

Python in Excel: Quick Wins Reference Guide

1. Sample random rows	<code>df.sample()</code> # Default <code>df.sample(20)</code> # Sample 20 rows
2. Number of missing values in each column	<code>df.isna().sum()</code> # Count missing per column <code>(df.isna().sum() / len(df))\</code> <code>.sort_values(ascending=False)</code> # % missing per column
3. Descriptive statistics	<code>df.describe()</code> # Default <code>df.describe(percentiles=[.10, .50, .90])</code> # Custom percentiles
4. Correlation matrix	<code>numeric_corr =</code> <code>df.select_dtypes(include="number").corr()</code> # Correlate all numeric variables <code>sns.heatmap(numeric_corr, annot=True)</code> # Visualize correlations
5. Frequency tables	<code>df["col_1"].value_counts()</code> # Frequency counts <code>pd.crosstab(df["col_1"], df["col_2"],</code> <code>normalize="all", margins=True)</code> # Two-way proportion
6. Resampling	<code>df_ts = df.set_index("date")</code> # Set datetime index <code>df_ts.resample("M").sum().head()</code> # Monthly totals <code>(df_ts.resample("H").ffill() / 24).head(72)</code> # Hourly forward-fill then scale



7. Index number by group	<pre>df["group_id"] = df.groupby("group_col").cumcount() + 1 # 1- based counter</pre>
8. Leading and lagging variables	<pre>df["lag_1"] = df["value"].shift(1) # Previous value df["lead_1"] = df["value"].shift(-1) # Next value df["pct_change_%"] = df["value"].pct_change() * 100 # % change</pre>
9. Rolling and cumulative aggregations	<pre>df["rolling_mean_7"] = df["value"].rolling(7).mean() # 7-period mean df["cum_sum"] = df["value"].cumsum() # Running total df["cum_mean"] = df["value"].expanding().mean() # Running average df["rolling_mean_7"] = df["value"].rolling(7, min_periods=1).mean() # Rolling w/ min</pre>
10. Conditionally format plots	<pre>sns.scatterplot(data=df, x="col_1", y="col_2", hue="group_col", alpha=.7) # Scatter w/ hue sns.barplot(data=df_mean, x="cat_col", y="metric", palette="Blues_r") # Bar chart with nonscaled palette</pre>
11. Pairplot	<pre>sns.pairplot(df) # Quick overview of all pair- wise relationships g = sns.pairplot(df, hue="group_col", diag_kind="kde", markers=["o", "s", "D"]) # Pairplot with hue, KDE diagonals, and custom markers</pre>



	<pre>g.fig.suptitle("Pairplot of Dataset by Group", y=1.02) # Add title</pre>
12.Jitterplot	<pre>sns.stripplot(data=df, x="cat_col_1", y="metric") # Default sns.stripplot(data=df, x="cat_col_1", y="metric", hue="cat_col_2", jitter=.2) # Custom jitter and hue</pre>
13.Pairplot	<pre>g = sns.FacetGrid(data=df, col="facet_col") # Column facets g.map(sns.scatterplot, "col_1", "col_2") # Map scatter g = sns.FacetGrid(data=df, col="facet_col", row="row_col", height=3) # Row + col facets g.map(sns.boxplot, "col_1") # Map boxplot g.set_titles(col_template="{col_name}", row_template="{row_name}") # Custom titles</pre>
14.Jointplot	<pre>sns.jointplot(data=df, x="col_x", y="col_y") # Default scatter + hist sns.jointplot(data=df, x="col_x", y="col_y", hue="group_col") # Add hue by category</pre>



15. Bubbleplot	<pre>sns.scatterplot(data=df, x="col_1", y="metric", hue="group_col", size="size_col", alpha=.6) # Bubble scatter sns.regplot(data=df, x="col_1", y="metric", scatter=False, color="black") # Trend line</pre>
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