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July 15, 2025

1 CellSexID: Single-Cell Sex Identification Tool

This notebook demonstrates a comprehensive pipeline for sex prediction in single-cell RNA-seq data using machine learning approaches. The analysis covers data preprocessing, quality control, feature selection, and model training with multiple algorithms to achieve robust sex classification.

1.1 Overview

This analysis pipeline begins with setting up the computational environment for single-cell RNA-seq data analysis. We utilize Scanpy, a leading single-cell analysis toolkit, alongside scikit-learn and XGBoost for machine learning applications. The workflow is designed to handle both mouse and human single-cell datasets for sex prediction tasks.

1.2 Dataset Preparation Strategy

1.2.1 Step 1: Individual Dataset Loading

The pipeline handles separate male and female single-cell RNA-seq datasets through a systematic approach:

- 1. **Matrix Loading**: Count matrices are loaded from MTX format files and properly oriented (cells × genes)
- 2. **Gene Annotation**: Gene metadata is imported from TSV files containing gene identifiers and symbols
- 3. Gene Assignment: Gene names are mapped to expression matrices using annotation files
- 4. **Uniqueness Enforcement**: Duplicate gene names are resolved to prevent downstream conflicts

This approach ensures data integrity and consistency across experimental conditions.

1.2.2 Step 2: Gene Space Harmonization

To enable meaningful comparison between male and female datasets, we perform gene space alignment:

- 1. Common Gene Identification: Calculate the intersection of gene features between datasets
- 2. Feature Space Harmonization: Subset both datasets to retain only shared genes
- 3. Consistency Verification: Ensure identical gene ordering across datasets

This step is critical for downstream analysis as it ensures that all samples are measured across the same feature space.

1.2.3 Step 3: Dataset Integration and Labeling

The final preprocessing step involves combining individual datasets into a unified analysis object:

- 1. Cell Identification: Assign unique identifiers to distinguish cells across datasets
- 2. Metadata Assignment: Add sex labels and batch information for downstream analysis
- 3. Dataset Concatenation: Merge datasets while preserving individual cell and gene identities
- 4. Quality Assurance: Verify proper integration and resolve any naming conflicts

This creates a single AnnData object containing all cells with appropriate metadata for classification tasks.

1.3 Dataset Integration

Combining male and female datasets into a single AnnData object with appropriate cell identifiers and metadata labels. The concatenation preserves individual cell identities while enabling joint analysis across conditions.

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     numba > = 0.56 - scanpy) (0.43.0)
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     resources>=3.2.0->matplotlib>=3.6->scanpy) (3.19.2)
     [notice] A new release of pip is
     available: 25.0.1 -> 25.1.1
     [notice] To update, run:
     /Applications/Xcode.app/Contents/Developer/usr/bin/python3 -m pip
     install --upgrade pip
     scanpy==1.10.2 anndata==0.10.8 umap==0.5.6 numpy==1.26.4 scipy==1.13.1
     pandas==2.2.2 scikit-learn==1.5.1 statsmodels==0.14.2 pynndescent==0.5.13
[40]: maledataRR = sc.read mtx('male.mtx')
      femaledataRR = sc.read mtx('female.mtx')
      fff=pd.read_csv('female.tsv',sep='\t',header=None)
      mmm=pd.read_csv('male.tsv',sep='\t',header=None)
      fff
      frames = [fff, mmm]
      name=pd.concat(frames,axis=0)
      femaledata=femaledataRR.T
      maledata=maledataRR.T
      femaledata.var.index = fff[1]
      maledata.var.index = mmm[1]
      femaledata.var
      femaledata.obs_names = [f"Cell {2685+i:d}" for i in range(femaledata.n obs)]
```

matplotlib>=3.6->scanpy) (2.9.0.post0)

```
maledata.obs_names = [f"Cell_{i:d}" for i in range(maledata.n_obs)]
[]: !pip install matplotlib==3.1.3
     import anndata
     fw=femaledata.X.todense()
     maledata
     fr=maledata.X.todense()
     !pip install AnnData
     import anndata
     x_combined = np.concatenate([fr,fw],axis=0)#male
     adata combined = anndata.AnnData(x combined) # np array anndata
     adata combined.var.index = fff[1]
     adata_combined.obs_names = [f"Cell_{i:d}" for i in range(adata_combined.n_obs)]
     c11 = np.ones(maledata.n_obs)
     c22= np.zeros(femaledata.n_obs)
     joined_list = [*c11, *c22]
     adata_combined.obs["gender"] = pd.Categorical(joined_list) # Categoricals are_
      ⇔preferred for efficiency
     adata_combined.obs
     adata=adata combined
     adata.var.index=fff[1]
     adata.X=np.asarray(adata.X)
     adata.var.index.name ="n"
     sc.pp.filter_genes(adata, min_cells=3)
     adata.var['mt'] = adata.var_names.str.startswith('mt-') # annotate the group_
      ⇔of mitochondrial genes as 'mt'
     sc.pp.calculate_qc_metrics(adata, qc_vars=['mt'], percent_top=None,_
      →log1p=False, inplace=True)
     adata.var_names_make_unique()
    Defaulting to user installation because normal site-packages is not writeable
    Collecting matplotlib==3.1.3
      Using cached matplotlib-3.1.3.tar.gz (40.9 MB)
      Preparing metadata (setup.py) ... done
    Requirement already satisfied: cycler>=0.10 in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (from
    matplotlib==3.1.3) (0.12.1)
    Requirement already satisfied: kiwisolver>=1.0.1 in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (from
    matplotlib==3.1.3) (1.4.5)
    Requirement already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (from
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    Requirement already satisfied: python-dateutil>=2.1 in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (from
    matplotlib==3.1.3) (2.9.0.post0)
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loper/Library/Frameworks/Python3.framework/Versions/3.9/lib/python3.9/site-
packages (from python-dateutil>=2.1->matplotlib==3.1.3) (1.15.0)
Building wheels for collected packages: matplotlib
 Building wheel for matplotlib (setup.py) ... error
 error: subprocess-exited-with-error
  x python setup.py bdist_wheel did not run successfully.
   exit code: 1
   > [518 lines of output]
______
     Edit setup.cfg to change the build options
     BUILDING MATPLOTLIB
       matplotlib: yes [3.1.3]
           python: yes [3.9.6 (default, Mar 12 2025, 20:22:46) [Clang
17.0.0
                       (clang-1700.0.13.3)]]
         platform: yes [darwin]
     OPTIONAL SUBPACKAGES
      sample_data: yes [installing]
            tests: no [skipping due to configuration]
     OPTIONAL BACKEND EXTENSIONS
              agg: yes [installing]
            tkagg: yes [installing; run-time loading from Python
Tcl/Tk]
           macosx: yes [installing, darwin]
     OPTIONAL PACKAGE DATA
             dlls: no [skipping due to configuration]
     running bdist_wheel
     running build
     running build_py
     creating build
     creating build/lib.macosx-10.9-universal2-3.9
     copying lib/pylab.py -> build/lib.macosx-10.9-universal2-3.9
     creating build/lib.macosx-10.9-universal2-3.9/mpl_toolkits
     copying lib/mpl_toolkits/__init__.py ->
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     creating build/lib.macosx-10.9-universal2-3.9/matplotlib
     copying lib/matplotlib/hatch.py ->
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```

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```

```
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```

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data/fonts/ttf
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
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data/fonts/pdfcorefonts
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-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/pdfcorefonts
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
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```

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ui-1.12.1/images/ui-icons_555555_256x240.png ->
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
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data/fonts/pdfcorefonts/ZapfDingbats.afm ->
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
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-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
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Roman.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
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ui-1.12.1/jquery-ui.structure.min.css ->
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```

```
ui-1.12.1
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ui-1.12.1
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build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/css
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build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/css
      copying lib/matplotlib/mpl-data/fonts/afm/pagdo8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/afm/pncbi8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/cmss10.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/qt4_editor_options.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/ttf/cmex10.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/afm/phvr8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Helvetica-
Bold.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/images/filesave large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/hand.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/pzcmi8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/afm/pplb8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/afm/pplbi8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      creating
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery
      creating
```

build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery/js

```
copying
lib/matplotlib/backends/web_backend/jquery/js/jquery.min.js ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery/js
      copying lib/matplotlib/mpl-data/images/home_large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/matplotlib_large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXSizFourSymReg.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSans-Bold.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/jquery-ui.structure.css ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/mpl-data/images/zoom_to_rect.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/move.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/phvro8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/cmtt10.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/home.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXSizFiveSymReg.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/home.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/backends/web_backend/css/page.css ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/css
      copying lib/matplotlib/mpl-data/fonts/afm/cmr10.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/backends/web_backend/css/boilerplate.css
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web backend/css
      copying lib/matplotlib/mpl-data/fonts/afm/putbi8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/sample_data/None_vs_nearest-
pdf.png -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/afm/phvl8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/help.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/cmex10.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-
```

```
data/sample_data/Minduka_Present_Blue_Pack.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSerif.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/afm/cmmi10.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/backends/web_backend/single_figure.html ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend
      copying lib/matplotlib/mpl-data/fonts/afm/psyr.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/jquery-ui.theme.css ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/mpl-data/stylelib/seaborn-
notebook.mplstyle -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/stylelib
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/AUTHORS.txt ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/mpl-data/images/filesave.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      creating
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/js
      copying lib/matplotlib/backends/web_backend/js/nbagg_mpl.js ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/js
      copying lib/matplotlib/mpl-data/sample_data/demodata.csv ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/stylelib/seaborn-poster.mplstyle
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/stylelib/seaborn-deep.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/sample_data/aapl.npz ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXGeneralBolIta.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/stylelib/grayscale.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/stylelib/bmh.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/sample_data/eeg.dat ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/sample_data/logo2.png ->
```

```
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/afm/putb8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSerif-Italic.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/subplots.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/ptmri8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/forward.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying
lib/matplotlib/backends/web_backend/ipython_inline_figure.html ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images/ui-icons_777620_256x240.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images
      copying lib/matplotlib/mpl-data/fonts/afm/cmtt10.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/forward.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/forward_large.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/back.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/back.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/index.html ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/mpl-data/images/zoom_to_rect.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXNonUniIta.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/afm/phvbo8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/hand.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/cmsy10.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/qt4_editor_options.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/zoom_to_rect.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXSizTwoSymBol.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
```

```
copying lib/matplotlib/mpl-data/images/qt4_editor_options.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/filesave.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/phvb8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/stylelib/ggplot.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/images/help_large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/backends/web_backend/jquery/js/jquery.js
->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery/js
      copying lib/matplotlib/mpl-data/sample_data/s1045.ima.gz ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/sample_data/grace_hopper.jpg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/backends/web_backend/nbagg_uat.ipynb ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Times-
BoldItalic.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/fonts/afm/pplr8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/cmmi10.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/afm/pbkl8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/hand.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/back_large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSansMono.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/afm/phvbo8an.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXGeneral.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/jquery-ui.css ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images/ui-icons_444444_256x240.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images
```

```
copying lib/matplotlib/mpl-data/images/matplotlib.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/seaborn-white.mplstyle
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/fonts/afm/phvr8an.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/readme.txt ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/fonts/afm/pcrr8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/sample_data/embedding_in_wx3.xrc
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/afm/pbkdi8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSansMono-
BoldOblique.ttf -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/zoom_to_rect_large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/seaborn-dark.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXNonUni.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/sample_data/grace_hopper.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Courier-
Oblique.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXGeneralBol.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/backends/web_backend/js/mpl_tornado.js ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/js
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images/ui-icons_fffffff_256x240.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web backend/jquery-
ui-1.12.1/images
      copying lib/matplotlib/mpl-data/sample_data/data_x_x2_x3.csv ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/afm/pagd8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/afm/pncr8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Times-Bold.afm
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/images/forward.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/seaborn-talk.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
```

```
copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Courier-
Bold.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/fonts/afm/pcrbo8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Times-
Italic.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/fonts/ttf/cmb10.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/stylelib/_classic_test.mplstyle
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXNonUniBol.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/home.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-
data/sample_data/percent_bachelors_degrees_women_usa.csv ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/images/zoom to rect.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/backends/web backend/jquery-
ui-1.12.1/jquery-ui.theme.min.css ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/mpl-data/fonts/afm/ptmb8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/jquery-ui.min.js ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/mpl-data/images/matplotlib.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSerif-Bold.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/LICENSE.txt ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Symbol.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/images/zoom_to_rect_large.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/seaborn.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/stylelib/dark_background.mplstyle
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/images/forward_large.png ->
```

```
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/ttf/LICENSE_DEJAVU ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/hand.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/Solarize_Light2.mplstyle
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/images/back_large.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/seaborn-
whitegrid.mplstyle -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/stylelib
      copying lib/matplotlib/mpl-data/fonts/afm/phvro8an.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSerif-
BoldItalic.ttf -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXSizThreeSymReg.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/stylelib/seaborn-pastel.mplstyle
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/images/forward.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/jquery-ui.js ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/mpl-data/sample_data/msft.csv ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      creating
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1/external
      creating
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1/external/jquery
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/external/jquery/jquery.js ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1/external/jquery
      copying lib/matplotlib/mpl-data/images/subplots.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Courier.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/sample_data/ct.raw.gz ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSans.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      creating build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
```

```
data/sample_data/axes_grid
      copying lib/matplotlib/mpl-
data/sample_data/axes_grid/bivariate_normal.npy ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data/axes_grid
      copying lib/matplotlib/mpl-data/images/subplots large.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/filesave.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/sample_data/membrane.dat ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/stylelib/seaborn-
colorblind.mplstyle -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/stylelib
      copying lib/matplotlib/mpl-data/fonts/afm/putr8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/help_large.ppm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/home_large.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSansMono-Bold.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSans-Oblique.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/matplotlib.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/pagk8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/move.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXNonUniBolIta.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/move_large.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/seaborn-
darkgrid.mplstyle -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/stylelib
      copying lib/matplotlib/mpl-data/fonts/afm/pcrb8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/afm/pbkli8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXSizTwoSymReg.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/afm/pzdr.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/back.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/matplotlibrc ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data
```

```
copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images/ui-icons_cc0000_256x240.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images
      copying lib/matplotlib/mpl-
data/images/qt4_editor_options_large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/sample_data/README.txt ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/stylelib/fast.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/images/help.ppm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/backends/web_backend/all_figures.html ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend
      copying lib/matplotlib/mpl-data/fonts/afm/pncb8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Helvetica-
Oblique.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/fonts/ttf/LICENSE_STIX ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/move.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXSizOneSymReg.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Helvetica-
BoldOblique.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSansMono-
Oblique.ttf -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXSizFourSymBol.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/subplots large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/seaborn-paper.mplstyle
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Courier-
BoldOblique.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/stylelib/seaborn-bright.mplstyle
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXSizOneSymBol.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/move_large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/backends/web_backend/jquery-
```

```
ui-1.12.1/images/ui-icons_777777_256x240.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images
      copying lib/matplotlib/mpl-data/sample_data/topobathy.npz ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample data
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSansDisplay.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/matplotlib.ppm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-
data/sample_data/jacksboro_fault_dem.npz ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/images/help.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/seaborn-dark-
palette.mplstyle -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/stylelib
      copying lib/matplotlib/backends/web_backend/js/mpl.js ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/js
      copying lib/matplotlib/mpl-data/images/subplots.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/tableau-
colorblind10.mplstyle -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/stylelib
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSans-
BoldOblique.ttf -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/ttf
      copying lib/matplotlib/mpl-data/stylelib/classic.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      UPDATING
build/lib.macosx-10.9-universal2-3.9/matplotlib/_version.py
      set build/lib.macosx-10.9-universal2-3.9/matplotlib/_version.py
to '3.1.3'
      running build_ext
      building 'matplotlib.ft2font' extension
      creating build/temp.macosx-10.9-universal2-3.9
      creating build/temp.macosx-10.9-universal2-3.9/src
      clang -Wno-unused-result -Wsign-compare -Wunreachable-code -fno-
common -dynamic -DNDEBUG -g -fwrapv -03 -Wall
-iwithsysroot/System/Library/Frameworks/System.framework/PrivateHeaders -iwithsy
sroot/Applications/Xcode.app/Contents/Developer/Library/Frameworks/Python3.frame
work/Versions/3.9/Headers -arch arm64 -arch x86_64 -Werror=implicit-function-
declaration -Wno-error=unreachable-code -Wno-error=unused-but-set-variable -Wno-
error=cast-function-type-mismatch -Wno-unknown-warning-option
-DFREETYPE_BUILD_TYPE=system
-DPY_ARRAY_UNIQUE_SYMBOL=MPL_matplotlib_ft2font_ARRAY_API
-DNPY_NO_DEPRECATED_API=NPY_1_7_API_VERSION -D__STDC_FORMAT_MACROS=1
-Iextern/agg24-svn/include -I/Users/haley/Library/Python/3.9/lib/python/site-
```

```
packages/numpy/core/include -I/Applications/Xcode.app/Contents/Developer/Library
/Frameworks/Python3.framework/Versions/3.9/include/python3.9 -c
src/checkdep_freetype2.c -o
build/temp.macosx-10.9-universal2-3.9/src/checkdep_freetype2.o
-I/opt/homebrew/opt/freetype/include/freetype2
-I/opt/homebrew/opt/libpng/include/libpng16
     src/checkdep freetype2.c:7:9: warning: Compiling with FreeType
version 2.13.3. [-W#pragma-messages]
         7 | #pragma message("Compiling with FreeType version " \
     1 warning generated.
     src/checkdep_freetype2.c:7:9: warning: Compiling with FreeType
version 2.13.3. [-W#pragma-messages]
         7 | #pragma message("Compiling with FreeType version " \
     1 warning generated.
     clang -Wno-unused-result -Wsign-compare -Wunreachable-code -fno-
common -dynamic -DNDEBUG -g -fwrapv -03 -Wall
-iwithsysroot/System/Library/Frameworks/System.framework/PrivateHeaders -iwithsy
sroot/Applications/Xcode.app/Contents/Developer/Library/Frameworks/Python3.frame
work/Versions/3.9/Headers -arch arm64 -arch x86 64 -Werror=implicit-function-
declaration -Wno-error=unreachable-code -Wno-error=unused-but-set-variable -Wno-
error=cast-function-type-mismatch -Wno-unknown-warning-option
-DFREETYPE_BUILD_TYPE=system
-DPY_ARRAY_UNIQUE_SYMBOL=MPL_matplotlib_ft2font_ARRAY_API
-DNPY NO DEPRECATED API=NPY 1 7 API VERSION -D STDC FORMAT MACROS=1
-Iextern/agg24-svn/include -I/Users/haley/Library/Python/3.9/lib/python/site-
packages/numpy/core/include -I/Applications/Xcode.app/Contents/Developer/Library
/Frameworks/Python3.framework/Versions/3.9/include/python3.9 -c src/ft2font.cpp
-o build/temp.macosx-10.9-universal2-3.9/src/ft2font.o
-I/opt/homebrew/opt/freetype/include/freetype2
-I/opt/homebrew/opt/libpng/include/libpng16
     src/ft2font.cpp:223:29: error: assigning to 'char *' from
'unsigned char *' converts between pointers to integer types where one is of the
unique plain 'char' type and the other is not
        223 |
                     tags = outline.tags + first;
                            src/ft2font.cpp:339:29: error: assigning to 'char *' from
'unsigned char *' converts between pointers to integer types where one is of the
unique plain 'char' type and the other is not
       339 |
                 tags = outline.tags + first;
                            2 errors generated.
     error: command '/usr/bin/clang' failed with exit code 1
      [end of output]
 note: This error originates from a subprocess, and is likely not a
```

problem with pip.

```
ERROR: Failed building wheel for matplotlib
  Running setup.py clean for matplotlib
Failed to build matplotlib
[notice] A new release of pip is
available: 25.0.1 -> 25.1.1
[notice] To update, run:
/Applications/Xcode.app/Contents/Developer/usr/bin/python3 -m pip
install --upgrade pip
ERROR: Failed to build installable wheels for some pyproject.toml based
projects (matplotlib)
Defaulting to user installation because normal site-packages is not
writeable
Requirement already satisfied: AnnData in
/Users/haley/Library/Python/3.9/lib/python/site-packages (0.10.8)
Requirement already satisfied: array-api-compat!=1.5,>1.4 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (1.7.1)
Requirement already satisfied: exceptiongroup in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (1.2.1)
Requirement already satisfied: h5py>=3.1 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (3.11.0)
Requirement already satisfied: natsort in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (8.4.0)
Requirement already satisfied: numpy>=1.23 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (1.26.4)
Requirement already satisfied: packaging>=20.0 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (24.1)
Requirement already satisfied: pandas!=2.1.0rc0,!=2.1.2,>=1.4 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (2.2.2)
Requirement already satisfied: scipy>1.8 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (1.13.1)
Requirement already satisfied: python-dateutil>=2.8.2 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
pandas!=2.1.0rc0,!=2.1.2,>=1.4->AnnData) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
pandas!=2.1.0rc0,!=2.1.2,>=1.4->AnnData) (2024.1)
Requirement already satisfied: tzdata>=2022.7 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
pandas!=2.1.0rc0,!=2.1.2,>=1.4->AnnData) (2024.1)
Requirement already satisfied: six>=1.5 in /Applications/Xcode.app/Contents/Deve
loper/Library/Frameworks/Python3.framework/Versions/3.9/lib/python3.9/site-
packages (from python-dateutil>=2.8.2->pandas!=2.1.0rc0,!=2.1.2,>=1.4->AnnData)
(1.15.0)
```

```
[notice] A new release of pip is
available: 25.0.1 -> 25.1.1
[notice] To update, run:
/Applications/Xcode.app/Contents/Developer/usr/bin/python3 -m pip
install --upgrade pip
filtered out 15132 genes that are detected in less than 3 cells
/Users/haley/Library/Python/3.9/lib/python/site-
packages/anndata/_core/anndata.py:1820: UserWarning: Variable names are not
unique. To make them unique, call `.var_names_make_unique`.
    utils.warn_names_duplicates("var")
```

1.3.1 Step 4: Quality Control Assessment

Quality control is essential for identifying and filtering low-quality cells that could bias downstream analysis:

- 1. **Mitochondrial Gene Detection**: Identify genes encoded by mitochondrial DNA as indicators of cell stress
- 2. Cellular Quality Metrics: Calculate key parameters including total transcript counts, gene detection rates, and mitochondrial gene percentages
- 3. **Data Visualization**: Generate comprehensive plots to assess data quality distributions and identify potential outliers

These metrics help distinguish healthy cells from dying or damaged cells that exhibit high mitochondrial gene expression.

```
[42]: !pip install matplotlib==3.1.3
  import anndata
  fw=femaledata.X.todense()
  maledata
  fr=maledata.X.todense()
  !pip install AnnData
  import anndata
  x_combined = np.concatenate([fr,fw],axis=0)#male
  adata_combined = anndata.AnnData(x_combined) # np array anndata
```

```
adata_combined.var.index = fff[1]
adata_combined.obs_names = [f"Cell_{i:d}" for i in range(adata_combined.n_obs)]
c11 = np.ones(maledata.n_obs)
c22= np.zeros(femaledata.n_obs)
joined_list = [*c11, *c22]
adata_combined.obs["gender"] = pd.Categorical(joined_list) # Categoricals are_
  ⇔preferred for efficiency
adata_combined.obs
adata=adata_combined
adata.var.index=fff[1]
adata.X=np.asarray(adata.X)
adata.var.index.name ="n"
sc.pp.filter_genes(adata, min_cells=3)
adata.var['mt'] = adata.var_names.str.startswith('mt-') # annotate the group_
  ⇔of mitochondrial genes as 'mt'
sc.pp.calculate_qc_metrics(adata, qc_vars=['mt'], percent_top=None,_
 →log1p=False, inplace=True)
adata.var_names_make_unique()
Defaulting to user installation because normal site-packages is not writeable
Collecting matplotlib==3.1.3
 Using cached matplotlib-3.1.3.tar.gz (40.9 MB)
 Preparing metadata (setup.py) ... done
Requirement already satisfied: cycler>=0.10 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
matplotlib==3.1.3) (0.12.1)
Requirement already satisfied: kiwisolver>=1.0.1 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
matplotlib==3.1.3) (1.4.5)
Requirement already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
matplotlib==3.1.3) (3.1.2)
Requirement already satisfied: python-dateutil>=2.1 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
matplotlib==3.1.3) (2.9.0.post0)
Requirement already satisfied: numpy>=1.11 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
matplotlib==3.1.3) (1.26.4)
Requirement already satisfied: six>=1.5 in /Applications/Xcode.app/Contents/Deve
loper/Library/Frameworks/Python3.framework/Versions/3.9/lib/python3.9/site-
packages (from python-dateutil>=2.1->matplotlib==3.1.3) (1.15.0)
Building wheels for collected packages: matplotlib
 Building wheel for matplotlib (setup.py) ... error
  error: subprocess-exited-with-error
  x python setup.py bdist_wheel did not run successfully.
```

```
exit code: 1
  > [518 lines of output]
     Edit setup.cfg to change the build options
     BUILDING MATPLOTLIB
        matplotlib: yes [3.1.3]
           python: yes [3.9.6 (default, Mar 12 2025, 20:22:46) [Clang
17.0.0
                        (clang-1700.0.13.3)]]
          platform: yes [darwin]
      OPTIONAL SUBPACKAGES
       sample_data: yes [installing]
             tests: no [skipping due to configuration]
      OPTIONAL BACKEND EXTENSIONS
               agg: yes [installing]
             tkagg: yes [installing; run-time loading from Python
Tcl/Tk]
           macosx: yes [installing, darwin]
      OPTIONAL PACKAGE DATA
              dlls: no [skipping due to configuration]
      running bdist_wheel
      running build
      running build_py
      creating build
      creating build/lib.macosx-10.9-universal2-3.9
      copying lib/pylab.py -> build/lib.macosx-10.9-universal2-3.9
      creating build/lib.macosx-10.9-universal2-3.9/mpl_toolkits
      copying lib/mpl_toolkits/__init__.py ->
build/lib.macosx-10.9-universal2-3.9/mpl toolkits
      creating build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/hatch.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/transforms.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/legend_handler.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/axis.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/textpath.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/quiver.py ->
```

build/lib.macosx-10.9-universal2-3.9/matplotlib

```
copying lib/matplotlib/backend_bases.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/units.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/pylab.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/colorbar.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/_version.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/_cm.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/backend_managers.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/mathtext.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/font_manager.py ->
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      copying lib/matplotlib/bezier.py ->
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      copying lib/matplotlib/figure.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/_init_.py ->
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build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/dviread.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/animation.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/type1font.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
```

```
copying lib/matplotlib/_mathtext_data.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
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build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/_color_data.py ->
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      copying lib/matplotlib/_pylab_helpers.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/tight_layout.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/ cm listed.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/dates.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/table.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
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      copying lib/matplotlib/pyplot.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
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build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/ticker.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
```

```
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      copying lib/matplotlib/contour.py ->
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      copying lib/matplotlib/ layoutbox.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/image.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib
      copying lib/matplotlib/scale.py ->
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      copying lib/matplotlib/_constrained_layout.py ->
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      copying lib/matplotlib/streamplot.py ->
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      creating
build/lib.macosx-10.9-universal2-3.9/mpl_toolkits/axes_grid1
      copying lib/mpl_toolkits/axes_grid1/inset_locator.py ->
build/lib.macosx-10.9-universal2-3.9/mpl_toolkits/axes_grid1
      copying lib/mpl_toolkits/axes_grid1/axes_grid.py ->
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      copying lib/mpl_toolkits/axes_grid1/colorbar.py ->
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      copying lib/mpl_toolkits/axes_grid1/anchored_artists.py ->
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      copying lib/mpl_toolkits/axes_grid1/mpl_axes.py ->
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      copying lib/mpl_toolkits/axes_grid1/axes_divider.py ->
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      creating
build/lib.macosx-10.9-universal2-3.9/mpl_toolkits/axes_grid
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build/lib.macosx-10.9-universal2-3.9/mpl_toolkits/axes_grid
```

```
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build/lib.macosx-10.9-universal2-3.9/mpl_toolkits/axes_grid
      copying lib/mpl toolkits/axes grid/parasite axes.py ->
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      copying lib/mpl_toolkits/axes_grid/axis_artist.py ->
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build/lib.macosx-10.9-universal2-3.9/mpl_toolkits/axes_grid
      copying lib/mpl_toolkits/axes_grid/__init__.py ->
build/lib.macosx-10.9-universal2-3.9/mpl_toolkits/axes_grid
      copying lib/mpl_toolkits/axes_grid/axes_size.py ->
build/lib.macosx-10.9-universal2-3.9/mpl_toolkits/axes_grid
      copying lib/mpl_toolkits/axes_grid/angle_helper.py ->
build/lib.macosx-10.9-universal2-3.9/mpl_toolkits/axes_grid
      copying lib/mpl_toolkits/axes_grid/grid_helper_curvelinear.py ->
build/lib.macosx-10.9-universal2-3.9/mpl_toolkits/axes_grid
      copying lib/mpl_toolkits/axes_grid/clip_path.py ->
build/lib.macosx-10.9-universal2-3.9/mpl_toolkits/axes_grid
      copying lib/mpl_toolkits/axes_grid/floating_axes.py ->
build/lib.macosx-10.9-universal2-3.9/mpl_toolkits/axes_grid
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      creating
build/lib.macosx-10.9-universal2-3.9/mpl_toolkits/axisartist
      copying lib/mpl_toolkits/axisartist/axislines.py ->
build/lib.macosx-10.9-universal2-3.9/mpl_toolkits/axisartist
```

```
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      copying lib/mpl toolkits/axisartist/parasite axes.py ->
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      creating build/lib.macosx-10.9-universal2-3.9/matplotlib/tri
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      copying lib/matplotlib/tri/triangulation.py ->
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build/lib.macosx-10.9-universal2-3.9/matplotlib/tri
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      copying lib/matplotlib/axes/_base.py ->
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      copying lib/matplotlib/axes/_subplots.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/axes
      copying lib/matplotlib/axes/__init__.py ->
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      copying lib/matplotlib/axes/_secondary_axes.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/axes
      creating
build/lib.macosx-10.9-universal2-3.9/matplotlib/sphinxext
      copying lib/matplotlib/sphinxext/__init__.py ->
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      copying lib/matplotlib/backends/backend_gtk3agg.py ->
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      copying lib/matplotlib/backends/backend_cairo.py ->
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      copying lib/matplotlib/backends/backend_pdf.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends
      copying lib/matplotlib/backends/backend_wxagg.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends
      copying lib/matplotlib/backends/backend_pgf.py ->
```

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      copying lib/matplotlib/backends/backend_mixed.py ->
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      copying lib/matplotlib/backends/backend_macosx.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends
      copying lib/matplotlib/backends/windowing.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends
      copying lib/matplotlib/backends/__init__.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends
      copying lib/matplotlib/backends/backend_qt4.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends
      copying lib/matplotlib/backends/backend_tkcairo.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends
      copying lib/matplotlib/backends/backend_qt4cairo.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends
      copying lib/matplotlib/backends/backend_gtk3.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends
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      copying lib/matplotlib/backends/backend_agg.py ->
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build/lib.macosx-10.9-universal2-3.9/matplotlib/style
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      copying lib/matplotlib/projections/__init__.py ->
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      copying lib/matplotlib/testing/decorators.py ->
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build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/qt_editor
      copying lib/matplotlib/backends/qt_editor/_formlayout.py ->
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      copying lib/matplotlib/backends/qt_editor/formlayout.py ->
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      copying lib/matplotlib/backends/qt_editor/__init__.py ->
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      copying lib/matplotlib/backends/qt_editor/formsubplottool.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/qt_editor
      copying lib/matplotlib/backends/qt_editor/figureoptions.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/qt_editor
      creating
build/lib.macosx-10.9-universal2-3.9/matplotlib/testing/jpl units
      copying lib/matplotlib/testing/jpl_units/UnitDblFormatter.py ->
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      copying lib/matplotlib/testing/jpl_units/UnitDbl.py ->
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      copying lib/matplotlib/testing/jpl_units/Duration.py ->
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build/lib.macosx-10.9-universal2-3.9/matplotlib/testing/jpl_units
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copying lib/matplotlib/testing/jpl_units/UnitDblConverter.py ->

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      copying lib/matplotlib/testing/jpl_units/StrConverter.py ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/testing/jpl units
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data/images
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data/stylelib
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ui-1.12.1
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ui-1.12.1/jquery-ui.structure.css ->
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ui-1.12.1
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data/fonts
      creating build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/afm/pzdr.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/filesave_large.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/jquery-ui.theme.css ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/mpl-data/images/help.ppm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      creating build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/sample_data
      copying lib/matplotlib/mpl-data/sample_data/demodata.csv ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/afm/putb8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/subplots.png ->
```

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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/home_large.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/back.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/back_large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
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-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
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ui-1.12.1/external/jquery
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ui-1.12.1/external/jquery
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-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
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data/fonts/ttf
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-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
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      copying lib/matplotlib/mpl-data/fonts/afm/ptmbi8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
```

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copying lib/matplotlib/mpl-data/fonts/afm/pbkdi8a.afm ->
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      copying lib/matplotlib/mpl-data/images/back.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/zoom to rect.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/home large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/subplots.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/tableau-
colorblind10.mplstyle -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/stylelib
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-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
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lib/matplotlib/backends/web_backend/jquery/js/jquery.min.js ->
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data/fonts/pdfcorefonts
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Bold.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
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Bold.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
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data/fonts/ttf
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Oblique.ttf -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/ttf
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
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copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Courier-
BoldOblique.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
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ui-1.12.1
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BoldItalic.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Times-
Italic.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
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ui-1.12.1/images/ui-icons_777777_256x240.png ->
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ui-1.12.1/images
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXNonUniBolIta.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/ttf/cmmi10.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
```

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build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/js
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXSizOneSymReg.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
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ui-1.12.1/images/ui-icons_cc0000_256x240.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Helvetica-
BoldOblique.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/stylelib/seaborn-
darkgrid.mplstyle -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/stylelib
      copying lib/matplotlib/mpl-data/sample_data/goog.npz ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/afm/phvbo8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/home.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/jquery-ui.css ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
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build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXSizThreeSymReg.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
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data/fonts/pdfcorefonts/ZapfDingbats.afm ->
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      copying lib/matplotlib/mpl-data/images/forward.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/seaborn-
whitegrid.mplstyle -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/stylelib
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/LICENSE.txt ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
      copying lib/matplotlib/mpl-data/sample_data/eeg.dat ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
```

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-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/fonts/ttf/LICENSE_DEJAVU ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-
data/sample_data/jacksboro_fault_dem.npz ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/stylelib/ggplot.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/fonts/afm/pbkl8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/filesave.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/back_large.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/pcrb8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/sample_data/README.txt ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/stylelib/seaborn.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/stylelib/fast.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSansMono.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/qt4_editor_options.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/pplb8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/afm/phvr8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/afm/pcrro8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/cmsy10.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/backends/web_backend/js/mpl_tornado.js ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/js
      copying lib/matplotlib/mpl-data/fonts/afm/cmmi10.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/subplots_large.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/ttf/cmb10.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/index.html ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/mpl-data/images/hand_large.gif ->
```

```
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/jquery-ui.js ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/mpl-data/fonts/afm/psyr.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/help_large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSerif-Italic.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/stylelib/seaborn-dark.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/images/forward_large.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/matplotlib.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Symbol.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/images/zoom_to_rect_large.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/move.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/phvro8an.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSans.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/stylelib/fivethirtyeight.mplstyle
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-
data/sample_data/percent_bachelors_degrees_women_usa.csv ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXSizFiveSymReg.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Helvetica.afm
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/images/help.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXSizFourSymBol.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXNonUniIta.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/afm/ptmri8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/home.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images/ui-icons_555555_256x240.png ->
```

```
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images
      copying lib/matplotlib/mpl-data/fonts/afm/phvb8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/stylelib/seaborn-deep.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/backends/web backend/js/mpl.js ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/js
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXNonUni.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/afm/phvb8an.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/back.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/bmh.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/jquery-ui.theme.min.css ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/mpl-data/stylelib/seaborn-dark-
palette.mplstyle -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/stylelib
      copying lib/matplotlib/mpl-data/fonts/afm/pncb8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/zoom_to_rect.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/cmtt10.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXGeneralItalic.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/stylelib/grayscale.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images/ui-icons ffffff 256x240.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images
      copying lib/matplotlib/mpl-data/images/move.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/jquery-ui.structure.min.css ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSerif-
BoldItalic.ttf -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/forward.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
```

```
copying lib/matplotlib/mpl-data/images/back.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/sample_data/topobathy.npz ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/images/move large.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/sample data/data x x2 x3.csv ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/afm/putri8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/home.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/package.json ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/mpl-data/images/home.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/pagko8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/afm/pncri8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/cmex10.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/jquery-ui.min.css ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/backends/web_backend/css/fbm.css ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/css
      copying lib/matplotlib/mpl-
data/sample_data/Minduka_Present_Blue_Pack.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSansMono-Bold.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/sample_data/embedding_in_wx3.xrc
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample data
      copying lib/matplotlib/mpl-data/images/hand.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/cmr10.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/afm/putr8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/filesave_large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/sample_data/grace_hopper.jpg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/stylelib/Solarize_Light2.mplstyle
```

```
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/images/zoom_to_rect.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/backends/web_backend/single_figure.html ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web backend
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXSizTwoSymReg.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXGeneralBolIta.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/sample_data/ct.raw.gz ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/stylelib/_classic_test.mplstyle
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/images/filesave.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Helvetica-
Oblique.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/sample_data/aapl.npz ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Times-Bold.afm
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/fonts/afm/pagk8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/jquery-ui.min.js ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1
      copying lib/matplotlib/mpl-data/fonts/afm/pbkli8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/afm/pbkd8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/sample_data/msft.csv ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Times-
Roman.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/images/zoom_to_rect_large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/seaborn-bright.mplstyle
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/fonts/afm/cmsy10.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/sample_data/None_vs_nearest-
pdf.png -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXGeneralBol.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/ttf/cmr10.ttf ->
```

```
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/stylelib/seaborn-
notebook.mplstyle -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/stylelib
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSans-
BoldOblique.ttf -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/afm/pcrbo8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/afm/phvr8an.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/afm/phvbo8an.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/backends/web_backend/nbagg_uat.ipynb ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images/ui-icons_444444_256x240.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images
      copying lib/matplotlib/mpl-data/stylelib/dark background.mplstyle
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/fonts/afm/phvlo8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/sample_data/ada.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
      copying lib/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images/ui-icons_777620_256x240.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery-
ui-1.12.1/images
      copying lib/matplotlib/mpl-data/images/forward.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/pncr8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/cmtt10.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/ttf/LICENSE_STIX ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/zoom_to_rect.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/backends/web_backend/css/boilerplate.css
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/css
      copying lib/matplotlib/mpl-data/fonts/pdfcorefonts/Courier-
Oblique.afm -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/fonts/pdfcorefonts
      copying lib/matplotlib/mpl-data/images/subplots_large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/sample_data/logo2.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample_data
```

```
copying lib/matplotlib/mpl-data/stylelib/seaborn-
colorblind.mplstyle -> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-
data/stylelib
      copying lib/matplotlib/mpl-data/images/subplots.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/hand.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/seaborn-talk.mplstyle ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-
data/images/qt4_editor_options_large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/hand.svg ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/pplri8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/forward.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/pncbi8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/backends/web_backend/all_figures.html ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend
      copying lib/matplotlib/mpl-data/images/matplotlib.ppm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/pplr8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/DejaVuSansDisplay.ttf
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/images/matplotlib_large.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/stylelib/seaborn-white.mplstyle
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/images/hand.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/phvro8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXNonUniBol.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying
lib/matplotlib/backends/web_backend/ipython_inline_figure.html ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend
      copying lib/matplotlib/mpl-data/images/qt4_editor_options.pdf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/help.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/afm/phvl8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/mpl-data/images/qt4_editor_options.svg ->
```

```
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/images/help_large.ppm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/sample_data/grace_hopper.png ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/sample data
      copying lib/matplotlib/mpl-data/images/move.gif ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/images
      copying lib/matplotlib/mpl-data/fonts/ttf/STIXSizTwoSymBol.ttf ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/ttf
      copying lib/matplotlib/mpl-data/fonts/afm/pagdo8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      copying lib/matplotlib/backends/web_backend/jquery/js/jquery.js
->
build/lib.macosx-10.9-universal2-3.9/matplotlib/backends/web_backend/jquery/js
      copying lib/matplotlib/mpl-data/stylelib/seaborn-poster.mplstyle
-> build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/stylelib
      copying lib/matplotlib/mpl-data/fonts/afm/pplbi8a.afm ->
build/lib.macosx-10.9-universal2-3.9/matplotlib/mpl-data/fonts/afm
      UPDATING
build/lib.macosx-10.9-universal2-3.9/matplotlib/ version.py
      set build/lib.macosx-10.9-universal2-3.9/matplotlib/_version.py
to '3.1.3'
      running build_ext
      building 'matplotlib.ft2font' extension
      creating build/temp.macosx-10.9-universal2-3.9
      creating build/temp.macosx-10.9-universal2-3.9/src
      clang -Wno-unused-result -Wsign-compare -Wunreachable-code -fno-
common -dynamic -DNDEBUG -g -fwrapv -03 -Wall
-iwithsysroot/System/Library/Frameworks/System.framework/PrivateHeaders -iwithsy
sroot/Applications/Xcode.app/Contents/Developer/Library/Frameworks/Python3.frame
work/Versions/3.9/Headers -arch arm64 -arch x86_64 -Werror=implicit-function-
declaration -Wno-error=unreachable-code -Wno-error=unused-but-set-variable -Wno-
error=cast-function-type-mismatch -Wno-unknown-warning-option
-DFREETYPE_BUILD_TYPE=system
-DPY ARRAY UNIQUE SYMBOL=MPL matplotlib ft2font ARRAY API
-DNPY_NO_DEPRECATED_API=NPY_1_7_API_VERSION -D__STDC_FORMAT_MACROS=1
-Iextern/agg24-svn/include -I/Users/haley/Library/Python/3.9/lib/python/site-
packages/numpy/core/include -I/Applications/Xcode.app/Contents/Developer/Library
/Frameworks/Python3.framework/Versions/3.9/include/python3.9 -c
src/checkdep_freetype2.c -o
build/temp.macosx-10.9-universal2-3.9/src/checkdep_freetype2.o
-I/opt/homebrew/opt/freetype/include/freetype2
-I/opt/homebrew/opt/libpng/include/libpng16
      src/checkdep_freetype2.c:7:9: warning: Compiling with FreeType
version 2.13.3. [-W#pragma-messages]
          7 | #pragma message("Compiling with FreeType version " \
      1 warning generated.
```

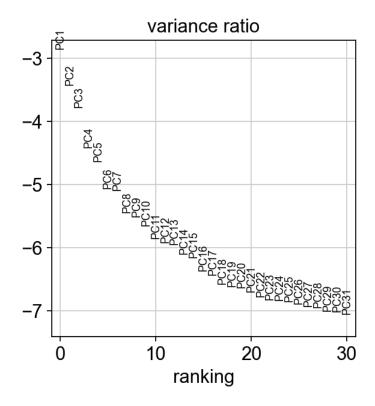
```
src/checkdep_freetype2.c:7:9: warning: Compiling with FreeType
version 2.13.3. [-W#pragma-messages]
         7 | #pragma message("Compiling with FreeType version " \
     1 warning generated.
     clang -Wno-unused-result -Wsign-compare -Wunreachable-code -fno-
common -dynamic -DNDEBUG -g -fwrapv -03 -Wall
-iwithsysroot/System/Library/Frameworks/System.framework/PrivateHeaders -iwithsy
sroot/Applications/Xcode.app/Contents/Developer/Library/Frameworks/Python3.frame
work/Versions/3.9/Headers -arch arm64 -arch x86_64 -Werror=implicit-function-
declaration -Wno-error=unreachable-code -Wno-error=unused-but-set-variable -Wno-
error=cast-function-type-mismatch -Wno-unknown-warning-option
-DFREETYPE_BUILD_TYPE=system
-DPY_ARRAY_UNIQUE_SYMBOL=MPL_matplotlib_ft2font_ARRAY_API
-DNPY_NO_DEPRECATED_API=NPY_1_7_API_VERSION -D__STDC_FORMAT_MACROS=1
-Iextern/agg24-svn/include -I/Users/haley/Library/Python/3.9/lib/python/site-
packages/numpy/core/include -I/Applications/Xcode.app/Contents/Developer/Library
/Frameworks/Python3.framework/Versions/3.9/include/python3.9 -c src/ft2font.cpp
-o build/temp.macosx-10.9-universal2-3.9/src/ft2font.o
-I/opt/homebrew/opt/freetype/include/freetype2
-I/opt/homebrew/opt/libpng/include/libpng16
     src/ft2font.cpp:223:29: error: assigning to 'char *' from
'unsigned char *' converts between pointers to integer types where one is of the
unique plain 'char' type and the other is not
       223 I
                     tags = outline.tags + first;
                            src/ft2font.cpp:339:29: error: assigning to 'char *' from
'unsigned char *' converts between pointers to integer types where one is of the
unique plain 'char' type and the other is not
        339 |
                     tags = outline.tags + first;
                            2 errors generated.
     error: command '/usr/bin/clang' failed with exit code 1
     [end of output]
 note: This error originates from a subprocess, and is likely not a
problem with pip.
 ERROR: Failed building wheel for matplotlib
 Running setup.py clean for matplotlib
Failed to build matplotlib
[notice] A new release of pip is
available: 25.0.1 -> 25.1.1
[notice] To update, run:
/Applications/Xcode.app/Contents/Developer/usr/bin/python3 -m pip
install --upgrade pip
```

```
ERROR: Failed to build installable wheels for some pyproject.toml based
projects (matplotlib)
Defaulting to user installation because normal site-packages is not
writeable
Requirement already satisfied: AnnData in
/Users/haley/Library/Python/3.9/lib/python/site-packages (0.10.8)
Requirement already satisfied: array-api-compat!=1.5,>1.4 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (1.7.1)
Requirement already satisfied: exceptiongroup in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (1.2.1)
Requirement already satisfied: h5py>=3.1 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (3.11.0)
Requirement already satisfied: natsort in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (8.4.0)
Requirement already satisfied: numpy>=1.23 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (1.26.4)
Requirement already satisfied: packaging>=20.0 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (24.1)
Requirement already satisfied: pandas!=2.1.0rc0,!=2.1.2,>=1.4 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (2.2.2)
Requirement already satisfied: scipy>1.8 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from AnnData) (1.13.1)
Requirement already satisfied: python-dateutil>=2.8.2 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
pandas!=2.1.0rc0,!=2.1.2,>=1.4->AnnData) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
pandas!=2.1.0rc0,!=2.1.2,>=1.4->AnnData) (2024.1)
Requirement already satisfied: tzdata>=2022.7 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
pandas!=2.1.0rc0,!=2.1.2,>=1.4->AnnData) (2024.1)
Requirement already satisfied: six>=1.5 in /Applications/Xcode.app/Contents/Deve
loper/Library/Frameworks/Python3.framework/Versions/3.9/lib/python3.9/site-
packages (from python-dateutil>=2.8.2->pandas!=2.1.0rc0,!=2.1.2,>=1.4->AnnData)
(1.15.0)
[notice] A new release of pip is
available: 25.0.1 -> 25.1.1
[notice] To update, run:
/Applications/Xcode.app/Contents/Developer/usr/bin/python3 -m pip
install --upgrade pip
filtered out 15132 genes that are detected in less than 3 cells
/Users/haley/Library/Python/3.9/lib/python/site-
packages/anndata/_core/anndata.py:1820: UserWarning: Variable names are not
unique. To make them unique, call `.var_names_make_unique`.
```

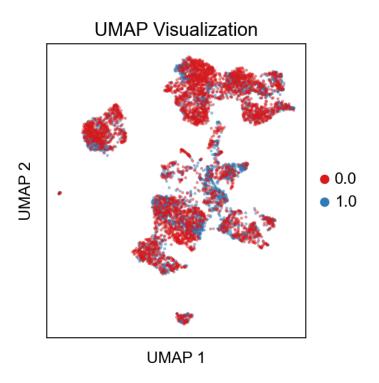
```
[43]: import scanpy as sc
      import matplotlib.pyplot as plt
      import seaborn as sns
      adata.var_names_make_unique()
      # Preprocessing
      sc.pp.normalize_total(adata, target_sum=1e4)
      sc.pp.log1p(adata)
      # sc.pp.highly variable genes(adata, min mean=0.0125, max mean=3, min disp=0.5)
      # sc.pl.highly_variable_genes(adata, save='5.pdf')
      # Dimensionality reduction
      sc.tl.pca(adata, svd_solver='arpack')
      sc.pl.pca_variance_ratio(adata, log=True, save='6.pdf')
      # Compute neighbors and UMAP (only once)
      sc.pp.neighbors(adata, n_neighbors=10, n_pcs=40)
      sc.tl.umap(adata)
      # Create the UMAP plot
      sc.pl.umap(adata,
                 color='gender',
                 palette={1: '#2c7bb6', 0: '#d7191c'},
                 size=20,
                 alpha=0.5,
                 frameon=True,
                 legend_loc='right margin',
                 title='UMAP Visualization',
                 legend_fontsize=12,
                 show=False)
      # Customize the plot
      ax = plt.gca()
      ax.set_title('UMAP Visualization', fontsize=14)
      plt.xlabel('UMAP 1', fontsize=12, labelpad=10)
      plt.ylabel('UMAP 2', fontsize=12, labelpad=10)
      plt.grid(False)
      ax.set facecolor('white')
      ax.spines['top'].set_visible(True)
      ax.spines['right'].set_visible(True)
      ax.spines['bottom'].set_visible(True)
      ax.spines['left'].set_visible(True)
      plt.tight_layout()
```

```
normalizing counts per cell
   finished (0:00:00)
computing PCA
  with n_comps=50
  finished (0:00:32)
```

WARNING: saving figure to file figures/pca_variance_ratio6.pdf



```
computing neighbors
   using 'X_pca' with n_pcs = 40
   finished: added to `.uns['neighbors']`
   `.obsp['distances']`, distances for each pair of neighbors
   `.obsp['connectivities']`, weighted adjacency matrix (0:00:00)
computing UMAP
   finished: added
   'X_umap', UMAP coordinates (adata.obsm) (0:00:06)
```



1.4 LOAD test data

```
[4]: !pip install pyranges
!pip install scanpy
import numpy as np
import pandas as pd
import scanpy as sc
import numpy as np
import pandas as pd
import scanpy as sc
neww=sc.read_h5ad("4_type.h5ad")
Defaulting to user installation because normal site-packages is not writeable
```

```
Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: pyranges in
/Users/haley/Library/Python/3.9/lib/python/site-packages (0.1.4)
Requirement already satisfied: pandas in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from pyranges) (2.2.2)
Requirement already satisfied: ncls>=0.0.63 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from pyranges)
(0.0.68)
Requirement already satisfied: tabulate in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from pyranges) (0.9.0)
Requirement already satisfied: sorted_nearest>=0.0.33 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from pyranges)
```

```
(0.0.39)
Requirement already satisfied: natsort in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from pyranges) (8.4.0)
Requirement already satisfied: numpy in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
ncls>=0.0.63->pyranges) (1.26.4)
Requirement already satisfied: python-dateutil>=2.8.2 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from pandas->pyranges)
(2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from pandas->pyranges)
Requirement already satisfied: tzdata>=2022.7 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from pandas->pyranges)
Requirement already satisfied: six>=1.5 in /Applications/Xcode.app/Contents/Deve
loper/Library/Frameworks/Python3.framework/Versions/3.9/lib/python3.9/site-
packages (from python-dateutil>=2.8.2->pandas->pyranges) (1.15.0)
[notice] A new release of pip is
available: 25.0.1 -> 25.1.1
[notice] To update, run:
/Applications/Xcode.app/Contents/Developer/usr/bin/python3 -m pip
install --upgrade pip
Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: scanpy in
/Users/haley/Library/Python/3.9/lib/python/site-packages (1.10.2)
Requirement already satisfied: anndata>=0.8 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (0.10.8)
Requirement already satisfied: get-annotations in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (0.1.2)
Requirement already satisfied: h5py>=3.1 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (3.11.0)
Requirement already satisfied: joblib in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (1.4.2)
Requirement already satisfied: legacy-api-wrap>=1.4 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (1.4)
Requirement already satisfied: matplotlib>=3.6 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (3.9.1)
Requirement already satisfied: natsort in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (8.4.0)
Requirement already satisfied: networkx>=2.7 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (3.2.1)
Requirement already satisfied: numba>=0.56 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (0.60.0)
Requirement already satisfied: numpy<2,>=1.23 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (1.26.4)
```

```
Requirement already satisfied: packaging>=21.3 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (24.1)
Requirement already satisfied: pandas>=1.5 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (2.2.2)
Requirement already satisfied: patsy in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (0.5.6)
Requirement already satisfied: pynndescent>=0.5 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (0.5.13)
Requirement already satisfied: scikit-learn>=0.24 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (1.5.1)
Requirement already satisfied: scipy>=1.8 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (1.13.1)
Requirement already satisfied: seaborn>=0.13 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (0.13.2)
Requirement already satisfied: session-info in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (1.0.0)
Requirement already satisfied: statsmodels>=0.13 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (0.14.2)
Requirement already satisfied: tqdm in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (4.66.4)
Requirement already satisfied: umap-learn!=0.5.0,>=0.5 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from scanpy) (0.5.6)
Requirement already satisfied: array-api-compat!=1.5,>1.4 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
anndata\geq 0.8- \approx (1.7.1)
Requirement already satisfied: exceptiongroup in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
anndata>=0.8->scanpy) (1.2.1)
Requirement already satisfied: contourpy>=1.0.1 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
matplotlib>=3.6->scanpy) (1.2.1)
Requirement already satisfied: cycler>=0.10 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
matplotlib>=3.6->scanpy) (0.12.1)
Requirement already satisfied: fonttools>=4.22.0 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
matplotlib >= 3.6 -> scanpy) (4.53.1)
Requirement already satisfied: kiwisolver>=1.3.1 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
matplotlib>=3.6->scanpy) (1.4.5)
Requirement already satisfied: pillow>=8 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
matplotlib>=3.6->scanpy) (10.4.0)
Requirement already satisfied: pyparsing>=2.3.1 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
matplotlib>=3.6->scanpy) (3.1.2)
Requirement already satisfied: python-dateutil>=2.7 in
/Users/haley/Library/Python/3.9/lib/python/site-packages (from
```

```
Requirement already satisfied: importlib-resources>=3.2.0 in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (from
    matplotlib>=3.6->scanpy) (6.4.0)
    Requirement already satisfied: llvmlite<0.44,>=0.43.0dev0 in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (from
    numba > = 0.56 - scanpy) (0.43.0)
    Requirement already satisfied: pytz>=2020.1 in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (from
    pandas>=1.5->scanpy) (2024.1)
    Requirement already satisfied: tzdata>=2022.7 in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (from
    pandas>=1.5->scanpy) (2024.1)
    Requirement already satisfied: threadpoolctl>=3.1.0 in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (from scikit-
    learn > = 0.24 - scanpy) (3.5.0)
    Requirement already satisfied: six in /Applications/Xcode.app/Contents/Developer
    /Library/Frameworks/Python3.framework/Versions/3.9/lib/python3.9/site-packages
    (from patsy->scanpy) (1.15.0)
    Requirement already satisfied: stdlib-list in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (from session-
    info->scanpy) (0.10.0)
    Requirement already satisfied: zipp>=3.1.0 in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (from importlib-
    resources>=3.2.0->matplotlib>=3.6->scanpy) (3.19.2)
    [notice] A new release of pip is
    available: 25.0.1 -> 25.1.1
    [notice] To update, run:
    /Applications/Xcode.app/Contents/Developer/usr/bin/python3 -m pip
    install --upgrade pip
[5]: import scanpy as sc
     import pandas as pd
     import matplotlib.pyplot as plt
     gene_ids = neww.var['gene_ids']
     # # Load the dataset
     adata = sc.read("4_type.h5ad")
     conditions = adata.obs['conditions'].unique()
     # Separating adata based on conditions
     # It's a good idea to check the actual condition values and modify as needed
     adata1 = adata[adata.obs['conditions'] == conditions[0], :]
     adata2 = adata[adata.obs['conditions'] == conditions[1], :]
     adata3 = adata[adata.obs['conditions'] == conditions[2], :]
     adata4 = adata[adata.obs['conditions'] == conditions[3], :]
```

matplotlib>=3.6->scanpy) (2.9.0.post0)

```
[6]: View of AnnData object with n_obs × n_vars = 8268 × 20902
        obs: 'dataset', 'sample', 'n_genes', 'n_genes_by_counts', 'total_counts',
     'total_counts_mt', 'pct_counts_mt', 'conditions'
        var: 'gene_ids', 'feature_types', 'n_cells', 'mt', 'n_cells_by_counts',
     'mean_counts', 'pct_dropout_by_counts', 'total_counts', 'highly_variable',
     'means', 'dispersions', 'dispersions_norm'
        uns: 'hvg', 'log1p'
[7]:
    conditions
[7]: ['WT_WT', 'mdx_WT', 'WT_mdx', 'mdx_mdx']
     Categories (4, object): ['WT_WT', 'mdx_WT', 'WT_mdx', 'mdx_mdx']
[8]: import scanpy as sc
     # Assuming adata1 is already defined and loaded
     # Plot the 20 highest expressed genes
     # sc.pl.highest_expr_genes(adata1, n_top=20)
     # Filter out genes that are detected in fewer than 3 cells
     # sc.pp.filter_genes(adata1, min_cells=3)
     # # # Normalize the data (scaling each cell to a total count of 1e4)
     # sc.pp.normalize_total(adata1, target_sum=1e4)
     # # # Logarithmically scale the data
     # sc.pp.log1p(adata1)
     # Convert the sparse matrix to a dense matrix and extract gene IDs
     matrix = adata1.X.todense() # Your matrix data
     gene_ids = adata1.var.index # Your gene IDs
     # Create a DataFrame
     df_adata1= pd.DataFrame(matrix, columns=gene_ids)
[9]: df_adata1
[9]:
          Xkr4 Gm1992 Rp1 Sox17
                                    Gm37587
                                                                     Tcea1 Rgs20 \
                                                Mrpl15
                                                          Lypla1
     0
           0.0
                    0.0 0.0
                                0.0
                                         0.0 0.000000 0.000000 1.067894
                                                                              0.0
     1
           0.0
                    0.0 0.0
                                0.0
                                        0.0 0.000000 0.000000
                                                                  1.189107
                                                                              0.0
     2
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                    0.0 0.0
                               0.0
                                        0.0 0.000000 0.000000
                                                                  1.751662
                                                                              0.0
           0.0
     3
                    0.0 0.0
                               0.0
                                        0.0 0.000000 0.000000
                                                                  1.616192
                                                                              0.0
           0.0
     4
                    0.0 0.0
                                         0.0 0.473922 0.473922
                                                                              0.0
                                0.0
                                                                 1.036327
```

[6]: adata1

```
8263
       0.0
               0.0 0.0
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                                     0.0 0.000000 0.000000
                                                               0.000000
                                                                           0.0
8264
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8267
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                                    AC132444.1
                                               AC132444.5 Csprs \
       Atp6v1h ...
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4
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                                                              2.204726
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8263
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                                                              3.339708
8264
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                                                              0.361726
8265
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                                                              0.831465
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             0.0
                                                 0.000000
                                                              1.166482
8267
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                  0.000000
                               0.0
                                      0.0
                                                 0.660885
                                                              0.000000
```

[8268 rows x 20902 columns]

```
[12]: extracted_columns_1
[12]:
                                                                       Kdm5d \
           Xist
                     Ddx3y
                             Gm42418
                                       Eif2s3y
                                                 Rps27rt Rp19-ps6
      0
             0.0
                 0.492508
                           4.253757
                                      0.492508
                                                0.492508
                                                          0.000000
                                                                    0.000000
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                                                          1.189107
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                                      0.000000
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                           5.250138
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                            4.308918
                                      0.000000
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                                                          0.473922
                                                                    0.000000
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                                      0.000000
                                                0.000000
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             0.0
      8264
             0.0
                 0.361726
                           6.557333
                                      0.626800
                                                0.000000
                                                          0.000000
                                                                    0.626800
             0.0 0.499769
      8265
                           5.148435
                                      0.499769
                                                0.000000
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                                                                    0.499769
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                                                         0.744476
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                           5.648857
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               Uba52
                         Rp135
                               Rpl36a-ps1
                                                 Uty
                                                         Wdr89
                                                                   Lars2
                                                                             Rps27
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            0.821026
                     2.749676
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                                            0.492508
                                                      0.000000
                                                                1.430808
                                                                          3.000345
      1
                                            0.000000
                                                      0.000000
                                                                1.717089
            1.189107
                     2.519368
                                  0.000000
                                                                          3.625577
      2
            1.218493
                     2.727364
                                  0.000000
                                            0.000000
                                                      0.000000
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                                                                          3.603629
                                 0.000000
                                                                1.616192 1.616192
      3
            0.000000 0.000000
                                                     0.000000
                                            1.616192
            1.231138
                     2.037198
                                  0.000000
                                            1.231138
                                                      0.000000
                                                                1.766484
                                                                          2.619722
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                                                                3.733287
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      8263 0.000000
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      8264
           0.626800
                     2.227895
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                                            0.836129
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                                                      0.000000
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                     2.172486
                                  0.000000
                                            0.000000
                                                                1.587201
                                                                          3.192131
      8266
           0.000000
                      1.875901
                                  0.000000
                                            0.744476
                                                      0.744476
                                                                2.167609
                                                                          2.489346
      8267
           0.000000
                     1.889949
                                  0.000000
                                            0.660885
                                                      0.000000
                                                                2.243742
                                                                          2.882401
      [8268 rows x 14 columns]
```

2 infer data

```
[85]: inf_data=sc.read_h5ad("IFNR_infer_sexpredect_v2(1).h5ad" )
# sc.pp.log1p(inf_data)
```

/Users/haley/Library/Python/3.9/lib/python/sitepackages/anndata/_core/anndata.py:1818: UserWarning: Observation names are not unique. To make them unique, call `.obs_names_make_unique`. utils.warn_names_duplicates("obs")

```
[86]: import scanpy as sc

# Suppose we have 'group_column' in inf_data.obs
# that can take values "groupA" and "groupB"

inf_IFNRKO = inf_data[inf_data.obs["group"] == "IFNRKO"].copy()
inf_CD45_1 = inf_data[inf_data.obs["group"] == "CD45_1"].copy()
```

```
/Users/haley/Library/Python/3.9/lib/python/site-
     packages/anndata/_core/anndata.py:1818: UserWarning: Observation names are not
     unique. To make them unique, call `.obs_names_make_unique`.
       utils.warn_names_duplicates("obs")
     /Users/haley/Library/Python/3.9/lib/python/site-
     packages/anndata/_core/anndata.py:1818: UserWarning: Observation names are not
     unique. To make them unique, call `.obs names make unique`.
       utils.warn_names_duplicates("obs")
[87]: def process_inf_data(inf_data):
          import scanpy as sc
          import pandas as pd
          # Apply log1p
          sc.pp.log1p(inf_data)
          # Create the DataFrame
          matrix = inf_data.X
          gene_ids = inf_data.var.index
          df = pd.DataFrame(matrix, columns=gene_ids)
          return df
[88]: inf_IFNRKO_df = process_inf_data(inf_IFNRKO)
      inf_CD45_1_df = process_inf_data(inf_CD45_1)
[89]: extracted_columns_IFNRKO = inf_IFNRKO_df[columns_to_extract]
      extracted_columns_CD45 = inf_CD45_1_df[columns_to_extract]
[90]: import numpy as np
      import pandas as pd
      import matplotlib.pyplot as plt
      from sklearn.preprocessing import StandardScaler
      from xgboost import XGBClassifier
      def predict_and_plot(
          df,
          model.
          scaler,
          output_prefix="output"
      ):
          Scales the DataFrame `df` using the provided `scaler`,
          predicts labels using `model`,
          saves the predictions to a CSV file,
          and generates a bar plot of predicted label proportions.
```

```
Parameters
   _____
  df : pd.DataFrame
      The test DataFrame to be predicted.
  model : a fitted sklearn-like model (e.g. XGBClassifier)
      The model used for prediction.
  scaler: an already fitted scaler (StandardScaler or similar)
      Used to transform the test data.
  output\_prefix : str
      A prefix for naming the output CSV and PNG files.
  # Scale the data
  df_scaled = scaler.transform(df)
  # Predict labels
  predicted_labels = model.predict(df_scaled)
  # Convert predictions to a DataFrame and save
  output_csv = f"{output_prefix}_predicted_labels.csv"
  pd.DataFrame({"Prediction": predicted_labels}).to_csv(output_csv,__
→index=False)
  print(f"Predicted labels saved to '{output_csv}'.")
  # Calculate percentages for the bar plot
  label_counts = pd.Series(predicted_labels).value_counts(normalize=True) *__
→100 # Convert to percentages
  # Create a bar plot
  ax = label_counts.plot(kind='bar', color=['blue', 'orange'], alpha=0.7)
  plt.xlabel("Prediction")
  plt.ylabel("Percentage")
  plt.title("Proportions of Predicted Labels")
  plt.xticks([0, 1], labels=['0', '1'], rotation=0)
  plt.ylim(0, 100) # Ensure the y-axis ranges from 0 to 100
  # Add percentage labels on the bars
  for i, v in enumerate(label_counts):
      plt.text(i, v + 1, f"{v:.2f}%", ha='center', va='bottom', fontsize=7)
  # Save and close the plot
  output_png = f"{output_prefix}_prediction_percentage_barplot.png"
  plt.savefig(output_png, dpi=300)
  plt.close() # Close the figure so we don't keep re-drawing
  print(f"Bar plot of predicted label percentages saved to '{output_png}'.\n")
```

```
[91]: # For reproducibility
      np.random.seed(42)
      # Example training data: full_X_train, full_Y_train
      # You have your model definition, for example:
      model = XGBClassifier(
          use label encoder=False,
          eval_metric='logloss',
          n estimators=300,
          learning_rate=0.01,
          max depth=6,
          random state=551
      )
      # Fit the scaler on training data
      scaler = StandardScaler()
      full_X_train_scaled = scaler.fit_transform(full_X_train)
      # Fit the model on the training data
      model.fit(full_X_train_scaled, full_Y_train)
     /Users/haley/Library/Python/3.9/lib/python/site-packages/xgboost/core.py:158:
     UserWarning: [21:14:19] WARNING:
     /Users/runner/work/xgboost/xgboost/src/learner.cc:740:
     Parameters: { "use_label_encoder" } are not used.
       warnings.warn(smsg, UserWarning)
[91]: XGBClassifier(base_score=None, booster=None, callbacks=None,
                    colsample_bylevel=None, colsample_bynode=None,
                    colsample_bytree=None, device=None, early_stopping_rounds=None,
                    enable categorical=False, eval metric='logloss',
                    feature_types=None, gamma=None, grow_policy=None,
                    importance type=None, interaction constraints=None,
                    learning_rate=0.01, max_bin=None, max_cat_threshold=None,
                    max cat to onehot=None, max delta step=None, max depth=6,
                    max_leaves=None, min_child_weight=None, missing=nan,
                    monotone_constraints=None, multi_strategy=None, n_estimators=300,
                    n_jobs=None, num_parallel_tree=None, random_state=551, ...)
[92]: # Suppose you have two DataFrames:
        inf_IFNRKO_df
        inf\_CD45\_1\_df
      # Call the function for the first DataFrame
      predict_and_plot(
          df=extracted columns IFNRKO ,
          model=model,
```

```
scaler=scaler,
    output_prefix="inf_IFNRKO" # This will produce_
 → "inf_IFNRKO_predicted_labels.csv" and_
 → "inf_IFNRKO_prediction_percentage_barplot.png"
# Call the function for the second DataFrame
predict_and_plot(
    df=extracted_columns_CD45,
    model=model,
    scaler=scaler,
    output_prefix="inf_CD45_1"
)
Predicted labels saved to 'inf_IFNRKO_predicted_labels.csv'.
Bar plot of predicted label percentages saved to
'inf_IFNRKO_prediction_percentage_barplot.png'.
Predicted labels saved to 'inf_CD45_1_predicted_labels.csv'.
Bar plot of predicted label percentages saved to
'inf_CD45_1_prediction_percentage_barplot.png'.
3 process validation data
```

```
[36]: adata
[36]: AnnData object with n_obs \times n_vars = 6648 \times 17153
         obs: 'gender', 'n_genes_by_counts', 'total_counts', 'total_counts_mt',
      'pct_counts_mt'
         var: 'n_cells', 'mt', 'n_cells_by_counts', 'mean_counts',
      'pct_dropout_by_counts', 'total_counts'
         uns: 'log1p', 'pca', 'neighbors', 'umap', 'gender_colors'
         obsm: 'X_pca', 'X_umap'
         varm: 'PCs'
         obsp: 'distances', 'connectivities'
[37]: rc = pd.DataFrame(adata.X, index=adata.obs names, columns=adata.var names)
      rc["y"] = pd.Categorical(joined_list)
[38]: rc
[38]: n
                Sox17
                       Mrpl15
                                  Lypla1
                                             Tceal Gm16041 Atp6v1h
                                                                        Rb1cc1 \
      Cell 0
                  0.0 0.00000 1.354815 0.000000
                                                        0.0 0.00000 0.000000
      Cell 1
                  0.0 0.00000 0.000000 1.223296
                                                        0.0 0.00000 0.788087
      Cell_2
                  0.0 0.00000 0.000000 0.000000
                                                        0.0 0.00000 0.000000
      Cell 3
                  0.0 0.00000 0.000000 2.327345
                                                        0.0 0.00000 0.000000
```

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0.0 0.94693 0.000000
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                                                             CAAA01147332.1
                                               Vamp7
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Cell_0
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                       1.0
Cell 1
             0.788087
                        1.0
Cell_2
             0.000000
                       1.0
Cell 3
             2.327345
                        1.0
Cell 4
                        1.0
             0.946930
Cell_6643
             0.000000
                        0.0
Cell_6644
             0.000000
                        0.0
Cell_6645
             0.000000 0.0
Cell_6646
                        0.0
             0.000000
Cell_6647
             1.885036
                        0.0
```

4 test data

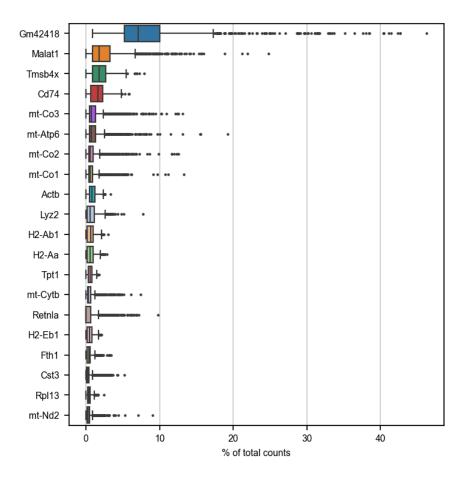
```
[18]: from sklearn.linear_model import LogisticRegression
      from xgboost import XGBRegressor
      from sklearn.svm import SVC
      from sklearn.pipeline import make_pipeline
      from sklearn.preprocessing import StandardScaler
      from sklearn.ensemble import RandomForestClassifier
      import numpy as np
      import pandas as pd
      import os
      import random
      import matplotlib
      import matplotlib.pyplot as plt
      from sklearn.datasets import load_svmlight_file
      from sklearn.preprocessing import StandardScaler
      from numpy.core.fromnumeric import transpose
      from sklearn.model_selection import train_test_split
      from sklearn.metrics import roc_curve, roc_auc_score
[21]: realmat = sc.read_mtx('matrix.mtx')
      realfeature=pd.read_csv('features.tsv',sep='\t',header=None)
      realmat=realmat[0:32285]
      realtag=pd.read_csv('tag.csv')
      actt=realfeature.loc[realfeature[2] == 'Gene Expression']
```

```
celltag=pd.read_csv('barcodes.tsv',sep='\t',header=None)
adata = anndata.AnnData(X=realmat.X.T)
lol = list(celltag[0]) # or pd.Series(actt[1])
adata.obs['cell'] = lol
lmao = list(actt[1]) # or pd.Series(actt[1])
adata.var['gene_ids'] = lmao
                                 # verbosity: errors (0), warnings (1),
sc.settings.verbosity = 3
 →info (2), hints (3)
sc.logging.print_header()
sc.settings.set_figure_params(dpi=80, facecolor='white', fontsize=7
sc.pp.filter_cells(adata, min_genes=100)
sc.pp.filter_genes(adata, min_cells=3)
adata
adata.var['mt'] = adata.var_names.str.startswith('mt-') # annotate the group_
⇔of mitochondrial genes as 'mt'
sc.pp.calculate_qc_metrics(adata, qc_vars=['mt'], percent_top=None,_
 ⇔log1p=False, inplace=True)
```

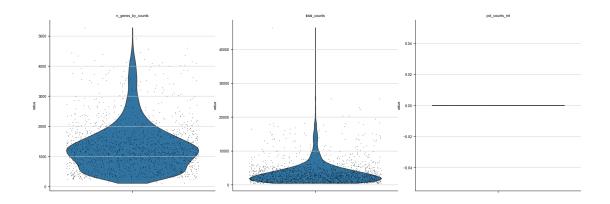
```
adata.var
# Remove headers and set the index to 'gene_ids'
adata.var.index = adata.var['gene_ids'].values
adata.var.index.name = None # This removes the header (name) of the index
adata.obs
#adata.var.index = actt[1].astype('string')
sc.pl.highest_expr_genes(adata, n_top=20, )
adata.var names make unique() # this is unnecessary if us
sc.pl.violin(adata, ['n_genes_by_counts', 'total_counts', 'pct_counts_mt'],
             jitter=0.4, multi_panel=True,save='8.pdf')
sc.pl.scatter(adata, x='total_counts', y='pct_counts_mt',save='9.pdf')
sc.pl.scatter(adata, x='total_counts', y='n_genes_by_counts',save='10.pdf')
sc.pp.normalize_total(adata, target_sum=1e4)
sc.pp.log1p(adata)
METT=adata.X
dense_matrix = METT.toarray()
#METT = METT[~METT['cell'].isin(realtag['cell_barcode'])
# Convert the dense NumPy array to a Pandas DataFrame
METT = pd.DataFrame(dense_matrix,columns=adata.var_names, index=adata.
 ⇔obs['cell'].values)
METT['cell'] = adata.obs['cell'].values
realtag.index=realtag['cell_barcode'].values
result = METT.join(realtag[['feature_call']], how='left')
print(METT.index.isin(realtag.index).sum())
result
# Remove rows where 'qender' is NaN
result = result.dropna(subset=['feature_call'])
print(result)
# Initialize the 'gender' column with NaN or some default value
result['gender'] = None
# Update 'gender' based on 'feature_call'
result.loc[result['feature_call'] == 'CMO305', 'gender'] = 1
result.loc[result['feature_call'] == 'CMO306', 'gender'] = 0
print(result)
columns_to_extract = ["Xist", "Ddx3y", "Gm42418", "Eif2s3y", "Rps27rt",
    "Rpl9-ps6", "Kdm5d", "Uba52", "Rpl35", "Rpl36a-ps1",
    "Uty", "Wdr89", "Lars2", "Rps27"]
extracted_df = result[columns_to_extract]
print(extracted df)
```

```
extracted_df['gender']=result['gender']
realtest=extracted_df.drop('gender',axis=1)
realtest_y=result['gender']
realtest_y = realtest_y.astype('float')
realtest_y
```

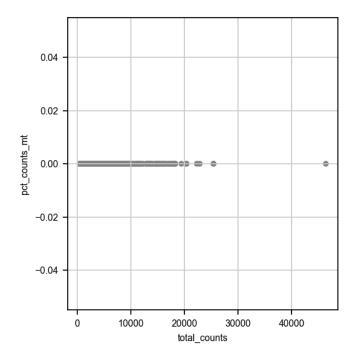
scanpy==1.10.2 anndata==0.10.8 umap==0.5.6 numpy==1.26.4 scipy==1.13.1
pandas==2.2.2 scikit-learn==1.5.1 statsmodels==0.14.2 pynndescent==0.5.13
filtered out 3 cells that have less than 100 genes expressed
filtered out 19261 genes that are detected in less than 3 cells
normalizing counts per cell
 finished (0:00:00)



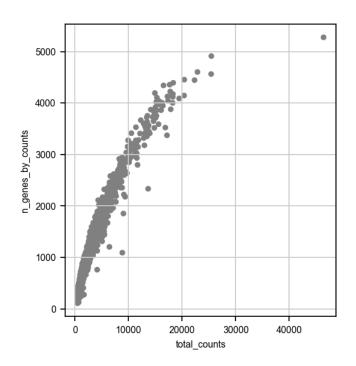
WARNING: saving figure to file figures/violin8.pdf



WARNING: saving figure to file figures/scatter9.pdf



WARNING: saving figure to file figures/scatter10.pdf



normalizing counts per cell finished (0:00:00) 1649

1043							
	Mrpl15	Lypla1	Tcea1	Atp6v1h	Rb1cc1	\	
AAACCCAAGACAGCTG-1	0.0	0.0	0.000000	0.0	0.000000		
AAACCCACAAAGACGC-1	0.0	0.0	0.000000	0.0	1.921457		
AAACGAAAGCACTCGC-1	0.0	0.0	0.000000	0.0	0.000000		
AAACGAACAAATGGTA-1	0.0	0.0	0.000000	0.0	0.000000		
AAAGGATGTGTGTTTG-1	0.0	0.0	0.000000	0.0	2.565319		
		•••					
TTTGGTTCAGTGGCTC-1	0.0	0.0	0.000000	0.0	0.000000		
TTTGGTTGTAACGTTC-1	0.0	0.0	0.000000	0.0	0.000000		
TTTGGTTTCAGCGTCG-1	0.0	0.0	1.767994	0.0	0.000000		
TTTGTTGGTACGTTCA-1	0.0	0.0	0.000000	0.0	0.000000		
TTTGTTGGTCCACTCT-1	0.0	0.0	1.086885	0.0	0.000000		
	4732440	D04Rik	Pcmtd1	Gm26901	Rrs1	2610203C22Rik	\
AAACCCAAGACAGCTG-1		0.0	0.744418	0.0	0.000000	0.0	
AAACCCACAAAGACGC-1		0.0	0.000000	0.0	0.000000	0.0	
AAACGAAAGCACTCGC-1		0.0	0.000000	0.0	0.000000	0.0	
AAACGAACAAATGGTA-1		0.0	0.000000	0.0	0.000000	0.0	
AAAGGATGTGTGTTTG-1		0.0	0.000000	0.0	0.000000	0.0	
		•••		•••		•••	
TTTGGTTCAGTGGCTC-1		0.0	1.859427	0.0	0.000000	0.0	
TTTGGTTGTAACGTTC-1		0.0	0.000000	0.0	0.000000	0.0	

```
TTTGGTTTCAGCGTCG-1
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                                     1.767994
                                                    0.0
                                                        0.000000
                                                                              0.0
TTTGTTGGTACGTTCA-1
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                                     0.000000
                                                    0.0
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TTTGTTGGTCCACTCT-1
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                                                         0.000000
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                            Vamp7
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[1649 rows x 13026 columns]
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                                                Atp6v1h
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                                         Tcea1
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TTTGGTTGTAACGTTC-1
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TTTGTTGGTCCACTCT-1
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                       AC125149.3
                                    AC168977.1 CAAA01118383.1
                                                                 Csprs
                                                                        Vamp7
AAACCCAAGACAGCTG-1
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                                                  TTTGGTTTCAGCGTCG-1
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                                       0.000000
                                                  TTTGTTGGTACGTTCA-1
TTTGTTGGTCCACTCT-1
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                                       1.086885
                                                  TTTGTTGGTCCACTCT-1
```

```
feature_call
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                                         1
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                           CM0306
                                         0
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                           CM0305
                                         1
TTTGTTGGTACGTTCA-1
                           CM0305
                                         1
TTTGTTGGTCCACTCT-1
                           CM0305
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[1649 rows x 13027 columns]
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                                           Gm42418
                                                     Eif2s3y
                                                                Rps27rt
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                                          Uba52
                                                    Rp135
                                                            Rpl36a-ps1
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TTTGGTTCAGTGGCTC-1
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                                                 2.471493
                                                                   0.0 0.0
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                                                 0.000000
TTTGGTTTCAGCGTCG-1
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                                      1.767994
                                                 2.745812
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                                                                        0.0
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                        Wdr89
                                  Lars2
                                             Rps27
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                               3.191448
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                                          2.538590
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```

```
TTTGGTTTCAGCGTCG-1 0.000000
                                   0.000000 2.745812
     TTTGTTGGTACGTTCA-1 0.000000
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     TTTGTTGGTCCACTCT-1 1.086885
                                   3.118624 2.181557
     [1649 rows x 14 columns]
     /var/folders/yw/8s12n9193cs3klhgvl8yg4p80000gn/T/ipykernel_92685/327623629.py:68
     : SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
       extracted_df['gender']=result['gender']
[21]: AAACCCAAGACAGCTG-1
                            1.0
                            1.0
      AAACCCACAAAGACGC-1
      AAACGAAAGCACTCGC-1
                            1.0
      AAACGAACAAATGGTA-1
                            1.0
      AAAGGATGTGTGTTTG-1
                            1.0
     TTTGGTTCAGTGGCTC-1
                            0.0
      TTTGGTTGTAACGTTC-1
                            0.0
      TTTGGTTTCAGCGTCG-1
                            1.0
      TTTGTTGGTACGTTCA-1
                            1.0
      TTTGTTGGTCCACTCT-1
                            1.0
     Name: gender, Length: 1649, dtype: float64
[22]: ffrc=rc.loc[:,columns to extract]
      ffrc["y"] = pd.Categorical(joined_list)
      female_exp = result[result['gender'] == 0]
      male_exp = result[result['gender'] == 1]
      female exp data= female exp.drop('gender',axis=1)
      male_exp_data= male_exp.drop('gender',axis=1)
[23]: result
[23]:
                          Mrpl15 Lypla1
                                             Tcea1 Atp6v1h
                                                               Rb1cc1 \
      AAACCCAAGACAGCTG-1
                             0.0
                                     0.0 0.000000
                                                        0.0
                                                             0.00000
                             0.0
      AAACCCACAAAGACGC-1
                                     0.0 0.000000
                                                        0.0
                                                             1.921457
                             0.0
      AAACGAAAGCACTCGC-1
                                     0.0 0.000000
                                                        0.0
                                                             0.000000
      AAACGAACAAATGGTA-1
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                                                             0.00000
      AAAGGATGTGTGTTTG-1
                             0.0
                                     0.0 0.000000
                                                        0.0
                                                             2.565319
      TTTGGTTCAGTGGCTC-1
                                     0.0 0.000000
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                             0.0
      TTTGGTTGTAACGTTC-1
                             0.0
                                     0.0 0.000000
                                                        0.0
                                                             0.00000
      TTTGGTTTCAGCGTCG-1
                             0.0
                                     0.0 1.767994
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                                                             0.00000
```

0.000000 2.131073

TTTGGTTGTAACGTTC-1

0.000000

TTTGTTGGTACGTTCA-1	0.0 0.0	0.000000	0.0 0.00000	0				
TTTGTTGGTCCACTCT-1	0.0 0.0	1.086885	0.0 0.00000	0				
	4732440D04Rik		m26901 Rrs1					
AAACCCAAGACAGCTG-1	0.0	0.744418	0.0 0.000000					
AAACCCACAAAGACGC-1	0.0	0.000000	0.0 0.000000					
AAACGAAAGCACTCGC-1	0.0	0.000000	0.0 0.000000					
AAACGAACAAATGGTA-1	0.0	0.000000	0.0 0.000000					
AAAGGATGTGTTTTG-1	0.0	0.000000	0.0 0.000000	0.0				
 TTTGGTTCAGTGGCTC-1	0.0	1.859427	0.0 0.000000	0.0				
TTTGGTTGTAACGTTC-1	0.0	0.000000	0.0 0.000000					
TTTGGTTTCAGCGTCG-1	0.0	1.767994	0.0 0.000000					
TTTGTTGGTACGTTCA-1	0.0	0.000000	0.0 2.698028					
TTTGTTGGTCCACTCT-1	0.0	0.000000	0.0 0.000000					
TITUTIUUTOOKOTOT T	0.0	0.00000	0.0 0.000000	0.0				
	AC125149.3	AC168977.1	CAAA01118383.1	Csprs Vamp7 \				
AAACCCAAGACAGCTG-1	0.0	0.0	0.744418	0.0 0.0				
AAACCCACAAAGACGC-1	0.0	0.0	0.00000	0.0 0.0				
AAACGAAAGCACTCGC-1	0.0	0.0	0.00000	0.0 0.0				
AAACGAACAAATGGTA-1	0.0	0.0	0.00000	0.0 0.0				
AAAGGATGTGTGTTTG-1	0.0	0.0	0.000000	0.0 0.0				
		•••		•••				
TTTGGTTCAGTGGCTC-1	0.0	0.0	0.000000	0.0 0.0				
TTTGGTTGTAACGTTC-1	0.0	0.0	0.000000	0.0 0.0				
TTTGGTTTCAGCGTCG-1	0.0	0.0	0.00000	0.0 0.0				
TTTGTTGGTACGTTCA-1	0.0	0.0	0.00000	0.0 0.0				
TTTGTTGGTCCACTCT-1	0.0	0.0	1.086885	0.0 0.0				
	GAAAAAA72220 4	10110000 1		77 \				
	CAAA01147332.1			cell \				
AAACCCAAGACAGCTG-1	0.0							
AAACCCACAAAGACGC-1	0.0		AAACCCACAAAGAC					
AAACGAAAGCACTCGC-1	0.0	0.000000	AAACGAAAGCACTC					
AAACGAACAAATGGTA-1	0.0							
AAAGGATGTGTTTTG-1	0.0	0.000000		TG-1				
TTTGGTTCAGTGGCTC-1	0.0		TTTGGTTCAGTGGC	TC-1				
TTTGGTTGTAACGTTC-1	0.0							
TTTGGTTTCAGCGTCG-1	0.0							
TTTGTTGGTACGTTCA-1	0.0							
TTTGTTGGTCCACTCT-1	0.0							
IIIGIIGGIOONOIOI-I	0.0	1.00000	TITGITGGTCCACT	01 1				
	feature_call	gender						
AAACCCAAGACAGCTG-1	CM0305	1						
AAACCCACAAAGACGC-1	CM0305	1						
AAACGAAAGCACTCGC-1	CM0305	1						
AAACGAACAAATGGTA-1	CM0305	1						

AAAGGATGTGTGTTTG-1	CM0305	1
•••		
TTTGGTTCAGTGGCTC-1	CM0306	0
TTTGGTTGTAACGTTC-1	CM0306	0
TTTGGTTTCAGCGTCG-1	CM0305	1
TTTGTTGGTACGTTCA-1	CM0305	1
TTTGTTGGTCCACTCT-1	CM0305	1

[1649 rows x 13027 columns]

[24]: male_exp_data

[24]:		Mrpli	15 L	ypla	.1 Tc	ea1	Atp6v	1h	Rb1c	c1 \		
	AAACCCAAGACAGCTG-1	0.		0000		000	-	.0	0.0000	00		
	AAACCCACAAAGACGC-1	0	.0 0.0	0000	0.000	000	0	.0	1.9214	57		
	AAACGAAAGCACTCGC-1	0	.0 0.0	0000	0.000	000	0	.0	0.0000	00		
	AAACGAACAAATGGTA-1	0	.0 0.0	0000	0.000	000	0	.0	0.0000	00		
	AAAGGATGTGTGTTTG-1	0	.0 0.0	0000	0.000	000	0	.0	2.5653	19		
		•••	•••		•••		•••					
	TTTGATCGTCCGTTTC-1	0	.0 0.4	1496	9 0.414	969	0	.0	0.0000	00		
	TTTGGAGCAGTGACCC-1	0	.0 0.0	0000	0.000	000	0	.0	0.0000	00		
	TTTGGTTTCAGCGTCG-1	0	.0 0.0	0000	0 1.767	994	0	.0	0.0000	00		
	TTTGTTGGTACGTTCA-1	0	.0 0.0	0000	0.000	000	0	.0	0.0000	00		
	TTTGTTGGTCCACTCT-1	0	.0 0.0	0000	0 1.086	885	0	.0	0.0000	00		
		47324	140D04R	ik	Pcmtd1	Gm	26901		Rrs1	26102030	22Rik	\
	AAACCCAAGACAGCTG-1		0	.0	0.744418	;	0.0	0.	000000		0.0	
	AAACCCACAAAGACGC-1		0	.0	0.000000)	0.0	0.	000000		0.0	
	AAACGAAAGCACTCGC-1		0		0.000000		0.0		000000		0.0	
	AAACGAACAAATGGTA-1		0	.0	0.000000)	0.0		000000		0.0	
	AAAGGATGTGTGTTTG-1		0	.0	0.000000)	0.0	0.	000000		0.0	
	•••					•••	•••			•••		
	TTTGATCGTCCGTTTC-1				0.000000		0.0		000000		0.0	
	TTTGGAGCAGTGACCC-1				0.000000		0.0		000000		0.0	
	TTTGGTTTCAGCGTCG-1				1.767994		0.0		000000		0.0	
	TTTGTTGGTACGTTCA-1				0.000000		0.0		698028		0.0	
	TTTGTTGGTCCACTCT-1		0	.0	0.000000)	0.0	0.	000000		0.0	
											_	
			nt-Cytb		125149.3		168977			118383.1	\	
	AAACCCAAGACAGCTG-1		. 274121		0.0			.0		0.744418		
	AAACCCACAAAGACGC-1		. 262269		0.0			.0		0.000000		
	AAACGAAAGCACTCGC-1		. 364882		0.0			.0		0.000000		
	AAACGAACAAATGGTA-1		. 365047		0.0			.0		0.000000		
	AAAGGATGTGTTTTG-1	5	.380296		0.0)	0	.0		0.000000		
						•••		_				
	TTTGATCGTCCGTTTC-1		.867098		0.0			.0		0.414969		
	TTTGGAGCAGTGACCC-1	4	.696054		0.0)	0	.0		0.000000		

```
TTTGGTTTCAGCGTCG-1
                        3.800684
                                          0.0
                                                       0.0
                                                                  0.00000
                                                       0.0
TTTGTTGGTACGTTCA-1
                        4.032499
                                          0.0
                                                                  0.00000
TTTGTTGGTCCACTCT-1
                        3.416918
                                          0.0
                                                       0.0
                                                                  1.086885
                                   CAAA01147332.1
                                                    AC149090.1
                     Csprs
                            Vamp7
AAACCCAAGACAGCTG-1
                       0.0
                              0.0
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                                                       0.00000
AAACCCACAAAGACGC-1
                       0.0
                              0.0
                                               0.0
                                                       1.921457
AAACGAAAGCACTCGC-1
                       0.0
                              0.0
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                                                       0.00000
AAACGAACAAATGGTA-1
                       0.0
                              0.0
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AAAGGATGTGTTTTG-1
                              0.0
                                                       0.00000
                       0.0
                                               0.0
TTTGATCGTCCGTTTC-1
                       0.0
                              0.0
                                               0.0
                                                       0.933333
TTTGGAGCAGTGACCC-1
                       0.0
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TTTGGTTTCAGCGTCG-1
                       0.0
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TTTGTTGGTACGTTCA-1
                       0.0
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                                                       0.000000
TTTGTTGGTCCACTCT-1
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                                               0.0
                                                       1.086885
                                    cell
                                          feature_call
AAACCCAAGACAGCTG-1
                     AAACCCAAGACAGCTG-1
                                                CM0305
AAACCCACAAAGACGC-1
                     AAACCCACAAAGACGC-1
                                                CM0305
AAACGAAAGCACTCGC-1
                     AAACGAAAGCACTCGC-1
                                                CM0305
AAACGAACAAATGGTA-1
                     AAACGAACAAATGGTA-1
                                                CM0305
AAAGGATGTGTTTTG-1
                     AAAGGATGTGTTTTG-1
                                                CM0305
TTTGATCGTCCGTTTC-1
                     TTTGATCGTCCGTTTC-1
                                                CM0305
TTTGGAGCAGTGACCC-1
                     TTTGGAGCAGTGACCC-1
                                                CM0305
TTTGGTTTCAGCGTCG-1
                     TTTGGTTTCAGCGTCG-1
                                                CM0305
TTTGTTGGTACGTTCA-1
                     TTTGTTGGTACGTTCA-1
                                                CM0305
TTTGTTGGTCCACTCT-1
                    TTTGTTGGTCCACTCT-1
                                                CM0305
```

[1427 rows x 13026 columns]

5 EXPDATA

[25]: realtest [25]: Xist Ddx3y Gm42418 Eif2s3y Rps27rt 0.000000 0.000000 5.790079 0.000000 AAACCCAAGACAGCTG-1 0.000000 0.000000 AAACCCACAAAGACGC-1 0.000000 6.888156 0.000000 0.000000 AAACGAAAGCACTCGC-1 0.000000 0.000000 5.788145 0.000000 0.000000 0.000000 AAACGAACAAATGGTA-1 0.000000 5.986185 0.000000 2.264689 AAAGGATGTGTTTTG-1 0.000000 0.000000 8.123958 0.000000 0.00000 TTTGGTTCAGTGGCTC-1 3.335779 0.000000 7.200241 0.000000 0.000000 TTTGGTTGTAACGTTC-1 3.969675 0.000000 6.050117 0.000000 0.000000 TTTGGTTTCAGCGTCG-1 0.000000 0.000000 6.049121 0.000000 0.000000

```
TTTGTTGGTACGTTCA-1 0.000000
                                    0.000000 7.608987
                                                        0.000000
                                                                  0.000000
      TTTGTTGGTCCACTCT-1
                          0.000000
                                    1.086885
                                             7.211483
                                                        1.086885
                                                                  0.00000
                          Rp19-ps6
                                    Kdm5d
                                              Uba52
                                                               Rpl36a-ps1
                                                        Rp135
                                                                           Uty \
      AAACCCAAGACAGCTG-1
                               0.0
                                      0.0 0.744418
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                                                                           0.0
                                                     1.690256
      AAACCCACAAAGACGC-1
                               0.0
                                      0.0
                                          2.538590
                                                     0.000000
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                                      0.0 0.000000
      AAACGAAAGCACTCGC-1
                                                     0.000000
      AAACGAACAAATGGTA-1
                               0.0
                                      0.0
                                           2.264689
                                                     0.000000
                                                                      0.0
                                                                           0.0
      AAAGGATGTGTTTTG-1
                               0.0
                                      0.0 0.000000
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                                                                      0.0 0.0
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      TTTGGTTCAGTGGCTC-1
                               0.0
                                      0.0
                                          1.859427
                                                     2.471493
      TTTGGTTGTAACGTTC-1
                               0.0
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      TTTGGTTTCAGCGTCG-1
                               0.0
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                                                     2.745812
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      TTTGTTGGTACGTTCA-1
                               0.0
                                      0.0
                                           2.070049
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                                                                      0.0 0.0
      TTTGTTGGTCCACTCT-1
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                                         1.595348
                                                     2.181557
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                                    2.167511
      AAACCCAAGACAGCTG-1
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      AAACCCACAAAGACGC-1 1.921457
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                                              2.538590
      AAACGAAAGCACTCGC-1
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      AAACGAACAAATGGTA-1
                          0.000000
                                    2.904508
                                              0.000000
      AAAGGATGTGTTTTG-1
                                    3.219260
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      TTTGGTTCAGTGGCTC-1
                          0.000000
                                    2.471493 0.000000
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      TTTGGTTGTAACGTTC-1
                          0.000000
                                              2.131073
      TTTGGTTTCAGCGTCG-1
                          0.000000
                                    0.000000
                                              2.745812
      TTTGTTGGTACGTTCA-1
                          0.000000
                                    3.356926
                                              3.573077
      TTTGTTGGTCCACTCT-1 1.086885
                                    3.118624 2.181557
      [1649 rows x 14 columns]
[26]: realtest_y
[26]: AAACCCAAGACAGCTG-1
                            1.0
      AAACCCACAAAGACGC-1
                            1.0
      AAACGAAAGCACTCGC-1
                            1.0
      AAACGAACAAATGGTA-1
                            1.0
      AAAGGATGTGTTTTG-1
                            1.0
      TTTGGTTCAGTGGCTC-1
                            0.0
      TTTGGTTGTAACGTTC-1
                            0.0
      TTTGGTTTCAGCGTCG-1
                            1.0
      TTTGTTGGTACGTTCA-1
                            1.0
      TTTGTTGGTCCACTCT-1
                            1.0
      Name: gender, Length: 1649, dtype: float64
```

```
[27]: full_train = rc.sample(frac=1, random_state=25)
      full_Y_train=full_train["y"]
      full_X_train=full_train.drop(["y"],axis=1)
      full_X_train = full_X_train[columns_to_extract]
[28]:
     full_X_train
[28]: n
                                         Eif2s3y
                                                   Rps27rt
                                                            Rp19-ps6
                                                                       Kdm5d
                                                                               Uba52
                  Xist
                        Ddx3y
                               Gm42418
      Cell_1678
                   0.0
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                                                       1.0
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      Cell_3705
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      Cell_5187
                  10.0
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      Cell_1921
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                                   19.0
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      Cell_1575
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      Cell_2934
                   7.0
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      Cell 2191
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      Cell 6618
                   7.0
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      Cell_318
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                   0.0
                                    8.0
                                                       6.0
      Cell_5252
                   1.0
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                                                  Lars2 Rps27
      Cell_1678
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      Cell_3705
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      Cell_5187
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      Cell_1921
                                              1.0
      Cell_1575
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      Cell_2934
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                                                             5.0
      Cell_2191
                    3.0
                                 1.0 0.0
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                                                     3.0
                                                             1.0
      Cell_6618
                    5.0
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                                              0.0
                                                     1.0
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      Cell 318
                    4.0
                                                     0.0
                                 0.0 0.0
                                              0.0
                                                             0.0
      Cell_5252
                    0.0
                                 1.0 0.0
                                              0.0
                                                     0.0
                                                             0.0
```

[6648 rows x 14 columns]

skip the following sev block straight to wtwt experimental

6 Performance with Scanorama

```
[30]: #!/usr/bin/env python3
"""

Train models on original training data, test on Scanorama-corrected data, save ROC/PRC plots, and export all curve data points for later use.
"""

import numpy as np
import pandas as pd
import scanorama
```

```
import pickle
import json
# First, let's check what's in your data
print("Checking original data...")
print(f"Type: {type(result)}")
print(f"Shape: {result.shape}")
# Check if data is already numeric
if isinstance(result, pd.DataFrame):
   print(f"Data types of first few columns: {result.dtypes[:5]}")
   print(f"Any non-numeric columns? {not all(result.dtypes.apply(lambda x: np.
 ⇒issubdtype(x, np.number)))}")
    # Check for actual NaN values
   print(f"NaN count per column (first 5): {result.isnull().sum()[:5].
 →tolist()}")
   print(f"Total NaN values: {result.isnull().sum().sum()}")
# Scanorama batch correction function
def scanorama_correction(result):
   Batch correction using Scanorama instead of ComBat.
    # Convert to numeric if needed
   if isinstance(result, pd.DataFrame):
        if all(result.dtypes.apply(lambda x: np.issubdtype(x, np.number))):
            print("Data is already numeric!")
            data_array = result.values
            gene_names = result.columns[:-1].tolist() # Get gene names
        else:
            print("Converting to numeric efficiently...")
            result numeric = result.apply(pd.to numeric, errors='coerce')
            data_array = result_numeric.values
            gene_names = result.columns[:-1].tolist()
    else:
        data_array = result.astype(float)
        gene_names = [f"Gene_{i}" for i in range(data_array.shape[1] - 1)]
    # Extract expression and gender
    expression_data = data_array[:, :-1]
   gender = data_array[:, -1]
   print(f"\nExpression shape: {expression_data.shape}")
   print(f"Gender values: {np.unique(gender[~np.isnan(gender)])}")
```

```
# Handle NaN values
  expr_nan_count = np.isnan(expression_data).sum()
  gender_nan_count = np.isnan(gender).sum()
  print(f"NaN in expression data: {expr_nan_count} out of {expression_data.
⇔size} values")
  print(f"NaN in gender: {gender_nan_count} out of {len(gender)} values")
  # Remove columns that are all NaN
  cols_all_nan = np.isnan(expression_data).all(axis=0)
  if cols_all_nan.any():
      print(f"Found {cols_all_nan.sum()} columns that are all NaN. Removing⊔
⇔them...")
      expression_data = expression_data[:, ~cols_all_nan]
      gene_names = [gene for gene, is_nan in zip(gene_names, cols_all_nan) if_
→not is_nan]
  # Replace remaining NaN with O (Scanorama doesn't handle NaN well)
  if np.isnan(expression_data).any():
      print("Replacing remaining NaN values with 0...")
      expression_data = np.nan_to_num(expression_data, nan=0)
  # Ensure gender is valid
  gender = np.nan_to_num(gender, nan=0).astype(int)
  print(f"\nFinal shapes - Expression: {expression_data.shape}, Gender: ___
→{len(gender)}")
  print(f"Gender distribution: {np.bincount(gender)}")
  # Split data by gender for Scanorama
  male_mask = gender == 0
  female_mask = gender == 1
  male_data = expression_data[male_mask]
  female_data = expression_data[female_mask]
  print(f"\nMale samples: {male_data.shape[0]}, Female samples: {female_data.
⇔shape [0] }")
  # Prepare data for Scanorama
  # Scanorama expects a list of datasets
  datasets = [male_data, female_data]
  genes_list = [gene_names, gene_names] # Same genes for both
  try:
      # Apply Scanorama correction
      print("\nApplying Scanorama batch correction...")
```

```
corrected_datasets, _ = scanorama.correct(datasets, genes_list,_
 →return_dimred=False)
        # Combine corrected datasets back together
       male_corrected = corrected_datasets[0]
        female_corrected = corrected_datasets[1]
        # Reconstruct full dataset in original order
        corrected_data = np.zeros_like(expression_data)
        corrected_data[male_mask] = male_corrected
        corrected_data[female_mask] = female_corrected
        print(f"Scanorama correction complete! Shape: {corrected data.shape}")
        return corrected_data
   except Exception as e:
       print(f"Scanorama failed: {e}")
       print("Falling back to simple batch correction...")
       return simple_batch_correction(expression_data, gender)
def simple_batch_correction(expression_data, gender):
   Simple batch correction as fallback.
   corrected = expression_data.copy()
   for col in range(expression_data.shape[1]):
        feature = expression_data[:, col]
        for g in np.unique(gender):
            mask = gender == g
            if mask.sum() > 0:
                corrected[mask, col] = feature[mask] - np.
 →nanmean(feature[mask]) + np.nanmean(feature)
   return corrected
# Apply Scanorama correction
realtest_scanorama = scanorama_correction(result)
if realtest_scanorama is not None:
   print(f"\nSuccess! Final shape: {realtest_scanorama.shape}")
# Now continue with the ML pipeline
from pathlib import Path
import matplotlib.pyplot as plt
from matplotlib import rcParams
```

```
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import (
   roc_curve,
    auc,
    precision_recall_curve,
    average_precision_score,
)
from sklearn.linear_model import LogisticRegression
from sklearn.svm import SVC
from sklearn.ensemble import RandomForestClassifier
from xgboost import XGBClassifier
    global config
rcParams.update({"font.family": "Arial", "font.size": 7})
OUTDIR = Path("/Users/haley/Desktop/send tooo/AAA final/scanorama")
OUTDIR.mkdir(exist_ok=True)
colors = ["#1f77b4", "#ff7f0e", "#2ca02c", "#d62728"]
models = \Gamma
    LogisticRegression(max_iter=1000, random_state=551),
    SVC(kernel="linear", probability=True, random_state=551),
    XGBClassifier(
        random state=551,
        use_label_encoder=False,
        eval metric="logloss",
        n_estimators=100,
        learning_rate=0.05,
        max_depth=10,
    ),
    RandomForestClassifier(max_depth=10, random_state=41),
]
    define columns to extract
columns_to_extract = [
    "Xist", "Ddx3y", "Gm42418", "Eif2s3y", "Rps27rt",
    "Rp19-ps6", "Kdm5d", "Uba52", "Rp135", "Rp136a-ps1",
    "Uty", "Wdr89", "Lars2", "Rps27"
]
    align features between train and test
print(f"\nOriginal shapes:")
print(f" Training: {full_X_train.shape}")
print(f" Test: {realtest_scanorama.shape}")
# Convert realtest_scanorama to DataFrame if it's an array
if isinstance(realtest_scanorama, np.ndarray):
```

```
# Need to get column names from somewhere - assuming they match the
 ⇔original data
    if 'result' in globals() and isinstance(result, pd.DataFrame):
        # Use column names from the original result DataFrame (minus the last_{f \sqcup}
 ⇒gender column)
        test_columns = result.columns[:-1]
        # Account for any removed all-NaN columns
        if realtest_scanorama.shape[1] < len(test_columns):</pre>
            # Some columns were removed, need to figure out which ones
            test_columns = test_columns[:realtest_scanorama.shape[1]]
        realtest_scanorama_df = pd.DataFrame(realtest_scanorama,_
 ⇔columns=test columns)
    else:
        print("ERROR: Cannot determine column names for test data")
        raise ValueError("Need column names to extract specific features")
else:
    realtest_scanorama_df = realtest_scanorama
# Extract only the columns that exist in the test data
available_columns = [col for col in columns_to_extract if col in_
 →realtest_scanorama_df.columns]
missing columns = [col for col in columns to extract if col not in,
 -realtest_scanorama_df.columns]
if missing_columns:
    print(f"WARNING: Missing columns in test data: {missing columns}")
print(f"Extracting {len(available_columns)} columns from test data")
realtest_scanorama_aligned = realtest_scanorama_df[available_columns]
print(f"\nAligned shapes:")
print(f" Training: {full_X_train.shape}")
print(f" Test: {realtest_scanorama_aligned.shape}")
# Verify they match
if realtest_scanorama_aligned.shape[1] != full_X_train.shape[1]:
    print(f"ERROR: Feature count mismatch! Train has {full_X_train.shape[1]},__
 →test has {realtest_scanorama_aligned.shape[1]}")
    print("Please check that the training data used the same columns")
    scale data
scaler = StandardScaler()
X train s = scaler.fit transform(full X train)
X_test_s = scaler.transform(realtest_scanorama_aligned)
    Initialize data storage for curves
```

```
curve_data = {
    'roc_curves': {},
    'prc_curves': {},
    'summary_metrics': {},
    'metadata': {
        'test_samples': len(realtest_y),
        'features_used': available_columns,
        'class_distribution': np.bincount(realtest_y).tolist(),
        'correction method': 'scanorama'
    }
}
   create plots
plt.figure(figsize=(12, 5))
roc_ax = plt.subplot(1, 2, 1)
pr_ax = plt.subplot(1, 2, 2)
   fit, predict, plot
for model, color in zip(models, colors):
    name = model.__class__._name__
    print(f"\nTraining {name}...")
    model.fit(X_train_s, full_Y_train)
    y_prob = model.predict_proba(X_test_s)[:, 1]
    fpr, tpr, roc_thresholds = roc_curve(realtest_y, y_prob)
    roc_auc = auc(fpr, tpr)
    roc_ax.plot(fpr, tpr, label=f'{name} (AUC={roc_auc:.3f})', color=color,_u
 →lw=2)
    # PR
    prec, rec, prc_thresholds = precision_recall_curve(realtest_y, y_prob)
    pr_auc = average_precision_score(realtest_y, y_prob)
    pr_ax.plot(rec, prec, label=f'{name} (AP={pr_auc:.3f})', color=color, lw=2)
    print(f" → AUROC: {roc_auc:.3f}, AUPRC: {pr_auc:.3f}")
        Store curve data
    # ROC curve data
    curve_data['roc_curves'][name] = {
        'fpr': fpr.tolist(),
        'tpr': tpr.tolist(),
        'thresholds': roc_thresholds.tolist(),
        'auc': float(roc auc)
    }
```

```
# PRC curve data
    curve_data['prc_curves'][name] = {
        'precision': prec.tolist(),
        'recall': rec.tolist(),
        'thresholds': prc_thresholds.tolist(),
        'average_precision': float(pr_auc)
    }
    # Summary metrics
    curve_data['summary_metrics'][name] = {
        'auroc': float(roc auc),
        'auprc': float(pr_auc),
        'color': color,
        'n_roc_points': len(fpr),
        'n_prc_points': len(prec)
    }
    format plots
roc_ax.plot([0, 1], [0, 1], 'k--', lw=1, alpha=0.5) # diagonal line
roc_ax.set_xlim([0, 1])
roc_ax.set_ylim([0, 1])
roc_ax.set_xlabel("False Positive Rate")
roc_ax.set_ylabel("True Positive Rate")
roc ax.set title("ROC Curves (Scanorama-corrected test set)")
roc_ax.legend(loc="lower right")
pr_ax.set_xlim([0, 1])
pr_ax.set_ylim([0, 1])
pr_ax.set_xlabel("Recall")
pr_ax.set_ylabel("Precision")
pr_ax.set_title("Precision-Recall Curves (Scanorama-corrected test set)")
pr_ax.legend(loc="lower left")
# Remove top and right spines
for ax in (roc_ax, pr_ax):
    ax.spines["top"].set_visible(False)
    ax.spines["right"].set_visible(False)
plt.tight_layout()
    save plots
plt.savefig(OUTDIR / "roc_prc_scanorama.png", dpi=300, bbox_inches='tight')
plt.savefig(OUTDIR / "roc_prc_scanorama.pdf", dpi=300, bbox_inches='tight')
plt.show()
    Save curve data in multiple formats
print(f"\n Saving curve data...")
```

```
# 1. Save as JSON (human-readable, cross-platform)
json_path = OUTDIR / "curve_data_scanorama.json"
with open(json_path, 'w') as f:
    json.dump(curve_data, f, indent=2)
            JSON: {json_path}")
print(f"
# 2. Save as pickle (Python-specific but preserves exact numpy arrays)
pickle_path = OUTDIR / "curve_data_scanorama.pkl"
with open(pickle_path, 'wb') as f:
    pickle.dump(curve data, f)
print(f"
            Pickle: {pickle_path}")
# 3. Save individual CSV files for each curve type
csv_dir = OUTDIR / "csv_curves"
csv_dir.mkdir(exist_ok=True)
# ROC curves CSV
roc_csv_data = []
for model_name, roc_data in curve_data['roc_curves'].items():
    for i, (fpr, tpr, threshold) in enumerate(zip(roc_data['fpr'],__
 →roc_data['tpr'], roc_data['thresholds'])):
        roc_csv_data.append({
            'model': model_name,
            'point_index': i,
            'fpr': fpr,
            'tpr': tpr,
            'threshold': threshold,
            'auc': roc_data['auc']
        })
roc_df = pd.DataFrame(roc_csv_data)
roc_csv_path = csv_dir / "roc_curves_scanorama.csv"
roc df.to csv(roc csv path, index=False)
print(f" ROC CSV: {roc_csv_path}")
# PRC curves CSV
prc csv data = []
for model_name, prc_data in curve_data['prc_curves'].items():
    for i, (prec, rec, threshold) in enumerate(zip(prc_data['precision'],__

¬prc_data['recall'], prc_data['thresholds'])):

        prc_csv_data.append({
            'model': model_name,
            'point_index': i,
            'precision': prec,
            'recall': rec,
            'threshold': threshold,
```

```
'average_precision': prc_data['average_precision']
       })
prc_df = pd.DataFrame(prc_csv_data)
prc_csv_path = csv_dir / "prc_curves_scanorama.csv"
prc_df.to_csv(prc_csv_path, index=False)
print(f" PRC CSV: {prc_csv_path}")
# 4. Save summary metrics as CSV
summary_df = pd.DataFrame(curve_data['summary_metrics']).T
summary csv path = OUTDIR / "summary metrics scanorama.csv"
summary_df.to_csv(summary_csv_path)
print(f"
            Summary CSV: {summary csv path}")
# 5. Create a simple loading script template
loading_script = '''#!/usr/bin/env python3
Example script to load and recreate the ROC/PRC curves from saved data
import json
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
from pathlib import Path
# Load the curve data
with open('curve_data_scanorama.json', 'r') as f:
    curve data = json.load(f)
# Or load from CSV
roc_df = pd.read_csv('csv_curves/roc_curves_scanorama.csv')
prc_df = pd.read_csv('csv_curves/prc_curves_scanorama.csv')
# Recreate the plots
fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(12, 5))
# Plot ROC curves from JSON data
for model_name, roc_data in curve_data['roc_curves'].items():
   ax1.plot(roc data['fpr'], roc data['tpr'],
             label=f"{model_name} (AUC={roc_data['auc']:.3f})", lw=2)
ax1.plot([0, 1], [0, 1], 'k--', alpha=0.5)
ax1.set_xlabel('False Positive Rate')
ax1.set_ylabel('True Positive Rate')
ax1.set_title('ROC Curves')
ax1.legend()
```

```
# Plot PRC curves from JSON data
for model_name, prc_data in curve_data['prc_curves'].items():
    ax2.plot(prc_data['recall'], prc_data['precision'],
             label=f"{model_name} (AP={prc_data['average_precision']:.3f})",__
 \hookrightarrow 1w=2)
ax2.set xlabel('Recall')
ax2.set_ylabel('Precision')
ax2.set_title('Precision-Recall Curves')
ax2.legend()
plt.tight_layout()
plt.show()
# Alternative: Plot from CSV data
# for model in roc_df['model'].unique():
      model_data = roc_df[roc_df['model'] == model]
      ax1.plot(model_data['fpr'], model_data['tpr'], label=model)
111
script_path = OUTDIR / "load_and_plot_curves.py"
with open(script_path, 'w') as f:
    f.write(loading_script)
print(f"
            Loading script: {script_path}")
print(f"\n All files saved to: {OUTDIR}")
print(f"\nFiles created:")
            Plots:")
print(f"
print(f"
                roc_prc_scanorama.png")
                roc prc scanorama.pdf")
print(f"
print(f"
            Curve Data:")
print(f"
                curve data scanorama.json (complete data)")
                curve_data_scanorama.pkl (Python binary)")
print(f"
print(f"
                summary metrics scanorama.csv (summary table)")
                csv_curves/roc_curves_scanorama.csv (ROC points)")
print(f"
                csv_curves/prc_curves_scanorama.csv (PRC points)")
print(f"
print(f"
            Script:")
                load_and_plot_curves.py (example loading code)")
print(f"
print(f"\n Data structure:")
print(f" • ROC: FPR, TPR, thresholds, AUC for each model")
print(f"
           • PRC: Precision, Recall, thresholds, AP for each model")
           • Metadata: Features used, sample counts, etc.")
print(f"
print(f"\n You can now recreate these plots exactly using the saved curve⊔
 ⇔points!")
  Print curve data summary
```

```
print(f"\n Curve Data Summary:")
for model_name in curve_data['summary_metrics'].keys():
    roc_points = curve data['summary_metrics'][model_name]['n_roc_points']
    prc_points = curve data['summary_metrics'][model_name]['n_prc_points']
    auroc = curve_data['summary_metrics'][model_name]['auroc']
    auprc = curve_data['summary_metrics'][model_name]['auprc']
               {model_name:20s}: {roc_points:3d} ROC points, {prc_points:3d}_
    print(f"
  ⇔PRC points, "
          f"AUROC={auroc:.3f}, AUPRC={auprc:.3f}")
Checking original data...
Type: <class 'pandas.core.frame.DataFrame'>
Shape: (1649, 13027)
Data types of first few columns: Mrpl15
                                            float32
Lypla1
           float32
Tcea1
           float32
Atp6v1h
           float32
Rb1cc1
           float32
dtype: object
Any non-numeric columns? True
NaN count per column (first 5): [0, 0, 0, 0, 0]
Total NaN values: 0
Converting to numeric efficiently...
Expression shape: (1649, 13026)
Gender values: [0. 1.]
NaN in expression data: 3298 out of 21479874 values
NaN in gender: 0 out of 1649 values
Found 2 columns that are all NaN. Removing them ...
Final shapes - Expression: (1649, 13024), Gender: 1649
Gender distribution: [ 222 1427]
Male samples: 222, Female samples: 1427
Applying Scanorama batch correction...
Found 13024 genes among all datasets
[[0.
            0.9954955]
 [0.
            0.
                     ]]
Processing datasets (0, 1)
Scanorama failed: setting an array element with a sequence.
Falling back to simple batch correction...
Success! Final shape: (1649, 13024)
Original shapes:
 Training: (6648, 14)
 Test: (1649, 13024)
```

Extracting 14 columns from test data

Aligned shapes:

Training: (6648, 14)
Test: (1649, 14)

Training LogisticRegression...

→ AUROC: 0.533, AUPRC: 0.850

Training SVC...

→ AUROC: 0.526, AUPRC: 0.837

Training XGBClassifier...

→ AUROC: 0.786, AUPRC: 0.949

Training RandomForestClassifier...

/Users/haley/Library/Python/3.9/lib/python/site-packages/xgboost/core.py:158:

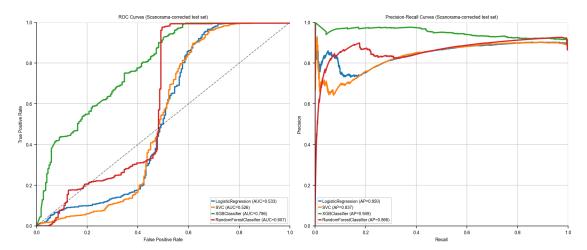
UserWarning: [18:04:10] WARNING:

/Users/runner/work/xgboost/xgboost/src/learner.cc:740:

Parameters: { "use_label_encoder" } are not used.

warnings.warn(smsg, UserWarning)

→ AUROC: 0.607, AUPRC: 0.866



Saving curve data...

JSON:

/Users/haley/Desktop/send_tooo/AAA_final/scanorama/curve_data_scanorama.json Pickle:

/Users/haley/Desktop/send_tooo/AAA_final/scanorama/curve_data_scanorama.pkl
ROC CSV: /Users/haley/Desktop/send_tooo/AAA_final/scanorama/csv_curves/roc_

```
curves_scanorama.csv
    PRC CSV: /Users/haley/Desktop/send_tooo/AAA_final/scanorama/csv_curves/prc_
curves_scanorama.csv
    Summary CSV:
/Users/haley/Desktop/send tooo/AAA final/scanorama/summary metrics scanorama.csv
    Loading script:
/Users/haley/Desktop/send tooo/AAA final/scanorama/load and plot curves.py
 All files saved to: /Users/haley/Desktop/send tooo/AAA final/scanorama
Files created:
    Plots:
        roc_prc_scanorama.png
        roc_prc_scanorama.pdf
    Curve Data:
        curve_data_scanorama.json (complete data)
        curve_data_scanorama.pkl (Python binary)
        summary_metrics_scanorama.csv (summary table)
        csv_curves/roc_curves_scanorama.csv (ROC points)
        csv_curves/prc_curves_scanorama.csv (PRC points)
        load_and_plot_curves.py (example loading code)
 Data structure:
   • ROC: FPR, TPR, thresholds, AUC for each model
   • PRC: Precision, Recall, thresholds, AP for each model
   • Metadata: Features used, sample counts, etc.
 You can now recreate these plots exactly using the saved curve points!
 Curve Data Summary:
  LogisticRegression : 275 ROC points, 1650 PRC points, AUROC=0.533,
AUPRC=0.850
  SVC
                       : 295 ROC points, 1644 PRC points, AUROC=0.526,
AUPRC=0.837
                       : 344 ROC points, 1136 PRC points, AUROC=0.786,
  XGBClassifier
  RandomForestClassifier: 352 ROC points, 1439 PRC points, AUROC=0.607,
AUPRC=0.866
```

7 figure 3-a cross validation

```
[20]: reboot_training_data = ffrc.sample(frac=0.6, random_state=25)
reboot_testing_data = ffrc.drop(reboot_training_data.index)

rbY_train=reboot_training_data["y"]
rbX_train=reboot_training_data.drop(["y"],axis=1)
```

```
rbY_test=reboot_testing_data["y"]
rbX_test=reboot_testing_data.drop(["y"],axis=1)
```

```
[21]: from sklearn.model_selection import cross_val_score, StratifiedKFold,
       ⇔cross_validate
      from sklearn.pipeline import Pipeline
      import numpy as np
      import matplotlib.pyplot as plt
      from sklearn.linear_model import LogisticRegression
      from sklearn.svm import SVC
      from sklearn.ensemble import RandomForestClassifier
      from xgboost import XGBClassifier
      from sklearn.preprocessing import StandardScaler
      # # Assuming rbX_train, rbY_train, rbX_test, rbY_test are your data
      # rbX = np.concatenate((rbX train, rbX test), axis=0)
      # rbY = np.concatenate((rbY_train, rbY_test), axis=0)
      # Your models
      models = [
          ('LR', Pipeline([('scaler', StandardScaler()), ('model', I
       →LogisticRegression(max_iter=1000, random_state=551))])),
          ('SVM', Pipeline([('scaler', StandardScaler()), ('model', ___
       →SVC(kernel='linear', probability=True, random_state=551))])),
          ('XGB', Pipeline([('scaler', StandardScaler()), ('model', XGBClassifier( __
       ⇔eval metric='logloss', n estimators=100, learning rate=0.05,
       →max_depth=10,random_state=551))])),
          ('RF', Pipeline([('scaler', StandardScaler()), ('model', L
       →RandomForestClassifier(max_depth=10, random_state=41))]))
      1
      scores = {name: {'accuracy': [], 'f1': []} for name, _ in models}
      # 5-Fold Cross-validation
      cv = StratifiedKFold(n_splits=5, shuffle=True, random_state=41)
      for name, model in models:
          # Change 'f1 macro' to 'f1 weighted' for multi-class scenario with
       ⇔consideration for class imbalance
          cv_results = cross_validate(model, full_X_train, full_Y_train, cv=cv,_

¬scoring=['accuracy', 'f1_weighted'])
          scores[name]['accuracy'] = cv_results['test_accuracy']
          scores[name]['f1'] = cv_results['test_f1_weighted']
```

```
# Print results
     for name in scores.keys():
         mean_accuracy = np.mean(scores[name]['accuracy'])
         std_accuracy = np.std(scores[name]['accuracy'])
         mean_f1 = np.mean(scores[name]['f1'])
         std_f1 = np.std(scores[name]['f1'])
         print(f"Model: {name}")
         print(f" Accuracy: Mean = {mean_accuracy:.2f}, Std = {std_accuracy:.2f}")
         print(f" F1 Score: Mean = {mean_f1:.2f}, Std = {std_f1:.2f}")
     Model: LR
       Accuracy: Mean = 0.90, Std = 0.01
       F1 Score: Mean = 0.90, Std = 0.01
     Model: SVM
       Accuracy: Mean = 0.90, Std = 0.01
       F1 Score: Mean = 0.90, Std = 0.01
     Model: XGB
       Accuracy: Mean = 0.90, Std = 0.01
       F1 Score: Mean = 0.90, Std = 0.01
     Model: RF
       Accuracy: Mean = 0.90, Std = 0.01
       F1 Score: Mean = 0.91, Std = 0.01
[23]: from sklearn.model_selection import cross_val_score, StratifiedKFold,
      ⇔cross_validate
     from sklearn.pipeline import Pipeline
     import numpy as np
     import matplotlib.pyplot as plt
     from sklearn.linear_model import LogisticRegression
     from sklearn.svm import SVC
     from sklearn.ensemble import RandomForestClassifier
     from xgboost import XGBClassifier
     from sklearn.preprocessing import StandardScaler
      # Assuming rbX train, rbY train, rbX test, rbY test are your data
      # Your models
     models = [
          ('LR', Pipeline([('scaler', StandardScaler()), ('model', L
       →LogisticRegression(max_iter=1000, random_state=551))])),
          ('SVM', Pipeline([('scaler', StandardScaler()), ('model', ___
       →SVC(kernel='linear', probability=True, random_state=551))])),
          ('XGB', Pipeline([('scaler', StandardScaler()), ('model', __
       →XGBClassifier(use_label_encoder=False, eval_metric='logloss', __
       n_estimators=100, learning_rate=0.05, max_depth=10,random_state=551))])),
```

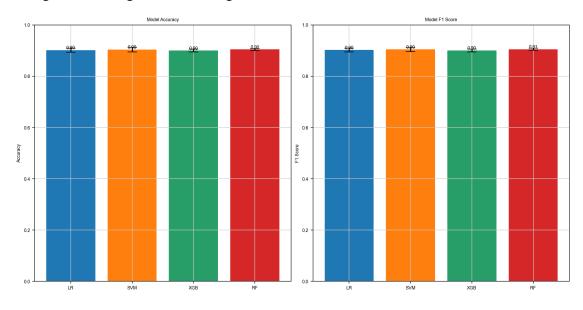
```
('RF', Pipeline([('scaler', StandardScaler()), ('model', ___
 →RandomForestClassifier(max_depth=10, random_state=41))]))
1
scores = {name: {'accuracy': [], 'f1': []} for name, _ in models}
# 5-Fold Cross-validation
cv = StratifiedKFold(n_splits=5, shuffle=True, random_state=551)
for name, model in models:
    # Change 'f1_macro' to 'f1_weighted' for multi-class scenario with
 ⇔consideration for class imbalance
    cv_results = cross_validate(model, full_X_train, full_Y_train, cv=cv,_
 ⇔scoring=['accuracy', 'f1_weighted'])
    scores[name]['accuracy'] = cv_results['test_accuracy']
    scores[name]['f1'] = cv_results['test_f1_weighted']
# Plotting
fig, ax = plt.subplots(1, 2, figsize=(12, 6))
# Accuracy Plot
for i, name in enumerate(scores.keys()):
   mean_accuracy = np.mean(scores[name]['accuracy'])
   std accuracy = np.std(scores[name]['accuracy'])
   bar = ax[0].bar(name, mean_accuracy, yerr=std_accuracy, capsize=7)
   ax[0].text(bar[0].get_x() + bar[0].get_width() / 2, bar[0].get_height(),

→f'{mean_accuracy:.2f}', ha='center', va='bottom', fontsize=7)
ax[0].set_title('Model Accuracy', fontsize=7)
ax[0].set_ylabel('Accuracy', fontsize=7)
ax[0].tick_params(axis='both', which='major', labelsize=7)
ax[0].set_ylim([0, 1])
# F1 Score Plot
for i, name in enumerate(scores.keys()):
   mean f1 = np.mean(scores[name]['f1'])
   std_f1 = np.std(scores[name]['f1'])
   bar = ax[1].bar(name, mean_f1, yerr=std_f1, capsize=7)
   ax[1].text(bar[0].get_x() + bar[0].get_width() / 2, bar[0].get_height(),

→f'{mean_f1:.2f}', ha='center', va='bottom', fontsize=7)
ax[1].set_title('Model F1 Score', fontsize=7)
ax[1].set ylabel('F1 Score', fontsize=7)
ax[1].tick_params(axis='both', which='major', labelsize=7)
ax[1].set_ylim([0, 1])
plt.tight_layout()
plt.show()
```

```
/Users/haley/Library/Python/3.9/lib/python/site-packages/xgboost/core.py:158:
UserWarning: [22:28:27] WARNING:
/Users/runner/work/xgboost/xgboost/src/learner.cc:740:
Parameters: { "use_label_encoder" } are not used.
  warnings.warn(smsg, UserWarning)
/Users/haley/Library/Python/3.9/lib/python/site-packages/xgboost/core.py:158:
UserWarning: [22:28:27] WARNING:
/Users/runner/work/xgboost/xgboost/src/learner.cc:740:
Parameters: { "use_label_encoder" } are not used.
 warnings.warn(smsg, UserWarning)
/Users/haley/Library/Python/3.9/lib/python/site-packages/xgboost/core.py:158:
UserWarning: [22:28:27] WARNING:
/Users/runner/work/xgboost/xgboost/src/learner.cc:740:
Parameters: { "use_label_encoder" } are not used.
 warnings.warn(smsg, UserWarning)
/Users/haley/Library/Python/3.9/lib/python/site-packages/xgboost/core.py:158:
UserWarning: [22:28:27] WARNING:
/Users/runner/work/xgboost/xgboost/src/learner.cc:740:
Parameters: { "use_label_encoder" } are not used.
  warnings.warn(smsg, UserWarning)
/Users/haley/Library/Python/3.9/lib/python/site-packages/xgboost/core.py:158:
UserWarning: [22:28:27] WARNING:
/Users/runner/work/xgboost/xgboost/src/learner.cc:740:
Parameters: { "use_label_encoder" } are not used.
```

warnings.warn(smsg, UserWarning)



8 figure 3-cd cross validation

```
[24]: import numpy as np
      import pandas as pd
      import matplotlib.pyplot as plt
      from matplotlib import rcParams
      from sklearn.linear_model import LogisticRegression
      from sklearn.svm import SVC
      from sklearn.ensemble import RandomForestClassifier
      from xgboost import XGBClassifier
      from sklearn.metrics import roc_auc_score, average_precision_score, roc_curve, u
       →precision_recall_curve
      from sklearn.preprocessing import StandardScaler
      from sklearn.model_selection import train_test_split
      from sklearn.pipeline import Pipeline
      # Set global font properties
      rcParams['font.family'] = 'Arial'
      rcParams['font.size'] = 7
      rcParams['font.weight'] = 'normal'
      # Split data into training and testing sets
      X_train, X_test, y_train, y_test = train_test_split(full_X_train, full_Y_train, __
       →test_size=0.2, random_state=551, stratify=full_Y_train)
      # Your models
      models = \Gamma
          ('LR', Pipeline([('scaler', StandardScaler()), ('model', )
       →LogisticRegression(max_iter=1000, random_state=551))])),
          ('SVM', Pipeline([('scaler', StandardScaler()), ('model', ___
       SVC(kernel='linear', probability=True, random_state=551))])),
          ('XGB', Pipeline([('scaler', StandardScaler()), ('model', __
       →XGBClassifier(use_label_encoder=False, eval_metric='logloss', __
       on_estimators=100, learning_rate=0.05, max_depth=10,random_state=551))])),
          ('RF', Pipeline([('scaler', StandardScaler()), ('model', L
       →RandomForestClassifier(max_depth=10, random_state=41))]))
      1
      # DataFrames to store all ROC and PRC data
      roc_data_list = []
      prc_data_list = []
```

```
# Plot settings
plt.figure(figsize=(12, 6))
roc_ax = plt.subplot(1, 2, 1)
prc_ax = plt.subplot(1, 2, 2)
for name, model in models:
   # Fit the model on the full training data
   model.fit(X_train, y_train)
   # Predict probabilities on the test set
   y_prob = model.predict_proba(X_test)[:, 1]
    # ROC Curve
   fpr, tpr, _ = roc_curve(y_test, y_prob)
   roc_auc = roc_auc_score(y_test, y_prob)
   roc_ax.plot(fpr, tpr, label=f'{name} (AUC = {roc_auc:.2f})')
   # Save ROC data
   roc_data_list.append(pd.DataFrame({'Model': name, 'fpr': fpr, 'tpr': tpr}))
   # Precision-Recall Curve
   precision, recall, _ = precision_recall_curve(y_test, y_prob)
   pr_auc = average_precision_score(y_test, y_prob)
   prc_ax.plot(recall, precision, label=f'{name} (AUPRC = {pr_auc:.2f})')
    # Save PRC data
   prc_data_list.append(pd.DataFrame({'Model': name, 'precision': precision,__

¬'recall': recall}))
# ROC Axis labels
roc_ax.set_title('ROC Curve Comparison')
roc_ax.set_xlabel('False Positive Rate')
roc_ax.set_ylabel('True Positive Rate')
roc_ax.legend(loc="lower right")
# PRC Axis labels
prc_ax.set_title('Precision-Recall Curve Comparison')
prc_ax.set_xlabel('Recall')
prc_ax.set_ylabel('Precision')
prc_ax.legend(loc="lower left")
plt.tight_layout()
plt.show()
# Combine all ROC and PRC data and save to CSV files
roc_data = pd.concat(roc_data_list, ignore_index=True)
prc_data = pd.concat(prc_data_list, ignore_index=True)
```

```
roc_data.to_csv('auroc_data.csv', index=False)
prc_data.to_csv('prc_data.csv', index=False)
```

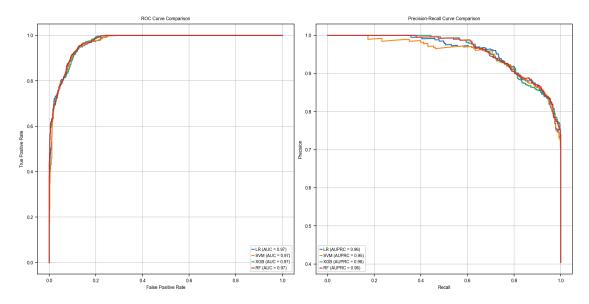
/Users/haley/Library/Python/3.9/lib/python/site-packages/xgboost/core.py:158:

UserWarning: [22:37:07] WARNING:

/Users/runner/work/xgboost/xgboost/src/learner.cc:740:

Parameters: { "use_label_encoder" } are not used.

warnings.warn(smsg, UserWarning)



```
[98]: import numpy as np
      import pandas as pd
      import matplotlib.pyplot as plt
      from matplotlib import rcParams
      from sklearn.linear_model import LogisticRegression
      from sklearn.svm import SVC
      from sklearn.ensemble import RandomForestClassifier
      from xgboost import XGBClassifier
      from sklearn.metrics import roc_auc_score, average_precision_score, roc_curve, u
       →precision_recall_curve
      from sklearn.preprocessing import StandardScaler
      from sklearn.model_selection import train_test_split
      from sklearn.pipeline import Pipeline
      # Set global font properties
      rcParams['font.family'] = 'Arial'
      rcParams['font.size'] = 7
```

```
rcParams['font.weight'] = 'normal'
# Split data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(full_X_train, full_Y_train,_
 stest_size=0.2, random_state=551, stratify=full_Y_train)
# Your models
models = \Gamma
    ('LR', Pipeline([('scaler', StandardScaler()), ('model', ___
 ('SVM', Pipeline([('scaler', StandardScaler()), ('model', ___
 →SVC(kernel='linear', probability=True, random_state=551))])),
    ('XGB', Pipeline([('scaler', StandardScaler()), ('model', ___
 ⇔XGBClassifier(use_label_encoder=False, eval_metric='logloss',⊔
 on_estimators=100, learning_rate=0.05, max_depth=10,random_state=551))])),
    ('RF', Pipeline([('scaler', StandardScaler()), ('model', ___
→RandomForestClassifier(max_depth=10, random_state=41))]))
1
# DataFrames to store all ROC and PRC data
roc_data_list = []
prc_data_list = []
# Dictionary to store metrics for printing
metrics_dict = {}
# Plot settings
plt.figure(figsize=(12, 6))
roc_ax = plt.subplot(1, 2, 1)
prc_ax = plt.subplot(1, 2, 2)
for name, model in models:
    # Fit the model on the full training data
   model.fit(X_train, y_train)
   # Predict probabilities on the test set
   y_prob = model.predict_proba(X_test)[:, 1]
   # ROC Curve
   fpr, tpr, _ = roc_curve(y_test, y_prob)
   roc_auc = roc_auc_score(y_test, y_prob)
   roc_ax.plot(fpr, tpr, label=f'{name} (AUC = {roc_auc})')
   # Save ROC data
   roc_data_list.append(pd.DataFrame({'Model': name, 'fpr': fpr, 'tpr': tpr}))
    # Precision-Recall Curve
```

```
precision, recall, _ = precision_recall_curve(y_test, y_prob)
    pr_auc = average_precision_score(y_test, y_prob)
    prc_ax.plot(recall, precision, label=f'{name} (AUPRC = {pr_auc})')
    # Save PRC data
    prc_data_list.append(pd.DataFrame({'Model': name, 'precision': precision,_

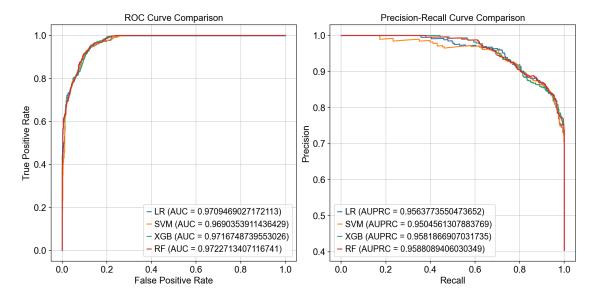
¬'recall': recall}))
    # Store metrics for printing
    metrics_dict[name] = {'AUROC': roc_auc, 'AUPRC': pr_auc}
    # Print metrics with full precision
    print(f'{name}: AUROC = {roc_auc}, AUPRC = {pr_auc}')
# ROC Axis labels
roc_ax.set_title('ROC Curve Comparison')
roc_ax.set_xlabel('False Positive Rate')
roc_ax.set_ylabel('True Positive Rate')
roc_ax.legend(loc="lower right")
# PRC Axis labels
prc_ax.set_title('Precision-Recall Curve Comparison')
prc_ax.set_xlabel('Recall')
prc_ax.set_ylabel('Precision')
prc_ax.legend(loc="lower left")
plt.tight_layout()
plt.show()
# Combine all ROC and PRC data and save to CSV files
roc_data = pd.concat(roc_data_list, ignore_index=True)
prc_data = pd.concat(prc_data_list, ignore_index=True)
roc data.to csv('auroc data.csv', index=False)
prc_data.to_csv('prc_data.csv', index=False)
# Save metrics to CSV with full precision
metrics_df = pd.DataFrame.from_dict(metrics_dict, orient='index').reset_index()
metrics_df.columns = ['Model', 'AUROC', 'AUPRC']
metrics_df.to_csv('model_metrics_full_precision.csv', index=False)
print("\nMetrics Summary:")
print(metrics_df)
```

LR: AUROC = 0.9709469027172113, AUPRC = 0.9563773550473652 SVM: AUROC = 0.9690353911436429, AUPRC = 0.9504561307883769 XGB: AUROC = 0.9716748739553026, AUPRC = 0.9581866907031735

```
/Users/haley/Library/Python/3.9/lib/python/site-packages/xgboost/core.py:158:
UserWarning: [08:54:32] WARNING:
/Users/runner/work/xgboost/xgboost/src/learner.cc:740:
Parameters: { "use_label_encoder" } are not used.
```

warnings.warn(smsg, UserWarning)

RF: AUROC = 0.9722713407116741, AUPRC = 0.9588089406030349



```
Model AUROC AUPRC

0 LR 0.971 0.956

1 SVM 0.969 0.950

2 XGB 0.972 0.958

3 RF 0.972 0.959

[25]: import numpy as np
```

Metrics Summary:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from matplotlib import rcParams
from sklearn.linear_model import LogisticRegression
from sklearn.svm import SVC
from sklearn.ensemble import RandomForestClassifier
from xgboost import XGBClassifier
from sklearn.metrics import roc_auc_score, average_precision_score, roc_curve,
precision_recall_curve
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
```

```
from sklearn.pipeline import Pipeline
# Set global font properties
rcParams['font.family'] = 'Arial'
rcParams['font.size'] = 7
rcParams['font.weight'] = 'normal'
# Split data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(full_X_train, full_Y_train, __
 →test_size=0.2, random_state=551, stratify= full_Y_train)
# Your models
models = \Gamma
    ('LR', Pipeline([('scaler', StandardScaler()), ('model', ___
 →LogisticRegression(max_iter=1000, random_state=551))])),
    ('SVM', Pipeline([('scaler', StandardScaler()), ('model', ___
 SVC(kernel='linear', probability=True, random_state=551))])),
    ('XGB', Pipeline([('scaler', StandardScaler()), ('model', __
 ⇔XGBClassifier(use_label_encoder=False, eval_metric='logloss',⊔
 on_estimators=100, learning_rate=0.05, max_depth=10,random_state=551))])),
    ('RF', Pipeline([('scaler', StandardScaler()), ('model', ...
→RandomForestClassifier(max_depth=10, random_state=41))]))
]
# Plot settings
plt.figure(figsize=(12, 6))
roc_ax = plt.subplot(1, 2, 1)
prc_ax = plt.subplot(1, 2, 2)
for name, model in models:
    # Fit the model on the full training data
    model.fit(X_train, y_train)
    # Predict probabilities on the test set
    y_prob = model.predict_proba(X_test)[:, 1]
    # ROC Curve
    fpr, tpr, _ = roc_curve(y_test, y_prob)
    roc_auc = roc_auc_score(y_test, y_prob)
    roc_ax.plot(fpr, tpr, label=f'{name} (AUC = {roc_auc:.2f})')
    # Precision-Recall Curve
    precision, recall, _ = precision_recall_curve(y_test, y_prob)
    pr_auc = average_precision_score(y_test, y_prob)
    prc_ax.plot(recall, precision, label=f'{name} (AUPRC = {pr_auc:.2f})')
```

```
# ROC Axis labels
roc_ax.set_title('ROC Curve Comparison')
roc_ax.set_xlabel('False Positive Rate')
roc_ax.set_ylabel('True Positive Rate')
roc_ax.legend(loc="lower right")

# PRC Axis labels
prc_ax.set_title('Precision-Recall Curve Comparison')
prc_ax.set_xlabel('Recall')
prc_ax.set_ylabel('Precision')
prc_ax.legend(loc="lower left")

plt.tight_layout()
plt.show()
```

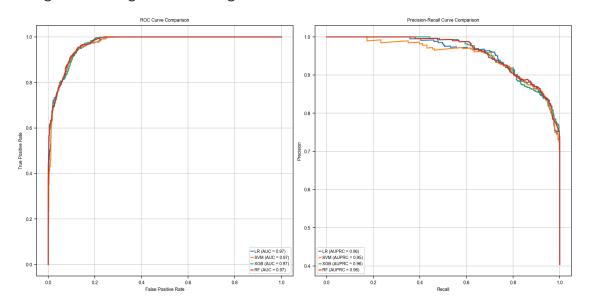
/Users/haley/Library/Python/3.9/lib/python/site-packages/xgboost/core.py:158:

UserWarning: [22:37:55] WARNING:

/Users/runner/work/xgboost/xgboost/src/learner.cc:740:

Parameters: { "use_label_encoder" } are not used.

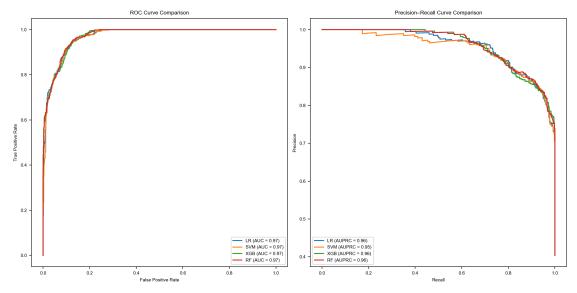
warnings.warn(smsg, UserWarning)



```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from matplotlib import rcParams
from pathlib import Path
from sklearn.linear_model import LogisticRegression
from sklearn.svm import SVC
from sklearn.ensemble import RandomForestClassifier
from xgboost import XGBClassifier
from sklearn.metrics import (
   roc_auc_score, average_precision_score,
   roc_curve, precision_recall_curve
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
from sklearn.pipeline import Pipeline
    global matplotlib settings
rcParams["font.family"] = "Arial"
rcParams["font.size"] = 7
rcParams["font.weight"] = "normal"
    data split
X_train, X_test, y_train, y_test = train_test_split(
   full_X_train, full_Y_train,
   test_size = 0.20,
   random_state= 551,
   stratify = full_Y_train
)
   model definitions
models = [
    ("LR",
    Pipeline([
         ("scaler", StandardScaler()),
         ("model", LogisticRegression(max_iter=1_000, random_state=551))
    ])),
    ("SVM",
    Pipeline([
         ("scaler", StandardScaler()),
         ("model", SVC(kernel="linear", probability=True, random_state=551))
    ])),
    ("XGB",
    Pipeline([
         ("scaler", StandardScaler()),
```

```
("model", XGBClassifier(
                       use_label_encoder=False,
                       eval_metric="logloss",
                       n_estimators=100,
                       learning_rate=0.05,
                       max_depth=10,
                       random_state=551))
    ])),
    ("RF",
    Pipeline([
         ("scaler", StandardScaler()),
         ("model", RandomForestClassifier(max_depth=10, random_state=41))
    ]))
]
# containers for saving curve points
roc_points = [] # list of DataFrames
pr_points = []
# plotting setup
fig, (roc_ax, prc_ax) = plt.subplots(1, 2, figsize=(12, 6))
for name, model in models:
   # Fit model
   model.fit(X_train, y_train)
   # Probability predictions
   y_prob = model.predict_proba(X_test)[:, 1]
   # ROC
   fpr, tpr, _ = roc_curve(y_test, y_prob)
   roc_auc = roc_auc_score(y_test, y_prob)
   roc_ax.plot(fpr, tpr, label=f"{name} (AUC = {roc_auc:.2f})")
   # Save ROC points
   roc_points.append(pd.DataFrame({
       "model": name,
       "fpr": fpr,
        "tpr": tpr
   }))
   precision, recall, _ = precision_recall_curve(y_test, y_prob)
                        = average_precision_score(y_test, y_prob)
   prc_ax.plot(recall, precision, label=f"{name} (AUPRC = {pr_auc:.2f})")
    # Save PR points
```

```
pr_points.append(pd.DataFrame({
        "model":
                     name,
        "recall":
                     recall,
        "precision": precision
   }))
    finalise plots
roc_ax.set_title("ROC Curve Comparison")
roc ax.set xlabel("False Positive Rate")
roc_ax.set_ylabel("True Positive Rate")
roc_ax.legend(loc="lower right")
prc_ax.set_title("Precision-Recall Curve Comparison")
prc_ax.set_xlabel("Recall")
prc_ax.set_ylabel("Precision")
prc_ax.legend(loc="lower left")
plt.tight_layout()
plt.show()
    save curve points to CSV
OUT_DIR = Path("/Users/haley/Desktop/send_tooo/AAA_final/fig3_curve_points")
OUT_DIR.mkdir(exist_ok=True)
pd.concat(roc_points, ignore_index=True).to_csv(OUT_DIR / "roc_curve_points.
 ⇔csv", index=False)
pd.concat(pr_points, ignore_index=True).to_csv(OUT_DIR / "pr_curve_points."
 ⇔csv", index=False)
print(f"Saved full ROC & PR curve coordinates to {OUT_DIR.resolve()}")
```



Saved full ROC & PR curve coordinates to /Users/haley/Desktop/send_tooo/AAA_final/fig3_curve_points

9 bootstrap auroc, aupre

```
[56]: #!/usr/bin/env python3
      """Bootstrapped evaluation of four classical classifiers on binary labels.
      Inputs (must be pre-loaded in the Python session):
          - full_X_train : pandas DataFrame, shape (n_samples, n_features)
          - full Y train : pandas Series or 1-D array of length n samples
          - realtest : pandas DataFrame, shape (n_test_samples, n_features)
          - realtest_y : pandas Series or 1-D array of length n_test_samples
      Outputs (saved in the working directory):
          - bootstrap_roc_auc.csv - mean AUROC + 95 % CI for each model
          - bootstrap_prc_auc.csv - mean AUPRC + 95 % CI for each model
      Each CSV has three rows (mean, lower CI, upper CI) and one column per model.
      from __future__ import annotations
      import warnings
      from pathlib import Path
      from typing import Tuple, List
      import numpy as np
      import pandas as pd
      import matplotlib.pyplot as plt # noqa: F401 - left here if you later addu
       \hookrightarrow plots
      from matplotlib import rcParams
      from sklearn.preprocessing import StandardScaler
      from sklearn.metrics import roc_auc_score, average_precision_score
      from sklearn.linear_model import LogisticRegression
      from sklearn.svm import SVC
      from sklearn.ensemble import RandomForestClassifier
      from xgboost import XGBClassifier
      # Global configuration
      N_BOOTSTRAPS: int = 100
```

```
RANDOM_SEED: int = 551
OUTPUT_DIR: Path = Path("/Users/haley/Desktop/send_tooo/AAA_final/fig4") #__
 → Change this if you want the CSVs elsewhere
# Matplotlib defaults - handy if you decide to visualise later
rcParams.update({
   "font.family": "Arial",
   "font.size": 7,
   "font.weight": "normal",
})
# Silence XGBoost warnings that clutter the console
warnings.filterwarnings("ignore", category=UserWarning, module="xgboost")
                      _____
# Utility functions
# -----
def bootstrap_metrics(
   model,
   X train: pd.DataFrame,
   y_train: pd.Series,
   X_test: pd.DataFrame,
   y_test: pd.Series,
   n_bootstraps: int = N_BOOTSTRAPS,
   seed: int = RANDOM_SEED,
) -> Tuple[Tuple[float, float, float], Tuple[float, float, float]]:
   """Compute bootstrapped AUROC and AUPRC for *model*.
   Returns
   roc_stats : (mean, lower_ci, upper_ci)
   prc_stats : (mean, lower_ci, upper_ci)
   11 11 11
   rng = np.random.default_rng(seed)
   roc scores: List[float] = []
   prc_scores: List[float] = []
   for _ in range(n_bootstraps):
       # Sample with replacement from the training set
       resample_idx = rng.choice(len(X_train), size=len(X_train), replace=True)
       X_resampled = X_train.iloc[resample_idx]
       y_resampled = y_train.iloc[resample_idx]
       # Standardise per-bootstrap to avoid information leakage
```

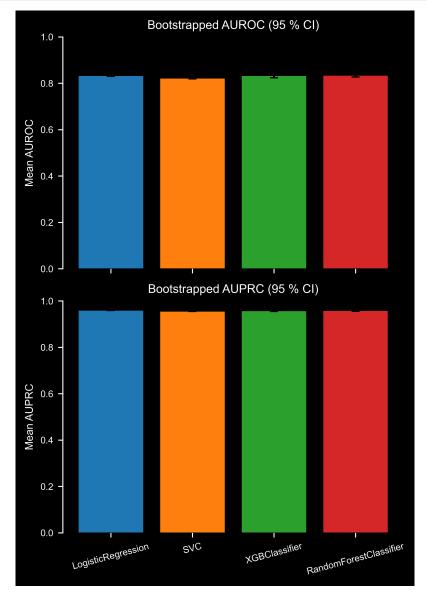
```
scaler = StandardScaler()
        X_resampled_scaled = scaler.fit_transform(X_resampled)
        X_test_scaled = scaler.transform(X_test)
        # Fit and predict
        model.fit(X_resampled_scaled, y_resampled)
        y_prob = model.predict_proba(X_test_scaled)[:, 1]
        roc_scores.append(roc_auc_score(y_test, y_prob))
        prc_scores.append(average_precision_score(y_test, y_prob))
    roc_arr = np.asarray(roc_scores)
    prc_arr = np.asarray(prc_scores)
    roc_stats = (
        float(roc_arr.mean()),
        float(np.percentile(roc_arr, 2.5)),
        float(np.percentile(roc_arr, 97.5)),
    )
    prc_stats = (
        float(prc_arr.mean()),
        float(np.percentile(prc_arr, 2.5)),
        float(np.percentile(prc_arr, 97.5)),
    )
    return roc_stats, prc_stats
# Main evaluation pipeline
def main() -> None:
    """Run bootstrap evaluation and save CSV summaries."""
    # These variables must already be defined elsewhere (e.g. notebook or \Box
 ⇒import)
   try:
        global full_X_train, full_Y_train, realtest, realtest_y # type: ignore
    except NameError as exc:
        raise RuntimeError(
            "full_X_train, full_Y_train, realtest, realtest_y need to be in_{\sqcup}
 ⇔scope "
            "before running this script."
        ) from exc
    models = \Gamma
        LogisticRegression(max_iter=1000, n_jobs=-1, random_state=RANDOM_SEED),
        SVC(kernel="linear", probability=True, random_state=RANDOM_SEED),
```

```
XGBClassifier(
            random_state=RANDOM_SEED,
            use_label_encoder=False,
            eval_metric="logloss",
            n_estimators=100,
            learning_rate=0.05,
            max_depth=10,
        ),
        RandomForestClassifier(max_depth=10, random_state=RANDOM_SEED),
    ]
    roc_results = {}
    prc results = {}
    for clf in models:
        roc_stats, prc_stats = bootstrap_metrics(
            clf, full_X_train, full_Y_train, realtest, realtest_y
        name = clf.__class__.__name__
        roc_results[name] = roc_stats
        prc_results[name] = prc_stats
        print(f"{name:>22}: AUROC={roc_stats[0]:.3f} AUPRC={prc_stats[0]:.3f}")
    # Save to CSV
    roc_df = pd.DataFrame(roc_results, index=["Mean AUROC", "Lower CI", "Upper_
    prc_df = pd.DataFrame(prc_results, index=["Mean AUPRC", "Lower CI", "Upper__
 ⇔CI"]).T
    OUTPUT_DIR.mkdir(parents=True, exist_ok=True)
    roc_df.to_csv(OUTPUT_DIR / "bootstrap_roc_auc.csv", index_label="Model")
    prc_df.to_csv(OUTPUT_DIR / "bootstrap_prc_auc.csv", index_label="Model")
    print("\nSaved bootstrap roc auc.csv and bootstrap prc auc.csv to",,,
 →OUTPUT_DIR.resolve())
if __name__ == "__main__":
    main()
   LogisticRegression: AUROC=0.833 AUPRC=0.960
                   SVC: AUROC=0.822 AUPRC=0.956
         XGBClassifier: AUROC=0.833 AUPRC=0.958
RandomForestClassifier: AUROC=0.834 AUPRC=0.958
```

Saved bootstrap_roc_auc.csv and bootstrap_prc_auc.csv to

/Users/haley/Desktop/send_tooo/AAA_final/fig4

```
[58]: #!/usr/bin/env python3
      Bar plots of AUROC and AUPRC (mean ± 95 % CI) on a **white** background.
      X-axis baseline and dashed y-axis grid are removed.
      import pandas as pd
      import matplotlib.pyplot as plt
      from pathlib import Path
      # 1. Read results
      roc_df = pd.read_csv(Path("bootstrap_roc_auc.csv"), index_col="Model")
      prc_df = pd.read_csv(Path("bootstrap_prc_auc.csv"), index_col="Model")
      models = roc_df.index.tolist()
      prc_df = prc_df.loc[models] # ensure same order
      def extract(df, col):
          mean = df[col].values
          err = [mean - df["Lower CI"].values,
                  df["Upper CI"].values - mean]
          return mean, err
      roc_mean, roc_err = extract(roc_df, "Mean AUROC")
      prc_mean, prc_err = extract(prc_df, "Mean AUPRC")
      # 2. Plotting setup
      colors = ["#1f77b4", "#ff7f0e", "#2ca02c", "#d62728"] # blue, orange, green, __
       \hookrightarrow red
      fig, axes = plt.subplots(2, 1, figsize=(4.2, 6), dpi=300, sharex=True)
      # Helper to style each subplot
      def style_axis(ax, ylabel, title, means, yerr):
          ax.bar(models, means, yerr=yerr, capsize=3,
                 color=colors, edgecolor="black", linewidth=0.6)
          ax.set_ylabel(ylabel, fontsize=8)
          ax.set_title(title, fontsize=9, pad=6)
          ax.set_ylim(0, 1)
          # Remove x-axis baseline and any grid lines
          ax.spines["bottom"].set_visible(False)
          ax.spines["top"].set_visible(False)
          ax.spines["right"].set_visible(False)
          # No grid
          ax.grid(False)
```



```
[33]:
    # Optional: Print results
    print("ROC AUC Data:")
```

```
print("\nPRC AUC Data:")
      print(prc_auc_df)
     ROC AUC Data:
                             Mean AUROC Lower CI Upper CI
     LogisticRegression
                                         0.829910
                                                   0.835949
                               0.833158
     SVC
                               0.822421
                                         0.818697
                                                   0.826163
     XGBClassifier
                               0.833681
                                         0.824071
                                                   0.843689
     RandomForestClassifier
                               0.835185
                                         0.827522
                                                   0.844919
     PRC AUC Data:
                             Mean AUPRC
                                         Lower CI
                                                   Upper CI
     LogisticRegression
                               0.959903
                                         0.958873
                                                   0.960731
     SVC
                               0.956253
                                         0.954741
                                                   0.957541
     XGBClassifier
                               0.957919
                                         0.954751
                                                   0.961573
     RandomForestClassifier
                               0.958576
                                        0.955422 0.961641
[49]: full_X_train
[49]: n
                    Rp135
                            Rps27rt
                                    Rp19-ps6
                                                  Rps27
                                                            Uba52
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      Cell_1678
                1.098612
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                                    0.000000
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                           1.386294
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      Cell_1575
                2.639057
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                                               1.098612 1.791759
                                                                   1.098612
      Cell_2934
                2.197225
                           1.945910 0.693147
                                               1.791759
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                 1.609438
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                                                          1.791759
      Cell_5187
                3.367296
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                                                          2.397895
      Cell_1921
                2.995732
                           0.0
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      Cell_1575
                3.496508
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                                           0.0
                                               0.000000
                                                          0.00000
      Cell_2934 2.708050
                           0.0
                                  0.0
                                           0.0 0.000000 2.079442
```

print(roc_auc_df)

```
Cell_2191
           3.332205
                     0.0
                            0.0
                                      0.0 0.000000
                                                     0.000000
Cell_6618
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Cell_5252
          0.693147
                     0.0
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                                          0.000000
                                                     0.693147
[6648 rows x 12 columns]
```

[55]: full_X_train

```
[55]: n
                     Xist
                              Ddx3y
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                                               Eif2s3y
                                                         Rps27rt
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                                     4.045352
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                                     4.331086
      Cell_5187
                3.324866
                                                   0.0
                                                         2.201503
                                                                               0.0
                           0.000000
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      Cell 1921
                0.000000
                           1.265771
                                     3.899357
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                                                        1.265771 1.806919
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                                     4.420726
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      Cell_2934
                 2.839474
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                                                         2.695019
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      Cell_2191
                0.000000
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                                                        1.313221
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                                     4.309315
                                                                   0.000000
      Cell_6618
                3.282452
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                                     4.087791
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                                                         2.119384
                                                                   1.539660
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      Cell_318
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      Cell_1678
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      Cell_1921
                1.806919
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      Cell_1575
                2.627344
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                                       2.163574
                                                 0.0
                                                       1.271835
                                                                 1.813976
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      Cell 2934
                0.000000
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      Cell 2191
                                       1.313221 0.0
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      Cell 6618
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```

[6648 rows x 14 columns]

```
[51]: scaler = StandardScaler()
full_X_train_scaled=scaler.fit_transform(full_X_train)
realtest_scaled=scaler.transform(realtest)
```

10 fig4-ab

```
[81]: #!/usr/bin/env python3
"""

Save *all* ROC and PR curve points in two master CSVs:
```

```
all_roc_points.csv # columns: model, fpr, tpr
      all_prc_points.csv # columns: model, recall, precision
It also produces a small summary (AUROC / AUPRC) table.
from pathlib import Path
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from matplotlib import rcParams
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import (
   roc_curve,
   auc,
   precision_recall_curve,
   average_precision_score,
from sklearn.linear_model import LogisticRegression
from sklearn.svm import SVC
from sklearn.ensemble import RandomForestClassifier
from xgboost import XGBClassifier # pip install xqboost if missing
   global config
rcParams.update({"font.family": "Arial", "font.size": 7})
OUTDIR = Path("/Users/haley/Desktop/send_tooo/AAA_final/curve_outputs")
OUTDIR.mkdir(exist_ok=True)
colors = ["#1f77b4", "#ff7f0e", "#2ca02c", "#d62728"]
models = [
   LogisticRegression(max_iter=1000, random_state=551),
   SVC(kernel="linear", probability=True, random_state=551),
   XGBClassifier(
       random state=551,
       use_label_encoder=False,
        eval metric="logloss",
       n_estimators=100,
       learning rate=0.05,
       max_depth=10,
   RandomForestClassifier(max_depth=10, random_state=41),
]
   scale data once
```

```
scaler = StandardScaler()
X train s = scaler.fit transform(full X train)
X_test_s = scaler.transform(realtest)
    containers
roc_rows, prc_rows = [], []
roc_summary, prc_summary = [], []
  fit, predict, collect points
for model in models:
   name = model.__class__._name__
   print("Training", name)
   model.fit(X_train_s, full_Y_train)
   y_prob = model.predict_proba(X_test_s)[:, 1]
   # ROC
   fpr, tpr, _ = roc_curve(realtest_y, y_prob)
   roc_auc = auc(fpr, tpr)
   roc_rows.append(pd.DataFrame({"model": name, "fpr": fpr, "tpr": tpr}))
   roc_summary.append({"model": name, "auc": roc_auc})
    # PR.
   prec, rec, _ = precision_recall_curve(realtest_y, y_prob)
   pr auc = average precision score(realtest y, y prob)
   prc_rows.append(pd.DataFrame({"model": name, "recall": rec, "precision": u
 ⇒prec}))
   prc_summary.append({"model": name, "auprc": pr_auc})
# write master CSVs
pd.concat(roc_rows, ignore_index=True).to_csv(OUTDIR / "all_roc_points.csv", __
 →index=False)
pd.concat(prc_rows, ignore_index=True).to_csv(OUTDIR / "all_prc_points.csv", __
 →index=False)
pd.DataFrame(roc_summary).to_csv(OUTDIR / "roc_auc_summary.csv", index=False)
pd.DataFrame(prc_summary).to_csv(OUTDIR / "prc_auc_summary.csv", index=False)
print(" Saved:")
for fn in ["all_roc_points.csv", "all_prc_points.csv",
           "roc_auc_summary.csv", "prc_auc_summary.csv"]:
   print(" ", OUTDIR / fn)
# (optional) quick plot from the master CSVs
plt.figure(figsize=(12, 5))
roc_ax = plt.subplot(1, 2, 1)
pr_ax = plt.subplot(1, 2, 2)
```

```
roc_pts = pd.read_csv(OUTDIR / "all_roc_points.csv")
prc_pts = pd.read_csv(OUTDIR / "all_prc_points.csv")
for name, color in zip([m.__class__.__name__ for m in models], colors):
    roc_sub = roc_pts[roc_pts.model == name]
    prc_sub = prc_pts[prc_pts.model == name]
    roc_ax.plot(roc_sub.fpr, roc_sub.tpr, label=name, color=color)
    pr_ax.plot(prc_sub.recall, prc_sub.precision, label=name, color=color)
roc ax.set title("ROC curves"); roc ax.set xlabel("FPR"); roc ax.
 ⇔set_ylabel("TPR")
pr_ax.set_title("PR curves");
                                pr_ax.set_xlabel("Recall"); pr_ax.
 ⇔set_ylabel("Precision")
for ax in (roc_ax, pr_ax):
    ax.spines["top"].set_visible(False)
    ax.spines["right"].set_visible(False)
    ax.legend(fontsize=7)
plt.tight_layout(); plt.show()
```

Training LogisticRegression

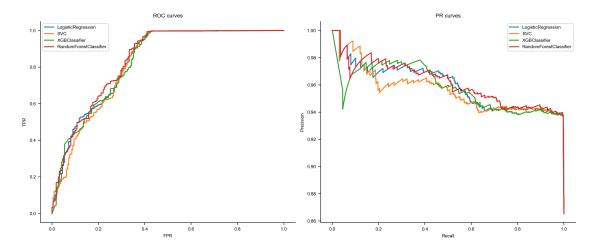
Training SVC

Training XGBClassifier

Training RandomForestClassifier

Saved:

/Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/all_roc_points.csv /Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/all_prc_points.csv /Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/roc_auc_summary.csv /Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/prc_auc_summary.csv



```
[47]: #!/usr/bin/env python3
      Save *all* ROC and PR curve points in two master CSVs:
            all_roc_points.csv # columns: model, fpr, tpr
            all_prc_points.csv # columns: model, recall, precision
      It also produces a small summary (AUROC / AUPRC) table.
      Additionally prints prediction counts and percentages for each model.
      from pathlib import Path
      import pandas as pd
      import numpy as np
      import matplotlib.pyplot as plt
      from matplotlib import rcParams
      from sklearn.preprocessing import StandardScaler
      from sklearn.metrics import (
          roc_curve,
          auc,
          precision_recall_curve,
          average_precision_score,
      )
      from sklearn.linear_model import LogisticRegression
      from sklearn.svm import SVC
      from sklearn.ensemble import RandomForestClassifier
      from xgboost import XGBClassifier # pip install xqboost if missing
          global config
      rcParams.update({"font.family": "Arial", "font.size": 7})
      OUTDIR = Path("/Users/haley/Desktop/send_tooo/AAA_final/curve_outputs")
      OUTDIR.mkdir(exist_ok=True)
      colors = ["#1f77b4", "#ff7f0e", "#2ca02c", "#d62728"]
      models = [
          LogisticRegression(max_iter=1000, random_state=551),
          SVC(kernel="linear", probability=True, random_state=551),
          XGBClassifier(
              random_state=551,
              use label encoder=False,
              eval_metric="logloss",
              n estimators=100,
              learning_rate=0.05,
              max_depth=10,
          ),
```

```
RandomForestClassifier(max_depth=10, random_state=41),
]
   scale data once
scaler = StandardScaler()
X_train_s = scaler.fit_transform(full_X_train)
X_test_s = scaler.transform(realtest)
   Calculate actual counts in test data
actual_female = (realtest_y == 0).sum() # Assuming 0 = female
actual_male = (realtest_y == 1).sum() # Assuming 1 = male
total_test = len(realtest_y)
print("=" * 70)
print("ACTUAL TEST DATA DISTRIBUTION")
print("=" * 70)
print(f"Total test samples: {total_test}")
print(f"Actual Female (0): {actual_female} ({actual_female/total_test*100:.
 →1f}%)")
print(f"Actual Male (1): {actual_male} ({actual_male/total_test*100:.1f}%)")
print()
  containers
roc_rows, prc_rows = [], []
roc_summary, prc_summary = [], []
# fit, predict, collect points
for i, model in enumerate(models):
   name = model.__class__._name__
   print("=" * 70)
   print(f"MODEL {i+1}: {name}")
   print("=" * 70)
    # Train model
   model.fit(X_train_s, full_Y_train)
   # Get predictions (not just probabilities)
   y_pred = model.predict(X_test_s)
   y_prob = model.predict_proba(X_test_s)[:, 1]
   # Count predictions
   pred_female = (y_pred == 0).sum()
   pred_male = (y_pred == 1).sum()
   # Print prediction results
   print(f"Predicted Female (0): {pred_female} ({pred_female/total_test*100:.
 →1f}%)")
```

```
print(f"Predicted Male (1): {pred_male} ({pred_male/total_test*100:.

41f}%)")

   print()
   # Calculate accuracy for each class
   correct female = ((y pred == 0) & (realtest y == 0)).sum()
   correct_male = ((y_pred == 1) & (realtest_y == 1)).sum()
   if actual_female > 0:
       female_accuracy = correct_female / actual_female * 100
       print(f"Female prediction accuracy: {correct_female}/{actual_female}_\_
 if actual_male > 0:
       male_accuracy = correct_male / actual_male * 100
       print(f"Male prediction accuracy: {correct_male}/{actual_male}__
 overall_accuracy = (correct_female + correct_male) / total_test * 100
   print(f"Overall accuracy: {overall_accuracy:.1f}%")
   # ROC
   fpr, tpr, _ = roc_curve(realtest_y, y_prob)
   roc_auc = auc(fpr, tpr)
   roc rows.append(pd.DataFrame({"model": name, "fpr": fpr, "tpr": tpr}))
   roc_summary.append({"model": name, "auc": roc_auc})
   print(f"ROC AUC: {roc auc:.3f}")
   # PR
   prec, rec, _ = precision_recall_curve(realtest_y, y_prob)
   pr_auc = average_precision_score(realtest_y, y_prob)
   prc_rows.append(pd.DataFrame({"model": name, "recall": rec, "precision": u
 →prec}))
   prc_summary.append({"model": name, "auprc": pr_auc})
   print(f"PR AUC: {pr auc:.3f}")
   print()
# write master CSVs
pd.concat(roc_rows, ignore_index=True).to_csv(OUTDIR / "all_roc_points.csv", __
 →index=False)
pd.concat(prc_rows, ignore_index=True).to_csv(OUTDIR / "all_prc_points.csv",_
pd.DataFrame(roc_summary).to_csv(OUTDIR / "roc_auc_summary.csv", index=False)
pd.DataFrame(prc_summary).to_csv(OUTDIR / "prc_auc_summary.csv", index=False)
```

```
print("=" * 70)
print(" Saved:")
for fn in ["all_roc_points.csv", "all_prc_points.csv",
          "roc_auc_summary.csv", "prc_auc_summary.csv"]:
   print(" ", OUTDIR / fn)
  (optional) quick plot from the master CSVs
plt.figure(figsize=(12, 5))
roc_ax = plt.subplot(1, 2, 1)
pr_ax = plt.subplot(1, 2, 2)
roc_pts = pd.read_csv(OUTDIR / "all_roc_points.csv")
prc_pts = pd.read_csv(OUTDIR / "all_prc_points.csv")
for name, color in zip([m. class . name for m in models], colors):
   roc_sub = roc_pts[roc_pts.model == name]
   prc_sub = prc_pts[prc_pts.model == name]
   roc_ax.plot(roc_sub.fpr, roc_sub.tpr, label=name, color=color)
   pr_ax.plot(prc_sub.recall, prc_sub.precision, label=name, color=color)
roc_ax.set_title("ROC curves"); roc_ax.set_xlabel("FPR"); roc_ax.
 ⇔set_ylabel("TPR")
⇔set_ylabel("Precision")
for ax in (roc_ax, pr_ax):
   ax.spines["top"].set visible(False)
   ax.spines["right"].set_visible(False)
   ax.legend(fontsize=7)
plt.tight_layout(); plt.show()
```

ACTUAL TEST DATA DISTRIBUTION

Total test samples: 1649
Actual Female (0): 222 (13.5%)
Actual Male (1): 1427 (86.5%)

MODEL 1: LogisticRegression

Predicted Female (0): 332 (20.1%) Predicted Male (1): 1317 (79.9%)

Female prediction accuracy: 146/222 (65.8%)
Male prediction accuracy: 1241/1427 (87.0%)

Overall accuracy: 84.1%

ROC AUC: 0.833 PR AUC: 0.960

MODEL 2: SVC

Predicted Female (0): 275 (16.7%) Predicted Male (1): 1374 (83.3%)

Female prediction accuracy: 143/222 (64.4%)
Male prediction accuracy: 1295/1427 (90.7%)

Overall accuracy: 87.2%

ROC AUC: 0.823 PR AUC: 0.957

MODEL 3: XGBClassifier

Predicted Female (0): 386 (23.4%) Predicted Male (1): 1263 (76.6%)

Female prediction accuracy: 146/222 (65.8%)
Male prediction accuracy: 1187/1427 (83.2%)

Overall accuracy: 80.8%

ROC AUC: 0.830 PR AUC: 0.957

MODEL 4: RandomForestClassifier

Predicted Female (0): 179 (10.9%) Predicted Male (1): 1470 (89.1%)

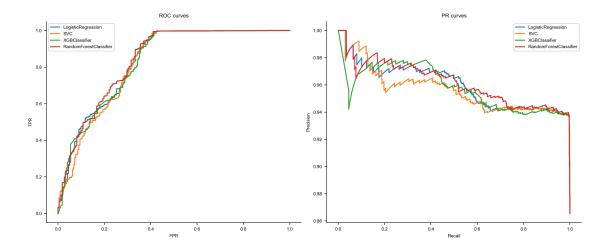
Female prediction accuracy: 134/222 (60.4%)
Male prediction accuracy: 1382/1427 (96.8%)

Overall accuracy: 91.9%

ROC AUC: 0.840 PR AUC: 0.962

Saved:

/Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/all_roc_points.csv /Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/all_prc_points.csv /Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/roc_auc_summary.csv /Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/prc_auc_summary.csv



11 misclassified labels

```
[105]: #!/usr/bin/env python3
       Identify and save misclassified samples for each model:
           - Males misclassified as females
           - Females misclassified as males
       Saves cell barcodes (not indices) and actual/predicted labels to CSV files.
       11 11 11
       from pathlib import Path
       import pandas as pd
       import numpy as np
       from sklearn.preprocessing import StandardScaler
       from sklearn.linear_model import LogisticRegression
       from sklearn.svm import SVC
       from sklearn.ensemble import RandomForestClassifier
       from xgboost import XGBClassifier # pip install xgboost if missing
           global config
       OUTDIR = Path("/Users/haley/Desktop/send_tooo/AAA_final/misclassified")
       OUTDIR.mkdir(parents=True, exist_ok=True)
       models = [
           LogisticRegression(max_iter=1000, random_state=551),
           SVC(kernel="linear", probability=True, random_state=551),
           XGBClassifier(
               random state=551,
```

```
use_label_encoder=False,
        eval_metric="logloss",
       n_estimators=100,
       learning_rate=0.05,
       max_depth=10,
   ),
   RandomForestClassifier(max_depth=10, random_state=41),
]
# scale data once
scaler = StandardScaler()
X_train_s = scaler.fit_transform(full_X_train)
X_test_s = scaler.transform(realtest)
    Calculate actual counts in test data
actual_female = (realtest_y == 0).sum() # 0 = female
actual_male = (realtest_y == 1).sum() # 1 = male
total_test = len(realtest_y)
print("=" * 70)
print("ACTUAL TEST DATA DISTRIBUTION")
print("=" * 70)
print(f"Total test samples: {total_test}")
print(f"Actual Female (0): {actual_female} ({actual_female/total_test*100:.
→1f}%)")
print(f"Actual Male (1): {actual_male} ({actual_male/total_test*100:.1f}%)")
print()
   fit, predict, save misclassified
for i, model in enumerate(models):
   name = model.__class__._name__
   print("=" * 70)
   print(f"MODEL {i+1}: {name}")
   print("=" * 70)
   # Train model
   model.fit(X_train_s, full_Y_train)
   # Get predictions
   y_pred = model.predict(X_test_s)
   # Count predictions
   pred_female = (y_pred == 0).sum()
   pred_male = (y_pred == 1).sum()
    # Print prediction results
```

```
print(f"Predicted Female (0): {pred_female} ({pred_female/total_test*100:.
→1f}%)")
  print(f"Predicted Male (1): {pred_male} ({pred_male/total_test*100:.
→1f}%)")
  print()
  # Calculate accuracy for each class
  correct_female = ((y_pred == 0) & (realtest_y == 0)).sum()
  correct_male = ((y_pred == 1) & (realtest_y == 1)).sum()
  if actual_female > 0:
      female_accuracy = correct_female / actual_female * 100
      print(f"Female prediction accuracy: {correct_female}/{actual_female}_\_
if actual male > 0:
      male_accuracy = correct_male / actual_male * 100
      print(f"Male prediction accuracy: {correct male}/{actual male}_\( \)
overall_accuracy = (correct_female + correct_male) / total_test * 100
  print(f"Overall accuracy: {overall_accuracy:.1f}%")
  # Find misclassified samples
  # Male misclassified as Female (actual=1, predicted=0)
  male_as_female_mask = (realtest_y == 1) & (y_pred == 0)
  # Female misclassified as Male (actual=0, predicted=1)
  female_as_male_mask = (realtest_y == 0) & (y_pred == 1)
  # Get cell barcodes instead of numeric indices
  # Assuming realtest has cell barcodes as index
  male_as_female_cells = realtest.index[male_as_female_mask].tolist()
  female_as_male_cells = realtest.index[female_as_male_mask].tolist()
  print(f"\nMisclassification Summary:")
  print(f" Males misclassified as Females: {len(male_as_female_cells)}")
  print(f" Females misclassified as Males: {len(female_as_male_cells)}")
      Save misclassified cells to CSV
  # Create dataframes with cell barcodes, actual label, and predicted label
  male_as_female_df = pd.DataFrame({
       'cell_barcode': male_as_female_cells,
      'actual_label': 1, # male
      'predicted_label': 0 # female
  })
```

```
female_as_male_df = pd.DataFrame({
        'cell_barcode': female_as_male_cells,
        'actual_label': 0, # female
        'predicted_label': 1 # male
   })
    # Save to files
   male_as_female_file = OUTDIR / f"{name}_male_misclassified_as_female.csv"
   female_as_male_file = OUTDIR / f"{name}_female_misclassified_as_male.csv"
   male as female df.to csv(male as female file, index=False)
   female_as_male_df.to_csv(female_as_male_file, index=False)
   print(f"\nSaved misclassified cell barcodes:")
                {male_as_female_file.name}")
   print(f"
                {female_as_male_file.name}")
   print(f"
        Also save a combined misclassification file
   all_misclassified = pd.concat([
        male_as_female_df.assign(misclassification_type='male_as_female'),
        female_as_male_df.assign(misclassification_type='female_as_male')
   ]).sort_values('cell_barcode')
   combined file = OUTDIR / f"{name} all misclassified.csv"
   all_misclassified.to_csv(combined_file, index=False)
   print(f"
                {combined file.name}")
   print()
# Create summary of all models' misclassifications
print("=" * 70)
print("CREATING MISCLASSIFICATION SUMMARY")
print("=" * 70)
summary_data = []
for model in models:
   name = model.__class__._name__
    # Read the saved files to get counts
   male_as_female_file = OUTDIR / f"{name}_male_misclassified_as_female.csv"
   female_as_male_file = OUTDIR / f"{name}_female_misclassified_as_male.csv"
   male_as_female_count = len(pd.read_csv(male_as_female_file))
   female_as_male_count = len(pd.read_csv(female_as_male_file))
    summary_data.append({
        'model': name,
        'male_as_female': male_as_female_count,
```

```
'female_as_male': female_as_male_count,
       'total misclassified': male as female count + female as male count
    })
summary_df = pd.DataFrame(summary_data)
summary_file = OUTDIR / "misclassification_summary.csv"
summary_df.to_csv(summary_file, index=False)
print(f" Saved misclassification summary: {summary file.name}")
print("\nSummary:")
print(summary_df.to_string(index=False))
______
ACTUAL TEST DATA DISTRIBUTION
_____
Total test samples: 1649
Actual Female (0): 222 (13.5%)
Actual Male (1): 1427 (86.5%)
MODEL 1: LogisticRegression
Predicted Female (0): 332 (20.1%)
Predicted Male (1): 1317 (79.9%)
Female prediction accuracy: 146/222 (65.8%)
Male prediction accuracy: 1241/1427 (87.0%)
Overall accuracy: 84.1%
Misclassification Summary:
 Males misclassified as Females: 186
 Females misclassified as Males: 76
Saved misclassified cell barcodes:
   LogisticRegression_male_misclassified_as_female.csv
   LogisticRegression_female_misclassified_as_male.csv
   LogisticRegression_all_misclassified.csv
MODEL 2: SVC
______
Predicted Female (0): 275 (16.7%)
Predicted Male (1): 1374 (83.3%)
Female prediction accuracy: 143/222 (64.4%)
Male prediction accuracy: 1295/1427 (90.7%)
Overall accuracy: 87.2%
```

Misclassification Summary: Males misclassified as Females: 132 Females misclassified as Males: 79 Saved misclassified cell barcodes: SVC_male_misclassified_as_female.csv SVC_female_misclassified_as_male.csv SVC_all_misclassified.csv MODEL 3: XGBClassifier ______ /Users/haley/Library/Python/3.9/lib/python/site-packages/xgboost/core.py:158: UserWarning: [10:50:50] WARNING: /Users/runner/work/xgboost/xgboost/src/learner.cc:740: Parameters: { "use_label_encoder" } are not used. warnings.warn(smsg, UserWarning) Predicted Female (0): 386 (23.4%) Predicted Male (1): 1263 (76.6%) Female prediction accuracy: 146/222 (65.8%) Male prediction accuracy: 1187/1427 (83.2%) Overall accuracy: 80.8% Misclassification Summary: Males misclassified as Females: 240 Females misclassified as Males: 76 Saved misclassified cell barcodes: XGBClassifier_male_misclassified_as_female.csv XGBClassifier_female_misclassified_as_male.csv XGBClassifier_all_misclassified.csv _____ MODEL 4: RandomForestClassifier ______ Predicted Female (0): 179 (10.9%) Predicted Male (1): 1470 (89.1%) Female prediction accuracy: 134/222 (60.4%) Male prediction accuracy: 1382/1427 (96.8%) Overall accuracy: 91.9%

 ${\tt Misclassification\ Summary:}$

Males misclassified as Females: 45 Females misclassified as Males: 88

Saved misclassified cell barcodes:

RandomForestClassifier_male_misclassified_as_female.csv RandomForestClassifier_female_misclassified_as_male.csv RandomForestClassifier_all_misclassified.csv

CREATING MISCLASSIFICATION SUMMARY

Saved misclassification summary: misclassification_summary.csv

Summary:

model	male_as_female	female_as_male	total_misclassified
${ t Logistic Regression}$	186	76	262
SVC	132	79	211
XGBClassifier	240	76	316
RandomForestClassifier	45	88	133

[121]: #!/usr/bin/env python3

```
Create AnnData object with RandomForest misclassification information
import pandas as pd
import numpy as np
import scanpy as sc
import anndata
from pathlib import Path
    Step 1: Load and process the original data (your existing code)
print("Loading original data...")
realmat = sc.read_mtx('matrix.mtx')
realfeature = pd.read_csv('features.tsv', sep='\t', header=None)
realmat = realmat[0:32285]
realtag = pd.read_csv('tag.csv')
actt = realfeature.loc[realfeature[2] == 'Gene Expression']
celltag = pd.read_csv('barcodes.tsv', sep='\t', header=None)
# Create AnnData object
adata = anndata.AnnData(X=realmat.X.T)
lol = list(celltag[0])
adata.obs['cell'] = lol
lmao = list(actt[1])
adata.var['gene_ids'] = lmao
# Scanpy settings
sc.settings.verbosity = 3
```

```
sc.logging.print_header()
sc.settings.set_figure_params(dpi=80, facecolor='white', fontsize=7)
# Filter cells and genes
sc.pp.filter_cells(adata, min_genes=100)
sc.pp.filter_genes(adata, min_cells=3)
# Mitochondrial genes QC
adata.var['mt'] = adata.var names.str.startswith('mt-')
sc.pp.calculate_qc_metrics(adata, qc_vars=['mt'], percent_top=None,_
 ⇔log1p=False, inplace=True)
# Set gene IDs as index
adata.var.index = adata.var['gene_ids'].values
adata.var.index.name = None
adata.var_names_make_unique()
# Normalize and log transform
sc.pp.normalize_total(adata, target_sum=1e4)
sc.pp.log1p(adata)
    Step 2: Process gender information
print("\nProcessing gender information...")
METT = adata.X
dense_matrix = METT.toarray()
METT = pd.DataFrame(dense_matrix, columns=adata.var_names, index=adata.
 ⇔obs['cell'].values)
METT['cell'] = adata.obs['cell'].values
# Join with tag information
realtag.index = realtag['cell_barcode'].values
result = METT.join(realtag[['feature_call']], how='left')
result = result.dropna(subset=['feature_call'])
# Initialize and update gender
result['gender'] = None
result.loc[result['feature_call'] == 'CMO305', 'gender'] = 1 # Male
result.loc[result['feature_call'] == 'CMO306', 'gender'] = 0 # Female
   Step 3: Load RandomForest misclassification results
print("\nLoading RandomForest misclassification data...")
OUTDIR = Path("/Users/haley/Desktop/send_tooo/AAA_final/misclassified")
# Read the RandomForest misclassification files
rf_all_misclass = pd.read_csv(OUTDIR /_

¬"RandomForestClassifier_all_misclassified.csv")
```

```
rf_male_as_female = pd.read_csv(OUTDIR /_

¬"RandomForestClassifier_male_misclassified_as_female.csv")

rf_female_as_male = pd.read_csv(OUTDIR /_
 → "RandomForestClassifier female misclassified as male.csv")
print(f"Total misclassified cells: {len(rf_all_misclass)}")
print(f" - Males misclassified as females: {len(rf_male_as_female)}")
print(f" - Females misclassified as males: {len(rf_female_as_male)}")
   Step 4: Add misclassification info to AnnData
print("\nAdding misclassification info to AnnData...")
# First, add the actual gender labels to all cells that have them
# Use NaN for numeric columns and empty string for string columns
adata.obs['actual_gender'] = np.nan # Use NaN instead of None for numeric
adata.obs['actual_gender_label'] = '' # Use empty string instead of None
# Map gender information from result dataframe
for idx, row in result.iterrows():
    if idx in adata.obs['cell'].values:
        cell_idx = adata.obs[adata.obs['cell'] == idx].index[0]
        adata.obs.loc[cell_idx, 'actual_gender'] = float(row['gender'])
        adata.obs.loc[cell_idx, 'actual_gender_label'] = 'Male' if__
 →row['gender'] == 1 else 'Female'
# Add RandomForest prediction and misclassification status
adata.obs['rf_predicted_gender'] = np.nan # Use NaN for numeric
adata.obs['rf_predicted_gender_label'] = '' # Use empty string
adata.obs['rf misclassified'] = False
adata.obs['rf_misclass_type'] = '' # Use empty string
# Mark misclassified cells
for _, row in rf_all_misclass.iterrows():
   cell_barcode = row['cell_barcode']
   if cell_barcode in adata.obs['cell'].values:
        cell_idx = adata.obs[adata.obs['cell'] == cell_barcode].index[0]
        adata.obs.loc[cell_idx, 'rf_predicted_gender'] = __

¬float(row['predicted_label'])
        adata.obs.loc[cell_idx, 'rf_predicted_gender_label'] = 'Male' if_
 →row['predicted_label'] == 1 else 'Female'
        adata.obs.loc[cell_idx, 'rf_misclassified'] = True
        adata.obs.loc[cell_idx, 'rf_misclass_type'] = __
 →row['misclassification_type']
# For correctly classified cells, predicted = actual
```

```
correct_mask = (adata.obs['actual_gender'].notna()) & (~adata.
 ⇔obs['rf misclassified'])
adata.obs.loc[correct_mask, 'rf_predicted_gender'] = adata.obs.
 →loc[correct_mask, 'actual_gender']
adata.obs.loc[correct_mask, 'rf_predicted_gender_label'] = adata.obs.
 →loc[correct_mask, 'actual_gender_label']
   Step 5: Add summary statistics to uns
print("\nAdding summary statistics...")
adata.uns['rf misclassification'] = {
    'total_misclassified': len(rf_all_misclass),
    'male_as_female': len(rf_male_as_female),
    'female_as_male': len(rf_female_as_male),
    'misclassified cells': {
        'male_as_female': rf_male_as_female['cell_barcode'].tolist(),
        'female_as_male': rf_female_as_male['cell_barcode'].tolist()
   }
}
# Count cells with gender info
has_gender = adata.obs['actual_gender'].notna().sum()
total_cells = len(adata.obs)
print(f"\nCells with gender information: {has_gender}/{total_cells}")
    Step 6: Convert data types before saving
print("\nConverting data types for h5ad compatibility...")
# Convert string columns to categorical to save space and ensure compatibility
string_cols = ['actual_gender_label', 'rf_predicted_gender_label', |
 for col in string_cols:
    # Replace empty strings with a proper category
   adata.obs[col] = adata.obs[col].replace('', 'Unknown')
    adata.obs[col] = pd.Categorical(adata.obs[col])
# Ensure numeric columns are float type
adata.obs['actual_gender'] = adata.obs['actual_gender'].astype('float32')
adata.obs['rf_predicted_gender'] = adata.obs['rf_predicted_gender'].
 ⇔astype('float32')
# Step 7: Save the AnnData object
   Step 7: Save the AnnData object
output_file = "adata_with_rf_misclassification.h5ad"
print(f"\nSaving AnnData to {output_file}...")
adata.write_h5ad(output_file)
   Step 8: Create visualization-ready subset
```

```
print("\nCreating subset with only cells that have gender labels...")
# Filter to only cells with gender information (not Unknown)
adata_gender = adata[adata.obs['actual_gender_label'] != 'Unknown'].copy()
# Add colors for visualization
adata_gender.uns['actual_gender_colors'] = ['#FF69B4', '#4169E1'] # Pink for_
 →Female, Blue for Male
adata_gender.uns['rf_misclassified_colors'] = ['#90EE90', '#FF6347'] # Light_
 → green for correct, Tomato for misclassified
output file gender = "adata gender_only_with_rf_misclass.h5ad"
print(f"Saving gender-only AnnData to {output file gender}...")
adata_gender.write_h5ad(output_file_gender)
   Print summary
print("\n" + "="*70)
print("SUMMARY")
print("="*70)
print(f"Total cells in AnnData: {len(adata.obs)}")
print(f"Cells with gender labels: {has_gender}")
print(f"Cells without gender labels: {total_cells - has gender}")
print(f"\nRandomForest misclassification:")
print(f" - Total misclassified: {len(rf_all_misclass)}")
print(f" - Males → Females: {len(rf_male_as_female)}")
print(f" - Females → Males: {len(rf_female_as_male)}")
print(f"\nOutput files created:")
print(f" 1. {output file} (all cells)")
print(f" 2. {output_file_gender} (only cells with gender labels)")
    Optional: Create a quick visualization
print("\nYou can now load the data and visualize misclassifications:")
print(">>> adata = sc.read_h5ad('adata_gender_only_with_rf_misclass.h5ad')")
print(">>> sc.pl.umap(adata, color=['actual_gender_label', 'rf_misclassified', u
 Loading original data...
scanpy==1.10.2 anndata==0.10.8 umap==0.5.6 numpy==1.26.4 scipy==1.13.1
pandas==2.2.2 scikit-learn==1.5.1 statsmodels==0.14.2 pynndescent==0.5.13
filtered out 3 cells that have less than 100 genes expressed
filtered out 19261 genes that are detected in less than 3 cells
normalizing counts per cell
   finished (0:00:00)
Processing gender information...
Loading RandomForest misclassification data...
```

Total misclassified cells: 133

```
- Females misclassified as males: 88
      Adding misclassification info to AnnData...
      Adding summary statistics...
      Cells with gender information: 1649/1649
      Converting data types for h5ad compatibility...
      Saving AnnData to adata_with_rf_misclassification.h5ad...
      Creating subset with only cells that have gender labels...
      Saving gender-only AnnData to adata_gender_only_with_rf_misclass.h5ad...
      SUMMARY
      _____
      Total cells in AnnData: 1649
      Cells with gender labels: 1649
      Cells without gender labels: 0
      RandomForest misclassification:
        - Total misclassified: 133
        - Males → Females: 45
        - Females → Males: 88
      Output files created:

    adata_with_rf_misclassification.h5ad (all cells)

        adata_gender_only_with_rf_misclass.h5ad (only cells with gender labels)
      You can now load the data and visualize misclassifications:
      >>> adata = sc.read_h5ad('adata_gender_only_with_rf_misclass.h5ad')
      >>> sc.pl.umap(adata, color=['actual_gender_label', 'rf_misclassified',
      'rf_misclass_type'])
[123]: import scanpy as sc
      import numpy as np
      # Load the AnnData
      adata = sc.read_h5ad('adata_with_rf_misclassification.h5ad')
      # Example 1: Get all misclassified cells
      misclassified_cells = adata[adata.obs['rf_misclassified'] == True]
      print(f"Number of misclassified cells: {len(misclassified_cells)}")
```

- Males misclassified as females: 45

```
# Example 2: Get males misclassified as females
male as female = adata[adata.obs['rf misclass type'] == 'male as female']
print(f"Males misclassified as females: {len(male_as_female)}")
# Example 3: View the actual vs predicted for a specific cell
cell_barcode = 'AAACCCAAGAAACCCA-1'  # Example - replace with a real barcode
if cell barcode in adata.obs['cell'].values:
    cell_data = adata[adata.obs['cell'] == cell_barcode].obs
   print(f"\nCell: {cell barcode}")
   print(f"Actual: {cell_data['actual_gender_label'].iloc[0]}")
   print(f"Predicted: {cell_data['rf_predicted_gender_label'].iloc[0]}")
   print(f"\nCell {cell barcode} not found in data")
# Example 4: Get summary statistics
summary = adata.uns['rf_misclassification']
print(f"\nTotal misclassified: {summary['total_misclassified']}")
print(f"Male → Female: {summary['male_as_female']}")
print(f"Female -> Male: {summary['female_as_male']}")
# Example 5: Get list of all cell barcodes that were misclassified (corrected)
# Convert to lists first, then concatenate
male_as_female_list = list(summary['misclassified_cells']['male_as_female'])
female as male list = list(summary['misclassified cells']['female as male'])
all_misclass_barcodes = male_as_female_list + female_as_male_list
print(f"\nTotal misclassified cell barcodes: {len(all_misclass_barcodes)}")
print(f"First 5 misclassified barcodes: {all_misclass_barcodes[:5]}")
# Alternative method using numpy concatenate
all_misclass_barcodes_np = np.concatenate([
    summary['misclassified_cells']['male_as_female'],
    summary['misclassified_cells']['female_as_male']
print(f"\nUsing numpy concatenate: {len(all_misclass_barcodes_np)} barcodes")
# Example 6: Get actual cell barcodes from the obs dataframe (more reliable)
print("\n--- Getting misclassified cells directly from obs ---")
misclass_barcodes_from_obs = adata.obs[adata.obs['rf_misclassified'] ==__
 Grue]['cell'].tolist()
print(f"Misclassified cells from obs: {len(misclass_barcodes_from_obs)}")
print(f"First 5: {misclass_barcodes_from_obs[:5]}")
# Example 7: Analyze misclassification patterns
print("\n--- Misclassification Analysis ---")
# Get gender distribution
print(f"Total females: {(adata.obs['actual_gender_label'] == 'Female').sum()}")
```

```
print(f"Total males: {(adata.obs['actual_gender_label'] == 'Male').sum()}")
# Calculate misclassification rates
total_females = (adata.obs['actual_gender_label'] == 'Female').sum()
total_males = (adata.obs['actual_gender_label'] == 'Male').sum()
female_misclass_rate = summary['female_as_male'] / total_females * 100
male_misclass_rate = summary['male_as_female'] / total_males * 100
print(f"\nMisclassification rates:")
print(f"Female → Male: {summary['female_as_male']}/{total_females}_⊔
 print(f"Male → Female: {summary['male_as_female']}/{total_males}_\( \)
 ⇔({male_misclass_rate:.1f}%)")
# Example 8: Look at a specific misclassified cell
print("\n--- Example misclassified cell ---")
if len(male_as_female) > 0:
    example_cell = male_as_female.obs.iloc[0]
    print(f"Cell barcode: {example_cell['cell']}")
    print(f"Actual gender: {example_cell['actual_gender_label']}__
 print(f"RF predicted: {example_cell['rf_predicted_gender_label']}_
 print(f"Misclass type: {example_cell['rf_misclass_type']}")
Number of misclassified cells: 133
Males misclassified as females: 45
Cell AAACCCAAGAAACCCA-1 not found in data
Total misclassified: 133
Male → Female: 45
Female → Male: 88
Total misclassified cell barcodes: 133
First 5 misclassified barcodes: ['AACCAACAGATACGAT-1', 'AACTTCTTCGCGTGCA-1',
'AAGTACCTCCTTCACG-1', 'AATGAAGGTGGAATGC-1', 'ACAAGCTAGATAACAC-1']
Using numpy concatenate: 133 barcodes
--- Getting misclassified cells directly from obs ---
Misclassified cells from obs: 133
First 5: ['AAAGGGCTCCTTCACG-1', 'AAATGGAAGTACCCTA-1', 'AACAGGGTCGAACTCA-1',
'AACCAACAGATACGAT-1', 'AACTTCTTCGCGTGCA-1']
--- Misclassification Analysis ---
Total females: 222
```

```
Total males: 1427
     Misclassification rates:
     Female → Male: 88/222 (39.6%)
     Male → Female: 45/1427 (3.2%)
     --- Example misclassified cell ---
     Cell barcode: AACCAACAGATACGAT-1
     Actual gender: Male (1.0)
     RF predicted: Female (0.0)
     Misclass type: male_as_female
 []:
 []:
[45]: #!/usr/bin/env python3
      Save *all* ROC and PR curve points in two master CSVs:
             all_roc_points.csv # columns: model, fpr, tpr
                                  # columns: model, recall, precision
             all_prc_points.csv
      It also produces a small summary (AUROC / AUPRC) table and confusion matrices \Box
       \hookrightarrow for each model.
      11 11 11
      from pathlib import Path
      import pandas as pd
      import numpy as np
      import matplotlib.pyplot as plt
      from matplotlib import rcParams
      import seaborn as sns
      from sklearn.preprocessing import StandardScaler
      from sklearn.metrics import (
          roc_curve,
          auc,
          precision_recall_curve,
          average_precision_score,
          confusion_matrix,
          classification_report,
      from sklearn.linear_model import LogisticRegression
      from sklearn.svm import SVC
      from sklearn.ensemble import RandomForestClassifier
      from xgboost import XGBClassifier # pip install xgboost if missing
```

```
global config
rcParams.update({"font.family": "Arial", "font.size": 7})
OUTDIR = Path("/Users/haley/Desktop/send_tooo/AAA_final/curve_outputs")
OUTDIR.mkdir(exist_ok=True)
colors = ["#1f77b4", "#ff7f0e", "#2ca02c", "#d62728"]
models = \Gamma
    LogisticRegression(max_iter=1000, random_state=551),
    SVC(kernel="linear", probability=True, random_state=551),
    XGBClassifier(
        random state=551,
        use_label_encoder=False,
        eval_metric="logloss",
        n_estimators=100,
        learning_rate=0.05,
        max_depth=10,
    ),
    RandomForestClassifier(max_depth=10, random_state=41),
]
   scale data once
scaler = StandardScaler()
X_train_s = scaler.fit_transform(full_X_train)
X test s = scaler.transform(realtest)
    containers
roc_rows, prc_rows = [], []
roc_summary, prc_summary = [], []
confusion_matrices = {}
   fit, predict, collect points
for model in models:
    name = model.__class__._name__
    print("Training", name)
    model.fit(X_train_s, full_Y_train)
    y_prob = model.predict_proba(X_test_s)[:, 1]
    y_pred = model.predict(X_test_s) # For confusion matrix
    # ROC
    fpr, tpr, _ = roc_curve(realtest_y, y_prob)
    roc_auc = auc(fpr, tpr)
    roc_rows.append(pd.DataFrame({"model": name, "fpr": fpr, "tpr": tpr}))
    roc_summary.append({"model": name, "auc": roc_auc})
    # PR
```

```
prec, rec, _ = precision_recall_curve(realtest_y, y_prob)
   pr_auc = average_precision_score(realtest_y, y_prob)
   prc_rows.append(pd.DataFrame({"model": name, "recall": rec, "precision": u
 →prec}))
   prc_summary.append({"model": name, "auprc": pr_auc})
    # Confusion Matrix
    cm = confusion_matrix(realtest_y, y_pred)
    confusion_matrices[name] = cm
    # Save confusion matrix as CSV
    cm_df = pd.DataFrame(cm,
                         index=['True Neg', 'True Pos'],
                         columns=['Pred Neg', 'Pred Pos'])
    cm_df.to_csv(OUTDIR / f"confusion_matrix_{name}.csv")
   # Print classification report
   print(f"\nClassification Report for {name}:")
   print(classification_report(realtest_y, y_pred))
# write master CSVs
pd.concat(roc_rows, ignore_index=True).to_csv(OUTDIR / "all_roc_points.csv",_
 →index=False)
pd.concat(prc_rows, ignore_index=True).to_csv(OUTDIR / "all_prc_points.csv", __
 →index=False)
pd.DataFrame(roc_summary).to_csv(OUTDIR / "roc_auc_summary.csv", index=False)
pd.DataFrame(prc_summary).to_csv(OUTDIR / "prc_auc_summary.csv", index=False)
print("\n Saved:")
for fn in ["all_roc_points.csv", "all_prc_points.csv",
           "roc_auc_summary.csv", "prc_auc_summary.csv"]:
   print(" ", OUTDIR / fn)
# plot confusion matrices
# Create a figure with 4 subplots (2x2 grid) for the 4 models
fig, axes = plt.subplots(2, 2, figsize=(12, 10))
axes = axes.ravel()
print("\n Creating confusion matrices for 4 models:")
for idx, name in enumerate(confusion_matrices.keys()):
    cm = confusion matrices[name]
   print(f" {idx+1}. {name}")
   # Plot confusion matrix
    sns.heatmap(cm, annot=True, fmt='d', cmap='Blues',
                xticklabels=['Negative', 'Positive'],
```

```
yticklabels=['Negative', 'Positive'],
                ax=axes[idx], cbar=True, square=True,
                annot_kws={"size": 12, "weight": "bold"})
   axes[idx].set_title(name, fontsize=12, weight='bold')
   axes[idx].set_xlabel('Predicted Label', fontsize=10)
   axes[idx].set_ylabel('True Label', fontsize=10)
plt.tight layout()
plt.savefig(OUTDIR / "confusion_matrices_all_4models.png", dpi=300,_
 ⇔bbox inches='tight')
plt.show()
print(" All 4 confusion matrices displayed and saved!")
  plot ROC and PR curves
plt.figure(figsize=(12, 5))
roc_ax = plt.subplot(1, 2, 1)
pr_ax = plt.subplot(1, 2, 2)
roc_pts = pd.read_csv(OUTDIR / "all_roc_points.csv")
prc_pts = pd.read_csv(OUTDIR / "all_prc_points.csv")
for name, color in zip([m.__class__._name__ for m in models], colors):
   roc_sub = roc_pts[roc_pts.model == name]
   prc_sub = prc_pts[prc_pts.model == name]
   roc_ax.plot(roc_sub.fpr, roc_sub.tpr, label=name, color=color)
   pr_ax.plot(prc_sub.recall, prc_sub.precision, label=name, color=color)
roc_ax.set_title("ROC curves"); roc_ax.set_xlabel("FPR"); roc_ax.
⇔set ylabel("TPR")
pr_ax.set_title("PR curves");
                                pr_ax.set_xlabel("Recall"); pr_ax.
 ⇔set ylabel("Precision")
for ax in (roc_ax, pr_ax):
   ax.spines["top"].set_visible(False)
   ax.spines["right"].set_visible(False)
   ax.legend(fontsize=7)
plt.tight layout()
plt.savefig(OUTDIR / "roc_pr_curves.png", dpi=300, bbox_inches='tight')
plt.show()
   create individual confusion matrix plots
for model_name, cm in confusion_matrices.items():
   plt.figure(figsize=(6, 5))
```

```
# Create a simple heatmap
    sns.heatmap(cm, annot=True, fmt='d', cmap='Blues',
                xticklabels=['Negative', 'Positive'],
                yticklabels=['Negative', 'Positive'],
                cbar=True, square=True,
                annot_kws={"size": 14})
   plt.title(model_name, fontsize=14, pad=20)
   plt.xlabel('Predicted Label', fontsize=12)
   plt.ylabel('True Label', fontsize=12)
   plt.tight_layout()
   plt.savefig(OUTDIR / f"confusion_matrix_{model_name}.png", dpi=300,__
 ⇔bbox_inches='tight')
   plt.close()
print(f"\n All confusion matrices saved to {OUTDIR}")
print("
         Individual plots and CSVs created for each model")
```

Training LogisticRegression

Classification Report for LogisticRegression:

	precision	recall	f1-score	support
0.0	0.44	0.66	0.53	222
1.0	0.94	0.87	0.90	1427
accuracy			0.84	1649
macro avg	0.69	0.76	0.72	1649
weighted avg	0.87	0.84	0.85	1649

Training SVC

Classification Report for SVC:

	precision	recall	f1-score	support
0.0	0.52	0.64	0.58	222
1.0	0.94	0.91	0.92	1427
accuracy			0.87	1649
macro avg	0.73	0.78	0.75	1649
weighted avg	0.89	0.87	0.88	1649

Training XGBClassifier

Classification Report for XGBClassifier:

precision recall f1-score support

0.0	0.38	0.66	0.48	222
1.0	0.94	0.83	0.88	1427
accuracy			0.81	1649
macro avg	0.66	0.74	0.68	1649
weighted avg	0.86	0.81	0.83	1649

 ${\tt Training}\ {\tt RandomForestClassifier}$

Classification Report for RandomForestClassifier:

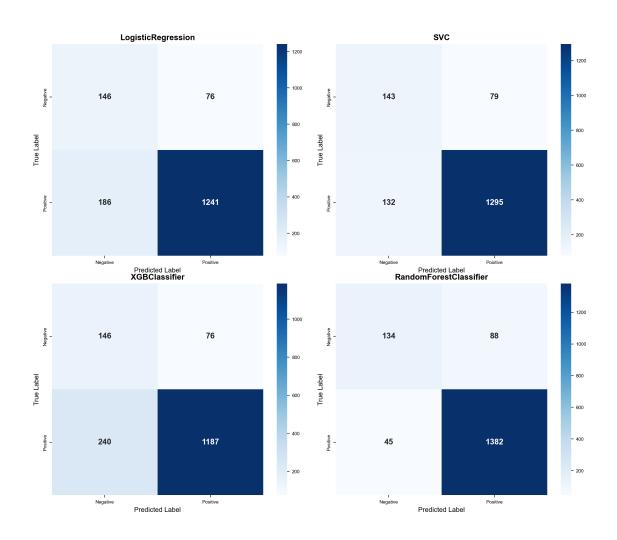
	precision	recall	f1-score	support
0.0	0.75	0.60	0.67	222
1.0	0.94	0.97	0.95	1427
accuracy			0.92	1649
macro avg	0.84	0.79	0.81	1649
weighted avg	0.91	0.92	0.92	1649

Saved:

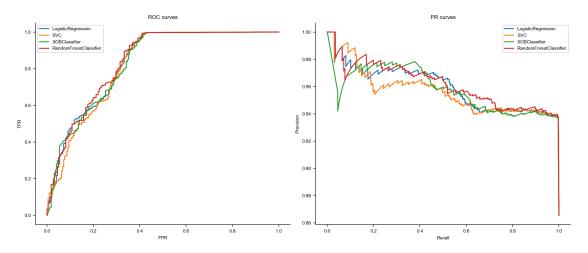
/Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/all_roc_points.csv /Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/all_prc_points.csv /Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/roc_auc_summary.csv /Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/prc_auc_summary.csv

Creating confusion matrices for 4 models:

- 1. LogisticRegression
- 2. SVC
- 3. XGBClassifier
- 4. RandomForestClassifier



All 4 confusion matrices displayed and saved!



All confusion matrices saved to
/Users/haley/Desktop/send_tooo/AAA_final/curve_outputs
Individual plots and CSVs created for each model

```
[86]: #!/usr/bin/env python3
      Save *all* ROC and PR curve points in two master CSVs, plus calculate
      class-specific metrics (F1, AUROC, AUPRC) for male (1) and female (0).
      Focus on Random Forest performance visualization.
            all_roc_points.csv # columns: model, fpr, tpr
            all_prc_points.csv # columns: model, recall, precision
            class_metrics.csv # columns: model, class, f1, auroc, auprc
      ,, ,, ,,
      from pathlib import Path
      import pandas as pd
      import numpy as np
      import matplotlib.pyplot as plt
      from matplotlib import rcParams
      from sklearn.preprocessing import StandardScaler
      from sklearn.metrics import (
          roc_curve,
          auc,
          precision_recall_curve,
          average_precision_score,
          f1_score,
          precision_score,
          recall_score,
          roc_auc_score,
      )
      from sklearn.linear_model import LogisticRegression
      from sklearn.svm import SVC
      from sklearn.ensemble import RandomForestClassifier
      from xgboost import XGBClassifier # pip install xqboost if missing
          global config
      rcParams.update({"font.family": "Arial", "font.size": 7})
      OUTDIR = Path("/Users/haley/Desktop/send tooo/AAA final/curve outputs")
      OUTDIR.mkdir(exist_ok=True)
      colors = ["#1f77b4", "#ff7f0e", "#2ca02c", "#d62728"]
      models = \Gamma
          LogisticRegression(max_iter=1000, random_state=551),
```

```
SVC(kernel="linear", probability=True, random_state=551),
   XGBClassifier(
       random_state=42,
       use_label_encoder=False,
       eval_metric="logloss",
       n_estimators=100,
       learning_rate=0.05,
       max_depth=10,
   RandomForestClassifier(max_depth=10, random_state=41),
1
   scale data once
scaler = StandardScaler()
X_train_s = scaler.fit_transform(full_X_train)
X_test_s = scaler.transform(realtest)
# containers
roc_rows, prc_rows = [], []
roc_summary, prc_summary = [], []
class_metrics = []
   fit, predict, collect points
for model in models:
   name = model.__class__.__name__
   print("Training", name)
   model.fit(X_train_s, full_Y_train)
   y_prob = model.predict_proba(X_test_s)[:, 1]
   y_pred = model.predict(X_test_s)
    # Overall ROC
   fpr, tpr, _ = roc_curve(realtest_y, y_prob)
   roc_auc = auc(fpr, tpr)
   roc_rows.append(pd.DataFrame({"model": name, "fpr": fpr, "tpr": tpr}))
   roc_summary.append({"model": name, "auc": roc_auc})
   # Overall PR
   prec, rec, _ = precision_recall_curve(realtest_y, y_prob)
   pr_auc = average_precision_score(realtest_y, y_prob)
   prc_rows.append(pd.DataFrame({"model": name, "recall": rec, "precision": u
   prc_summary.append({"model": name, "auprc": pr_auc})
   # Class-specific metrics
   for class_label, class_name in [(0, "Female"), (1, "Male")]:
        # Create binary mask for current class
        y_true_binary = (realtest_y == class_label).astype(int)
```

```
y_pred_binary = (y_pred == class_label).astype(int)
        # F1 score for this class
        f1 = f1_score(y_true_binary, y_pred_binary)
        # AUROC for this class (one-vs-rest)
        # For class 0 (female), we need prob of being female = 1 - prob of_{\square}
 ⇔being male
        if class_label == 0:
            y_prob_class = 1 - y_prob
        else:
            y_prob_class = y_prob
        auroc = roc_auc_score(y_true_binary, y_prob_class)
        # AUPRC for this class
        auprc = average_precision_score(y_true_binary, y_prob_class)
        # Store metrics
        class_metrics.append({
            "model": name,
            "class": class name,
            "class_label": class_label,
            "f1": f1,
            "auroc": auroc,
            "auprc": auprc,
            "precision": precision_score(y_true_binary, y_pred_binary),
            "recall": recall_score(y_true_binary, y_pred_binary)
        })
# write master CSVs
pd.concat(roc_rows, ignore_index=True).to_csv(OUTDIR / "all_roc_points.csv", __
 →index=False)
pd.concat(prc rows, ignore index=True).to csv(QUTDIR / "all prc points.csv", |

    index=False)

pd.DataFrame(roc_summary).to_csv(OUTDIR / "roc_auc_summary.csv", index=False)
pd.DataFrame(prc_summary).to_csv(OUTDIR / "prc_auc_summary.csv", index=False)
# Save class-specific metrics
class_metrics_df = pd.DataFrame(class_metrics)
class_metrics_df.to_csv(OUTDIR / "class_metrics.csv", index=False)
print(" Saved:")
for fn in ["all_roc_points.csv", "all_prc_points.csv",
           "roc_auc_summary.csv", "prc_auc_summary.csv", "class_metrics.csv"]:
    print(" ", OUTDIR / fn)
```

```
# Create 3 separate bar plots for Random Forest only
# Filter for Random Forest only
rf_metrics = class_metrics_df[class_metrics_df["model"] ==__

¬"RandomForestClassifier"]

metrics = ["f1", "auroc", "auprc"]
metric_names = ["F1 Score", "AUROC", "AUPRC"]
gender_colors = {"Female": "#FF69B4", "Male": "#4169E1"} # Pink for female,
 ⇒blue for male
# Create 3 separate figures
for metric, metric_name in zip(metrics, metric_names):
   fig, ax = plt.subplots(figsize=(8, 6))
   # Prepare data
   x = np.arange(2) # Two bars: Female and Male
   values = []
   labels = []
   colors_list = []
   for class_name in ["Female", "Male"]:
       value = rf metrics[rf metrics["class"] == class_name][metric].values[0]
       values.append(value)
       labels.append(class_name)
        colors_list.append(gender_colors[class_name])
    # Create bars
   bars = ax.bar(x, values, color=colors_list, width=0.6)
   # Add value labels on bars
   for i, (bar, v) in enumerate(zip(bars, values)):
        ax.text(bar.get_x() + bar.get_width()/2, v + 0.01,
                f'{v:.3f}', ha='center', va='bottom', fontsize=12,

¬fontweight='bold')
    # Customize plot
   ax.set_title(f"Random Forest {metric_name} by Gender", fontsize=16,__

→fontweight='bold')
   ax.set_xlabel("Gender", fontsize=14)
   ax.set_ylabel(metric_name, fontsize=14)
   ax.set_xticks(x)
   ax.set_xticklabels(labels, fontsize=12)
   ax.set_ylim(0.5, 1.1)
   ax.grid(True, alpha=0.3, axis='y')
    # Remove top and right spines
```

```
ax.spines["top"].set_visible(False)
   ax.spines["right"].set_visible(False)
    # Make y-axis labels larger
   ax.tick_params(axis='y', labelsize=11)
   plt.tight_layout()
   plt.savefig(OUTDIR / f"rf_{metric}_by_gender.png", dpi=300,_
 ⇔bbox_inches='tight')
   plt.show()
# Print Random Forest summary
print("\n Random Forest Class-Specific Metrics Summary:")
print("=" * 60)
print(rf_metrics[["class", "f1", "auroc", "auprc", "precision", "recall"]].
⇔to_string(index=False))
# Also create combined visualization
fig, axes = plt.subplots(1, 3, figsize=(15, 5))
fig.suptitle("Random Forest Performance by Gender", fontsize=16, ___

¬fontweight='bold')
for ax, (metric, metric_name) in zip(axes, zip(metrics, metric_names)):
    # Prepare data
   x = np.arange(2)
   values = []
   labels = []
   colors_list = []
   for class_name in ["Female", "Male"]:
        value = rf_metrics[rf_metrics["class"] == class_name][metric].values[0]
       values.append(value)
       labels.append(class_name)
        colors_list.append(gender_colors[class_name])
    # Create bars
   bars = ax.bar(x, values, color=colors_list, width=0.6)
    # Add value labels
   for i, (bar, v) in enumerate(zip(bars, values)):
        ax.text(bar.get_x() + bar.get_width()/2, v + 0.01,
                f'{v:.3f}', ha='center', va='bottom', fontsize=11,

¬fontweight='bold')
    # Customize
   ax.set_title(metric_name, fontsize=14)
   ax.set_xlabel("Gender", fontsize=12)
```

```
ax.set_ylabel(metric_name, fontsize=12)
    ax.set_xticks(x)
    ax.set_xticklabels(labels, fontsize=11)
    ax.set_ylim(0.5, 1.1)
    ax.grid(True, alpha=0.3, axis='y')
    ax.spines["top"].set_visible(False)
    ax.spines["right"].set_visible(False)
plt.tight layout()
plt.savefig(OUTDIR / "rf_all_metrics_combined.png", dpi=300,_
 ⇔bbox inches='tight')
plt.show()
    Original ROC and PR curves (all models)
plt.figure(figsize=(12, 5))
roc_ax = plt.subplot(1, 2, 1)
pr_ax = plt.subplot(1, 2, 2)
roc_pts = pd.read_csv(OUTDIR / "all_roc_points.csv")
prc_pts = pd.read_csv(OUTDIR / "all_prc_points.csv")
for name, color in zip([m.__class__._name__ for m in models], colors):
    roc_sub = roc_pts[roc_pts.model == name]
    prc_sub = prc_pts[prc_pts.model == name]
    roc_ax.plot(roc_sub.fpr, roc_sub.tpr, label=name, color=color)
    pr_ax.plot(prc_sub.recall, prc_sub.precision, label=name, color=color)
roc_ax.set_title("ROC curves"); roc_ax.set_xlabel("FPR"); roc_ax.

set_ylabel("TPR")

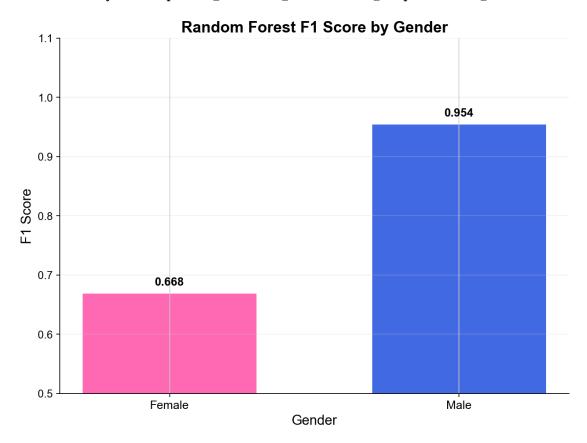
⇔set_ylabel("Precision")
for ax in (roc_ax, pr_ax):
    ax.spines["top"].set_visible(False)
    ax.spines["right"].set_visible(False)
    ax.legend(fontsize=7)
plt.tight layout()
plt.savefig(OUTDIR / "roc_pr_curves.png", dpi=300, bbox_inches='tight')
plt.show()
Training LogisticRegression
Training SVC
Training XGBClassifier
Training RandomForestClassifier
/Users/haley/Library/Python/3.9/lib/python/site-packages/xgboost/core.py:158:
UserWarning: [17:32:31] WARNING:
```

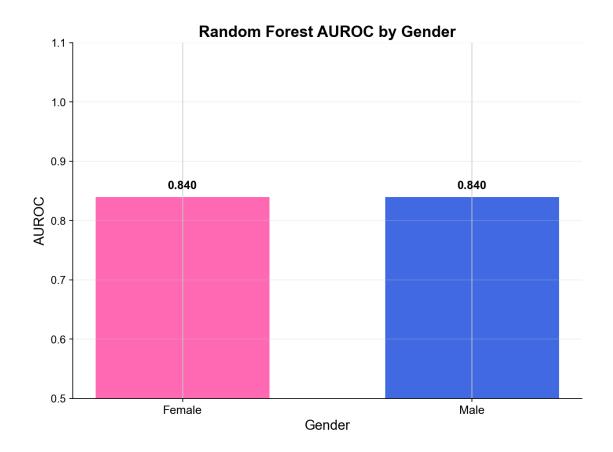
/Users/runner/work/xgboost/xgboost/src/learner.cc:740: Parameters: { "use_label_encoder" } are not used.

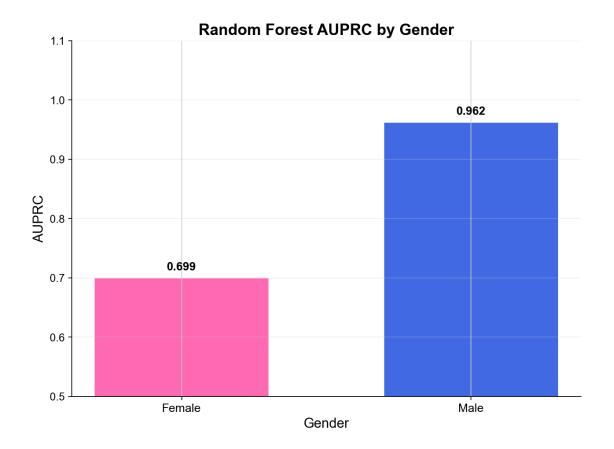
warnings.warn(smsg, UserWarning)

Saved:

/Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/all_roc_points.csv /Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/all_prc_points.csv /Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/roc_auc_summary.csv /Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/prc_auc_summary.csv /Users/haley/Desktop/send_tooo/AAA_final/curve_outputs/class_metrics.csv

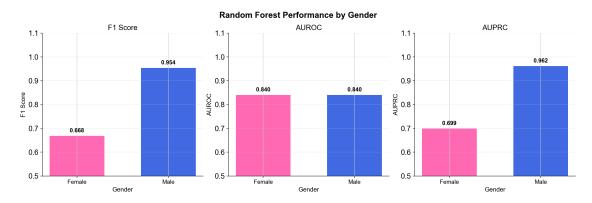


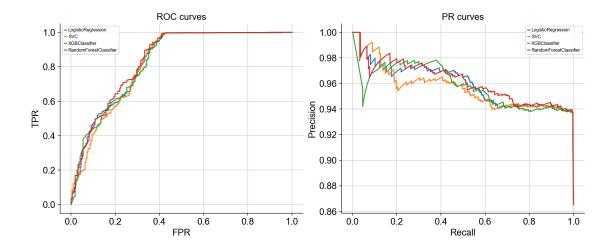




Random Forest Class-Specific Metrics Summary:

class f1 auroc auprc precision recall
Female 0.668329 0.839839 0.698761 0.748603 0.603604
Male 0.954090 0.839839 0.961600 0.940136 0.968465



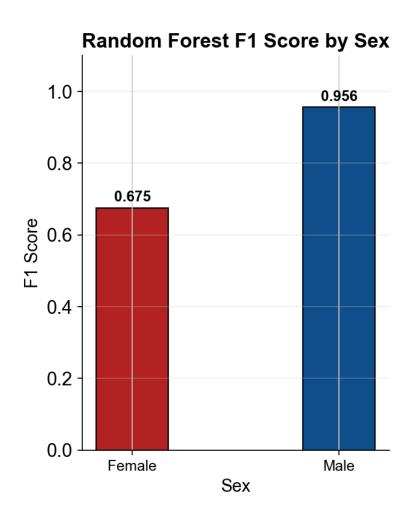


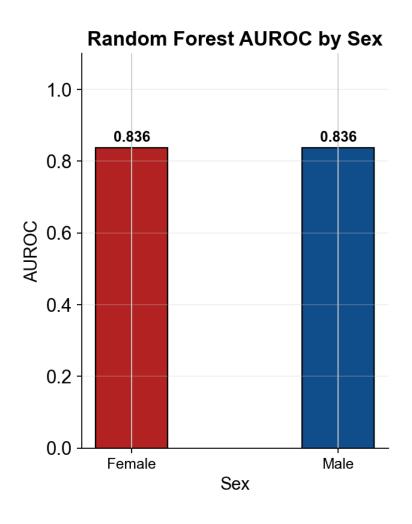
```
[95]: #!/usr/bin/env python3
     Random-Forest (seed=97; 600×depth10×leaf2)
       overall / female AUROC & AUPRC
         (Sex, , Y 0)
       full_X_train, full_Y_train, realtest, realtest_y
     from pathlib import Path
     import numpy as np, pandas as pd
     import matplotlib.pyplot as plt
     from sklearn.preprocessing import StandardScaler
     from sklearn.metrics import (precision_score, recall_score, f1_score,
                                 roc_auc_score, average_precision_score)
     from sklearn.ensemble import RandomForestClassifier
     # ======= 1. =======
     # +++
      # full_X_train, full_Y_train, realtest, realtest_y
     # 111
     scaler
              = StandardScaler()
     X_train_s = scaler.fit_transform(full_X_train)
     X_test_s = scaler.transform(realtest)
     y_train = full_Y_train.astype(int)
                = realtest_y.astype(int)
     y_test
     # ======= 2. RF =======
     rf = RandomForestClassifier(
             n_{estimators} = 600,
             max_depth
                            = 10,
```

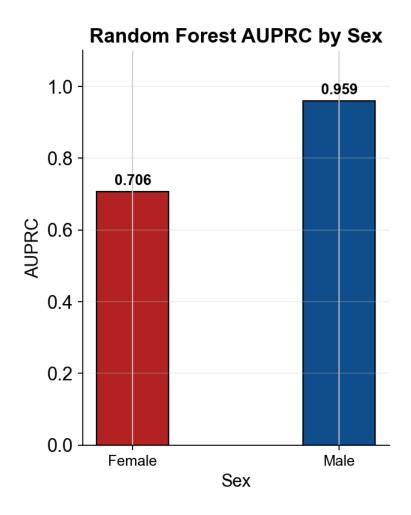
```
min_samples_leaf = 2,
       random_state = 97,
       n jobs
                        = -1)
rf.fit(X_train_s, y_train)
# ======= 3. & =======
prob_male = rf.predict_proba(X_test_s)[:, 1]
prob_fem = 1 - prob_male
       = (prob_male >= 0.5).astype(int)
y pred
prec_f = precision_score(y_test, y_pred, pos_label=0)
rec_f = recall_score (y_test, y_pred, pos_label=0)
f1_f = f1_score
                       (y_test, y_pred, pos_label=0)
overall_auc = roc_auc_score(y_test, prob_male)
overall_aupr = average_precision_score(y_test, prob_male)
           = roc_auc_score((y_test==0).astype(int), prob_fem)
fem_auc
            = average_precision_score((y_test==0).astype(int), prob_fem)
fem_aupr
print("\n====
                ====")
print(f"Overall AUROC={overall_auc:.3f} AUPRC={overall_aupr:.3f}")
print(f"Female AUROC={fem_auc:.3f} AUPRC={fem_aupr:.3f}")
print("\n==== (0.5) Female ====")
print(f"Precision={prec_f:.3f} Recall={rec_f:.3f} F1={f1_f:.3f}")
# ====== 4.
OUTDIR = Path("./curve_outputs"); OUTDIR.mkdir(exist_ok=True)
gender_colors = {"Female": "#B22222", "Male": "#104E8B"}
metrics = {
    "F1 Score": (f1_f,
                f1_score(y_test, y_pred, pos_label=1)),
    "AUROC":
                (fem_auc, overall_auc),
    "AUPRC":
               (fem_aupr, overall_aupr)
}
for metric_name, (val_f, val_m) in metrics.items():
   fig, ax = plt.subplots(figsize=(4, 5))
   bars = ax.bar([0, 1], [val_f, val_m], width=0.35,
                 color=[gender_colors["Female"], gender_colors["Male"]],
                 edgecolor="black")
   for bar, v in zip(bars, [val_f, val_m]):
       ax.text(bar.get_x()+bar.get_width()/2, v+0.01,
               f"{v:.3f}", ha="center", va="bottom",
               fontsize=11, weight="bold")
```

```
ax.set_xticks([0, 1])
   ax.set_xticklabels(["Female", "Male"], fontsize=11)
   ax.set_xlabel("Sex", fontsize=13)
   ax.set_ylabel(metric_name, fontsize=13)
   ax.set_title(f"Random Forest {metric_name} by Sex",
                 fontsize=15, weight="bold")
                                   # ← Y O
   ax.set_ylim(0, 1.1)
   ax.grid(axis="y", alpha=0.3)
   ax.spines[['top','right']].set_visible(False)
   plt.tight_layout()
   plt.savefig(OUTDIR / f"rf_{metric_name.replace(' ','_').lower()}_by_sex.
 ⇒png",
                dpi=300, bbox_inches="tight")
   plt.show()
               {OUTDIR.resolve()}")
print(f"\n
```

```
==== 0verall AUROC=0.836 AUPRC=0.959
Female AUROC=0.836 AUPRC=0.706
==== (0.5) Female ====
Precision=0.773 Recall=0.599 F1=0.675
```







/Users/haley/Desktop/send_tooo/curve_outputs

```
[97]: #!/usr/bin/env python3
      n n n
      Train 4 models \rightarrow 0.5
                            → Female/Male
         True_Counts
        pred_counts.csv
      11 11 11
      from pathlib import Path
      import numpy as np, pandas as pd
      from sklearn.preprocessing import StandardScaler
      from sklearn.metrics import f1_score, roc_auc_score, average_precision_score
      from sklearn.linear_model import LogisticRegression
      from sklearn.svm
                                  import SVC
      from sklearn.ensemble
                                  import RandomForestClassifier
      from xgboost
                                  import XGBClassifier
```

```
# 111
# full_X_train, full_Y_train, realtest, realtest_y
# 111
scaler
        = StandardScaler()
X_train_s = scaler.fit_transform(full_X_train)
X_test_s = scaler.transform(realtest)
y_train = full_Y_train.astype(int)
        = realtest_y.astype(int)
y_test
# =======
              _____
model dict = {
   "LogReg":
                  LogisticRegression(max_iter=1000, random_state=42),
                  SVC(kernel="linear", probability=True, random_state=42),
   "LinearSVC":
   "XGBoost":
                  XGBClassifier(
                      eval_metric="logloss", random_state=42,
                      n_estimators=100, learning_rate=0.05, max_depth=10),
   "RandomForest": RandomForestClassifier(max_depth=10, random_state=42)
}
rows = []
# ----- True Counts -----
rows.append({
   "Model": "True Counts",
   "Female": int((y_test == 0).sum()),
   "Male": int((y test == 1).sum()),
   "AUROC": np.nan,
   "AUPRC":
             np.nan,
   "F1":
             np.nan
})
# -----
for name, clf in model_dict.items():
   clf.fit(X_train_s, y_train)
   prob_male = clf.predict_proba(X_test_s)[:, 1]
           = (prob_male >= 0.5).astype(int)
   y_pred
   rows.append({
       "Model":
                 name,
       "Female": int((y_pred == 0).sum()),
       "Male": int((y_pred == 1).sum()),
       "AUROC": roc_auc_score(y_test, prob_male),
       "AUPRC": average_precision_score(y_test, prob_male),
       "F1": f1_score(y_test, y_pred)
   })
```

```
# =======
                    _____
             = pd.DataFrame(rows)
    out_path = Path("./pred_counts.csv")
    out_df.to_csv(out_path, index=False)
    print(out_df.to_string(index=False, float_format="%.3f"))
    print(f"\n
                    {out_path.resolve()}")
           Model Female Male AUROC AUPRC
                                                F1
     True Counts
                     222 1427
                                  NaN
                                               NaN
                                         NaN
          LogReg
                     332 1317 0.833 0.960 0.905
       LinearSVC
                     297
                          1352 0.823 0.957 0.918
         XGBoost
                     386 1263 0.830 0.957 0.883
    RandomForest
                     208
                         1441 0.837 0.959 0.946
          /Users/haley/Desktop/send_tooo/pred_counts.csv
[]:
[]:
[]:
    !pip install lightgbm scikit-learn
    Defaulting to user installation because normal site-packages is not writeable
    Collecting lightgbm
      Downloading lightgbm-4.6.0-py3-none-macosx_12_0_arm64.whl.metadata (17 kB)
    Requirement already satisfied: scikit-learn in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (1.5.1)
    Requirement already satisfied: numpy>=1.17.0 in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (from lightgbm)
    (1.26.4)
    Requirement already satisfied: scipy in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (from lightgbm)
    Requirement already satisfied: joblib>=1.2.0 in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (from scikit-learn)
    Requirement already satisfied: threadpoolctl>=3.1.0 in
    /Users/haley/Library/Python/3.9/lib/python/site-packages (from scikit-learn)
    (3.5.0)
    Downloading lightgbm-4.6.0-py3-none-macosx_12_0_arm64.whl (1.6 MB)
                             1.6/1.6 MB
    24.4 MB/s eta 0:00:00
    Installing collected packages: lightgbm
    Successfully installed lightgbm-4.6.0
```

```
[notice] A new release of pip is
     available: 25.0.1 -> 25.1.1
     [notice] To update, run:
     /Applications/Xcode.app/Contents/Developer/usr/bin/python3 -m pip
     install --upgrade pip
[30]: roc data
[30]:
                             Model
                                         fpr
                                                   tpr
                                                             auc
      0
                LogisticRegression 0.000000 0.000000
                                                        0.832816
      1
                LogisticRegression
                                    0.000000 0.000701
                                                        0.832816
      2
                LogisticRegression
                                    0.000000 0.030834 0.832816
                LogisticRegression
                                    0.004505 0.030834 0.832816
                                    0.004505 0.065172 0.832816
      4
                LogisticRegression
      999
           RandomForestClassifier 0.963964 1.000000 0.839839
      1000 RandomForestClassifier 0.968468 1.000000 0.839839
      1001 RandomForestClassifier 0.977477 1.000000 0.839839
      1002 RandomForestClassifier 0.990991 1.000000
                                                        0.839839
      1003 RandomForestClassifier 1.000000 1.000000 0.839839
      [1004 rows x 4 columns]
[82]: #!/usr/bin/env python3
      Train Random Forest on full train set, test on full test set.
      Print true and predicted label counts (female/male).
      11 11 11
      import pandas as pd
      import numpy as np
      from sklearn.ensemble import RandomForestClassifier
         model setup
      rf_model = RandomForestClassifier(max_depth=10, random_state=41)
          train on full training set
      print("Training Random Forest...")
      rf_model.fit(full_X_train, full_Y_train)
         predict on full test set
      print("Making predictions...")
      y_pred = rf_model.predict(realtest)
          count true labels
      true female count = np.sum(realtest y == 0)
      true_male_count = np.sum(realtest_y == 1)
```

```
# count predicted labels
     pred_female_count = np.sum(y_pred == 0)
     pred_male_count = np.sum(y_pred == 1)
     # print results
     print("\n" + "="*50)
     print("TRUE LABEL COUNTS:")
     print(f" Female (0): {true_female_count}")
     print(f" Male (1): {true_male_count}")
                          {true_female_count + true_male_count}")
     print(f" Total:
     print("\nPREDICTED LABEL COUNTS:")
     print(f" Female (0): {pred_female_count}")
     print(f" Male (1): {pred_male_count}")
     print(f" Total:
                       {pred_female_count + pred_male_count}")
     print("="*50)
         optional: accuracy
     accuracy = np.mean(realtest_y == y_pred)
     print(f"\nAccuracy: {accuracy:.4f}")
     Training Random Forest...
     Making predictions...
     _____
     TRUE LABEL COUNTS:
       Female (0): 222
      Male (1): 1427
      Total:
                  1649
     PREDICTED LABEL COUNTS:
      Female (0): 179
      Male (1): 1470
       Total:
                  1649
     Accuracy: 0.9193
[85]: #!/usr/bin/env python3
     Draw a confusion matrix for the Random Forest classification results.
     import matplotlib.pyplot as plt
     import seaborn as sns
     import numpy as np
```

```
from sklearn.metrics import confusion_matrix, classification_report
# create confusion matrix
cm = confusion_matrix(realtest_y, y_pred)
# plot confusion matrix
plt.figure(figsize=(8, 6))
# Plot heatmap with counts only
sns.heatmap(cm,
           annot=True,
           fmt='d',
           cmap='Blues',
           cbar=True,
           square=True,
           xticklabels=['Female ', 'Male '],
           yticklabels=['Female ', 'Male '],
           annot_kws={'size': 14})
plt.title('Confusion Matrix - Random Forest', fontsize=14, pad=20)
plt.xlabel('Predicted Label', fontsize=12)
plt.ylabel('True Label', fontsize=12)
plt.tight_layout()
   print detailed metrics
print("\nCONFUSION MATRIX:")
print(cm)
print(f"\nTrue Negatives (TN): {cm[0,0]}")
print(f"False Positives (FP): {cm[0,1]}")
print(f"False Negatives (FN): {cm[1,0]}")
print(f"True Positives (TP): {cm[1,1]}")
   calculate additional metrics
tn, fp, fn, tp = cm.ravel()
accuracy = (tp + tn) / (tp + tn + fp + fn)
precision = tp / (tp + fp) if (tp + fp) > 0 else 0
recall = tp / (tp + fn) if (tp + fn) > 0 else 0
specificity = tn / (tn + fp) if (tn + fp) > 0 else 0
⇔recall) > 0 else 0
print(f"\nMETRICS:")
print(f"Accuracy: {accuracy:.4f}")
print(f"Precision: {precision:.4f}")
print(f"Recall: {recall:.4f}")
print(f"Specificity: {specificity:.4f}")
print(f"F1-Score: {f1_score:.4f}")
```

CONFUSION MATRIX:

[[134 88] [45 1382]]

True Negatives (TN): 134
False Positives (FP): 88
False Negatives (FN): 45
True Positives (TP): 1382

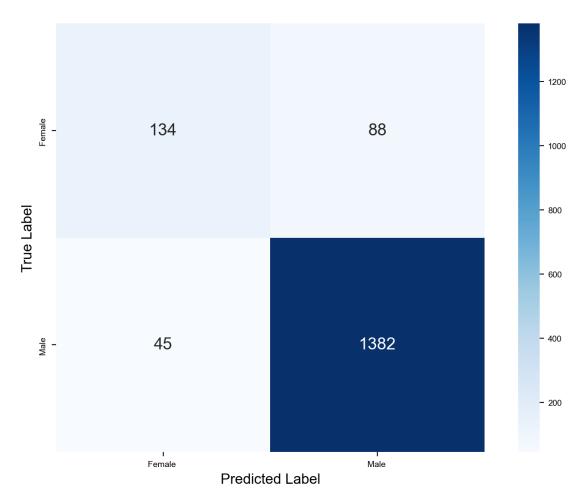
METRICS:

Accuracy: 0.9193
Precision: 0.9401
Recall: 0.9685
Specificity: 0.6036
F1-Score: 0.9541

CLASSIFICATION REPORT:

	precision	recall	II-score	support
Female (0) Male (1)	0.7486 0.9401	0.6036 0.9685	0.6683 0.9541	222 1427
accuracy macro avg	0.8444	0.7860	0.9193	1649 1649
weighted avg	0.9144	0.9193	0.9156	1649

Confusion Matrix - Random Forest



12 fig1 drawing mouse

```
[452]: import matplotlib.pyplot as plt
import numpy as np

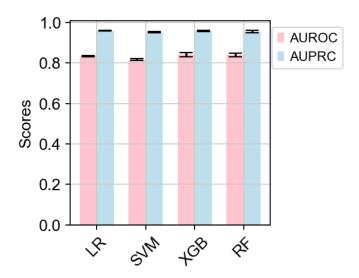
# ROC AUC Data
auroc_data = {
    'LogisticRegression': [0.833374, 0.830398, 0.835730],
    'SVC': [0.816254, 0.811984, 0.821324],
    'XGBClassifier': [0.842595, 0.831601, 0.851632],
    'RandomForestClassifier': [0.839546, 0.829966, 0.849857]
}

# PRC AUC Data
```

```
auprc_data = {
    'LogisticRegression': [0.959603, 0.958579, 0.960437],
    'SVC': [0.951278, 0.949247, 0.954626],
    'XGBClassifier': [0.959140, 0.955415, 0.962364],
    'RandomForestClassifier': [0.957209, 0.950502, 0.961562]
}
# Changing the model names
new names = {
    'LogisticRegression': 'LR',
    'SVC': 'SVM',
    'XGBClassifier': 'XGB',
    'RandomForestClassifier': 'RF'
}
auroc_data = {new_names.get(k, k): v for k, v in auroc_data.items()}
auprc_data = {new_names.get(k, k): v for k, v in auprc_data.items()}
# Prepare data for plotting
model_names = list(auroc_data.keys())
auroc_means = [auroc_data[name][0] for name in model_names]
auroc_errors = [(auroc_data[name][0] - auroc_data[name][1], auroc_data[name][2]_u
 →- auroc_data[name][0]) for name in model_names]
auprc_means = [auprc_data[name][0] for name in model_names]
auprc_errors = [(auprc_data[name][0] - auprc_data[name][1], auprc_data[name][2]_
 → auprc_data[name][0]) for name in model_names]
# Plot settings
bar_width = 0.35
opacity = 0.8
error_config = {'capsize': 5}
# Creating the bar plot with the original figure size
fig, ax = plt.subplots(figsize=(4, 2.83)) # Set the figure size here as_
 \rightarrowprovided
bar1 = ax.bar(np.arange(len(model_names)) - bar_width/2, auroc_means, bar_width,
              alpha=opacity, color='lightpink', yerr=np.array(auroc_errors).T,
              error_kw=error_config, label='AUROC')
bar2 = ax.bar(np.arange(len(model_names)) + bar_width/2, auprc_means, bar_width,
              alpha=opacity, color='lightblue', yerr=np.array(auprc_errors).T,
              error_kw=error_config, label='AUPRC')
# Add some text for labels and axes ticks
ax.set_ylabel('Scores', fontsize=12)
ax.set_xticks(np.arange(len(model_names)))
ax.set_xticklabels(model_names, fontsize=12)
```

```
ax.tick_params(axis='y', labelsize=12)
ax.legend(loc='upper right', bbox_to_anchor=(1.4,1), fontsize=10)

plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

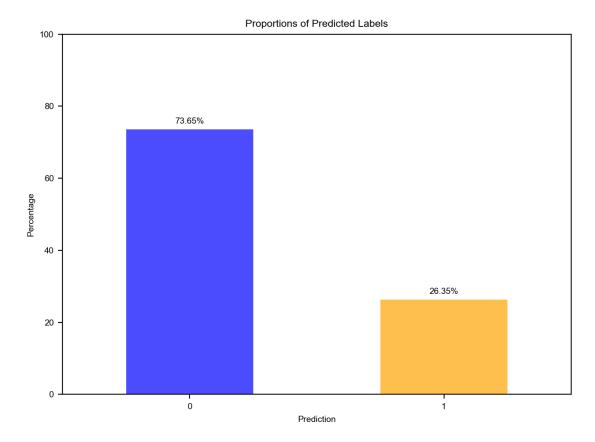


13 wtwt experimental(test data performance)

```
[48]: adata1
[48]: View of AnnData object with n_obs \times n_vars = 8268 \times 20902
          obs: 'dataset', 'sample', 'n_genes', 'n_genes_by_counts', 'total_counts',
      'total_counts_mt', 'pct_counts_mt', 'conditions'
          var: 'gene_ids', 'feature_types', 'n_cells', 'mt', 'n_cells_by_counts',
      'mean_counts', 'pct_dropout_by_counts', 'total_counts', 'highly_variable',
      'means', 'dispersions', 'dispersions_norm'
          uns: 'hvg', 'log1p'
[35]: import numpy as np
      import pandas as pd
      import matplotlib.pyplot as plt
      from matplotlib import rcParams
      from xgboost import XGBClassifier
      from sklearn.preprocessing import StandardScaler
      from sklearn.metrics import roc_auc_score, precision_recall_curve,_
       →average_precision_score, roc_curve, auc
      import matplotlib.pyplot as plt
```

```
plt.style.use("default")
# Set global font properties
rcParams['font.family'] = 'Arial'
rcParams['font.size'] = 7
rcParams['font.weight'] = 'normal'
# Define model
# model = RandomForestClassifier(max depth=10, random state=41)
model = RandomForestClassifier(max_depth=10, random_state=41)
np.random.seed(42) # For reproducibility
# Scale the training data
scaler = StandardScaler()
full_X_train_scaled = scaler.fit_transform(full_X_train)
# Scale the test data
adata1_scaled = scaler.transform(extracted_columns_1)
# Fit the model on the training data
model.fit(full_X_train_scaled, full_Y_train)
# Predict labels for adata1
predicted_labels = model.predict(adata1_scaled)
# Save predictions to a CSV file
output_df = pd.DataFrame({"Prediction": predicted_labels})
output_df.to_csv("predicted_labels_jun16.csv", index=False)
print("Predicted labels saved to 'predicted_labels_jun16.csv'.")
# Calculate percentages for the bar plot
label_counts = pd.Series(predicted_labels).value_counts(normalize=True) * 100 _
 →# Convert to percentages
# Create a bar plot
label_counts.plot(kind='bar', color=['blue', 'orange'], alpha=0.7)
plt.xlabel("Prediction")
plt.ylabel("Percentage")
plt.title("Proportions of Predicted Labels")
plt.xticks([0, 1], labels=['0', '1'], rotation=0)
plt.ylim(0, 100) # Ensure the y-axis ranges from 0 to 100
plt.tight_layout()
# Add percentage labels on the bars
for i, v in enumerate(label counts):
```

Predicted labels saved to 'predicted_labels_jun16.csv'.



Bar plot of predicted label percentages saved to 'prediction_percentage_barplot.png'.

```
[46]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from matplotlib import rcParams
from sklearn.ensemble import RandomForestClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.svm import LinearSVC
```

```
from xgboost import XGBClassifier
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import roc_auc_score, precision_recall_curve, u
 ⇒average_precision_score, roc_curve, auc
from sklearn.model_selection import cross_val_score, cross_val_predict
from sklearn.metrics import accuracy score, precision score, recall score,
 ⊶f1 score
import os
import warnings
warnings.filterwarnings('ignore')
# Set style and global font properties
plt.style.use("default")
rcParams['font.family'] = 'Arial'
rcParams['font.size'] = 7
rcParams['font.weight'] = 'normal'
# Create output directory
output_dir = "/Users/haley/Desktop/send_tooo/AAA_final/wtwt"
os.makedirs(output_dir, exist_ok=True)
print(f"Output directory created/verified: {output_dir}")
# Define number of cores for parallel processing
n_cores = -1  # Use all available cores
# Define models
models = {
    "LogisticRegression": LogisticRegression(
       max_iter=1000,
       n_jobs=n_cores,
       random_state=41
   ),
    "LinearSVC": LinearSVC(
        dual=False, # Better for many features
       random_state=41
   ),
    "XGBClassifier": XGBClassifier(
        eval_metric='logloss',
       n_estimators=100,
       learning_rate=0.05,
       max_depth=10,
       n_jobs=n_cores,
       random_state=41
   ),
    "RandomForestClassifier": RandomForestClassifier(
       max_depth=10,
       n_jobs=n_cores,
```

```
random_state=66
   ),
}
# Scale the training data
scaler = StandardScaler()
full_X_train_scaled = scaler.fit_transform(full_X_train)
# Scale the test data
adata1_scaled = scaler.transform(extracted_columns_1)
# Dictionary to store results
results = {}
predictions = {}
probabilities = {}
cv_scores = {}
# Train each model and make predictions
print("\nTraining models and making predictions...\n")
for model_name, model in models.items():
   print(f"Training {model_name}...")
    # Fit the model
   model.fit(full_X_train_scaled, full_Y_train)
    # Make predictions
   predicted_labels = model.predict(adata1_scaled)
   predictions[model_name] = predicted_labels
    # Get prediction probabilities (if available)
    if hasattr(model, "predict_proba"):
       predicted_probs = model.predict_proba(adata1_scaled)[:, 1]
   elif hasattr(model, "decision_function"):
        # For SVM, convert decision function to probabilities using sigmoid
        decision_values = model.decision_function(adata1_scaled)
       predicted_probs = 1 / (1 + np.exp(-decision_values))
    else:
        predicted_probs = predicted_labels.astype(float)
   probabilities[model_name] = predicted_probs
    # Perform cross-validation on training data to get performance metrics
   cv_accuracy = cross_val_score(model, full_X_train_scaled, full_Y_train,_
 ⇔cv=5, scoring='accuracy')
    cv_precision = cross_val_score(model, full_X_train_scaled, full_Y_train,_u
 ⇔cv=5, scoring='precision')
```

```
cv_recall = cross_val_score(model, full_X_train_scaled, full_Y_train, cv=5,__
 ⇔scoring='recall')
    cv_f1 = cross_val_score(model, full_X_train_scaled, full_Y_train, cv=5,_
 ⇔scoring='f1')
    cv_roc_auc = cross_val_score(model, full_X_train_scaled, full_Y_train,_
 ⇔cv=5, scoring='roc_auc')
    # Store cross-validation scores
    cv_scores[model_name] = {
        'accuracy': cv_accuracy,
        'precision': cv_precision,
        'recall': cv_recall,
        'f1': cv_f1,
        'roc_auc': cv_roc_auc
    }
    # Store results
    results[model_name] = {
        'predictions': predicted_labels,
        'probabilities': predicted_probs,
        'cv_accuracy': cv_accuracy.mean(),
        'cv_precision': cv_precision.mean(),
        'cv recall': cv recall.mean(),
        'cv_f1': cv_f1.mean(),
        'cv_roc_auc': cv_roc_auc.mean()
    }
    # Save predictions and probabilities to CSV in the specified folder
    output_df = pd.DataFrame({
        "Prediction": predicted_labels,
        "Probability": predicted_probs
    })
    output_file = os.path.join(output_dir,__

¬f"predicted_labels_probs_{model_name}_jun16.csv")

    output df.to csv(output file, index=False)
    print(f" - Predictions and probabilities saved to '{output_file}'")
    print(f" - CV Accuracy: {cv_accuracy.mean():.4f} (+/- {cv_accuracy.std() *_
 \Rightarrow 2:.4f)")
    print(f" - CV ROC-AUC: {cv_roc_auc.mean():.4f} (+/- {cv_roc_auc.std() * 2:.
 \hookrightarrow 4f)\n")
# Create performance comparison table
print("\nModel Performance Summary (Cross-Validation):")
print("-" * 70)
performance_df = pd.DataFrame({
    model_name: {
```

```
'Accuracy': f"{results[model_name]['cv_accuracy']:.4f}",
        'Precision': f"{results[model_name]['cv_precision']:.4f}",
        'Recall': f"{results[model_name]['cv_recall']:.4f}",
        'F1-Score': f"{results[model_name]['cv_f1']:.4f}",
        'ROC-AUC': f"{results[model_name]['cv_roc_auc']:.4f}"
   } for model_name in models.keys()
}).T
print(performance_df)
# Save performance summary to CSV in the output folder
performance_file = os.path.join(output_dir, "model_performance_summary.csv")
performance_df.to_csv(performance_file)
print(f"\nPerformance summary saved to '{performance_file}'")
# Create visualizations
# 1. Bar plot comparing prediction distributions
fig, axes = plt.subplots(2, 2, figsize=(10, 8))
axes = axes.ravel()
for idx, (model_name, preds) in enumerate(predictions.items()):
   ax = axes[idx]
   label_counts = pd.Series(preds).value_counts(normalize=True) * 100
   bars = ax.bar([0, 1], label_counts.sort_index(), color=['blue', 'orange'],
 ⇒alpha=0.7)
   ax.set_xlabel("Prediction")
   ax.set_ylabel("Percentage")
   ax.set_title(f"{model_name} - Predicted Labels")
   ax.set_xticks([0, 1])
   ax.set_xticklabels(['0', '1'])
   ax.set_ylim(0, 100)
   # Add percentage labels on bars
   for bar, count in zip(bars, label counts.sort index()):
       height = bar.get_height()
        ax.text(bar.get_x() + bar.get_width()/2., height + 1,
                f'{count:.1f}%', ha='center', va='bottom', fontsize=7)
plt.tight_layout()
plt.savefig(os.path.join(output_dir, "all_models_prediction_distributions.")
 →png"), dpi=300)
plt.show()
# 2. Performance metrics comparison plot
metrics = ['cv_accuracy', 'cv_precision', 'cv_recall', 'cv_f1', 'cv_roc_auc']
metric_names = ['Accuracy', 'Precision', 'Recall', 'F1-Score', 'ROC-AUC']
```

```
fig, ax = plt.subplots(figsize=(10, 6))
x = np.arange(len(metric_names))
width = 0.2
for i, model_name in enumerate(models.keys()):
    values = [results[model_name][metric] for metric in metrics]
    ax.bar(x + i*width, values, width, label=model_name)
ax.set xlabel('Metrics')
ax.set_ylabel('Score')
ax.set_title('Model Performance Comparison (Cross-Validation)')
ax.set_xticks(x + width * 1.5)
ax.set_xticklabels(metric_names)
ax.legend()
ax.set_ylim(0, 1.1)
# Add value labels on bars
for i, model_name in enumerate(models.keys()):
    values = [results[model_name] [metric] for metric in metrics]
    for j, v in enumerate(values):
        ax.text(j + i*width, v + 0.01, f'\{v:.3f\}', ha='center', va='bottom', 

→fontsize=6)
plt.tight_layout()
plt.savefig(os.path.join(output_dir, "model_performance_comparison.png"),
 →dpi=300)
plt.show()
# 3. Box plot for cross-validation scores
fig, axes = plt.subplots(1, 5, figsize=(15, 4))
metrics_for_box = ['accuracy', 'precision', 'recall', 'f1', 'roc_auc']
metric_names_for_box = ['Accuracy', 'Precision', 'Recall', 'F1-Score', __

¬'ROC-AUC']

for idx, (metric, metric_name) in enumerate(zip(metrics_for_box,_
 →metric_names_for_box)):
    ax = axes[idx]
    data_to_plot = [cv_scores[model_name][metric] for model_name in models.
 ⇒keys()]
    bp = ax.boxplot(data_to_plot, labels=list(models.keys()), patch_artist=True)
    # Color the boxes
    colors = ['lightblue', 'lightgreen', 'lightcoral', 'lightyellow']
    for patch, color in zip(bp['boxes'], colors):
        patch.set_facecolor(color)
    ax.set_title(metric_name)
```

```
ax.set_ylabel('Score')
   ax.tick_params(axis='x', rotation=45)
   ax.set_ylim(0, 1.1)
plt.tight_layout()
plt.savefig(os.path.join(output_dir, "cv_scores_boxplot.png"), dpi=300)
plt.show()
# 4. Create ensemble predictions (majority voting) and average probabilities
ensemble_predictions = np.zeros_like(predictions['LogisticRegression'])
ensemble_probabilities = np.zeros_like(probabilities['LogisticRegression'])
for model_preds in predictions.values():
    ensemble_predictions += model_preds
for model_probs in probabilities.values():
    ensemble_probabilities += model_probs
# Majority voting for predictions
ensemble_predictions = (ensemble_predictions >= len(models) / 2).astype(int)
# Average probabilities
ensemble_probabilities = ensemble_probabilities / len(models)
# Save ensemble predictions and probabilities
ensemble_df = pd.DataFrame({
    "Prediction": ensemble predictions,
    "Probability": ensemble_probabilities
})
ensemble_file = os.path.join(output_dir, "predicted_labels_probs_ensemble_jun16.
 GCSV")
ensemble_df.to_csv(ensemble_file, index=False)
print(f"\nEnsemble predictions (majority voting) and probabilities saved to⊔
 # Plot ensemble predictions
plt.figure(figsize=(6, 4))
ensemble_counts = pd.Series(ensemble_predictions).value_counts(normalize=True)_u
bars = plt.bar([0, 1], ensemble_counts.sort_index(), color=['blue', 'orange'],
 \rightarrowalpha=0.7)
plt.xlabel("Prediction")
plt.ylabel("Percentage")
plt.title("Ensemble Model - Predicted Labels (Majority Voting)")
plt.xticks([0, 1], labels=['0', '1'])
plt.ylim(0, 100)
# Add percentage labels
```

```
for bar, count in zip(bars, ensemble_counts.sort_index()):
   height = bar.get_height()
   plt.text(bar.get_x() + bar.get_width()/2., height + 1,
            f'{count:.1f}%', ha='center', va='bottom', fontsize=7)
plt.tight_layout()
plt.savefig(os.path.join(output_dir, "ensemble_prediction_barplot.png"),_

dpi=300)
plt.show()
# 5. Add probability distribution plots
fig, axes = plt.subplots(2, 2, figsize=(10, 8))
axes = axes.ravel()
for idx, (model_name, probs) in enumerate(probabilities.items()):
   ax = axes[idx]
   ax.hist(probs, bins=50, alpha=0.7, color='purple', edgecolor='black')
   ax.set_xlabel("Probability")
   ax.set ylabel("Frequency")
   ax.set_title(f"{model_name} - Probability Distribution")
   ax.axvline(x=0.5, color='red', linestyle='--', alpha=0.5)
   # Add statistics
   ax.text(0.95, 0.95, f'Mean: {np.mean(probs):.3f}\nStd: {np.std(probs):.3f}',
            transform=ax.transAxes, verticalalignment='top', u
 ⇔horizontalalignment='right',
            bbox=dict(boxstyle='round', facecolor='white', alpha=0.8),,,
 ⇔fontsize=7)
plt.tight_layout()
plt.savefig(os.path.join(output_dir, "probability_distributions.png"), dpi=300)
plt.show()
# Plot ensemble probability distribution
plt.figure(figsize=(6, 4))
plt.hist(ensemble_probabilities, bins=50, alpha=0.7, color='green',_
 ⇔edgecolor='black')
plt.xlabel("Probability")
plt.ylabel("Frequency")
plt.title("Ensemble Model - Probability Distribution")
plt.axvline(x=0.5, color='red', linestyle='--', alpha=0.5)
# Add statistics
plt.text(0.95, 0.95, f'Mean: {np.mean(ensemble probabilities):.3f}\nStd: {np.

¬std(ensemble_probabilities):.3f}',
         transform=plt.gca().transAxes, verticalalignment='top',
 ⇔horizontalalignment='right',
```

```
bbox=dict(boxstyle='round', facecolor='white', alpha=0.8), fontsize=7)
plt.tight_layout()
plt.savefig(os.path.join(output_dir, "ensemble_probability_distribution.png"), __
 ⊶dpi=300)
plt.show()
print(f"\nAll files have been saved to: {output_dir}")
print("\nSaved files:")
print("- predicted_labels_probs_LogisticRegression_jun16.csv")
print("- predicted_labels_probs_LinearSVC_jun16.csv")
print("- predicted_labels_probs_XGBClassifier_jun16.csv")
print("- predicted labels_probs_RandomForestClassifier_jun16.csv")
print("- predicted_labels_probs_ensemble_jun16.csv")
print("- model_performance_summary.csv")
print("- all_models_prediction_distributions.png")
print("- model_performance_comparison.png")
print("- cv_scores_boxplot.png")
print("- ensemble prediction barplot.png")
print("- probability_distributions.png")
print("- ensemble probability distribution.png")
print("\nAnalysis complete!")
```

Output directory created/verified: /Users/haley/Desktop/send_tooo/AAA_final/wtwt

Training models and making predictions...

Training LogisticRegression...

- Predictions and probabilities saved to '/Users/haley/Desktop/send_tooo/AAA_f inal/wtwt/predicted_labels_probs_LogisticRegression_jun16.csv'
 - CV Accuracy: 0.9009 (+/- 0.0182)
 - CV ROC-AUC: 0.9680 (+/- 0.0082)

Training LinearSVC...

- Predictions and probabilities saved to '/Users/haley/Desktop/send_tooo/AAA_f inal/wtwt/predicted_labels_probs_LinearSVC_jun16.csv'
 - CV Accuracy: 0.9012 (+/- 0.0178)
 - CV ROC-AUC: 0.9678 (+/- 0.0084)

Training XGBClassifier...

- Predictions and probabilities saved to '/Users/haley/Desktop/send_tooo/AAA_f inal/wtwt/predicted_labels_probs_XGBClassifier_jun16.csv'
 - CV Accuracy: 0.9045 (+/- 0.0088)
 - CV ROC-AUC: 0.9684 (+/- 0.0054)

Training RandomForestClassifier...

- Predictions and probabilities saved to '/Users/haley/Desktop/send_tooo/AAA_f

inal/wtwt/predicted_labels_probs_RandomForestClassifier_jun16.csv'

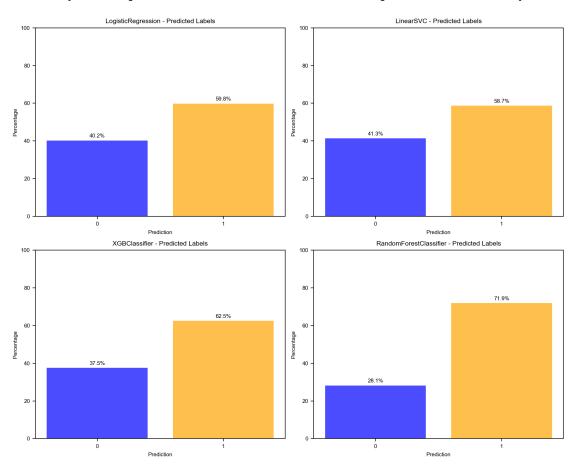
- CV Accuracy: 0.9058 (+/- 0.0059)
- CV ROC-AUC: 0.9693 (+/- 0.0068)

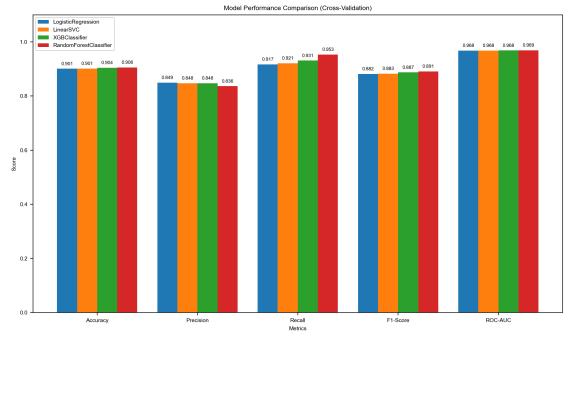
Model Performance Summary (Cross-Validation):

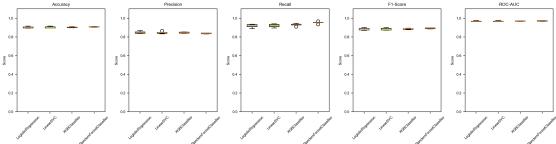
	Accuracy	Precision	Recall	F1-Score	ROC-AUC
LogisticRegression	0.9009	0.8494	0.9173	0.8820	0.9680
LinearSVC	0.9012	0.8476	0.9210	0.8827	0.9678
XGBClassifier	0.9045	0.8475	0.9311	0.8873	0.9684
RandomForestClassifier	0.9058	0.8364	0.9534	0.8910	0.9693

Performance summary saved to

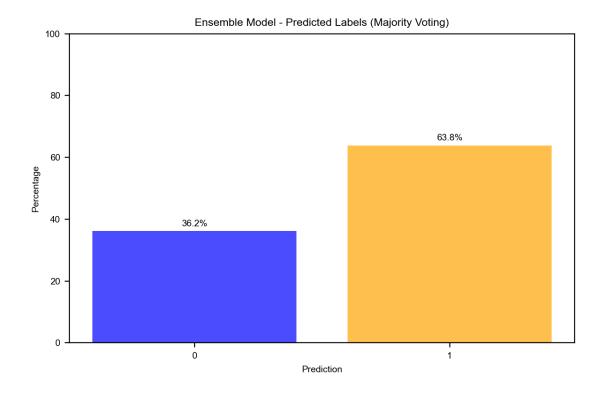
'/Users/haley/Desktop/send_tooo/AAA_final/wtwt/model_performance_summary.csv'

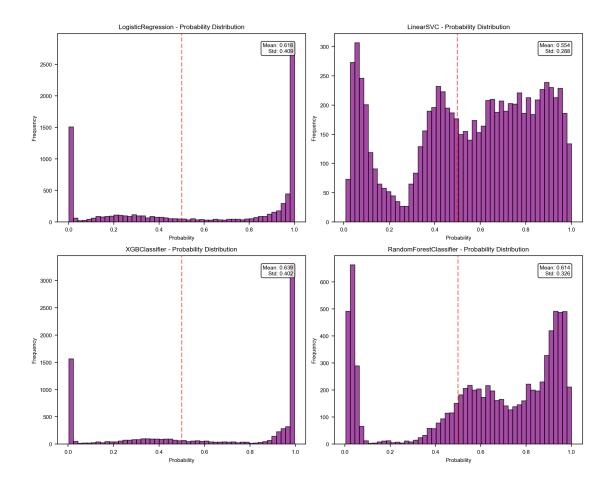




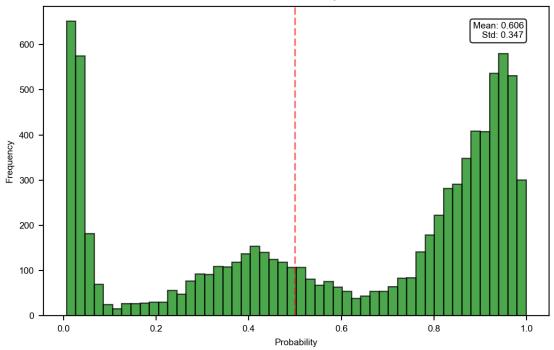


Ensemble predictions (majority voting) and probabilities saved to '/Users/haley/Desktop/send_tooo/AAA_final/wtwt/predicted_labels_probs_ensemble_jun16.csv'









All files have been saved to: /Users/haley/Desktop/send_tooo/AAA_final/wtwt

Saved files:

- predicted_labels_probs_LogisticRegression_jun16.csv
- predicted_labels_probs_LinearSVC_jun16.csv
- predicted_labels_probs_XGBClassifier_jun16.csv
- predicted_labels_probs_RandomForestClassifier_jun16.csv
- predicted_labels_probs_ensemble_jun16.csv
- model_performance_summary.csv
- all_models_prediction_distributions.png
- model_performance_comparison.png
- cv_scores_boxplot.png
- ensemble_prediction_barplot.png
- probability_distributions.png
- ensemble_probability_distribution.png

Analysis complete!

```
[106]: import numpy as np
  import pandas as pd
  import matplotlib.pyplot as plt
  from matplotlib import rcParams
```

```
from sklearn.ensemble import RandomForestClassifier # Added missing import
from xgboost import XGBClassifier
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import roc_auc_score, precision_recall_curve,_
 →average_precision_score, roc_curve, auc
import matplotlib.pyplot as plt
plt.style.use("default")
# Set global font properties
rcParams['font.family'] = 'Arial'
rcParams['font.size'] = 7
rcParams['font.weight'] = 'normal'
# Define model
# model = RandomForestClassifier(max_depth=10, random_state=41)
model = RandomForestClassifier(max_depth=10, random_state=41)
np.random.seed(42) # For reproducibility
# Scale the training data
scaler = StandardScaler()
full X train scaled = scaler.fit transform(full X train)
# Scale the test data
adata1_scaled = scaler.transform(extracted_columns_1)
# Fit the model on the training data
model.fit(full_X_train_scaled, full_Y_train)
# Predict probabilities for adata1
predicted_probabilities = model.predict_proba(adata1_scaled)
# Predict labels for adata1
predicted_labels = model.predict(adata1_scaled)
# Create output DataFrame with both probabilities and predictions
output_df = pd.DataFrame({
   "Probability_Class_0": predicted_probabilities[:, 0], # Probability of_
    "Probability_Class_1": predicted_probabilities[:, 1], # Probability of □
 ⇔class 1
    "Predicted_Label": predicted_labels
})
# Save to CSV file
output_df.to_csv("predicted_probabilities_jun16.csv", index=False)
print("Predicted probabilities and labels saved to_{\sqcup}

¬'predicted_probabilities_jun16.csv'.")
```

```
# Display summary statistics
print("\nSummary Statistics:")
print(f"Mean probability for class 1: {predicted_probabilities[:, 1].mean():.

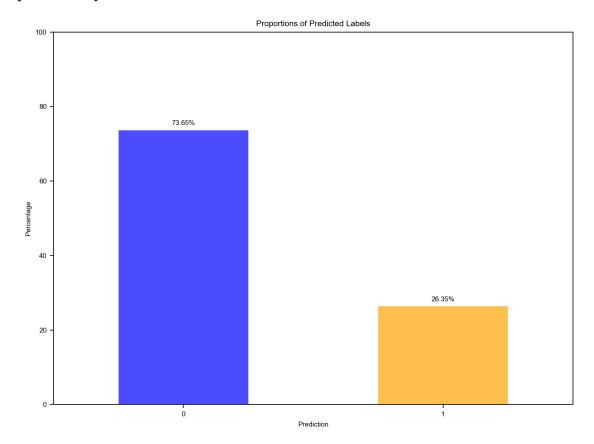
4f}")
print(f"Min probability for class 1: {predicted probabilities[:, 1].min():.4f}")
print(f"Max probability for class 1: {predicted_probabilities[:, 1].max():.4f}")
# Calculate percentages for the bar plot
label_counts = pd.Series(predicted_labels).value_counts(normalize=True) * 100 __
 ⇔# Convert to percentages
# Create a bar plot
plt.figure(figsize=(8, 6))
label_counts.plot(kind='bar', color=['blue', 'orange'], alpha=0.7)
plt.xlabel("Prediction")
plt.ylabel("Percentage")
plt.title("Proportions of Predicted Labels")
plt.xticks([0, 1], labels=['0', '1'], rotation=0)
plt.ylim(0, 100) # Ensure the y-axis ranges from 0 to 100
plt.tight_layout()
# Add percentage labels on the bars
for i, v in enumerate(label_counts):
   plt.text(i, v + 1, f"{v:.2f}%", ha='center', va='bottom', fontsize=7)
# Save and show the plot
plt.savefig("prediction_percentage_barplot.png", dpi=300)
plt.show()
print("Bar plot of predicted label percentages saved to⊔
 ⇔'prediction_percentage_barplot.png'.")
# Optional: Create histogram of probabilities for class 1
plt.figure(figsize=(8, 6))
plt.hist(predicted probabilities[:, 1], bins=50, alpha=0.7, color='green',
 ⇔edgecolor='black')
plt.xlabel("Probability of Class 1")
plt.ylabel("Frequency")
plt.title("Distribution of Predicted Probabilities for Class 1")
plt.grid(True, alpha=0.3)
plt.tight_layout()
plt.savefig("probability_distribution_class1.png", dpi=300)
plt.show()
print("Probability distribution plot saved to 'probability_distribution_class1.

¬png'.")
```

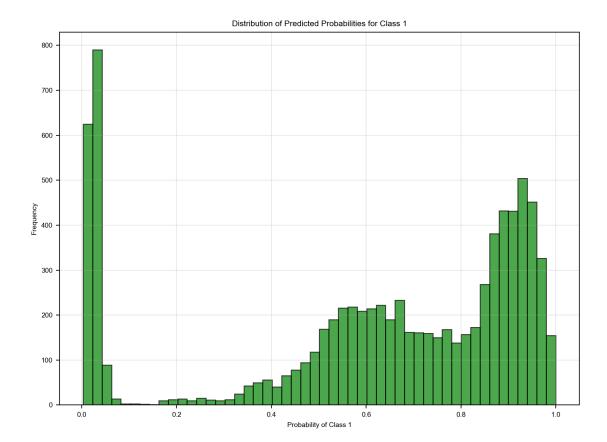
Predicted probabilities and labels saved to 'predicted_probabilities_jun16.csv'.

Summary Statistics:

Mean probability for class 1: 0.6145 Min probability for class 1: 0.0031 Max probability for class 1: 0.9995



Bar plot of predicted label percentages saved to 'prediction_percentage_barplot.png'.



Probability distribution plot saved to 'probability_distribution_class1.png'.

```
[89]: # Predict labels for adata1
predicted_labels = model.predict(adata1_scaled)

# Save predictions to a CSV file
output_df = pd.DataFrame({"Prediction": predicted_labels})
output_df.to_csv("predicted_labels.csv", index=False)
print("Predicted labels saved to 'dec_3_predicted_labels.csv'.")
```

Predicted labels saved to 'dec_3_predicted_labels.csv'.

```
[]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from matplotlib import rcParams
from xgboost import XGBClassifier
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import roc_auc_score, precision_recall_curve,
→average_precision_score, roc_curve, auc
```

```
# Set global font properties
rcParams['font.family'] = 'Arial'
rcParams['font.size'] = 7
rcParams['font.weight'] = 'normal'
# LogisticRegression(max_iter=1000, random_state=551),
          SVC(kernel='linear', probability=True, random_state=551),
          XGBClassifier(
              use label encoder=False,
              eval metric='logloss',
#
              n_{estimators=100},
#
              learning_rate=0.05,
#
              max_depth=10,
#
              random_state=551
          ),
          RandomForestClassifier(max_depth=10, random_state=41)
# Define model
# model = XGBClassifier(
     use label encoder=False,
      eval_metric='logloss',
#
      n estimators=300, # set large n estimators but use early stopping
      learning_rate=0.01,
      max depth=6,
      random state=551
# )500
              0.05
model =XGBClassifier(
use_label_encoder=False,
            eval_metric='logloss',
            random_state=551,
    n_estimators=500, # set large n_estimators but use early_stopping
    learning_rate=0.05,
    max_depth=6
np.random.seed(42) # For reproducibility
# Scale the training data
scaler = StandardScaler()
full_X_train_scaled = scaler.fit_transform(full_X_train)
# Fit the model on the training data
model.fit(full_X_train_scaled, full_Y_train)
# Function to process each dataset
```

```
def process_and_save(extracted_columns, name):
    # Scale the test data
    adata_scaled = scaler.transform(extracted_columns)
    # Predict labels
    predicted_labels = model.predict(adata_scaled)
    # Save predictions to a CSV file
    output_df = pd.DataFrame({"Prediction": predicted_labels})
    output_file = f"predicted_labels_{name}.csv"
    output df.to csv(output file, index=False)
    print(f"Predicted labels saved to '{output_file}'.")
    # Calculate percentages for the bar plot
    label_counts = pd.Series(predicted_labels).value_counts(normalize=True) *__
 →100 # Convert to percentages
    # Create a bar plot
    label_counts.plot(kind='bar', color=['blue', 'orange'], alpha=0.7)
    plt.xlabel("Prediction")
    plt.ylabel("Percentage")
    plt.title(f"Proportions of Predicted Labels - {name}")
    plt.xticks([0, 1], labels=['0', '1'], rotation=0)
    plt.ylim(0, 100) # Ensure the y-axis ranges from 0 to 100
    plt.tight_layout()
    # Add percentage labels on the bars
    for i, v in enumerate(label_counts):
        plt.text(i, v + 1, f"{v:.2f}%", ha='center', va='bottom', fontsize=7)
    # Save and show the plot
    plot_file = f"prediction_percentage_barplot_{name}.png"
    plt.savefig(plot_file, dpi=300)
    plt.close() # Close the plot to avoid overlap in subsequent iterations
    print(f"Bar plot of predicted label percentages saved to '{plot_file}'.")
# Process each dataset
process_and_save(extracted_columns_1, "dataset_1")
process_and_save(extracted_columns_2, "dataset_2")
process_and_save(extracted_columns_3, "dataset_3")
process_and_save(extracted_columns_4, "dataset_4")
Predicted labels saved to 'predicted_labels_dataset_1.csv'.
Bar plot of predicted label percentages saved to
'prediction_percentage_barplot_dataset_1.png'.
Predicted labels saved to 'predicted labels dataset 2.csv'.
Bar plot of predicted label percentages saved to
'prediction_percentage_barplot_dataset_2.png'.
```

```
Predicted labels saved to 'predicted_labels_dataset_3.csv'. Bar plot of predicted label percentages saved to 'prediction_percentage_barplot_dataset_3.png'. Predicted labels saved to 'predicted_labels_dataset_4.csv'. Bar plot of predicted label percentages saved to 'prediction_percentage_barplot_dataset_4.png'.
```

```
[211]: import numpy as np
       import pandas as pd
       import matplotlib.pyplot as plt
       from matplotlib import rcParams
       from xgboost import XGBClassifier
       from sklearn.preprocessing import StandardScaler
       from sklearn.metrics import roc_auc_score, precision_recall_curve, u
       ⇒average_precision_score, roc_curve, auc
       # Set global font properties
       rcParams['font.family'] = 'Arial'
       rcParams['font.size'] = 7
       rcParams['font.weight'] = 'normal'
       # Define model
       model = XGBClassifier(
           use label encoder=False,
           eval_metric='logloss',
           random_state=551,
           n_estimators=500,
           learning_rate=0.15,
          max_depth=6
       )
       np.random.seed(42) # For reproducibility
       # Scale the training data
       scaler = StandardScaler()
       full_X_train_scaled = scaler.fit_transform(full_X_train)
       # Fit the model on the training data
       model.fit(full_X_train_scaled, full_Y_train)
       # Function to process each dataset and print male percentage
       def process_and_save(extracted_columns, name):
           # Scale the test data
           adata_scaled = scaler.transform(extracted_columns)
           # Predict labels
           predicted_labels = model.predict(adata_scaled)
```

```
# Save predictions to a CSV file
    output_df = pd.DataFrame({"Prediction": predicted_labels})
    output_file = f"predicted_labels_{name}.csv"
   output_df.to_csv(output_file, index=False)
   print(f"Predicted labels saved to '{output_file}'.")
   # Calculate percentages for male (label 0)
   label_counts = pd.Series(predicted_labels).value_counts(normalize=True) *__
 →100 # Convert to percentages
   male_percentage = label_counts.get(0, 0) # Get male percentage (0), u
 →default to 0 if missing
   female_percentage = label_counts.get(1, 0) # Get female percentage (1),
 ⇔default to 0 if missing
   print(f"{name} - Male Percentage: {male percentage:.2f}%, Female Percentage:
 # Create a bar plot
   label counts.plot(kind='bar', color=['blue', 'orange'], alpha=0.7)
   plt.xlabel("Prediction")
   plt.ylabel("Percentage")
   plt.title(f"Proportions of Predicted Labels - {name}")
   plt.xticks([0, 1], labels=['Male', 'Female'], rotation=0)
   plt.ylim(0, 100) # Ensure the y-axis ranges from 0 to 100
   plt.tight_layout()
   # Add percentage labels on the bars
   for i, v in enumerate(label_counts):
       plt.text(i, v + 1, f"{v:.2f}%", ha='center', va='bottom', fontsize=7)
    # Save and close the plot
   plot_file = f"prediction_percentage_barplot_{name}.png"
   plt.savefig(plot_file, dpi=300)
   plt.close() # Close the plot to avoid overlap in subsequent iterations
   print(f"Bar plot of predicted label percentages saved to '{plot_file}'.")
# Process each dataset and print male percentages
process_and_save(extracted_columns_1, "dataset_1")
process_and_save(extracted_columns_2, "dataset_2")
process_and_save(extracted_columns_3, "dataset_3")
process_and_save(extracted_columns_4, "dataset_4")
```

Predicted labels saved to 'predicted_labels_dataset_1.csv'. dataset_1 - Male Percentage: 34.41%, Female Percentage: 65.59% Bar plot of predicted label percentages saved to 'prediction_percentage_barplot_dataset_1.png'.

Predicted labels saved to 'predicted_labels_dataset_2.csv'.
dataset_2 - Male Percentage: 43.95%, Female Percentage: 56.05%
Bar plot of predicted label percentages saved to
'prediction_percentage_barplot_dataset_2.png'.
Predicted labels saved to 'predicted_labels_dataset_3.csv'.
dataset_3 - Male Percentage: 28.82%, Female Percentage: 71.18%
Bar plot of predicted label percentages saved to
'prediction_percentage_barplot_dataset_3.png'.
Predicted labels saved to 'predicted_labels_dataset_4.csv'.
dataset_4 - Male Percentage: 32.60%, Female Percentage: 67.40%
Bar plot of predicted label percentages saved to
'prediction_percentage_barplot_dataset_4.png'.