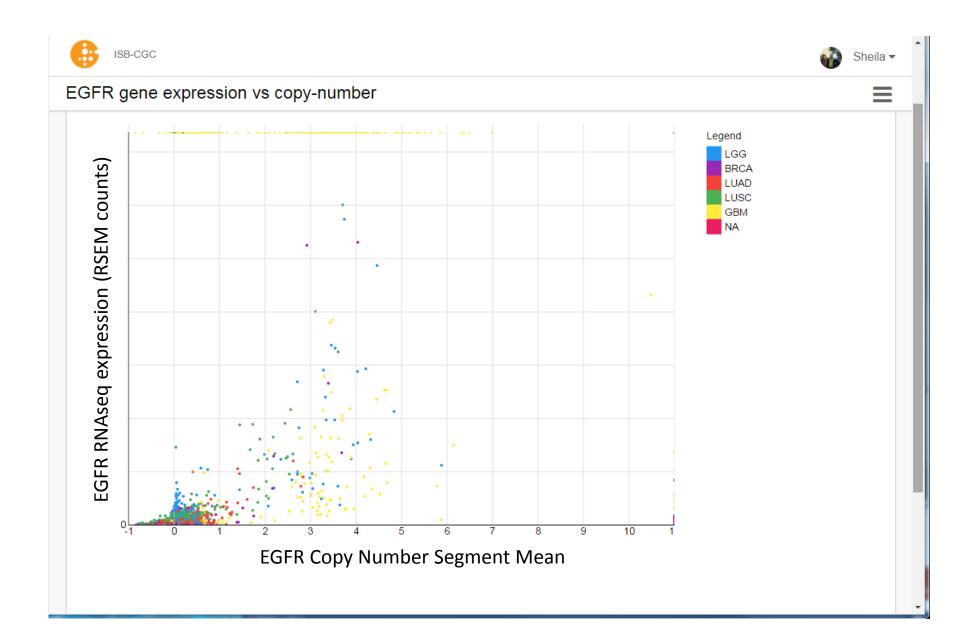
Selected Filters			Clear All
Project: TCGA 🗙	Study: GBM 🗙 Study: LGG	X SampleTypeCo	de: 01 🗙
Clinical Features			
Study	Vital Status	Sample Type	Tumor Tissue Site
	Age at Initial		
Gender	Pathologic Diagnosis	Loss	
	Show	Less	

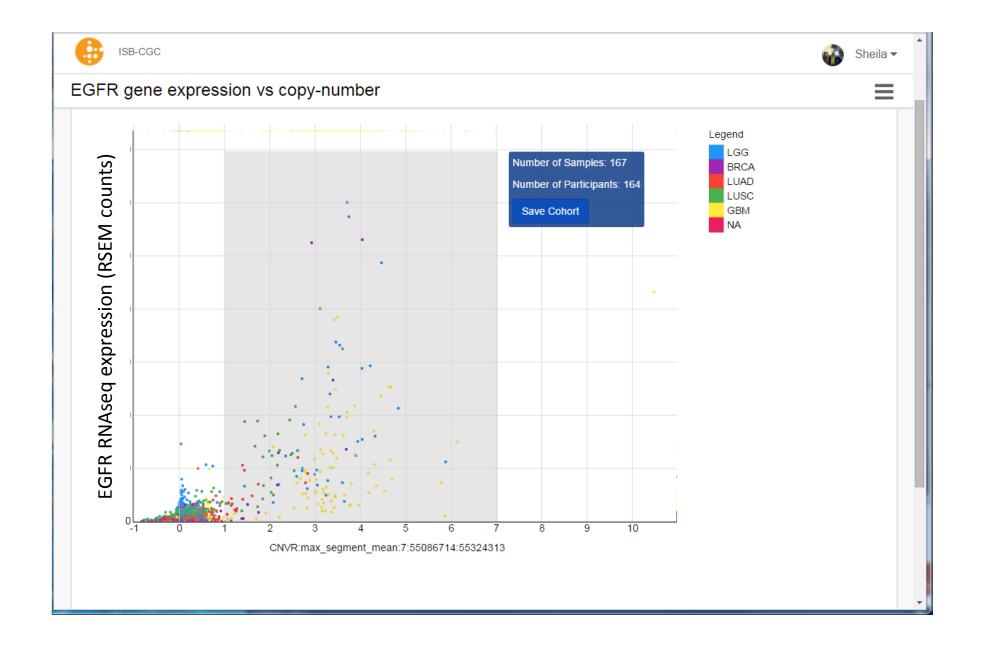
ISB-(•••	ISB-CGC	Create Col	nort		×	Sheila
Create C	Create Coho	Create Cohort	Name:]		Save As New Cohort
Don	Donor	Donor	Data Selected Filters:	Project: TCGA 🗙	Study: GBM X Study: LGG X		Clear All
▼ PROJECT	▶ PROJECT	▶ PROJECT		SampleTypeCode: 01 >	:		
TCGA (2	► STUDY	▶ STUDY				leTypeCod	le: 01 🗙
CCLE (12	▼ VITAL STATUS	✓ VITAL STATUS			Cr	eate Cohort	
STUDY	🗆 Alive (493)	Alive (493)	_	Clinical Features			
VITAL STA	Dead (625)	Dead (625)					
GENDER	None (10)	None (10)		Study	Vital Status	Sample Type	Tumor Tissue Site
	▶ GENDER	• GENDER					
AGE AT DI	AGE AT DIAGNO	AGE AT DIAGNOSIS					
SAMPLETY	▶ SAMPLETYPECC	▶ SAMPLETYPECODE					
TUMOR TI:	TUMOR TISSUE	TUMOR TISSUE SITE					
HISTOLOG							
PRIOR DIA	HISTOLOGICAL 1	HISTOLOGICAL TYPE		Gender	Age at Initial Pathologic Diagnosis		
PATHOLO	PRIOR DIAGNOS	PRIOR DIAGNOSIS					
TUMOR ST	PATHOLOGIC ST	PATHOLOGIC STAGE					
	► TUMOR STATUS	TUMOR STATUS					
NEW TUMC	▶ NEW TUMOR EV	► NEW TUMOR EVENT AFTE	R INITIAL TREATMENT				
HISTOLOG	HISTOLOGICAL	► HISTOLOGICAL GRADE					
RESIDUAL	RESIDUAL TUMC	▶ RESIDUAL TUMOR			Shov	v Less	

ISB-(ISB-CGC	ISB-CGC	ISB-CGC	Search Cohorts and Visualizations O Share Cohort ×	6
Create C	Create Coho	Create Coh	+ Create	🛍 🕜 EGFR study 🗙	
Don	Donor	Donor	Cohorts	Please select the users you would like to share these cohorts with: Phyliss Lee (phyliss.lee@gmail.com)	Last Modified
▼ PROJECT	▶ PROJECT	▶ PROJECT	Visualizations	Phyliss Lee (plee@systemsbiology.org) David Pot (david_pot@sra.com) Zack Rodebaugh (zrodebau@systemsbiology.org)	11/18/2015 4:34 p.m.
TCGA (2	▶ STUDY	STUDY	SeqPeek Plots		11/18/2015 4:22 p.m.
	✓ VITAL STATUS	✓ VITAL STATUS	ocqr cex r lots	Share Cohort	11/18/2015 4:02 p.m.
► STUDY	Alive (493)	Alive (493)		All TCGA Data 24891 isb@test.com	11/09/2015 2:14 a.m.
	Dead (625)	Dead (625)			
► VITAL STA	None (10)	None (10)			
► GENDER					
AGE AT DI	▶ GENDER	▶ GENDER			
► SAMPLET	AGE AT DIAGNO	AGE AT DIAGNC			
► SAIVIPLET	▶ SAMPLETYPECC	► SAMPLETYPEC			
TUMOR TIS		► TUMOR TISSUE			
▶ HISTOLOG	TUMOR TISSUE	► TUMOR TISSUE			
▶ PRIOR DIA	HISTOLOGICAL 1	HISTOLOGICAL			
	▶ PRIOR DIAGNOS	► PRIOR DIAGNOS			
PATHOLOC	PATHOLOGIC ST	▶ PATHOLOGIC S			
TUMOR ST	PATHOLOGIC ST				
► NEW TUM	TUMOR STATUS	TUMOR STATUS			
	► NEW TUMOR EV	► NEW TUMOR EV			
► HISTOLOG	► HISTOLOGICAL (HISTOLOGICAL			
▶ RESIDUAL					
	RESIDUAL TUMC	► RESIDUAL TUM			

Create C	Create Coho	Create Coh		Share Cohort ×	
		Create Con	+ Create	EGFR study ×	
Don	Donor	Donor	Cohorts	Please select the users you would like to share these cohorts with: Phyliss Lee (phyliss.lee@gmail.com)	Last Modified
▼ PROJECT	PROJECT	► PROJECT	Visualizations	Phyliss Lee (plee@systemsbiology.org) David Pot (david_pot@sra.com) Zack Rodebaugh (zrodebau@systemsbiology.org)	11/18/2015 4:34
TCGA (2	▶ STUDY	▶ STUDY	SeqPeek Plots		11/18/2015 4:22
CCLE (12	▼ VITAL STATUS	✓ VITAL STATUS		EC Share Cohort	11/18/2015 4:02
STUDY	🗆 Alive (493)	Alive (493)		All TCGA Data 24891 isb@test.com	11/09/2015 2:14
▶ VITAL STA	Dead (625)	Dead (625)			
GENDER	None (10)	None (10)			
	• GENDER	► GENDER	Add	itional Cohort operations include:	
AGE AT DI	AGE AT DIAGNO	► AGE AT DIAGNC		•	
► SAMPLETY	▶ SAMPLETYPECC	▶ SAMPLETYPEC	•	set operations (union, intersection,	
TUMOR TIS				(
HISTOLOG	TUMOR TISSUE	TUMOR TISSUE		complement)	
PRIOR DIA	HISTOLOGICAL 1	HISTOLOGICAL	•	comment	
PATHOLOG	PRIOR DIAGNOS	PRIOR DIAGNOS		comment	
	PATHOLOGIC ST	▶ PATHOLOGIC S	•	clone	
TUMOR ST	► TUMOR STATUS	► TUMOR STATUS		dalata	
▶ NEW TUM(▶ NEW TUMOR EV	▶ NEW TUMOR EV	•	delete	
HISTOLOG					
▶ RESIDUAL	HISTOLOGICAL	HISTOLOGICAL			
	RESIDUAL TUMC	► RESIDUAL TUM			







Python, R, and SQL for the Computational Scientist:

IPython Interactive Computing







ISB Cancer Genomics Cloud

The ISB-CGC is providing access to TCGA data and computation on the Google Cloud Platform.

📮 Repositories

🛱 People 32 🛛 🖓 Teams 5 🔹 🔅 Settings

ISB-CGC-Webapp

Java Script 🔺 0 🦻 1

Python ★ 0 🎾 1

HTML ★4 ₽2

ISB CGC Webapp Updated 22 hours ago

ISB-CGC-data-proc

code for uploading cancer data into GCS and $\ensuremath{\mathsf{BigQuery}}$

Updated 23 hours ago

examples-R

Analysis examples based on the ISB-CGC hosted TCGA data, using R and R Markdown.

Updated 23 hours ago

examples-Python

★7 ₽2

Analysis examples based on the ISB-CGC hosted TCGA data, using Python and IPython Notebooks.

Updated 3 days ago

https://github.com/isb-cgc/examples-R

https://github.com/isb-cgc/examples-Python

GitHub

SRA ISB	
data	

ISB Cancer Genomics Cloud

The ISB-CGC is providing access to TCGA data and computation on the G

E Repositories	💮 People 32 🖾 Teams 5 🔅 Sett	ings	
ISB-CGC-Webap ISB CGC Webapp Updated 22 hours ago	p	Java Script	★0 ¥1
ISB-CGC-data-pr	OC er data into GCS and BigQuery	Python	★0 §21
Updated 23 hours ago			
examples-R Analysis examples based Markdown.	I on the ISB-CGC hosted TCGA data, using R and R $% \left({{\left({{\left({{\left({{\left({{\left({{\left({{\left($	HTML	★4 ¥2
Updated 23 hours ago			
examples-Pytho Analysis examples based IPython Notebooks.	n I on the ISB-CGC hosted TCGA data, using Python and	1	★ 7 ₽2
Updated 3 days ado			

E README.md

examples-Python

This repository contains analysis examples based on the ISB-CGC hosted TCGA data in BigQuery, using Python, IPython Notebooks, and Google Cloud Datalab.

Where to start?

You can find an overview of the BigQuery tables in this notebook and from there, we suggest that you look at the two "Creating TCGA cohorts" notebooks (part 1 and part 2) which describe and make use of the Clinical and Biospecimen tables. From there you can delve into the various molecular data tables as well as the Annotations table. For now these sample notebooks are intentionally relatively simple and do not do any analysis that integrates data from multiple tables but once you have a grasp of how to use the data, developing your own more complex analyses should not be difficult. You could even contribute an example back to our github repository! You are also welcome to submit bug reports, comments, and featurerequests as github issues.

How to run the notebooks

- 1. Launch your own Cloud Datalab instance in the cloud or run it locally.
- 2. Work through the introductory notebooks that are pre-installed on Cloud Datalab.
- 3. Run git clone https://github.com/isb-cgc/examples-Python.git On your local file system to download the notebooks.
- 4. Import the ISB-CGC notebooks into your Cloud Datalab instance by navigating to the notebook list page and uploading them.

If you are running in the cloud, be sure to shut down Cloud Datalab when you are no longer using it. Shut down instructions and other tips are here.

GitHub



ISB Cancer Genomics Cloud

The ISB-CGC is providing access to TCGA data and computation on the G

🛱 People 🔞 🥻 Teams 5 🛛 🧔 Settings

SB-CGC-Webapp	
SB CGC Webapp	
Jpdated 22 hours ago	

ISB-CGC-data-proc

Python ★ 0 🞉 1

JavaScript 🔺 0 🖗 1

code for uploading cancer data into GCS and BigQuery Updated 23 hours ago

examples-R

HTML ★4 🖗2

Analysis examples based on the ISB-CGC hosted TCGA data, using R and R Markdown.

Updated 23 hours ago

examples-Python

★7 ₽2

Analysis examples based on the ISB-CGC hosted TCGA data, using Python and IPython Notebooks.

Updated 3 days ago

README.md

examples-R

Analysis examples based on the ISB-CGC hosted TCGA data, using R and R Markdown.

To install:

examples-Py

This repository contains analy

Notebooks, and Google Cloud

You can find an overview of th "Creating TCGA cohorts" note

From there you can delve into

notebooks are intentionally rel

have a grasp of how to use the

contribute an example back to

How to run the not

1. Launch your own Cloud E

2. Work through the introduc

3. Run git clone https://g

4. Import the ISB-CGC notel

them.

requests as github issues.

Where to start?

require(devtools) || install.packages("devtools")
install_github("isb-cgc", "examples-R", build_vignettes=TRUE)

To view and run the vignettes.

help(package="ISBCGCExamples")

There are vignettes for each TCGA data type, and more elaborate examples involving analyzing genomizene expression and methylation, and correlating protein and mRNA levels.

The vignettes as **R-markdown** can be found in the examples-R/inst/doc directory, which can serve as exbuiltin BigQuery functions like Pearson correlation, or even how to implement more complex functions like correlation. Queries can be simple character vectors, or standalone files. Results are returned as data.fr bigrquery package to interact with the servers.

The **SQL** files used in the vignettes can be found at examples-R/inst/sql. These are parsed and dispatch using the DisplayAndDispatchQuery function, found in the file of the same name in examples-R/R.

If you have trouble with the OAuth, see examples-R/inst/doc/BigQueryIntroduction.html for some instruction.

Docker

Bioconductor provides an excellent set of docker containers which include R, RStudio Server, and the se packages appropriate for certain use cases.

This R package is also available in a Docker container derived from bioconductor/release_core :

b.gcr.io/isb-cgc-public-docker-images/r-examples

It can be run like so:

docker run -p 8787:8787 -v YOUR_LOCAL_DIRECTORY:/home/rstudio/data \
 b.gcr.io/isb-cgc-public-docker-images/r-examples:latest

If you are running in the cloud, and other tips are here.

JGitHub E README.md



ISB Cancer Genomics Cloud

The ISB-CGC is providing access to TCGA data and computation on the G http://www.isb-cgc.org

Repositories

Reople 32 Teams 5 Settings

ISB-CGC-Webapp	
ISB CGC Webapp	

Updated 22 hours ago

ISB-CGC-data-proc

Python ★ 0 🎾 1

JavaScript 🔺 0 🖗 1

code for uploading cancer data into GCS and BigQuery Updated 23 hours ago

examples-R

HTML ★4 ¥2

Analysis examples based on the ISB-CGC hosted TCGA data, using R and R Markdown.

Updated 23 hours ago

examples-Python

★7 ¥2

Analysis examples based on the ISB-CGC hosted TCGA data, using Python and IPython Notebooks

Updated 3 days ago

E README.md

examples-R

Analysis examples based on the ISB-CGC hosted TCGA data, using R and R Markdown.

To install:

examples-Py

This repository contains analy

Notebooks, and Google Cloud

You can find an overview of th

"Creating TCGA cohorts" note

From there you can delve into

notebooks are intentionally rel

have a grasp of how to use the

contribute an example back to

How to run the not

1. Launch your own Cloud E

2. Work through the introduc

3. Run git clone https://g

4. Import the ISB-CGC notel

If you are running in the cloud,

and other tips are here.

them.

requests as github issues.

Where to start?

require(devtools) || install.pack install github("isb-cgc", "examp)

To view and run the vignettes.

help(package="ISBCGCExamples")

Depends: There are vignettes for each TCGA Imports: gene expression and methylation, ar

Suggests: The vignettes as R-markdown can b Published: builtin BigQuery functions like Pears Author: correlation. Queries can be simple cl bigrquery package to interact with th

The SQL files used in the vignettes using the DisplayAndDispatchQuery

If you have trouble with the OAuth

Docker

Bioconductor provides an excellent packages appropriate for certain use

This R package is also available in a

Author: Cassie Doll [aut], Nicole Deflaux [aut], Siddha

b.gcr.io/isb-cgc-public-docker-images/r-examples

It can be run like so:

bigrquery: An Interface to Google's BigQ Easily talk to Google's BigQuery database from R. 0.1.0 Version: $R (\geq 3.1.0)$

The Comprehensive R Archiv

<u>httr</u>, jsonlite, assertthat, <u>R6</u> (\geq 2.0.0 testthat 2015-01-13 Hadley Wickham [aut, cre], RStud



R Client for Google Genomics API

Bioconductor version: Release (3.2)

Provides an R package to interact with the Google Ger

docker run -p 8787:8787 -v YOUR_LOCAL_DIRECTORY:/home/rstudio/data \ b.gcr.io/isb-cgc-public-docker-images/r-examples:latest

A Google Cloud Datalab Copy Number segments (autosaved)

1 🔒 🗩 🚯

🔊 Notebook → 🔹 🗳 Add Code 🗳 Add Markdown 🚍 Delete 🔺 Move Up 🔸 Move Down 🕨 Run → 🖂 Clear → 📿 Reset Session

Copy Number segments (Broad)

The goal of this notebook is to introduce you to the Copy Number (CN) segments BigQuery table.

This table contains all available TCGA Level-3 copy number data produced by the Broad Institute using the Affymetrix Genome Wide SNP6 array, as of October 2015. (Actual archive dates range from April 2011 to October 2014.) The most recent archives (*eg* broad.mit.edu_UCEC.Genome_Wide_SNP_6.Level_3.143.2013.0) for each of the 33 tumor types was downloaded from the DCC, and data extracted from all files matching the pattern %_nocnv_hg19.seg.txt. Each of these segmentation files has six columns: Sample, Chromosome, Start, End, Num_Probes, and Segment_Mean. During ETL the sample identifer contained in the segmentation files was mapped to the TCGA aliquot barcode based on the SDRF file in the associated mage-tab archive.

In order to work with BigQuery, you need to import the python bigquery module (gcp.bigquery) and you need to know the name(s) of the table(s) you are going to be working with:

import gcp.bigquery as bq cn_BQtable = bq.Table('isb-cgc:tcga_201510_alpha.Copy_Number_segments')

 $\label{eq:stable} From now on, we will refer to this table using this variable ($cn_BQtable), but we could just as well explicitly give the table name each time.$

Let's start by taking a look at the table schema:

%bigquery schema --table \$cn_BQtable

name	type	mode	description
ParticipantBarcode	STRING		
SampleBarcode	STRING		
SampleTypeLetterCode	STRING		
AliquotBarcode	STRING		
Study	STRING		
Platform	STRING		
Chromosome	STRING		
Start	INTEGER		
End	INTEGER		
Num_Probes	INTEGER		
Segment_Mean	FLOAT		

Unlike most other molecular data types in which measurements are available for a common set of genes, CpG probes, or microRNAs, this data is produced using a data-driven approach for each aliquot independently. As a result, the number, sizes and positions of these segments can vary widely from one sample to another.

Navigation Help

Help for Python APIs

You can enter class? or member? within a code cell in the notebook to get help on a Python API.

For example, try str? to get help information on the built-in Python method to convert a value to its string representation.

Additional help topics and links are also available from the menu off the Help icon on the top of the page.

Docs and Samples

The Datalab Guide featuring documentation and sample notebooks is also a great way to check out how you can use Datalab.

Google Cloud Datalab	Cor A Google Cloud Datalab Copy Number segments (autosaved)	0 🛦 🗩 8
🕼 Notebook 🗸 🔹 🖬 Add (Navigation Help
	Now we'll use matplotlib to create some simple visualizations.	Help for Python APIs You can enter class? or member?
Copy Number	import numpy as np import matplotlib.pyplot as plt	within a code cell in the notebook to get help on a Python API.
The goal of this notebook is	o in For the segment means, let's invert the log-transform and then bin the values to see what the distribution looks like:	For example, try str? to get help information on the built-in Python
This table contains all availa Genome Wide SNP6 array, a		method to convert a value to its string representation.
recent archives (<i>eg</i> broad.m types was downloaded from	it. SELECT lin_bin,	Additional help topics and links are also available from the menu off the Help
Each of these segmentation	iles FROM (icon on the top of the page.
During ETL the sample ident the SDRF file in the associate	dn (2.*POW(2,Segment_Mean)) AS lin_CN,	Docs and Samples The Datalab Guide featuring documentation and sample notebooks
In order to work with BigQu the name(s) of the table(s) yo	FROM	is also a great way to check out how you can use Datalab.
<pre>import gcp.bigquer cn_BQtable = bq.Ta</pre>		
From now on, we will refer to table name each time.	HAVING (n > 2000) ORDER BY lin_bin ASC	
Let's start by taking a look at		
% bigquery schema -	pretices(contact ini_oni j+ots,contact ini_oni j),	
	<pre>plt.title('Histogram of Average Copy-Number'); plt.ylabel('# of segments'); plt.xlabel('integer copy-number');</pre>	
•	TRIN 900000 Histogram of Average Copy-Number	
SampleTypeLetterCode S		
-	TRI ¹ 700000	
	TRIN 600000	
	ITE, H	
	0 1 2 3 4 5 6 7 8 9 10 integer copy-number	
Unlike most other molecular microRNAs, this data is prod		
sizes and positions of these s		

Ä Google Cloud Datalab Cor	ä Google Cloud Datalab Copy Number segmen	Google Cloud Datalab Copy Number segments (autosaved)	0 🛓 🗭 0
Notebook - Add Code	🖉 Notebook 🗸 🔹 Add Cor	B Notebook → Add Code C Add Markdown C Delete A Move Up → Move Down > Run → C Clear → O Reset Session	Navigation Help
	Now we'll use matplotlib to create some simple visual	bin ORDER BY bin ASC	Help for Python APIs You can enter class? or member? within a code cell in the notebook to get
Copy Number se	import matplotlib.pyplot as plt	%%sqlmodule getSLhist_1k_amp	help on a Python API.
The goal of this notebook is to in	For the segment means, let's invert the log-transform	SELECT	For example, try str? to get help information on the built-in Python
This table contains all available Genome Wide SNP6 array, as of recent archives (<i>eg</i> broad.mit. types was downloaded from the Each of these segmentation files During ETL the sample identifer the SDRF file in the associated m In order to work with BigQuery, the name(s) of the table(s) you an import gcp.bigquery a cn_BQtable = bq.Table	SELECT lin_bin, COUNT(*) AS n FROM (SELECT Segment_Mean, (2.*POW(2,Segment_Mean)) AS lin_ INTEGER(((2.*POW(2,Segment_Mean)) FROM St WHERE ((End-Start+1)>1000 AND SampleT GROUP BY	GROUP BY bin ORDER BY bin ASC	method to convert a value to its string representation. Additional help topics and links are also available from the menu off the Help icon on the top of the page. Docs and Samples The Datalab Guide featuring documentation and sample notebooks is also a great way to check out how you can use Datalab.
From now on, we will refer to thi table name each time. Let's start by taking a look at the	(n > 2000) ORDER BY lin_bin ASC	<pre>plt.plot(SLhist_1k['bin'],SLhist_1k['n'],'ro:'); plt.plot(SLhistDel['bin'],SLhistDel['n'],'bo-') plt.plot(SLhistAmp['bin'],SLhistDel['n'],'go-',alpha=0.3) plt.xscale('log'); plt.yscale('log');</pre>	
‰bigquery schemata	<pre>plt.bar(CNhist['lin_bin']+0.1,CNhist plt.xticks(CNhist['lin_bin']+0.5,CNF</pre>	<pre>plt.ylabel('# of Segments'); plt.isla('isla('isla'));</pre>	
name type	<pre>plt.title('Histogram of Average Copy plt.ylabel('# of segments');</pre>	Distribution of Segment Lengths	
ParticipantBarcode STRIN	<pre>plt.xlabel('integer copy-number');</pre>	10 ⁶ Discribution of Segment Lengths	
SampleBarcode STRIM	900000 Histogram of Average Co	105	
SampleTypeLetterCode STRIN	800000		
AliquotBarcode STRIN	700000	2	
Study STRIN	600000		
Platform STRIN	와 500000		
Chromosome STRIN	5	4 10 ²	
Start INTE	*		
End INTE			
Num_Probes INTE	20000	10°	
Segment_Mean FLOA		10 ⁹ 10 ¹ 10 ² 10 ³ Segment length (Kb)	
Unlike most other molecular dat microRNAs, this data is produce	integer copy-num	The amplification and deletion distributions are nearly identical and still seem to roughly follow a power-law distribution. We can also infer from this graph that a majority of the segments less than 10Kb in length are either amplifications or deletions, while ~90% of the segments of lengths >100Kb are copy-number neutral.	

+

Unlike most other molecular dat microRNAs, this data is produce sizes and positions of these segn

The histogram illustrates that the vast majority of the either side representing deletions (left) and amplificat

A Google Cloud Datalab Cop	ä Google Cloud Datalab Copy Number segmen	🛓 Google Cloud Datalab Copy Number segmen	A Google Cloud Datalab Copy Number segments (autosaved)
🖉 Notebook 🗸 🛛 🖬 Add Code	Add Core Now we'll use matplotlib to create some simple visual	Ø Notebook → SAdd Cool	Notebook - Add Code Add Markdown Delete A Move Up V Move Down PRun - Clear - O Rese
Copy Number se	import numpy as np	ORDER BY bin ASC %%sqlmodule get5Lhist_1k_amp	And now we'll take a look at histograms of the average copy-number for these three genes. TP53 (in green) shows a significant number of partial deletions (CN<0), while MYC (in blue) shows some partial amplifications more frequently than EGFR, while EGFR (pale red) shows a few extreme amplifications (log2(CN/2) > 2). The final figure shows the same histograms on a semi-log plot to bring up the rarer events.
The goal of this notebook is to ir This table contains all available Genome Wide SNP6 array, as of recent archives (<i>eg</i> broad.mit.	90%sqlmodule getCNhist	SELECT bin, COUNT(*) AS n FROM (SELECT	<pre>binWidth = 0.2 binVals = np.arange(-2+(binWidth/2.), 6-(binWidth/2.), binWidth) plt.hist(tp53CN['avgCN'],bins=binVals,normed=False,color='green',alpha=0.9,label='TP53'); plt.hist(mycCN ['avgCN'],bins=binVals,normed=False,color='blue',alpha=0.7,label='MYC'); plt.hist(egfrCN['avgCN'],bins=binVals,normed=False,color='red',alpha=0.5,label='EGFR');</pre>
types was downloaded from the Each of these segmentation files During ETL the sample identifer the SDRF file in the associated n	COUNT(*) AS n S FROM (S SELECT	(END-Start+1) AS segLength, INTEGER((END-Start+1)/1000) AS b FROM \$t WHERE	plt.legend(loc='upper right'); 6000 5000 5000 EGFR
In order to work with BigQuery, the name(s) of the table(s) you a import gcp.bigquery a	INTEGER(((2.*POW(2,Segment_Mean) , FROM % \$t WHERE ((End-Start+1)>1000 AND SampleT	(END-Start+1)<1000000 AND Sample GROUP BY bin ORDER BY bin ASC	4000
From now on, we will refer to th table name each time.	HAVING	<pre>SLhistDel = bq.Query(getSLhist_1k_de SLhistAmp = bq.Query(getSLhist_1k_am plt.plot(SLhist_1k['bin'],SLhist_1k[plt.plot(SLhistDel['bin'],SLhistDel[plt.plot(SLhistAmp['bin'],SLhistDel[</pre>	
Let's start by taking a look at the	<pre>CNhist = bq.Query(getCNhist,t=cn_BQt bar_width=0.80</pre>	<pre>plt.xscale('log'); plt.yscale('log');</pre>	00
% bigquery schemata	<pre>plt.bar(CNhist['lin_bin']+0.1,CNhist plt.xticks(CNhist['lin_bin']+0.5,CNh plt.title('Histogram of Average Copy</pre>	<pre>plt.xlabel('Segment length (Kb)'); plt.ylabel('# of Segments'); plt.title('Distribution of Segment L</pre>	<pre>plt.hist(tp53CN['avgCN'],bins=binVals,normed=False,color='green',alpha=0.9,label='TP53'); plt.hist(mycCN ['avgCN'],bins=binVals,normed=False,color='blue',alpha=0.7,label='MYC'); plt.hist(egfrCN['avgCN'],bins=binVals,normed=False,color='red',alpha=0.5,label='EGFR');</pre>
name type ParticipantBarcode STRII	<pre>plt.ylabel('# of segments'); plt.xlabel('integer_copy-number');</pre>	10 ⁶ Distribution of Segment Le	<pre>plt.yscale('log'); plt.legend(loc='upper right');</pre>
SampleBarcode STRII SampleTypeLetterCode STRII	900000 Histogram of Average Co	10 ⁵	10 ⁴ TP53 MYC EGFR
AliquotBarcode STRII Study STRII	600000 20	10 ⁴	
Platform STRII Chromosome STRII	6,5000	5 * 10 ²	

10¹

10° _2

-1

1

2

0

4

3

5

6

Unlike most other molecular dat microRNAs, this data is produce sizes and positions of these segn

INTE

INTE

INTE

FLOA

Start

End

Num_Probes

Segment_Mean

The histogram illustrates that the vast majority of the either side representing deletions (left) and amplificat

*#

300000

200000

100000

0

0 1 2 3 4 5 6 integer copy-numb

The amplification and deletion distributions are nearly from this graph that a majority of the segments less th lengths >100Kb are copy-number neutral.

10¹

Segment length (Kb)

10¹

10⁰

10⁰

F D

7:24 PM (23 minutes ago) 📩 Zack Rodebaugh <appengine.noreply@google.com> ***** to me 💌

Hello, dr.breteuil@gmail.com

I invite you to join the Google Developers Console project "ISB-CGC-01-0001". Please click this link to accept my invitation:

https://console.developers.google.com/project/isb-cgc-01-0001/rsvp/?account=dr breteuil@gmail.com

Thanks Zack Rodebaugh zrodebau@systemsbiology.org

Home

Permission

APIs & auth Monitoring

Source Cod

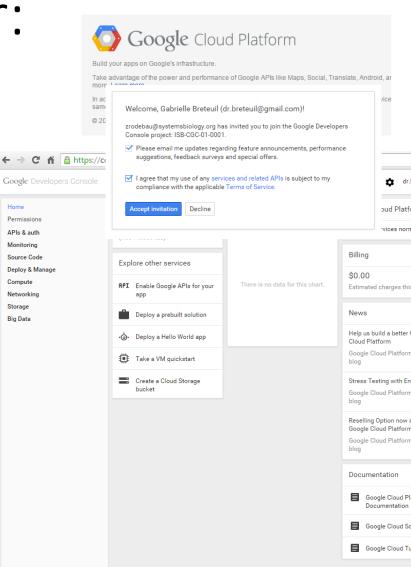
Compute

Storage

Big Data

Networking

Deploy & Manag



programmatic access for the Algorithm Developer:

>your own Google Cloud Project , with automatic access to:

- Cloud Storage
- ➢ BigQuery
- ➢ Google Genomics
- \succ all Google Compute technologies, including:
 - Compute Engine: anything you can do on your laptop/desktop you can do on a VM
 - Container Engine: fully managed and hosted container orchestration - create and deploy clusters in seconds
 - Dataflow: successor to MapReduce

the ISB-CGC API provides programmatic access to the same functionality as the web-app and more:

Cloud Endpoints API (backed by App Engine)

- > authenticate from the command-line
- > make requests to Endpoints API, eg:
 - get list of my cohorts
 - ➤ get cohort details
 - save a new cohort
 - > get list of data files associated with a cohort

plantain:~ kiverson\$ python isb_auth.py Your browser has been opened to visit:

https://accounts.google.com/o/oauth2/auth?redirect_uri=http%3A%2F%2Flocalhost%3A& 80%2F&response_type=code&client_id=907668440978-0ol0griu70qkeb6k3gnn2vipfa5mgl60.app .googleusercontent.com&scope=https%3A%2F%2Fwww.googleapis.com%2Fauth%2Fuserinfo.emai &approval_prompt=force&access_type=offline

If your browser is on a different machine then exit and re-run this application with the command-line parameter

--noauth_local_webserver

Authentication successful. plantain:~ kiverson\$

```
Authentication Status
                localhost:8080/?code=4/pMvbcc2gNPmS...
  Apps
            🛨 Bookmarks 📃 genomics
                                                            Ot
                                          auth
  The authentication flow has completed
 ntain:~ kiverson$ python isb_curl.py https://isb-cgc.ar
v1/cohorts_list?cohort_id=12
                age_at_initial_pathologic_diagnosis".
        te_saved": "2015-11-19 00:05:12",
  email": "kiverson@svstemsbiologv.org".
 kind": "cohort_api#coh
```

Summary

ISB-CGC Phase 1

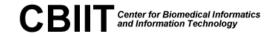
- Low-level sequence and SNP array data as *files* in Cloud Storage
- High-level data and annotations as tables in BigQuery
- Multiple access modes and interfaces:
 - Interactive web-application
 - R, Python, SQL, and JavaScript
 - Endpoint APIs

ISB-CGC Phase 2

- Low-level sequence data in **Google Genomics**
- Variant calls in **Google Genomics** and **BigQuery**









This project has been funded in whole with Federal funds from the National Cancer Institute, National Institutes of Health, Department of Health and Human Services, under Contract No. HHSN261201400007C.



ISB Cancer ISB Genomics Cloud

Questions?

www.isb-cgc.org info@isb-cgc.org

