

# Data Commons

# Context

Data powers everything

- policy,
- journalism,
- health,
- science

How do we make it easier?

# Problem is not a shortage of data

Demographics: ACS, Housing Survey, Community Survey,

Economics: BLS, BEA, World Bank, OECD, ...

Health: CDC Wonder, CDC Diabetes, County Health, ...

Climate: NOAA, Hurricanes, Flooding, ...

Genomics: NCBI, ENCODE, ...

Too many formats, schemas, ...

# Using Data: current model

Forage for data sources, track down assumptions/provenance

Clean it up, compile data sources, figure out storage, ...

High upfront costs, sparse ecosystem, few tools, ...

# Google maps made satellite imagery part of everyone's life

**NASA Landsat Science**

Home About Landsat 8 Landsat 9 News How Landsat Helps Education Images Data

**USGS**

**Where to Get Data**

1981-1982	1983-1984	1985-1986	1987-1988	1989-1990	1991-1992	1993-1994	1995-1996	1997-1998	1999-2000	2001-2002	2003-2004	2005-2006	2007-2008	2009-2010	2011-2012	2013-2014	2015-2016	2017-2018	2019-2020	2021-2022	2023-2024
-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------

**Landsat in Action**

**Landsat Collection 1 Reprocessing Status**  
As of January 3, 2013



Google Maps - 2901 Cowper Street 94308

http://maps.google.com/

2901 Cowper Street

2901 Cowper St, Palo Alto, CA 94308

## How do I search for and download Landsat data?

Landsat data products held in the USGS archives can be searched and downloaded at no charge from a variety of sources. Visit the [Landsat Data Access](#) web page for information about how Landsat data products can be downloaded individually or in bulk.

Landsat imagery not found in the USGS archive might have been collected by one of the USGS International Cooperator ground stations, each of which are the primary source of distributing data collected at their location. Station information and contact details can be found at the [USGS International Ground Station \(IGS\) Network](#) website.

**Related Content**

FAQ Publications News Multimedia

Filter Total Items: 7

Select Topic Select Location Select Year Release Date ... Apply Filter Reset

**What are the band designations for the Landsat satellites?**

The sensors aboard each of the Landsat satellites were designed to acquire data in different ranges of frequencies along the electromagnetic spectrum ( View Bandpass Wavelengths for all Landsat Sensors )

The Multispectral Scanner (MSS) carried on Landsat 1,2,3,4 and 5 collected data in four ranges (bands); Thematic Mapper (TM) sensor on...



# We want to do that for data

Navigation: Data, Tools, News, Research, Resources, About, Help

What is the Interactive Data Application?  
BEA's interactive data application is the one-stop shop for accessing BEA data on the fly. The interactive application makes it easier to access and use our statistics by providing a common look and feel for users across the national, international, regional

United States Census Bureau

Search

AMERICAN COMMUNITY SURVEY

DATA TABLES & TOOLS

Data Profiles Selector  
Data Profiles consist of four tables (Social, Economic, Housing, Demographic) to give a broad statistical view of a particular geography. We provide an easy to use Data Profiles geography selector for the most popular geographies: state, county and place. Our full data website, data.census.gov, provides additional geographies for this table type and many more.

IES NCES National Center for Education Statistics

- 2005 Data Products
- 2003 Data Products
- 2001 Data Products
- 1999 Data Products
- 1991-1996 Data Products

2016 DATA PRODUCTS

User's Manual

- NHES:2016 Data File User's Manual (14.4 MB)

Data Files

Early Childhood Program Participation (ECPP)

- ECPP SAS file (5.04 MB)
- ECPP SPSS file (4.27 MB)
- ECPP Stata file (4.12 MB)
- ECPP R file (4.28 MB)
- ECPP ASCII file (4.53 MB)
- ECPP CSV file (15.1 MB)
- ECPP SAS public-use setup file

Public-Use Data Files and Documentation

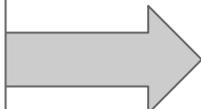
The National Center for Health Statistics (NCHS) is pleased to offer downloadable public-use data files through the Centers for Disease Control and Prevention's (CDC) FTP file server. Users of this service have access to data sets, documentation, and questionnaires from NCHS surveys and data collection systems. Downloading instructions are available in "readme" files.

Public-use data files are prepared and disseminated to provide access to the full scope of the data. This allows researchers to manipulate the data in a format appropriate for their analyses. NCHS makes every effort to release data collected through its surveys and data systems in a timely manner.

Users of NCHS public-use data files must comply with data use restrictions to ensure that the information will be used solely for statistical analysis or reporting purposes.

National Health and Nutrition Examination Survey (NHANES)

for statistics about people and households located in geographies with mid-sized populations? Check out our new Supplemental Tables! These simplified tables provide the most detailed statistics at a lower population threshold than the standard 1-year Detailed Tables. They are the only source of 1-year data for geographies with populations of 20,000 to 50,000. Use the geography selector to get links to the tables on data.census.gov.



From search for dataset,  
download, clean, normalize,  
join ...

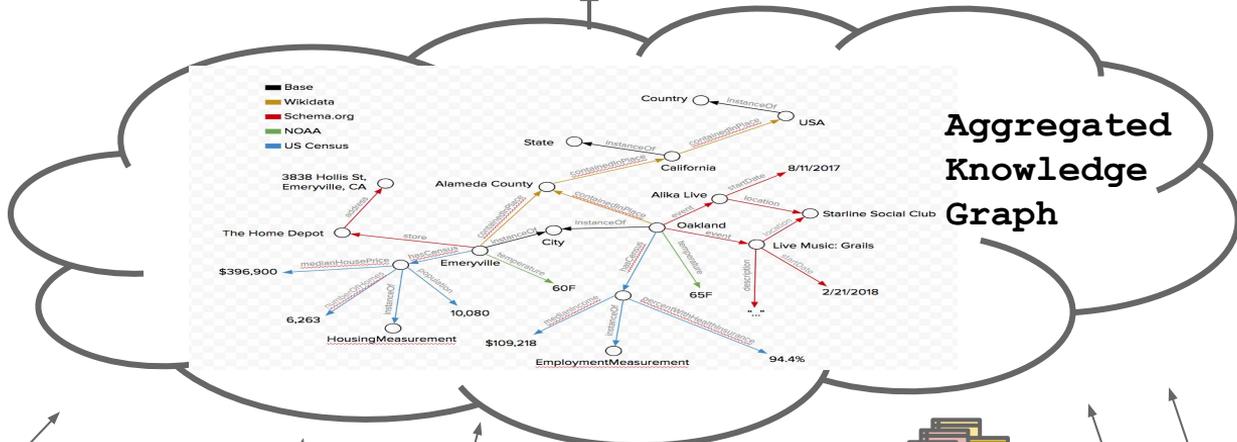
to

Just ask Google



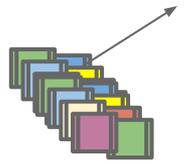
Journalists      Students      Search, News      Researchers      Cloud

APIS



Medicare

Sequence data



CDC



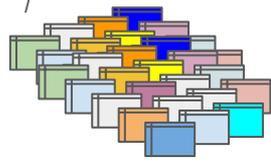
NOAA



FBI



BLS



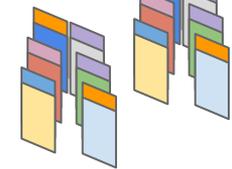
Census (ACS)



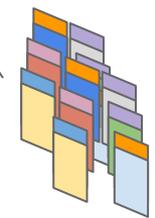
Wikidata



EPA



Landsat



Grid

# Data Commons vs collection of datasets

Collections of datasets (ala NIH Data Commons). Solves the problem of finding the dataset.

But the remaining problems --- cleaning, joining ... remain

Datacommons --- a single KG built by cleaning, normalizing, joining all these datasets

# Data Commons v 0.9

- People, places, ...: Integrated view of Census (ACS), CDC, BLS, BEA, FEC, NOAA, DEA, DOL,... on average 5k variables for every state, county, city, zip, school district, congressional district, ...
- Education: College Scoreboard, NCES, ...
- Disasters: earthquakes, hurricanes, floods, fires
  
- Scientific Collections: Bronx Botanical Gardens, ENCODE, parts of NCBI, GTEx

# Data Commons v 0.9

Applications for 4 categories

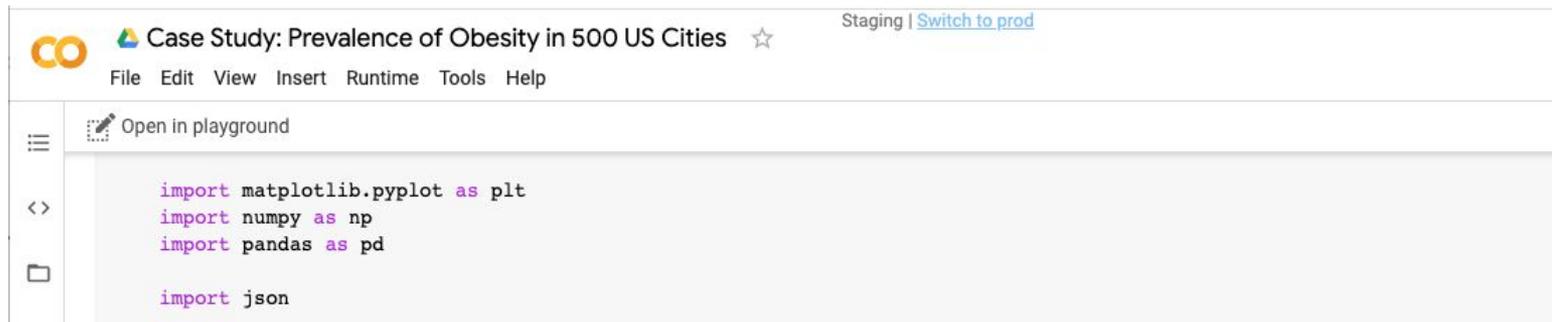
- Researchers
- Students
- Journalists
- Google Users

# APIs

## APIs in

- REST
- Python / Python Notebooks
- SPARQL
- SQL against Big Query
- Google Sheets

# Python Notebooks for students & researchers



The screenshot shows a Jupyter Notebook interface. At the top, there is a title bar with the text "Case Study: Prevalence of Obesity in 500 US Cities" and a star icon. Below the title bar is a menu bar with options: File, Edit, View, Insert, Runtime, Tools, Help. The main area of the notebook contains a code cell with the following Python code:

```
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd

import json
```

Below the code cell, there is a section header: **Case Study: Prevalence of Obesity in 500 US Cities**. The text below the header discusses the correlation of obesity with health and economic factors, citing the CDC and the US Bureau of Labor Statistics. It also mentions the US Census and the Data Commons knowledge graph. The text concludes with a list of variables to be used in a linear regression model.

Obesity is well known to correlate with health factors such as high blood pressure, but is also known to correlate with economic factors such as low-income, unemployment, etc [1][2]. The Center for Disease Control (CDC) provides prevalence percentages on health conditions such as [obesity](#), [high blood pressure](#), and [high cholesterol](#) for approximately 500 major cities in the US (e.g. [San Francisco](#), [New York](#), and [Austin](#)). Meanwhile, the US Bureau of Labor Statistics provides [unemployment rates](#) while the US Census provides [poverty rates](#) for most cities across the United States.

Even though these statistics come from different datasets across different government agencies with different storage formats, Data Commons surfaces each of these in a single, uniform knowledge graph. In fact, you can see this in the [browser](#) by looking at the *provenance* column. Let's use the data in Data Commons to create a linear regression model that incorporates variables:

- Prevalence of high blood pressure
- Unemployment rate
- Percent of population living with income below the poverty line

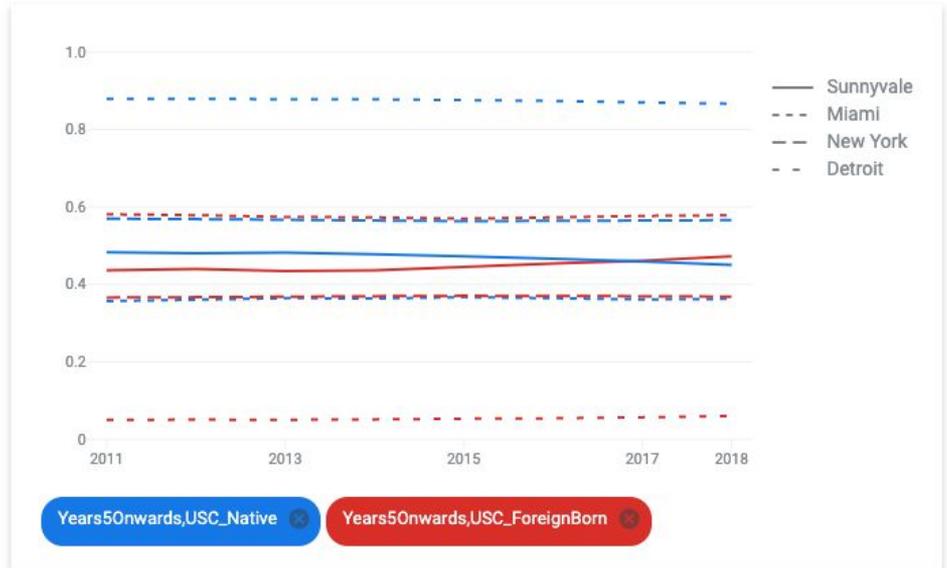
to predict the prevalence of obesity in the 500 cities that the CDC provides data for. One thing you may note is that the US Census also provides employment statistics (you can see this by navigating to the "employment" and "employmentStatus" sections for [San Francisco](#) and observing the different provenances). Our choice of using statistics from the Bureau of Labor Statistics is purely demonstrative, but it would be interesting to see if similar results can be reproduced using US Census employment statistics.

### Select Variables

Per capita

- Demographics<sup>(1610)</sup> >
- Population
- Median age
- Median income
- Age<sup>(402)</sup> >
- Citizenship<sup>(5)</sup> >
- Gender<sup>(1196)</sup> >
- Income<sup>(42)</sup> >
- Marital Status<sup>(67)</sup> >
- Nativity<sup>(104)</sup> ▾
  - Foreign Born<sup>(52)</sup>
  - Native<sup>(52)</sup>
- Place Of Birth<sup>(108)</sup> >
- Poverty Status<sup>(238)</sup> >
- Race<sup>(1064)</sup> >
- Veteran Status<sup>(14)</sup> >
- Educational

Add another place



# Simple Charting Tool

# Simple Correlation Tool

Select Variables

- Income<sup>(42)</sup> >
- Marital Status<sup>(67)</sup> ^
  - Divorced<sup>(19)</sup> >
  - Married And Not Separated<sup>(5)</sup> >
  - Never Married<sup>(19)</sup> >
  - Separated<sup>(5)</sup> >
  - Widowed<sup>(19)</sup> >
- Nativity<sup>(104)</sup> >
- Place Of Birth<sup>(108)</sup> >
- Poverty Status<sup>(238)</sup> >
- Race<sup>(1064)</sup> >
- Veteran Status<sup>(14)</sup> >
- Educational Attainment<sup>(38)</sup> >
- Level Of School<sup>(43)</sup> >
- Health<sup>(1208)</sup> >
- Crime<sup>(11)</sup> ^
  - Crime Type<sup>(11)</sup> ^
    - Arson
    - Combined Crime
    - Property<sup>(4)</sup> >

Select two variables from the left menu, then select a place type.

Plot all the  in

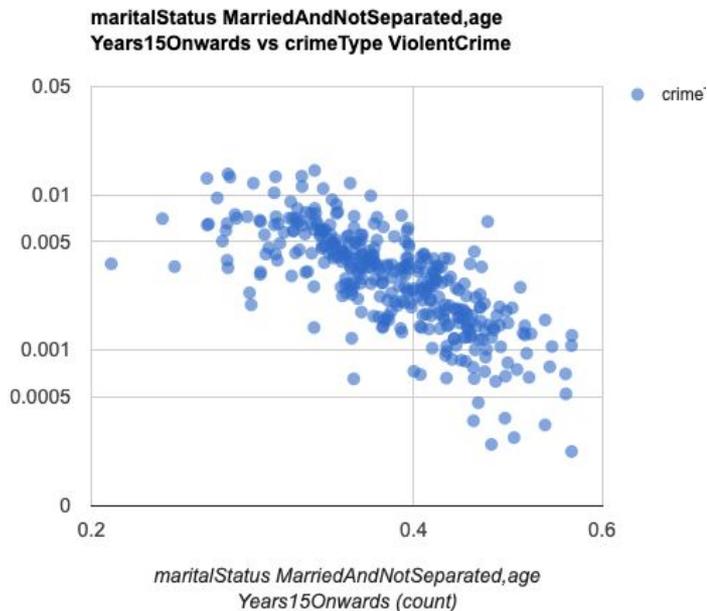
Plot locations with populations between  and  people.

Year:   Per capita  Swap x and y axes  Plot x axis on log scale  Plot y axis on log scale

Mean x: 0.385 | Mean y: 0.004 | Standard deviation x: 0.063 | Standard deviation y: 0.000

Number of datapoints dropped: 85

crimeType ViolentCrime (count)



# Google Sheets API



cwierbonski Copy of Data Commons API Sheets function demo ☆

File Edit View Insert Format Data Tools Add-ons Help All changes saved in Drive



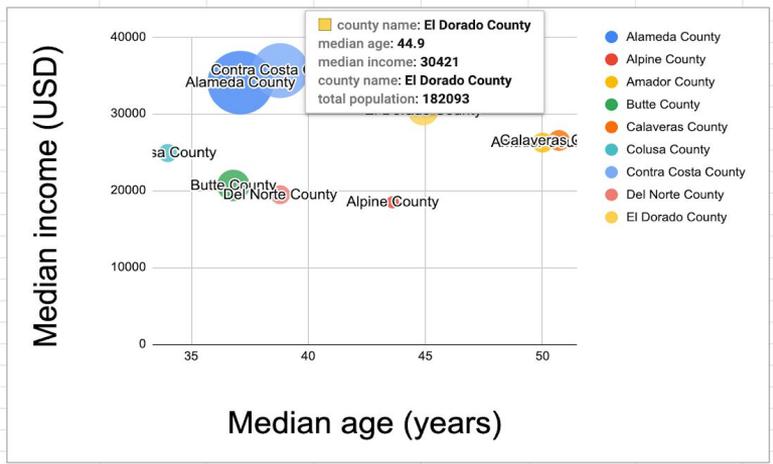
Share



100% % 0.00 123 Default (Ari... 10 B I S A

`=DCGETNAME(D2:D)`

	state	state dcid	county dcid	county name	total population	median age	median income	below poverty line
2	California	geold/06	geold/06001	Alameda County	1584980	37.1	34052	194639
3	Texas	geold/48	geold/06003	Alpine County	1131	43.6	18482	195
4	New York	geold/36	geold/06005	Amador County	36995	50	26246	4191
5	Massachusetts	geold/25	geold/06007	Butte County	222564	36.8	20668	47532
6			geold/06009	Calaveras Count	44767	50.7	26527	5590
7			geold/06011	Colusa County	21396	34	24889	3171
8			geold/06013	Contra Costa Co	1096070	38.8	35618	117915
9			geold/06015	Del Norte County	27788	38.8	19470	5207
10			geold/06017	El Dorado Count	182093	44.9	30421	18831
11			geold/06019	Fresno County	956749	31.4	20870	252187
12			geold/06021	Glenn County	28029	37.2	21344	5721
13			geold/06023	Humboldt County	135034	37.4	21018	28158
14			geold/06025	Imperial County	178206	32.2	16068	40114
15			geold/06027	Inyo County	18373	45.7	26094	2019
16			geold/06029	Kern County	865736	31.1	21458	195744
17			geold/06031	Kings County	150998	31.4	23559	29900
18			geold/06033	Lake County	64158	45.5	20264	15597
19			geold/06035	Lassen County	32645	37	22625	3563
20			geold/06037	Los Angeles Cou	10038400	35.6	24966	1800260
21			geold/06039	Madera County	153187	33.5	20217	33968
22			geold/06041	Marin County	258349	45.4	43504	20964
23			geold/06043	Mariposa County	17789	50.6	26191	2651
24			geold/06045	Mendocino Coun	87544	42.2	21495	17527
25			geold/06047	Merced County	263885	30.4	20681	67417
26			geold/06049	Modoc County	9184	47.8	20607	1673
27			geold/06051	Mono County	14146	38.5	30041	684
28			geold/06053	Monterey County	428441	33.5	23491	69475
29			geold/06055	Napa County	140295	40.3	30918	14020



Low-level functions demo

Other functions demo

Sum: 3454531.7

Explore

# **Biomedical Data Commons**



# Biomedical Data Commons Datasets

- CDC - 500 Cities
- CDC - Diabetes Atlas
- CDC - Wonder
- ChEMBL\*
- ClinVar
- dbSNP
- Dartmouth Medicare Atlas
- Disease Ontology\*
- FDA - Pharmacologic Class\*
- GTEx
- ENCODE
- Entrez Gene
- MeSH\*
- NY Times - COVID-19 Cases + Deaths
- SIDER\*
- SPOKE\*
- UCSC Genome Browser
- US Census - SAHIE
- UniProt\*
- WHO - COVID-19 Cases + Deaths
- WHO - ICD-10 Codes

\*Source: UCSF SPOKE

# Eg: Extract 3 data points on given variants

Given a list of genetic variants

Find:

1. Clinical Significance
2. Functional Category
3. Significant Gene Associations in Whole Blood

# Current Methods to find information

## Option 1: Site Search

Use needed  
database browser  
and search all  
variants  
individually

## Option 2: Download and Analyze Data

Download data,  
read it into  
memory, and  
parse it for the  
needed info  
programmatically

## Option 3: Data Commons

Run 1 query on  
Data Commons

# More on Option 2

- I want to find out more about a list of genetic variants
- Download ClinVar and analyze clinical significance of the variants - **a few hours**
- Download dbSNP and identify the type of genomic region these SNPs are located - **a couple of days**
- Download GTEx and identify which genes a genetic variant is significantly associated in a given tissue - **a week**

# Example Data Commons Query

Issue Query to Data Commons

```
SELECT ?gv ?p
WHERE {
    ?chr dcid "bio/hg38_chr21" .
    ?gv inChromosome ?chr .
    ?gv typeOf GeneticVariant .
    ?gv geneSymbol ?RUNX1 .
    ?gv hg38GenomicPosition ?p
}
```

Time to run query: 21 seconds

# Many Other Commons ...

- Economics
- Covid
- Energy
- ....

# Technical Challenges

- Representation
- Infrastructure
- Inference
  
- General thoughts on KGs ...

# Representational Issues

Representing time, geo, measures, provenance, ...

Stat. aggregates exacerbate the problem of long predicates:  
maleLatinoPopulationUnder...

Triples are easy to understand, relatively widely understood

But real systems need more expressiveness

# Infrastructure

Data Commons has about 50B triples

Bio part is another 100B triples

Wide range of queries and latency requirements

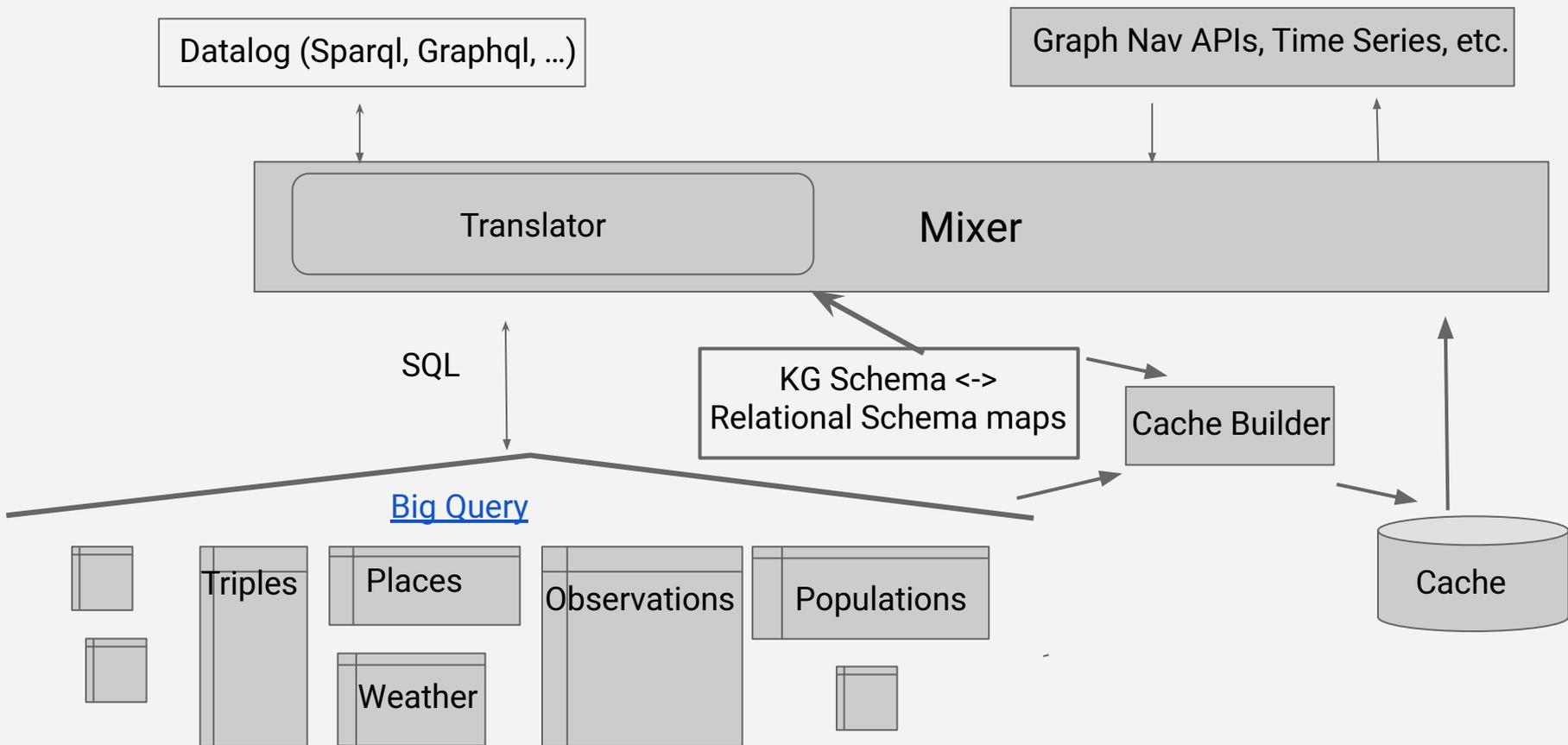
- simple queries that require millisecond responses
- complex sparql queries
- everything in between

# Data Commons approach

Multiple ‘backing’ stores, that support different classes of queries with different latencies

- Big Query ... in memory hash tables
- KG is a view on top of more native encodings
- DLG + function terms + contexts + simple kinds of inference is the real ‘language’ --- serves as the ‘epistemological/knowledge level’

# Serving System



# Ending thoughts

- Vocabulary creep: Google KG, Cyc, Wikidata all have many tens of thousands of 'schema' terms. Human language manages with few thousand terms. How do we bring the compositionality of NL to KR?
- The problems of old AI haven't been solved. They just can't be expressed clearly in today's ML formalisms. e.g., World's tallest mountain yesterday?