An Introduction to the ISB-CGC Cancer Resources in the Google Cloud with BigQuery as a Statistical Tool

David Pot and Fabian Seidl for ISB-CGC 2024-03-27



Outline of today's agenda

- Introduction to ISB-CGC and the data commons
- ISB-CGC's approach to derived data
- Data exploration in BigQuery ('Excel-like data tables in the cloud')
- Hands on demonstration of the Google Cloud with focus on BigQuery

The ISB-CGC homepage isb-cgc.org

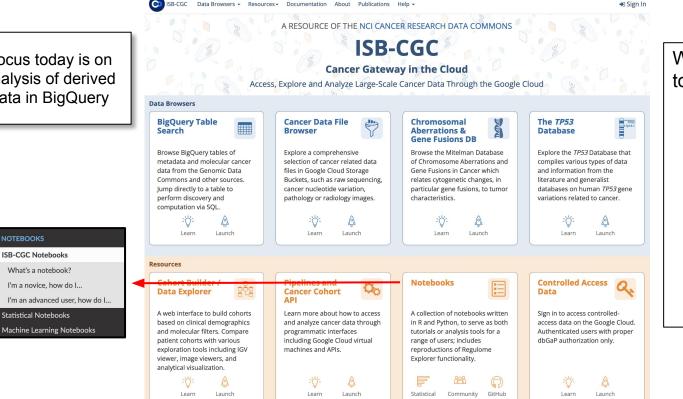
Focus today is on analysis of derived data in **BigQuery**

NOTEBOOKS

ISB-CGC Notebooks

What's a notebook?

Statistical Notebooks



Ways to use data and tools:

- Explore datasets
- Create cohorts
- Run pipelines
- Specialized DBs
 - Mitelman DB of Ο Chromosomal Aberrations & Gene Fusions
 - The TP53 DB \cap
 - caNanoLab Ο

Home

Search Q

- **Cases Cytogenetics**
- Gene Fusions
- Clinical Associations
- Recurrent Chromosome Aberrations
- References
- User Guide About
- Contact

Check out the new View Karyotype Details link for each karyotype in your Cases Cytogenetics search results. Clicking this link displays gains and losses affecting the chromosomes and their genomic coordinates. You can also view the net imbalances in chromosomes for your search result, with charts and images with the click of a button! Read more at User Guide: About Genomic Imbalances in Cytogenetic Cases

Mitelman Database Chromosome Aberrations and Gene Fusions in Cancer

This site has been funded by: National Cancer Institute | Swedish Cancer Society | Swedish Childhood Cancer Foundation

https://mitelmandatabase.isb-cgc.org/

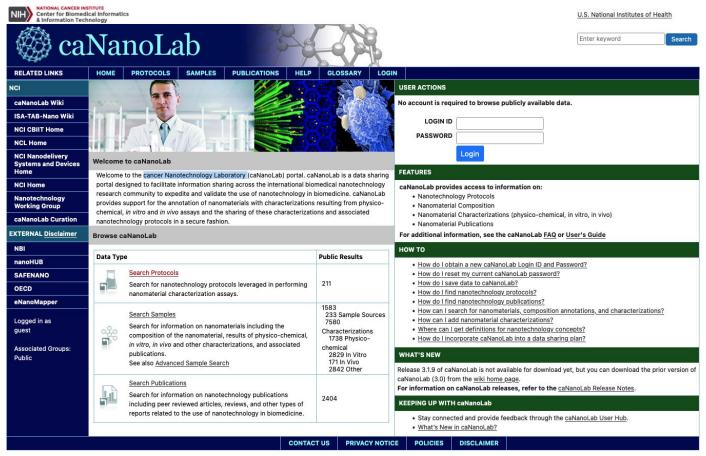
The TP53 Database About User Manual Other Resources Events Release Notes

The *TP53* Database compiles various types of data and information from the literature and generalist databases on human *TP53* gene variations related to cancer. The database is hosted by the National Cancer Institute (NCI) of the United States. The content reflects the R20, July 2019 version

Upcoming *TP53* meetings and conferences can be found in the Events menu. Did you find any issues? Please submit a report.

Functional / Structural Data Ð **Tumor Variants** Ð **Germline Variants H** Explore functional and structural data and frequency Explore data for TP53 tumor variants identified in Explore data for individuals that are carriers of a TP53 statistics of all possible single nucleotide substitutions germline variant and families in which at least one human tumor samples. Includes data on the type and in TP53 exonic sequences, other variants reported in position of variants, detailed information on the tumor family member has been identified as a carrier of a human samples, and validated polymorphisms. in which the variants have been found, and on various germline variant in the TP53 gene. characteristics of the patients in which the tumor developed. Ð **Cell Lines** (\pm) **Mouse Models Experimentally Induced Variants H** Explore data for cell-lines that have been screened for Explore data for mouse models with engineered p53 Explore data for variants in the human TP53 gene TP53 variant and have been published in the scientific that are compiled in the caMOD database or reported in obtained from mutagenicity assays in the Hupki mouse literature, in the Sanger cell-line database, or the Broad the scientific literature. model (MEF cells treated with the indicated carcinogen Cancer cell-line Encyclopedia. agent) or in a yeast assay.

https://tp53.isb-cgc.org/



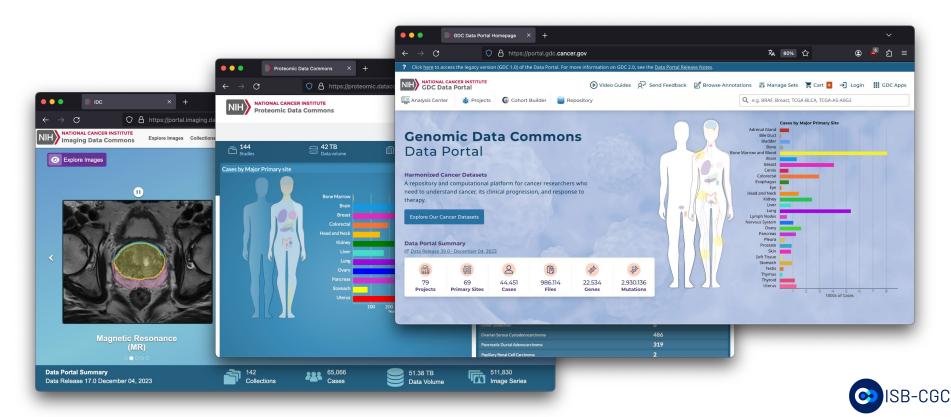
Department of Health and Human Services | National Institutes of Health | National Cancer Institute | USA.gov | Vulnerability Disclosure

NIH...Turning Discovery Into Health®

caNanoLab Release 3.1.9 Build cananolab-3.1.9-0f58a02

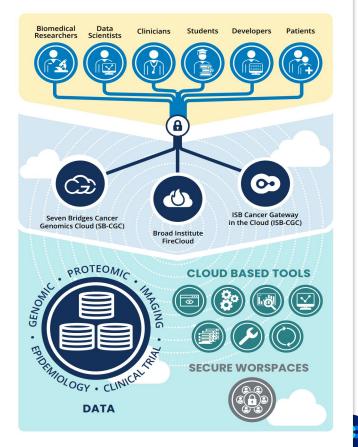
https://cananolab.cancer.gov

The three Data Commons host and control access to different types of cancer data



ISB-CGC's approach to enabling data science in the cloud

- Moving Excel files into the cloud
- Derived molecular data available for query as you need, updated frequently
- Tooling examples provided to enable data mining and Machine Learning of your data
- Sharing of results with those you choose
- Maximum flexibility of scripting and compute for those who desire it



ISB-CGC

ISB-CGC Focuses on Derived Data via BigQuery

	Projects	Clinical/ Biospecimen	File Metadata	Gene Expression	Somatic Mutation	Copy Number	miRNA Expression	DNA Methylation	Protein Expression*	Acetylome	Glycoproteome	Phosphoproteome	Ubiquitylome
	GDC Metadata		х										
	APOLLO	Х	X						X				
	BEATAML1.0	Х	х	х	х								
	CCLE	Х	х	Х	X	X							
	CDDP EAGLE	Х	х		х	х							
	CGCI	Х	X	Х	х	X							
	CMI	Х	х	х	х								
	CPTAC	Х	х	Х	х	X	Х		Х				
	CTSP	Х	х	х									
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	FM	Х	Х								- ا مراغان ، مم		
GDC	GENIE	х	Х								muitiple	well-kno	wn can
GDC	HCMI	Х	х	х	х	Х					-		
	MATCH	Х	X									datasets	5
	MMRF	Х	х	х	х					S			
	MP2PRT	X	х			Х							
	NCICCR	Х	х	х									
	OHSU	Х	X	Х									
	ORGANOID	Х	х	х									
	REBC	Х	х			Х							
	TARGET	Х	х	х	Х	х	Х						
	TCGA	Х	X	Х	х	X	Х	Х	х				
	TRIO	Х	х										
	VAREPOP	X	х										
	WCDT	Х	Х	Х									
	PDC metadata		Х										
	APOLLO	Х	х									Х	
	BROAD	х	Х										
	CBTTC	Х	х						Х			Х	
	CPTAC	Х	Х							Х	Х	Х	X
PDC	Georgetown Proteomics Research												
	Program	х	Х										
	ICPC	Х	Х						х			Х	
	Quantitative Digital Maps of												
	Tissue	Х	Х									7.0	
	TCGA	X	Х				1					Х	



Data wrangling can be onerous, for example GDC has 24,944 individual transcriptome files for just TCGA

Filters								
+ Add a Custom Filter				🛃 Manifest View Images	;)	Add All Files to	Cart 🔋 🕆 Remov	e All From Cart
Experimental Strategy	୯ 🕼 ୨	JSON	TSV	Total of 24,944 Files ± 20,925 Cases	a 105.4	7 GB Q Se	earch	
Name 🔺	Files 🗘							
RNA-Seq scRNA-Seq	24,944 (2.53%) 74 (0.01%)	Cart	Access 🌲	File Name 🍦	Cases	Project 🌲	Data Category 🍦	Data Format 🍦
	show less		Open	J d0ee5ff7-a49a-4633-93a6-40c9e29fb0b7.rna_seq.augmented_star_gene_counts.tsv	<mark>1</mark>	TCGA-BRCA	Transcriptome Profiling	TSV
Wgs Coverage	ৎ [] গ	T	Open	S c58a5583-7b04-4b67-9372-e161e18d7de1.rna_seq.augmented_star_gene_counts.tsv	<mark>月</mark> 1	J TCGA-BRCA	Transcriptome Profiling	TSV
Name A No data for this field	Files 🗘		Open	269c35f0-a4f7-4e30-a69f-f1f3b7b5dace.rna_seq.augmented_star_gene_counts.tsv	<mark>月</mark> 1	TCGA-BRCA	Transcriptome Profiling	TSV
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Aligned Reads Gene Expression Quantification Splice Junction Quantification	73,550 (7.46%) 24,944 (2.53%) 24,944 (2.53%)		Open	be813beb-9b35-4063-9d61-9f49a7fd7706.rna_seq.augmented_star_gene_counts.tsv	<mark>月</mark> 1	J TCGA-BRCA	Transcriptome Profiling	TSV
	93,175 (9.45%)		Open	D 01661d94-fc16-4456-95cf-a5fa4e1e196c.rna_seq.augmented_star_gene_counts.tsv	<mark>Я 1</mark>	DITCGA-BRCA	Transcriptome Profiling	TSV
	show less						Transcriptome	

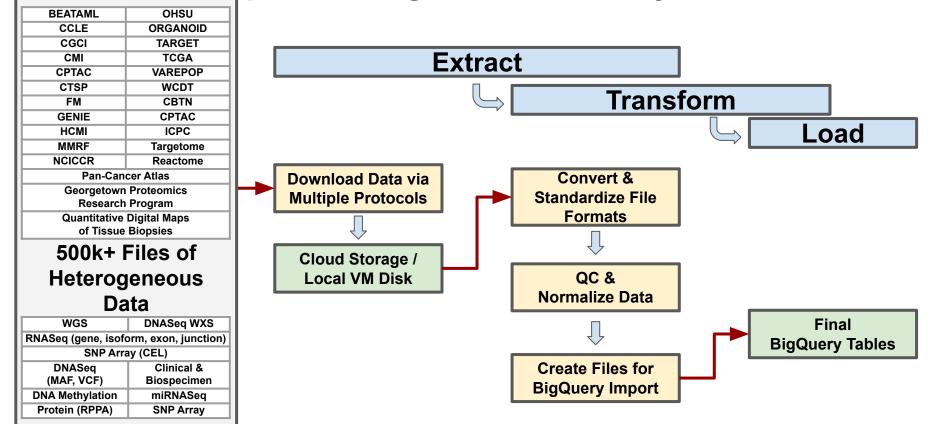


The data commons host a wealth of data from 20 cancer initiative programs

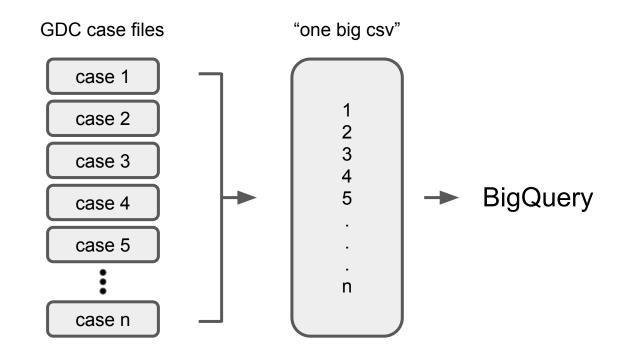
MAJOR PROJECT CASE FILE FILE PRIMARY COUNT SEQ EXP SIZE ID SNV CNV METH CLINICAL BIO COUNT SITES TCGA-BRCA 78.93 TB Adrenal Gland 569 18.723 📕 582 🕨 **Bile Duct** Lung Adenocarcinoma 579 -----Bladder TCGA-LUAD 585 585 585 Bone TCGA-UCEC 519 _ Bone Marrow TCGA-KIRC Brain TCGA-HNSC 518 Breast TCGA-LGG 1000 Cervix TCGA-THCA _ Colorectal TCGA-LUSC Esophagus TCGA-PRAD Eye TCGA-SKCM Head and Neck TCGA-COAD Kidnev TCGA-STAD TCGA-BLCA Liver TCGA-LIHC Luna TCGA-CESC Lymph Nodes TCGA-KIRP Nervous System TCGA-SARC Not Reported TCGA-LAML Other and Ill-defined Sites TCGA-PAAD Ovary TCGA-ESCA Pancreas TCGA-PCPG Pleura TCGA-READ Prostate TCGA-TGCT . Skin TCGA-THYM Soft Tissue TCGA-KICH Stomach TCGA-ACC Testis TCGA-MESO Thymus TCGA-UVM TCGA-DLBC Thyroid TCGA-UCS Uterus TCGA-CHOL 11.3k 10.6k 10.4k 11.1k 10.9k 11.3k 346k 536T 11.0k 11.3k

Case count per Data Category

ISB-CGC runs ETL pipelines to reduce the processing barrier of entry

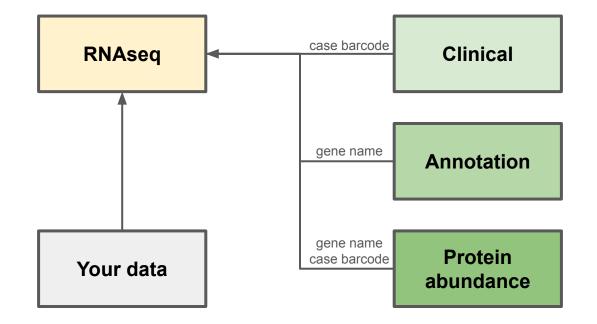


The Google Cloud offers tools to simply host derived data by concatenating these files into a single BQ table





BigQuery enables simple and efficient links between data types





BigQuery enables simple and efficient linking between tables

Que	ery results		▲ SAVE RESULTS ▼			MEXPLORE DATA -			
<	JOB INFORMATION	RESUL	TS	CHART	JSON	EXEC	UTION D	ETAILS	>
Row	case_barcode 🔻		fpkm_u	q_unstranded					
1	TCGA-05-4402			15.8591					
2	TCGA-05-4403			7.5686					
3	TCGA-05-4396			5.1522					
4	TCGA-05-4405			5.5745					
5	TCGA-05-4397			54.251					
6	TCGA-05-4398			17.5632					
7	TCGA-05-4249			4.651					
8	TCGA-05-4250			45.5012					
		Results	per page:	50 🔻	1 – 50 of	598 <	<	>	Х

SELECT

<fields>
FROM `isb-cgc-bq.TCGA.RNAseq_hg38_gdc_current` rna
WHERE <conditionals>



BigQuery enables simple and efficient linking between tables

Que	ery results		🛓 SAVE RES	ULTS 🝷 🛛 🎢	EXPLOR	E DATA	-	\$
<	JOB INFORMATION	RESULT	CHART	JSON	EXECU	TION DE	TAILS	>
Row	case_barcode 🔻		fpkm_uq_unstranded	expcigarettes_p	er_			
1	TCGA-50-6592		36.7756	nui	11			
2	TCGA-50-6591		66.5771	nui	11			
3	TCGA-50-6590		41.5093	2.739726027397				
4	TCGA-50-5946		38.8077	nui	11			
5	TCGA-44-3918		28.5915	1.095890410958				
6	TCGA-44-3917		61.4633	0.876712328767				
7	TCGA-44-3917		15.0178	0.876712328767				
8	TCGA-44-3918		6.2086	1.095890410958				
		Results p	oer page: 50 ▼	1 – 50 of 537	<	<	>	×

SELECT

<fields> FROM `isb-cgc-bq.TCGA.RNAseq_hg38_gdc_current` rna JOIN `isb-cgc-bq.TCGA.clinical_gdc_current` clin WHERE <conditionals>



BigQuery enables simple and efficient linking between tables

<	JOB INFORMATION	RESUL	TS CHART	JSON EX	ECUTION DETAILS	1
Row	case_barcode 👻		fpkm_uq_unstranded	exp_cigarettes_per_	protein_expression	
1	TCGA-86-A4P7		8.9616	nuli	-0.6541634745	
2	TCGA-91-6829		14.4956	5.178082191780	0.399470062	
3	TCGA-91-6828		12.558	nuli	-0.174617102	
4	TCGA-86-A4P8		1.947	nuli	-0.4681542825	
5	TCGA-38-4629		40.1784	5.479452054794	0.739331331	
6	TCGA-38-6178		10.7342	nuli	-0.013338784	
7	TCGA-78-7166		32.8658	2.082191780821	-0.0290081565	
8	TCGA-78-7167		3.7353	3.506849315068	-0.438169383	

SELECT

<fields>
FROM `isb-cgc-bq.TCGA.RNAseq_hg38_gdc_current` rna
JOIN `isb-cgc-bq.TCGA.clinical_gdc_current` clin
JOIN `isb-cgc-bq.TCGA.protein_expression_hg38_gdc_current` prot
WHERE <conditionals>



There is a substantial overlap of data from the same cases across the Commons

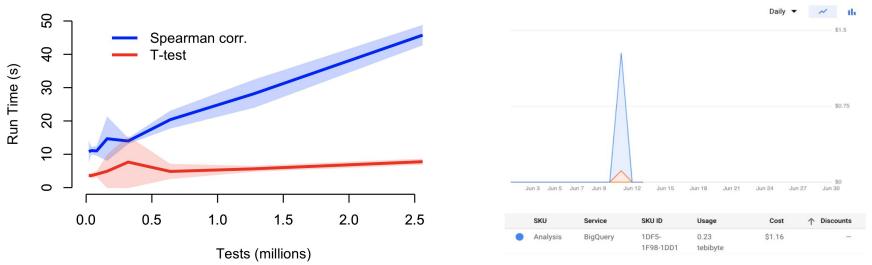
1 sql_full = """ 2 with gdc as (3 with cases as (
4 SELECT 5 distinct 6 case_barcode, 7 case_gdc_id, 8 project_id 9 FROM `isb-cgc-bg.GDC_case_file_metadata.caseData_current` 10), 11 files as (
<pre>13 case_gdc_id, 14 associated_entities_entity_submitter_id, 15 data_type 16 FROM `isb-cgc-bg.GDC_case_file_metadata.fileData_active_current` 17) 18 SELECT</pre>			
<pre>19 cases.case_barcode gdc_id, 20 cases.project_id, 21 array_agg(distinct files.data_type) gdc_data 22 FROM cases JOIN files ON cases.case_gdc_id = files.case_gdc_id 23 GROUP BY cases.case_barcode, cases.project_id 24), 25 pdc as (26 SELECT</pre>			
27 distinct case_submitter_id pdc 1 ids = [] 28 FROM `isb-cgc-bq.PDC_metadata.ali 2 for index, row in df.iterrows(): 1 ids = [] 29 idc as (3 id = set([x for x in row[[(0,3,4]] if x is not None]) 1 id = set([x for x in row[[(0,3,4]] if x is not None]) 31 PatientID idc_id, 4 if len(id) != 1: print(len(id)) 5 id = aspend(id) 32 array_agg(distinct has_derived) 5 id = id.pop() ids.append(id) 7 df['id'] = ids IDC 34 array_agg(distinct has_quantia 7 df['id'] = ids 8 df.head() IDC			
37) 38 SELECT * from gdc	ative	id	
39 FULL OUTER JOIN pdc ON gdc.gdc_id = 40 FULL OUTER JOIN idc ON gdc.gdc_id = 0 TCGA-31-1951 TCGA-OV [Copy Number Segment, Masked Copy Number Segme None None []	0	TCGA-31-1951	
41 42 full_query = client.query(sql_full) 1 TCGA-FS-A1Z7 TCGA-SKCM [Copy Number Segment, Masked Copy Number Segme None None [] []	-	TCGA-FS-A1Z7	
43 df = full_guery.result().to_datafra 2 TCGA-FS-A1ZQ TCGA-SKCM [Copy Number Segment, Masked Copy Number Segment. None [] [] 44 df.head() 2 TCGA-SKCM [Copy Number Segment, Masked Copy Number Segment. None [] []		TCGA-FS-A1ZQ	
3 TCGA-60-2716 TCGA-LUSC [Copy Number Segment, Masked Copy Number Segmen. None TCGA-60-2716 [False] [False] 4 TCGA-77-8153 TCGA-LUSC [Copy Number Segment, Masked Copy Number Segmen None None None []		TCGA-60-2716 TCGA-77-8153	3

B-CGC

BigQuery is a powerful statistical tool that can run hundreds of millions of tests in seconds

Testing BigQuery compute time with statistical tests

- Millions of tests in 40 seconds
- 6.6 billion correlations for \$1.16





How to run statistics inside BigQuery

Big data is hard. Statistics is even harder. Doing statistics on big data is mind-blowingly hard. We are going to provide some tools to start you on a road to making statistics on big data, if not easy, at least possible.





Collaborators:

Ian Mathews, Redivis; Boris Aguilar, Institute for Systems Biology

https://medium.com/@jrossthomson/how-to-run-statistics-inside-bigquery-95c0c6864f23

I'll show hands on navigation of working in the Google Cloud Console

- BigQuery Search Tool
- Google Cloud Console
 - VMs and pipelines
 - Navigating BigQuery
- Create a project
- Upload a small table
- Introductory exploration
- Notebooks section





The benefits of working with ISB-CGC in the cloud

- Multiple specialized databases such as Mittleman
- Easy exploration of existing GDC and PDC data
- Access Virtual Machines and controlled data for customized pipelines
- BigQuery as a tool with scaling Excel functionality
 - Affordable storage and sharing of tabular data
 - Data exploration and quick statistics
 - Derived data from well known reference NCI datasets and annotations
 - Fast links between diverse data types
 - Advanced statistical analyses using Python, R, and Bioconductor
 - Rapidly able to expand to Machine Learning



Redback@isb-cgc.org



ISB-CGC Office Hours

Do you need assistance with getting started? Questions on merging your research with cancer data in the cloud? Or possibly help with troubleshooting?

We have **virtual Office Hours on Tuesdays and Thursdays** for any questions on ISB-CGC functionality or data that you may have. We look forward to speaking with you.

Day of the Week	Time	Host	Link
Tuesday	2:00pm – 3:00pm Eastern	Poojitha Gundluru	http://meet.google.com /jkg-cxke-yzs
Thursday	11:00am – 12:00pm Eastern	Poojitha Gundluru	http://meet.google.com /jai-kgkg-sii



The ISB-CGC team



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David Pot Danna Huffman Fabian Seidl Jacob Wilson Poojitha Gundluru Prema Venkatesan Deena Bleich

