

# R API Client Demo Notebook

This is an R Markdown (<http://rmarkdown.rstudio.com>) Notebook. When you execute code within the notebook, the results appear beneath the code. To execute chunks, click the *Run* button within the chunk, or by placing your cursor inside it and press *Cmd+Shift+Enter*.

**FEEDBACK REQUEST** thanks for checking out our API demo. After using this, we would greatly appreciate your feedback. You can send an issue request (<https://github.com/datacommonsorg/api-r/issues>) and mark the issue using the **feedback** label!

**DISCLAIMER** this notebook uses an experimental version of the Data Commons R API Client. The semantics and availability of this API may be subject to change without prior notice!

## Using the R API Client to Analyse Statistics in Data Commons

This tutorial introduces the Data Commons open knowledge graph and discusses how to programmatically access its data through the R API Client.

If you already installed the client (<https://github.com/datacommonsorg/api-r/blob/master/README.md#installing>), you can skip this chunk and just run the next one. If you still need to install the client, make sure you have the dependencies listed in the devtools guide (<https://www.rstudio.com/products/rpackages/devtools/>), then uncomment and run the following chunk.

```
## SKIP THIS CHUNK IF YOU'VE ALREADY INSTALLED THIS CLIENT
# if(!require(devtools)) install.packages("devtools", quiet=TRUE)
# library(devtools, quiet=TRUE)
# devtools::install_github("datacommonsorg/api-r", quiet=TRUE)
```

Let's load the client.

```
library(datacommons, quiet=TRUE)
```

```
## — Attaching packages ————— tidyverse
1.2.1 —
```

```
## ✓ ggplot2 3.2.1    ✓ purrr 0.3.2
## ✓ tibble 2.1.3     ✓ dplyr 0.8.3
## ✓ tidyr 0.8.3      ✓ stringr 1.4.0
## ✓ readr 1.3.1      ✓ forcats 0.4.0
```

```
## — Conflicts — tidyverse_conflicts() —
## ✖ dplyr::filter() masks stats::filter()
## ✖ dplyr::lag() masks stats::lag()
```

```
##
## Attaching package: 'jsonlite'
```

```
## The following object is masked from 'package:purrr':
##
## flatten
```

Using the Data Commons API requires you to provision an API key on GCP. Follow these steps to create your own key. To get an API key, follow the instructions here (<https://datacommons.readthedocs.io/en/latest/started.html#creating-an-api-key>) to get your API key.

Replace `Sys.getenv("API_KEY")` with your API key, such as “123456”. Then run the following block.

```
REPLACE_WITH_YOUR_API_KEY = Sys.getenv("API_KEY")

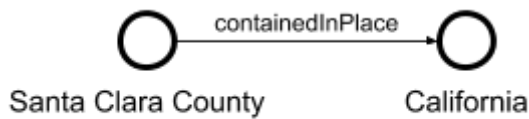
SetApiKey(REPLACE_WITH_YOUR_API_KEY)
```

## What is Data Commons?

Data Commons is an open knowledge graph of structured data. It contains statements about real world objects such as

- Santa Clara County (<https://browser.datacommons.org/kg?dcid=geold/06085>) is contained in the State of California (<https://browser.datacommons.org/kg?dcid=geold/06>)
- The latitude of Berkeley, CA (<https://browser.datacommons.org/kg?dcid=geold/0606000>) is 37.8703
- The population of all persons in Maryland (<https://browser.datacommons.org/kg?dcid=dc/o/6w1c9qk7hxjch>) has a total count of 5,996,080.

In the graph, *entities* (<https://en.wikipedia.org/wiki/Entity>) like Santa Clara County (<https://browser.datacommons.org/kg?dcid=geold/06085>) are represented by nodes. Every node is uniquely identified by its **dcid** (Data Commons Identifier) and has a **type** corresponding to what the node represents. For example, California (<https://browser.datacommons.org/kg?dcid=geold/06>) is identified by the dcid `geoId/06` and is of type `State` (<https://browser.datacommons.org/kg?dcid=State>). *Relations* between entities are represented by directed edges between these nodes. For example, the statement “Santa Clara County is contained in the State of California” is represented in the graph as two nodes: “Santa Clara County” and “California” with an edge labeled “containedInPlace” (<https://schema.org/containedInPlace>) pointing from Santa Clara to California. This can be visualized by the following diagram.



A portion of the Data Commons graph

Here, we call the edge label, “containedInPlace”, the *property label* (or *property* for short) associated with the above relation. We may also refer to “California” as the *property value* associated with Santa Clara County along property “containedInPlace”.

Notice that the direction is important! One can say that “Santa Clara County” is containedInPlace of “California”, but “California” is certainly not contained in “Santa Clara County”! In general, how Data Commons models data is similar to the Schema.org (<https://schema.org>) Data Model as Data Commons leverages schema.org to provide a common set of types and properties. For a broader discussion on how data is modeled in Data Commons, one can refer to documentation on the Schema.org data model (<https://schema.org/docs/datamodel.html>).

## The Data Commons Browser

Throughout this tutorial, we will be using the Data Commons browser (<https://browser.datacommons.org>). The browser provides a human readable way of navigating nodes within the knowledge graph. This is particularly useful for discovering what parameters to pass into the R API Client in order to correctly query for nodes in the graph.

## The R API Client

The R API Client provides functions for users to programmatically access nodes in the Data Commons open knowledge graph. In this tutorial, we will be demonstrating how to use the API access nodes in the Data Commons graph and store their information in a tibble (the tidyverse data frame).

## Using the API To Plot Unemployment Data

The Bureau of Labor Statistics (<https://www.bls.gov>) provides a monthly count for number of individuals who are employed at the State, County, and City level. This data is surfaced in the Data Commons; for example, one can find employment statistics associated with Santa Clara County here (<https://browser.datacommons.org/kg?dcid=dc/p/y6xm2mny8mck1&db=>). Our task for this tutorial will be to extract employment data associated with counties in California from Data Commons using the R API Client and view it in a tibble. We will focus on how functions such as

- `GetPropertyValues`
- `GetPlacesIn`
- `GetPopulations`
- `GetObservations`

operate when using a tibble, as well as how statistical observations are modeled within the Data Commons graph. To see a full list of functions from the Data Commons R API Client, you can use R’s `ls` function:

```
ls(pos = "package:datacommons")
```

```
## [1] "GetObservations"    "GetPlacesIn"         "GetPopulations"
## [4] "GetPropertyLabels"  "GetPropertyValues"   "GetTriples"
## [7] "Query"              "SetApiKey"
```

To begin, we will initialize a tibble with the dcid associated with California: geold/06 (<https://browser.datacommons.org/kg?dcid=geold/06>).

```
# Initialize the tibble
data = tibble('state' = 'geoId/06')

# View the frame
data
```

```
## # A tibble: 1 x 1
##   state
##   <chr>
## 1 geoId/06
```

## Accessing Properties of a Node

For all properties, one can use **GetPropertyValues** to get the associated values. We would like to know that the dcid we have in our data frame belongs to California by getting the name of the node identified by “geold/06”.

To see the documentation for any function, you can bring up the R doc in the Help panel using the built-in `help` function or its `?` shortcut:

```
?GetPropertyValues
```

`GetPropertyValues` accepts the following parameters.

- **dcids** - A vector OR single-column tibble/data frame of string(s) that identify node(s) to get property values for.
- **prop** - The property to get property values for.
- **outgoing** [=TRUE] - An optional flag that indicates the property is oriented away from the given nodes if true.
- **valueType** [=NULL] - An optional parameter which filters property values by the given type.
- **limit** [=100] - An optional parameter which limits the total number of property values returned *aggregated over all given nodes*.

When the dcids are given as a column, the returned list of property values is a corresponding column where the i-th entry corresponds to property values associated with the i-th given dcid. Some properties, like `containedInPlace` (<https://browser.datacommons.org/kg?dcid=containedInPlace>), may have many

property values. Consequently, the cells of the returned series will always contain a list of property values. Let's take a look:

```
# Call GetPropertyValues. Because the return value will be a column, we can
# assign it directly to a new column in our frame.
data$state_name = GetPropertyValues(select(data, state), 'name')

# Display the frame
data
```

```
## # A tibble: 1 x 2
##   state      state_name
##   <chr>      <named list>
## 1 geoId/06 <chr [1]>
```

For each list in the returned column, we may need to expand each element of the list into its own row. This can easily be achieved by calling `unnest`.

```
data = unnest(data)
data
```

```
## # A tibble: 1 x 2
##   state      state_name
##   <chr>      <chr>
## 1 geoId/06 California
```

## Working with Places in Data Commons

We would now like to get all Counties contained in California. This can be achieved by calling `GetPropertyValues`, but because a large fraction of use cases will involve accessing geographical locations, the API also implements a function, **`GetPlacesIn`**, to get places contained within a list of given places. `GetPlacesIn` accepts the following parameters.

- **`dcids`** - A vector or a single-column tibble/data frame of dcid strings identifying places to get places in.
- **`placeType`** - The type of places contained in the given dcids to query for.

When `dcids` is specified as a column, the return value is a column with the same format as that returned by `GetPropertyValues`. Let's call `GetPlacesIn` to get all counties within California.

```

# Call GetPlacesIn to get all counties in California. Here the type we use is
# "County". See https://browser.datacommons.org/kg?dcid=County for examples
# of
# other nodes in the graph with type "County".
data$county = GetPlacesIn(select(data, state), 'County')

# Display the frame
data

```

```

## # A tibble: 1 x 3
##   state    state_name county
##   <chr>    <chr>      <named list>
## 1 geoId/06 California <chr [58]>

```

Notice that both `state_name` and `county` are columns with lists. `unnest` will automatically flatten all columns that have lists in their cells.

```

# Call flatten frame
data = unnest(data)

# Display the frame
data

```

```

## # A tibble: 58 x 3
##   state    state_name county
##   <chr>    <chr>      <chr>
## 1 geoId/06 California geoId/06001
## 2 geoId/06 California geoId/06003
## 3 geoId/06 California geoId/06005
## 4 geoId/06 California geoId/06007
## 5 geoId/06 California geoId/06009
## 6 geoId/06 California geoId/06011
## 7 geoId/06 California geoId/06013
## 8 geoId/06 California geoId/06015
## 9 geoId/06 California geoId/06017
## 10 geoId/06 California geoId/06019
## # ... with 48 more rows

```

Let's now get the names for the county column, as we did for the state column.

```
# Get the names of all counties in California.
data$county_name = GetPropertyValues(select(data, county), 'name')
data = unnest(data)

# Display the frame
data
```

```
## # A tibble: 58 x 4
##   state   state_name county   county_name
##   <chr>   <chr>      <chr>      <chr>
## 1 geoId/06 California geoId/06001 Alameda County
## 2 geoId/06 California geoId/06003 Alpine County
## 3 geoId/06 California geoId/06005 Amador County
## 4 geoId/06 California geoId/06007 Butte County
## 5 geoId/06 California geoId/06009 Calaveras County
## 6 geoId/06 California geoId/06011 Colusa County
## 7 geoId/06 California geoId/06013 Contra Costa County
## 8 geoId/06 California geoId/06015 Del Norte County
##
## 9 geoId/06 California geoId/06017 El Dorado County
## 10 geoId/06 California geoId/06019 Fresno County
## # ... with 48 more rows
```

## Working with Statistical Observations

Finally, we are ready to query for Employment statistics in Data Commons. Before proceeding, we discuss briefly about how Data Commons models statistics.

Statistical observations can be separated into two types of entities: the *statistical population* that the observation is describing, and the *observation* itself. A statistic such as

The number of employed individuals living in Santa Clara in January 2018 was 1,015,129

can thus be represented by two entities: one capturing the population of *all persons who are unemployed* and another capturing the observation 1,015,129 made in January 2018. Data Commons represents these two entity types as `StatisticalPopulation` and `Observation`. Let's now focus on each separate one.

### StatisticalPopulations

Consider the node's browser page (<https://browser.datacommons.org/kg?dcid=dc/p/y6xm2mny8mck1&db=>) representing the `StatisticalPopulation` of all employed persons living in Santa Clara County.

About: dc/p/y6xm2mny8mck1

☒ Raw Graph View  
☐ Chart View

Property	Value	Provenance
<a href="#">typeOf</a>	<a href="#">StatisticalPopulation</a>	<a href="#">bls.gov</a>
<a href="#">populationType</a>	<a href="#">Person</a>	<a href="#">bls.gov</a>
<a href="#">location</a>	<a href="#">Santa Clara County</a>	<a href="#">bls.gov</a>
<a href="#">numConstraints</a>	1	<a href="#">bls.gov</a>
<a href="#">employment</a>	BLS_Employed	<a href="#">bls.gov</a>
<a href="#">localCuratorLevelId</a>	LAUS_POP_CN060850000000005	<a href="#">bls.gov</a>
<a href="#">dcid</a>	dc/p/y6xm2mny8mck1	<a href="#">datacommons.org</a>

Observation of count

<a href="#">observedNode of</a>	<a href="#">2010-05, count=829149</a>	<a href="#">bls.gov</a>
<a href="#">observedNode of</a>	<a href="#">1990-11, count=805614</a>	<a href="#">bls.gov</a>

The population of employed individuals in Santa Clara County

At the top of the browser page, we are presented a few properties of this node to consider:

- The **typeOf** this node is `StatisticalPopulation` as expected.
- The **populationType** is `Person` telling us that this is a statistical population describing persons.
- The **location** of this node is Santa Clara County (<https://browser.datacommons.org/kg?dcid=geold/06085>) telling us that the persons in this statistical population live in Santa Clara County.
- The **dcid** property tells us the dcid of this node
- The **localCuratorLevelId** tells us information about how this node was uploaded to the graph. For the purposes of this tutorial, we can ignore this field.

There are two other properties defined: `numConstraints` and `employment`. These two properties help us describe entities contained in this statistical population. Properties used to describe the entities captured by a `StatisticalPopulation` are called *constraining properties*. In the example above, `employment=BLS_Employed` is a constraining property that tells us the Statistical Population captures employed persons. `numConstraints` denotes how many constraining properties there are, and in the example above, `numConstraints=1` tells us that `employment` is the only constraining property.

To query for `StatisticalPopulation`s using the Data Commons R API Client, we call **GetPopulations**. The function accepts the following parameters.

- **dcids** - A list or Pandas Series of dcids denoting the locations of populations to query for.
- **populationType** - The `populationType` of the `StatisticalPopulation`
- **constraintsPVMMap** - An optional map from constraining property to the value that the `StatisticalPopulation` should be constrained by.

When a column is provided to `dcids`, the return value is a column with each cell populated by a single dcid and not a list. This is because the combination of `dcids`, `populationType`, and optionally `constraintsPVMMap` always map to a unique statistical population `dcid` if it exists.

Let's call `GetPopulations` to get the populations of all employed individuals living in counties specified by the `county` column of our tibble.



```
# First we create the constraintsPVMMap
props <- list(employment = 'BLS_Employed')

# We now call get_populations.
data$employed_pop <- GetPopulations(select(data, county), 'Person', constraintsPVMMap=props)

# Display the DataFrame. Notice that we don't need to flatten the frame.
print(head(data))
```

```
## # A tibble: 6 x 5
##   state   state_name county      county_name   employed_pop
##   <chr>   <chr>      <chr>      <chr>         <chr>
## 1 geoId/06 California geoId/06001 Alameda County dc/p/69mjbvx3c6m38
## 2 geoId/06 California geoId/06003 Alpine County   dc/p/38cz2egkw07rb
## 3 geoId/06 California geoId/06005 Amador County  dc/p/d0tgg3hcqwevh
## 4 geoId/06 California geoId/06007 Butte County   dc/p/fn9p8qqppptdf
## 5 geoId/06 California geoId/06009 Calaveras County dc/p/cgf7kpwcz56b
## 6 geoId/06 California geoId/06011 Colusa County  dc/p/d1118skxd9gl2
```

## Observations

At the bottom of the page describing the `StatisticalPopulation` of all employed persons living in Santa Clara County is a list of observations made of that population. This is the browser page (<https://browser.datacommons.org/kg?dcid=dc/o/b5ylgwwh1d5s1>) for the node representing the observed count of this population for January 2018.

About: [dc/o/b5ylgwwh1d5s1](#)

Property	Value	Provenance
<a href="#">typeOf</a>	<a href="#">Observation</a>	<a href="#">bls.gov</a>
<a href="#">measuredProperty</a>	<a href="#">count</a>	<a href="#">bls.gov</a>
<a href="#">observationDate</a>	2018-01	<a href="#">bls.gov</a>
<a href="#">observationPeriod</a>	P1M	<a href="#">bls.gov</a>
<a href="#">measuredValue</a>	1.015129e+06	<a href="#">bls.gov</a>
<a href="#">measurementMethod</a>	<a href="#">BLSSeasonallyUnadjusted</a>	<a href="#">bls.gov</a>
<a href="#">observedNode</a>	<a href="#">dc/p/y6xm2mny8mck1</a>	<a href="#">bls.gov</a>
<a href="#">dcid</a>	<a href="#">dc/o/b5ylgwwh1d5s1</a>	<a href="#">datacommons.org</a>

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The number of employed individuals in Santa Clara County

In this page, there are a few properties to consider:

- The **typeOf** this node is `Observation` as expected.
- The **measuredProperty** is `count` telling us that this observation measures the number of persons in this statistical population.
- The **measurementMethod** is `BLSSeasonallyUnadjusted` to indicate how the observation was adjusted. We can click on that link to see what `BLSSeasonallyUnadjusted` means.
- The **observationDate** is `2018-01`. This date is formatted by ISO8601 ([https://en.wikipedia.org/wiki/ISO\\_8601](https://en.wikipedia.org/wiki/ISO_8601)) standards.
- The **observedPeriod** is `P1M` to denote that this observation was carried over a period of 1 month.
- The **observedNode** which tells us what the node is being observed by the observation. Here its value is `dc/p/y6xm2mny8mck1` which is the dcid of the population of employed persons living in Santa Clara.

The final property of interest is **measuredValue**. This property tells us that the raw value observed by the observation (whose value is 1,015,129) in this case. The `measuredValue` is also a *statistic type* associated with the observation. For a single observation, there could be many statistics that describe it. One would be the raw value represented by `measuredValue`, while others include `meanValue`, `medianValue`, `marginOfError`, and more.

These parameters are useful for deciding what values to provide to the API. To query for `Observation`s using the R API Client, we call **GetObservations** which accepts the following parameters.

- **dcids** - A vector or single column tibble/data frame of dcids of nodes that are observed by observations being queried for.
- **measuredProperty** - The `measuredProperty` of the observation.
- **statsType** - The statistical type of the observation.
- **observationDate** - The `observationDate` of the observation.
- **observationPeriod** [=NULL] - An optional parameter specifying the `observationPeriod` of the observation.
- **measurementMethod** [=NULL] - An optional parameter specifying the `measurementMethod` of the observation.

One thing to note is that not all `Observation`s will have a property value for `observationPeriod` and `measurementMethod`. For example, the number of housing units with 3 bedrooms in Michigan (<https://browser.datacommons.org/kg?dcid=dc/o/6x1emqkzvjq2&db=>) does not have `observationPeriod` while number of Other Textile Mills (<https://browser.datacommons.org/kg?dcid=dc/o/mc6yc1v1004y5&db=>) does not have `measurementMethod`. These parameters are thus optional arguments to the `GetObservations` API function.

When a column is provided to `dcids`, the return value is again a column with each cell populated by the statistic and not a list of statistics. The combination of the above parameters always map to a unique observation if it exists. If the statistic does not exist for the given parameters, then the cell will contain `NaN`.

Let's get the `measuredValue` of observations of the column of populations we just queried for.

```
# Call GetObservations We are passing into the parameters the values that we
# saw in the link above.
data$employed_count <- GetObservations(
  select(data, employed_pop),
  'count',
  'measuredValue',
  '2018-01',
  observationPeriod='P1M',
  measurementMethod='BLSSeasonallyUnadjusted')

# Display the DataFrame
print(head(data))
```

```
## # A tibble: 6 x 6
##   state   state_name county   county_name   employed_pop   employed_cou
##   <chr>   <chr>      <chr>      <chr>          <chr>          <db
## 1 geoId/... California geoId/06... Alameda Coun... dc/p/69mjbvx3c...      8197
## 2 geoId/... California geoId/06... Alpine County  dc/p/38cz2egkw...        5
## 3 geoId/... California geoId/06... Amador County  dc/p/d0tgg3hcq...     138
## 4 geoId/... California geoId/06... Butte County   dc/p/fn9p8qqpp...     967
## 5 geoId/... California geoId/06... Calaveras Co... dc/p/cgf7kpwcz...     202
## 6 geoId/... California geoId/06... Colusa County  dc/p/dll18skxd...      85
```

## Wrapping Things Up

Now that we have our tibble populated, we can analyse the data. Let's plot the counts in a bar chart to see what we find.

```

# Make a deep copy
final_data <- data
# Sort by employment count
final_data <- final_data[order(final_data$employed_count, decreasing = TRUE
),]
# make county_name an ordered factor so ggplot2 doesn't try to sort it...
final_data$county_name <- factor(final_data$county_name, levels = final_data
$county_name)

# Plot the bar chart
# final_data.plot.bar(x='county_name', y='employed_count', rot=90, figsize=
(12, 8))
# plt.show()

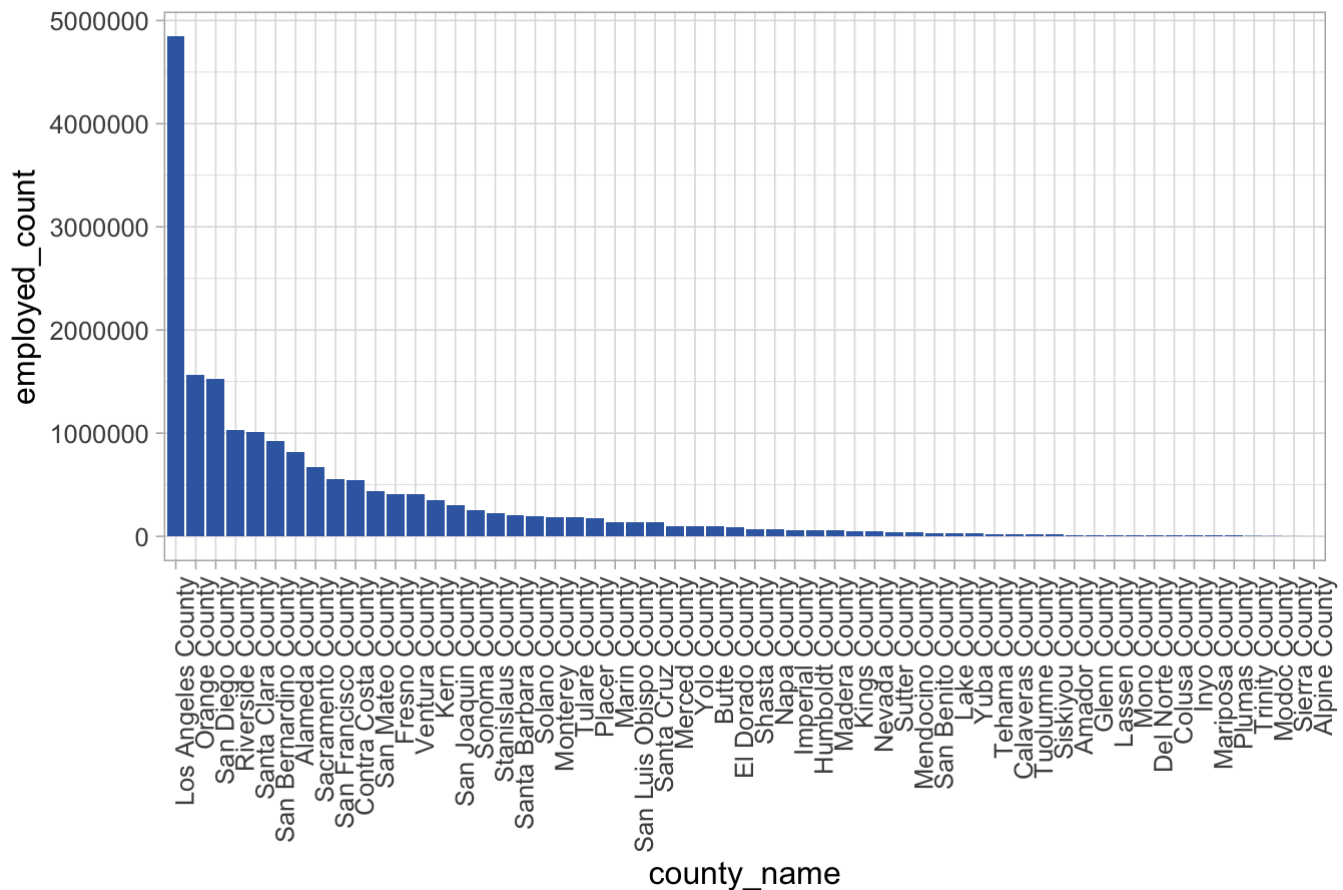
# barplot(final_data$employed_count, names.arg = final_data$county_name)

theme_set(theme_light(base_size = 12))
options(scipen=999)
g <- ggplot(final_data, aes(x=county_name, y=employed_count)) +
  geom_bar(stat = "identity", fill = "#396AB1") +
  theme(axis.text.x = element_text(angle = 90, hjust = 1)) +
  ggtitle("Number of Employed Individuals in California Counties, January 20
18")

g

```

## Number of Employed Individuals in California Counties, January 2018



This wraps up our tutorial on how to use the Data Commons R API Client to access statistics in the knowledge graph and view it in a tibble. From this tutorial we should now know

- How statistics are modeled in the graph
- What Node API methods are available for accessing these statistics
  - `GetPropertyValues`
  - `GetPlacesIn`
  - `GetPopulations`
  - `GetObservations`
- How to use API methods to access these statistics

We did not cover 2 Node API methods:

- `GetTriples`: get all triples (subject-predicate-object) where the specified node is either a subject or an object.
- `GetPropertyLabels`: get property labels of each specified node.

Remember that you can learn more about these functions, or any function, using the R `help(function)` command or the `?function` shortcut. To see a list of all functions in a package, use the R `ls(pos = "package:datacommons")` command.

Lastly, one may wish to perform a more nuanced analysis of employment. For example, to really understand employment rates in a given county, we may wish to normalize the counts we queried for by the counts of various populations such as

- All persons in a county.
- All persons of working age in a county.
- All persons with or without a disability in a county.

We hope that with the tools provided in this notebook, such analysis will be easier and quicker to perform.  
Happy hunting!