## Arsenic-Total-Urine in file UTAS J

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## 0.0.1 Arsenic-Total-Urine in file UTAS\_J

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Gets rid of all rows that contain NA values and outputs the cleaned data to "Clean UTAS\_J.xlsx"

```
[1]: import os
     import numpy as np
     import pandas as pd
     from scipy import stats
     import seaborn as sns
     import statsmodels.api as sm
     %matplotlib inline
     result_dir = '~/Environmental Health Project - Part 2/result'
     data_dir = '~/Environmental Health Project - Part 2/data'
     in_file_name = 'UTAS_J.xlsx'
     out_file_name = 'Clean UTAS_J.xlsx'
     in_file_full_name = os.path.join(data_dir, in_file_name)
     out_file_full_name = os.path.join(result_dir, out_file_name)
     data_in = pd.read_excel(in_file_full_name)
     data = data_in.drop(['Unnamed: 0'], axis = 1)
     data_no_na = data.dropna()
     data_no_na.to_excel(out_file_full_name)
```

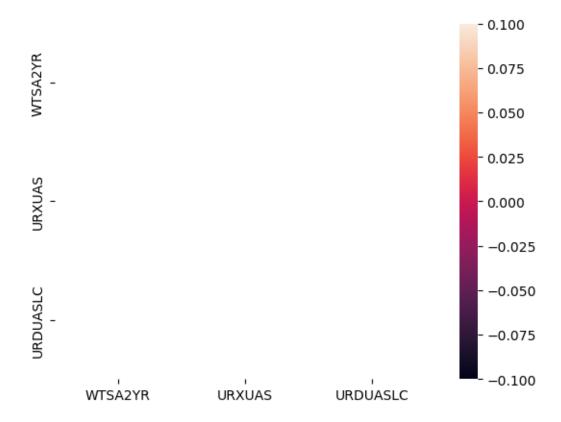
C:\Users\kecoc\anaconda3\lib\site-packages\openpyxl\styles\stylesheet.py:226:
UserWarning: Workbook contains no default style, apply openpyxl's default
warn("Workbook contains no default style, apply openpyxl's default")

Computes the pearson correlation coefficient between every pair of variables and exports coefficients that are greater than 0.5 and less than 0.999 as a CSV file.

```
[2]: in_file_name = 'Clean UTAS_J.xlsx'
out_file_name = "UTAS_J_Filtered_Pearson_Correlation_Coefficient.csv"
```

```
in_file_full_name = os.path.join(result_dir, in_file_name)
out_file_full_name = os.path.join(result_dir, out_file_name)
data_in = pd.read_excel(in_file_full_name)
data = data_in.drop(['Unnamed: 0', 'SEQN'], axis = 1)
column_points = data.columns #Gets the labels for the final data frame
length = len(column_points) #Gets the amount of labels of the dataframe for the
 ⇔column and index so the final dataframe can be shaped correctly
values = [] #Creates an empty list that the pearson correlation coefficients \Box
 ⇔will be stored in
for h in column_points: #Calculates the pearson correlation coefficient for
 ⇔every pair of columns
    for k in column_points:
        res = stats.pearsonr(data[h], data[k])
        values.append(res.statistic) #Stores the pearson correlation
 →coefficient for each pair of columns in the values list
reshaped values = np.array(values).reshape(length,length) #Creates a reshaped
 →numpy array using the values list
matrix = pd.DataFrame(reshaped_values, index = column_points, columns = __
 →column_points) #Creates a dataframe of the pearson correlation coefficient
 with the variables as the index and column
filtered_matrix = matrix[(matrix > 0.50) & (matrix < 0.999)] #Filters the_
 matrix to only show the values greater than 0.50 and less than 0.999
print("Heatmap of the filtered pearson correlation coefficient values")
print(sns.heatmap(filtered_matrix)) #Creates a heatmap of the filtered values
filtered_matrix.to_csv(out_file_full_name) #Outputs the filtered matrix to a__
  ⇔csv file
Heatmap of the filtered pearson correlation coefficient values
AxesSubplot(0.125,0.11;0.62x0.77)
```

C:\Users\kecoc\anaconda3\lib\site-packages\seaborn\matrix.py:198:
RuntimeWarning: All-NaN slice encountered
 vmin = np.nanmin(calc\_data)
C:\Users\kecoc\anaconda3\lib\site-packages\seaborn\matrix.py:203:
RuntimeWarning: All-NaN slice encountered
 vmax = np.nanmax(calc\_data)



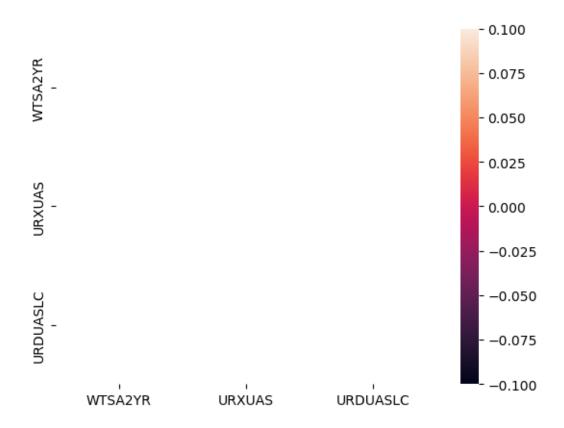
Computes the p-values for the F statistic as well as the  $R^2$  values for each pair of variables and exports the  $R^2$  values that are greater than 0.5 and less than 0.999 as a CSV file.

```
[4]: out_file_name = "UTAS_J_Filtered_R^2_values.csv"
     out_file_full_name = os.path.join(result_dir, out_file_name)
     P_values = [] #Creates an empty list that the P-values of the F statistic will_
      ⇔be stored in
     R_squared_values = [] #Creates an empty list that the R^2 values will be stored.
      \hookrightarrow in
     for h in column_points: #Calculates the pvalue of the f-statistic and r^2 value_
      →for each pair of variables
         for k in column_points:
             X2 = sm.add_constant(data[h])
             est = sm.OLS(data[k],X2)
             est2 = est.fit()
             P_values.append(est2.f_pvalue) #Appends the pvalue of the f-statistic_
      →to the p_values list
             R_squared_values.append(est2.rsquared) #Appends the r^2 value to the
      \hookrightarrow R_squared_values list
```

```
reshaped_p_values = np.array(P_values).reshape(length,length) #Reshapes the_
 \hookrightarrow P_value\ list\ into\ a\ numpy\ array\ of\ the\ correct\ shape
p_value_matrix = pd.DataFrame(reshaped_p_values, index = column_points, columns_
 →= column_points) #Creates a dataframe of the P_values array with the
 ⇔variable names as the columns and indexs
reshaped_R_squared_values = np.array(R_squared_values).reshape(length,length)_
  →#Reshapes the R_squared_values list into a numpy array of the correct shape
R_squared_values_matrix = pd.DataFrame(reshaped_R_squared_values, index = __
 ⇒column_points, columns = column_points) #Creates a dataframe of the
 \neg R squared_values array with the variable names as the columns and indexs
filtered_matrix = R_squared_values_matrix[(R_squared_values_matrix > 0.50) &__
 →(R_squared_values_matrix < 0.999)] #Filters the R_squared_values matrix to_
 →only show values greater than 0.50 and less than 0.999
print("HeatMap of filtered R^2 values")
print(sns.heatmap(filtered_matrix))
filtered_matrix.to_csv(out_file_full_name) #Outputs the filtered_matrix to a_
  ⇔csv file
HeatMap of filtered R^2 values
AxesSubplot(0.125,0.11;0.62x0.77)
C:\Users\kecoc\anaconda3\lib\site-packages\seaborn\matrix.py:198:
RuntimeWarning: All-NaN slice encountered
  vmin = np.nanmin(calc data)
C:\Users\kecoc\anaconda3\lib\site-packages\seaborn\matrix.py:203:
```

RuntimeWarning: All-NaN slice encountered

vmax = np.nanmax(calc\_data)



The data indicates that there isn't a strong correlation between any of the variables in the UTAS\_J file. Each measured pearson correlation coefficient and  $r^2$  value is either under 0.50 or above 0.999.