### Background

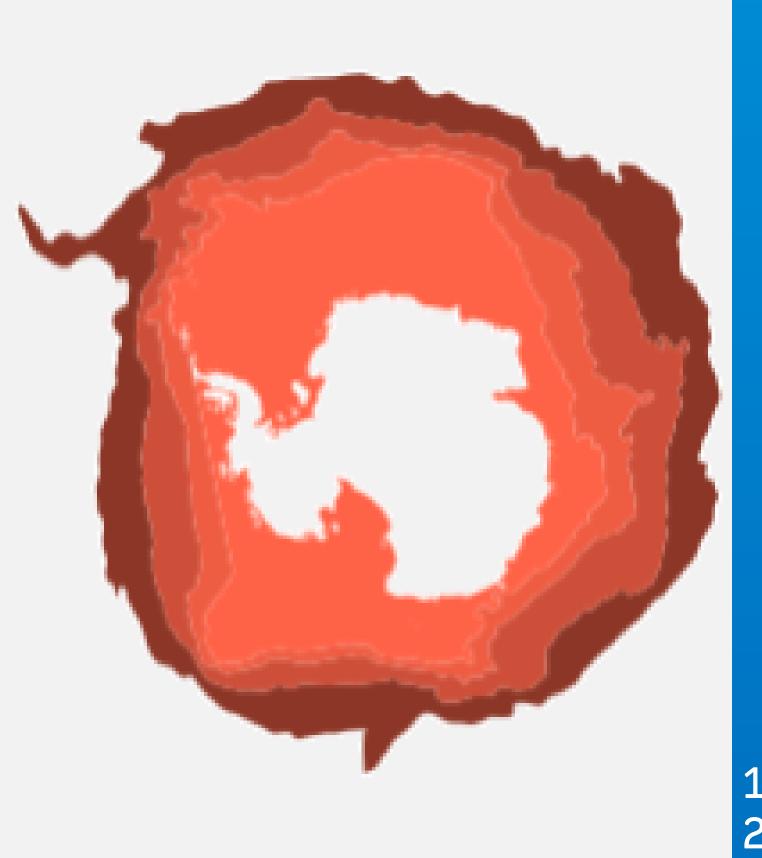
The Southern Ocean is considered a large part of the global carbon cycle and phytoplankton are a key driver by converting  $CO_2$  to carbon in the deep ocean. Current databases examine phytoplankton presence and CO2 flux but don't take community species composition into account.

We aim to determine the relationship between phytoplankton species and biological production

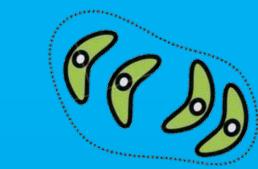
### Methods

I added a frontal zone label to all the data by reading in a fixed front definition (Kim and Orsi, 2014), converting to polygon shapefiles in python, and performing a geospatial join.

Researchers commonly filter data by the frontal zone and sector, so the join is performed at insert time (rather than runtime) which greatly increases query speed for the user



# SOPhy

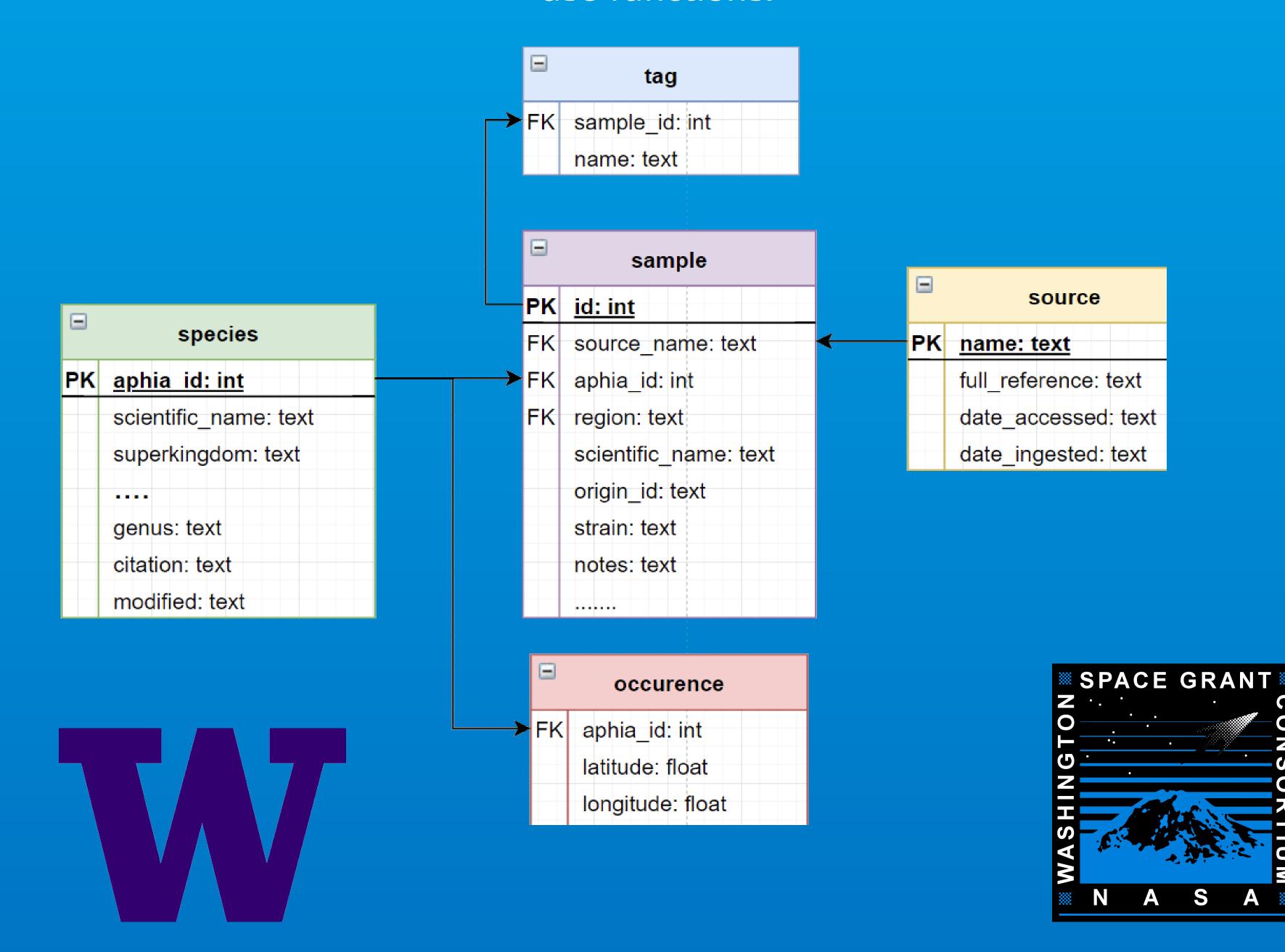


## Southern Ocean Phytoplankton Database

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Combining in situ datasets from cruises and adding features such as full taxonomy and labeled Southern Ocean fronts and sectors.

Database schema and backend in SQLite. Used Python to transform the data and add new features such as visualizations, full taxonomy records, and ease of use functions.



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### Schema

The first step of this project was to design the schema and normalize the dataset. This is the heart of the database and an evolving document on version 8.

Below is a sample query that counts the total amount of genera across the data. Genus is not present in the main table however the data can be joined using the aphia\_id key which represents the exact species of phytoplankton

select count(\*) as entries, genus

from sample, microscopy

where sample.aphia\_id = microscopy.aphia\_i

group by genus

order by entries desc;

### entries 
### genus 
### 4491 Chaetoceros

2009 Fragilariopsis

1937 Tripos

1893 Corethron

#### Future Work

Next, we will use this data to train a machine learning model to help us study the relationship between phytoplankton species and biological production. We will get data from Argo and BGC-Argo floats to create clusters. We will study the correlation between the clusters and community composition