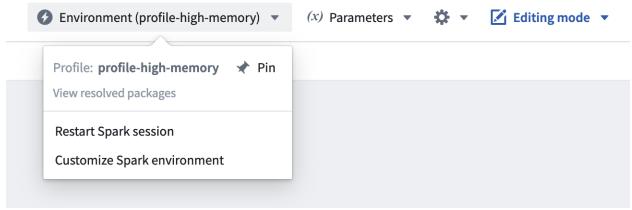
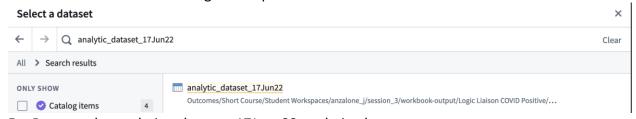
Session 4 Exercises: Data Analysis in R and Python

Exercise 1: R

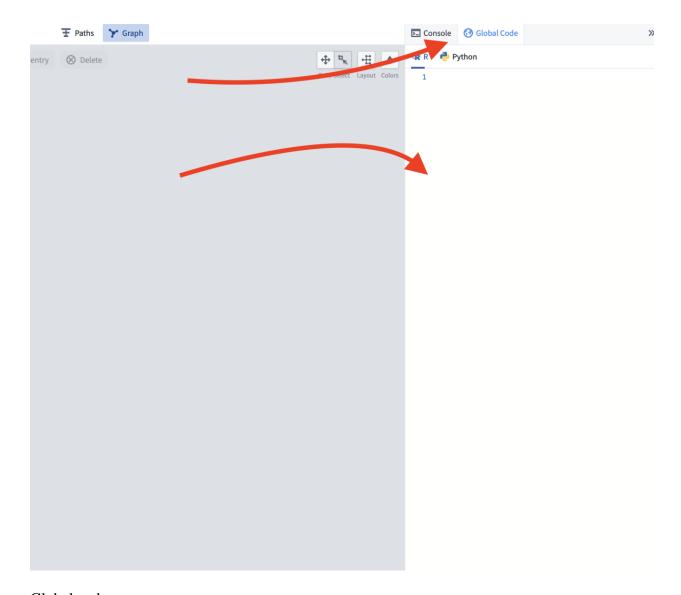
- 1. Create a new folder in your workspace and name it 'Session 4'
- 2. Create a new code workbook in the Session 4 folder and name it 'Malnutrition Data Analysis'
- 3. Change the environment to 'profile-high-memory'



4. Import the dataset titled, 'analytic_dataset_17Jun22.' You can do this by searching for it in the search bar after clicking the 'Import dataset' button:



- 5. Rename the analytics_dataset_17June22 analytic_dataset.
- 6. Select analytic_dataset and select "new transform" then select "R code." Name the transform exercise_1.
- 7. Copy the code below into the global code section of the code workbook:



Global code:

```
library(gtsummary)
library(dplyr)
```

```
# Use dplyr to select data elements of interest
local_df <- analytic_dataset %>%
    dplyr::select(age, age_group, gender, race_ethnicity, BMI, BMI_category,
        CCI, CCI_category, diabetes, MI, CHF, PVD, CVD, dementia, pulmonary,
        AIRD, liver, PUD, paralysis, renal, cancer, HIV, smoking, COVID_death, IMV,
        ARDS, ECMO, HAPI, LOS, malnutrition_status)

# Set factors: this defines the order of appearance
local_df$malnutrition_status <- factor(local_df$malnutrition_status, levels = c("No Documented Malnutrition",
        "Hx of Malnutrition", "Hospital-Acquired Malnutrition"))</pre>
```

local_df\$age_group <- factor(local_df\$age_group, levels = c("<30", "30-49",

```
"50-64", "65+"))

local_df$gender <- factor(local_df$gender, levels = c("FEMALE", "MALE"))

local_df$race_ethnicity <- factor(local_df$race_ethnicity, levels = c("White Non-Hispanic",
    "Black or African American Non-Hispanic", "Hispanic or Latino Any Race",
    "Other", "Unknown"))

local_df$BMI_category <- factor(local_df$BMI_category, levels = c("<18.5",
    "18.5-24.9", "25.0-29.9", "30+"))

local_df$CCI_category <- factor(local_df$CCI_category, levels = c("<1",
    "1-3", ">3"))

# Set references

local_df$race_ethnicity=relevel(as.factor(local_df$race_ethnicity),ref="White Non-Hispanic")

local_df$gender=relevel(as.factor(local_df$gender),ref="FEMALE")

local_df$malnutrition_status=relevel(as.factor(local_df$malnutrition_status),ref="No Documented Malnutrition")
```

8. Copy the code below into the exercise_1 transformation. Toggle the 'save as dataset' button and run the transformation.

R code:

```
descriptive_statistics_by_malnutrition_status <- function(analytic_dataset) {

# Use gtsummary to create descriptive statistics, stratified by

# malnutrition status

table1 <- local_df %>%

    tbl_summary(by = malnutrition_status) %>%
    add_p()

print(table1)

# Convert to tibble to coerce to a dataframe

table1 <- as_tibble(table1, col_labels = FALSE)

# Return summary stats. Note: column labels are excluded because they

# cannot be coerced into a data frame in this platform.

return(table1)

}
```

Exercise 2: Python

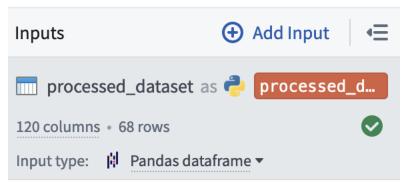
- In the Malnutrition Data Analysis code workbook, select new transform and click 'Python code.'
- 2. Copy the code below into the new transform and toggle 'Save as dataset.'

```
def processed_dataset(analytic_dataset, manifest_safe_harbor):
  "Visualizes the distribution of numeric values and the density of non-numeric values."
  # data_partner_id field is mandatory in input dataframe
  if 'data_partner_id' not in manifest_safe_harbor.columns:
     raise Exception('Input dataframe must contain "data_partner_id" column')
  # cdm_name will be joined in later, so drop if it exists
  df = analytic_dataset.drop('cdm_name')
  # Convert non-numeric columns to binary (null/notnull)
  nonnumeric_cols = [c for c,d in df.dtypes if 'int' not in d]
  for c in nonnumeric cols:
     df = df.withColumn(c, F.when(F.col(c).isNotNull(),1).otherwise(0))
  # Aggregate mean values at the site level
  cols = list(df.columns)
  cols.remove('data_partner_id')
  df = (df)
     .groupBy('data_partner_id')
     .agg(*[F.avg(c).alias(c) for c in cols])
     .toPandas()
  )
  # Score the overall density of indicator columns
  numeric_df = df.select_dtypes('number')
  indicator_cols = numeric_df.columns[numeric_df.max(axis=0)<1]
  print('indicator_cols',indicator_cols)
  df.insert(0, 'overall', df[indicator_cols].sum(axis=1))
  df['overall'] = df['overall'] / df['overall'].max()
  cols = cols + ['overall']
  for c in cols:
     diff_to_median_col = c + "_diff_to_median"
     dmd_col = c + "_dmd"
     # Calculate median absolute deviation (MAD)
     median = df[c].median()
     print(c, 'median =\t', median)
     df[diff_to_median_col] = (df[c] - median)
     MAD = df[diff_to_median_col].abs().median()
```

```
print(c, 'MAD =\t', MAD)
    # Calculate how many MADs value is from median
    df[dmd_col] = (df[diff_to_median_col]/MAD) if MAD > 0 else 0
    df.drop(columns=[diff_to_median_col], errors='ignore', inplace=True)
  # Join in cdm_name for visualizations
  sites = (manifest_safe_harbor
    .select(F.col('data_partner_id').cast('int'),'cdm_name')
    .toPandas()
  )
  return df.merge(sites, on='data_partner_id')
## Global imports and functions included below ##
import pyspark.sql.functions as F
from pyspark.sql.functions import abs as _abs
import seaborn as sns
import matplotlib.pyplot as plt
from matplotlib.patches import Patch
```

- 3. Import the manifest_safe_harbor table from the data catalog, which is located in the De-Identified Data folder.
- 4. Select new transform and click 'Python code.'
- 5. Copy the code below into the new transform and toggle 'Save as dataset.'

6. Go to the import tab in the transform and change the input type to Pandas dataframe.



Code for python transform courtesy of Evan French. Source: https://unite.nih.gov/workspace/module/view/latest/ri.workshop.main.module.3ab34203-d7f3-482e-adbd-f4113bfd1a2b?id=KO-9901C7E&view=focus