Single-cell Morphology Quality Control (coSMicQC)

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Scan for coSMicQC!

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I. Erroneous outliers and analysis

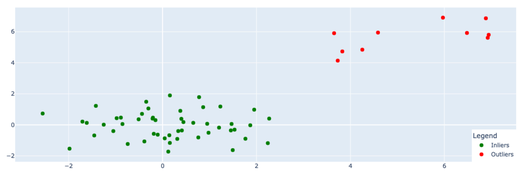

Figure 1: *Erroneous outlier anomalies can measure outside our expectations and impact analysis.*

Single-cell morphology data from high-throughput microscopy provide critical insights into disease mechanisms and therapeutic efficacy. However, ***segmentation errors*** during image analysis such as misidentifying cell compartments or artifacts as cells can lead to inaccurate single-cell measurements and ***erroneous anomalies*** within the data.

Researchers often resort to ***error-prone, bespoke filtering methods*** or aggregate data into bulk profiles to avoid discrepancies caused by anomaly outliers. These techniques fail to perform ***quality control*** on the data, often compromising the quality of single-cell profiles and impeding the potential for meaningful discoveries.

II. Single-cell quality control package

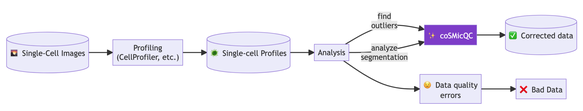

Figure 2: *coSMicQC enables high-quality data outcomes by checking for outliers.*

To address these challenges, we introduce ***coSMicQC (Single-cell Morphology Quality Control)***, an open source Python package designed to enhance the accuracy of single-cell morphology analysis. **coSMicQC** offers default and customizable thresholds for quality control, integrating seamlessly into both command line and Python API workflows.

coSMicQC features interactive visualizations that help users identify outlier distributions, and it introduces the ***CytoDataFrame*** — a novel data format that links single-cell measurements with their corresponding images and segmentation masks in real-time, enriching data analysis and interpretation.

III. Getting started with coSMicQC

☆ 1) *Installation*

```
# pip install from pypi
pip install coSMicQC

# install directly from source
pip install git+https://github.com/
WayScience/coSMicQC.git
```

coSMicQC may be installed from PyPI or source.

☆ 2) *Finding outliers*

```
import cosmicqc

# load a parquet file with single-cell
profile data and find outliers
scdf = cosmicqc.analyze.find_outliers(
    df="single-cell-profiles.parquet",
    metadata_columns=[
      "Metadata_ImageNumber",
      "Image_Metadata_Plate_x"
    ],
    feature_thresholds={
      "Nuclei_AreaShape_Area": -1
    },
)
```

```
Number of outliers: 328
Outliers Range:
Nuclei_AreaShape_Area Min: 734.0
Nuclei_AreaShape_Area Max: 1904.0
      Nuclei_AreaShape_Area  Metadata_ImageNumber Image_Metadata_Plate_x
23                    921.0                     2               Plate_2
28                    845.0                     2               Plate_2
29                   1024.0                     2               Plate_2
32                    787.0                     2               Plate_2
37                   1347.0                     2               Plate_2
...                     ...                   ...                   ...
```

Figure 3: *Reports in **coSMicQC** provide quality control feedback for analysis.*

The `find_outliers` function in **coSMicQC** uses single-cell feature thresholds to provide a report on how many outliers were detected. We use ***z-scores*** to help define thresholds used throughout coSMicQC. `find_outliers` may be used through Python or CLI (see below).

```
# CLI interface for coSMicQC find_outliers
$ cosmicqc find_outliers \
  --df single-cell-profiles.parquet \
  --metadata_columns \[Metadata_ImageNumber\]
\
  --feature_thresholds
'{"Nuclei_AreaShape_Area": -1}'

Number of outliers: 328
Outliers Range:
Nuclei_AreaShape_Area Min: 734.0
Nuclei_AreaShape_Area Max: 1904.0
...
```

☆ 3) *Visualizing outlier distributions*

```
import cosmicqc

# label outliers within the profiles
scdf = cosmicqc.analyze.label_outliers(
    df="single-cell-profiles.parquet",
    include_threshold_scores=True,
# show outlier histogram plots
).show_report()
```

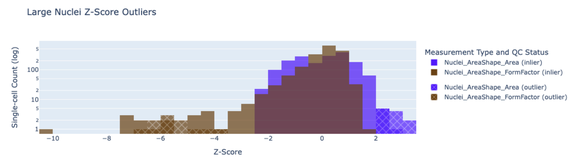

Figure 4: *Histograms from **coSMicQC** help scientists identify where outliers occur in the distribution.*

Deep erroneous anomaly analysis is enabled within **coSMicQC** through the ***`label_outliers`*** function, which appends z-score data for features, and the ***`CytoDataFrame.show_report`*** method to visualize where outliers are detected within the dataset.

☆ 4) *Understanding outlier segmentation issues*

```
import cosmicqc

# passing image and mask dirs to display
images
cosmicqc.CytoDataFrame(
    data="single-cell-profiles.parquet",
    data_context_dir="./image_directory/",
    data_mask_context_dir="./
mask_directory/",
)
```

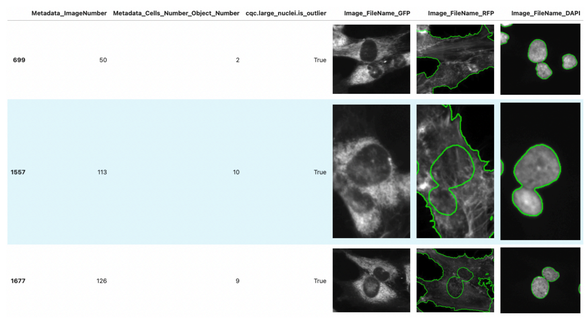

Figure 5: *Displaying outlier and image data increases analysis iteration within familiar interfaces.*

CytoDataFrames returned by all **coSMicQC** operations enable researchers to analyze outlier status alongside single-cell images directly in a Jupyter environment.

IV. Real-world applications

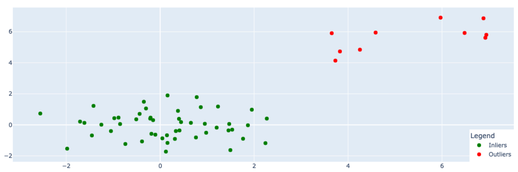

Figure 6: *Etiam viverra sollicitudin velit.*

Lorem ipsum dolor sit amet, consectetur adipiscing elit. In vel augue ante. Duis placerat ex id turpis consequat, at aliquet arcu vehicula.

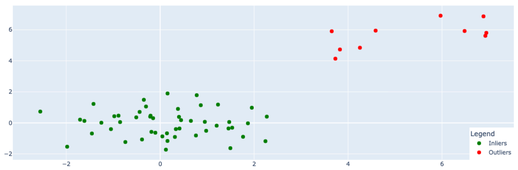

Figure 7: *Ut ut metus at diam hendrerit dictum.*

Donec bibendum ex vitae egestas ornare. Cras ut vulputate diam. Nullam feugiat feugiat purus, eu various metus maximus sit amet.

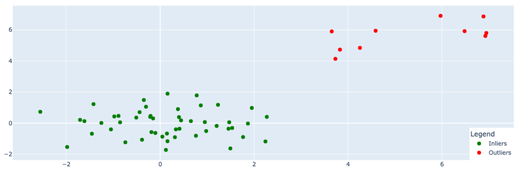

Figure 8: *Praesent eu sem id nibh viverra finibus.*

Fusce ac mi eu augue suscipit sodales. Morbi quis gravida dui. Name tellus elit, ullamcorper ut ipsum ut, eleifend facilisis dui.

V. Future steps

We plan to strengthen **coSMicQC** through further erroneous anomaly detection techniques, integrate with existing single-cell pipelines technologies, and expand image format compatibility.