# vivainsights Python library: CHEAT SHEET

### **Basics**



<u>vivainsights</u> is a Python library that offers a set of tools and functions for analysing and visualising data from Microsoft Viva Insights

You can install or update the package from PyPI by running: pip install vivainsights



The latest development version and documentation can be found on our GitHub repository: <a href="https://github.com/microsoft/vivainsights-py/">https://github.com/microsoft/vivainsights-py/</a>

### **Example – loading and running the library:**

import vivainsights as vi pq df = vi.import query("path/pq df.csv") vi.create bar(pq df, metric = 'Email hours')

# Data import / export

Use our handy functions below which are optimized for best practice forgetting data in and out of Python

- import\_query() Import CSV queries faster and pre-formatted for **vivainsights** functions (instead of read.csv())
- export() Copy a data frame to clipboard, or write as a CSV, or a ggplot object as PNG or SVG

### **Inbuilt datasets**

Explore vivainsights by using inbuilt demo datasets



### load\_pq\_data()

Loads a sample Person Query data to the environment



### load\_mt\_data()

Loads a sample Standard meeting query data to the environment



## load\_g2g\_data()

Loads a sample group-to-group query data to the environment



### p2p\_data\_sim()

Person to Person query / edge list based on the graph

### **Data validation**

Validate and understand your data prior to starting a piece of analysis



# identify datefreq()

Identify date frequency based on a series of dates



### identify\_holidayweeks()

Identify likely holiday weeks (for the entire pop) where collaboration hours lie far outside the mean



### identify\_inactiveweeks()

Identify likely person-weeks where collaboration hours lie far outside the mean relative to the population average



### identify\_nkw()

Identify likely non-knowledge workers where average person collaboration hours lie far outside the mean

# Flexible analysis

Flexible analysis functions are versatile, allowing you to pass any metric as a string parameter, e.g., metric = 'Email\_hours'



### create bar()

Returns a bar plot showing the average of a selected metric by default. This function creates a bar chart directly from the aggregated / summarized data



### create\_boxplot()

Analyzes a selected metric and returns a box plot by default



### create inc()

Returns a heatmap for the generated incidence analysis



By default, returns a line chart for the defined metric



### create\_sankey()

Create a 'networkD3' style sankey chart based on a long count table with

### Full documentation site:

# https://microsoft.github.io/vivainsights-py/





# **Network analysis**

Analyze edge list datasets (e.g., Person-to-person, Group-to-group) from Viva Insights



### network\_g2g()

Pass a data frame containing a group-to-group query and return a network plot



# network\_p2p()

Pass a data frame containing a person-to-person query and return a network visualization

# **Exploratory analysis**

Explore the data and surface initial hypotheses



### keymetrics scan()

Aggregate multiple key metrics by an org attribute and return a heatmap plot by default, with options to return a data frame;



### create\_rank()

Returns a plot by default, with an option to return a table with all groups (across multiple HR attributes) ranked by the specified metric

# Other analysis



### create\_IV()

Specify an outcome variable and return IV outputs



# p\_test()

Specify an outcome variable and return p-test outputs. All numeric variables in the dataset are used as predictor variables



# compute\_gini()

Compute the Gini coefficient, a measure of statistical dispersion to represent inequality



# create\_lorenz()

This function computes the Gini coefficient and plots the Lorenz curve based on a selected metric from a Person Query data frame

