

Getting started with the tutorial data set

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Creating a new project and importing the tutorial data set

The tutorial data set is available through Partek® Flow®.

- Click your **avatar** (Figure 1)

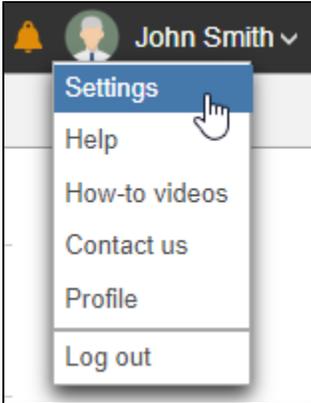


Figure 17. Location of the Settings link on the main page of Partek Flow

- Click **Settings**

On the *System information* page, the *Download tutorial data* section includes pre-loaded data sets used by Partek Flow tutorials (Figure 2).

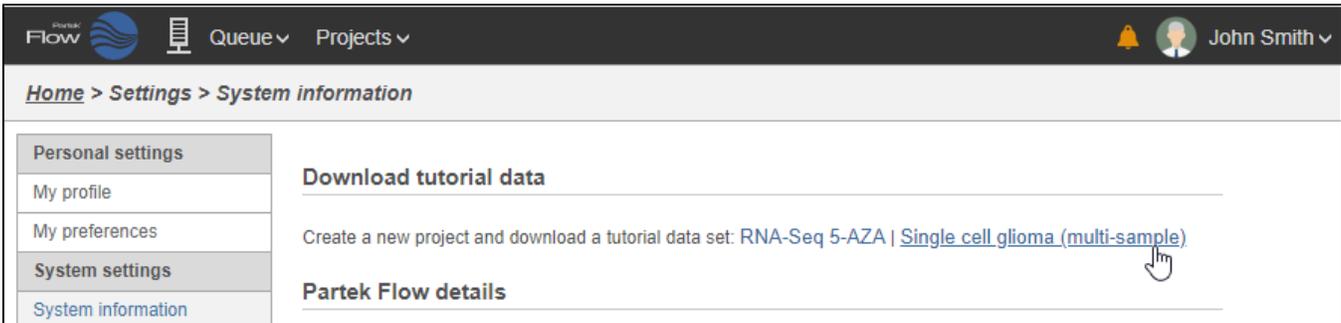


Figure 18. Tutorial data sets available through Partek Flow

- Click **Single cell glioma (multi-sample)**

The tutorial data set will be downloaded onto your Partek Flow server and a new project, *Glioma (multi-sample)*, will be created. You will be directed to the *Data* tab of the new project. Because this is a tutorial project, there is no need to click on *Import data*, as the import is handled automatically (Figure 3).

The tutorial data is being downloaded and imported in the background. You can wait a few minutes and refresh the page.

Analyses **Data** Log Project settings Notebook Data Viewer Attachments

Data

Import

Sample attributes

Manage

Assign values from file

You'll need some samples before you can run an analysis.

To get started, click the "Import" menu option.

Figure 19. The data tab during tutorial data import

You can wait a few minutes for the download to complete, or check the download progress by selecting **Queue** then **View queued tasks...** to view the *Queue* (Figure 4).

Home > Queued tasks

There is 1 task in the queue.

Status	Task	Project	User	Submitted	End	Workers	Cancel
<input type="checkbox"/>	Download Tutorial Data	↑ Glioma (multi-sample)	Paul Fullerton	20 Mar 2018, 09:28 AM CDT	20 Mar 2018, 09:28 AM CDT		✖

View recent activity

T - Waiting for upstream tasks to complete R - Waiting for system resources ⚠ - Cannot run with current system configuration

Time estimates are being continuously updated and will become more accurate.

Figure 20. Viewing the queue

Once the download completes, the sample table will appear in the *Data* tab, with one row per sample (Figure 5).

Home > Glioma (multi-sample) (Project owner)

Analyses	Data	Log	Project settings	Notebook	Data Viewer	Attachments																																																			
<table border="1"> <thead> <tr> <th>Data</th> </tr> </thead> <tbody> <tr> <td>Import</td> </tr> <tr> <td>Sample attributes</td> </tr> <tr> <td>Manage</td> </tr> <tr> <td>Assign values</td> </tr> <tr> <td>Assign values from file</td> </tr> <tr> <td>Add system-wide attribute</td> </tr> </tbody> </table>		Data	Import	Sample attributes	Manage	Assign values	Assign values from file	Add system-wide attribute	<table border="1"> <thead> <tr> <th></th> <th>Sample name</th> <th colspan="2">Attributes</th> </tr> <tr> <th></th> <th></th> <th>Subtype</th> <th># Cells</th> </tr> </thead> <tbody> <tr> <td>1</td> <td>MGH36</td> <td>Oligodendroglioma</td> <td>788</td> </tr> <tr> <td>2</td> <td>MGH42</td> <td>Astrocytoma</td> <td>682</td> </tr> <tr> <td>3</td> <td>MGH45</td> <td>Astrocytoma</td> <td>608</td> </tr> <tr> <td>4</td> <td>MGH53</td> <td>Oligodendroglioma</td> <td>861</td> </tr> <tr> <td>5</td> <td>MGH54</td> <td>Oligodendroglioma</td> <td>1225</td> </tr> <tr> <td>6</td> <td>MGH56</td> <td>Astrocytoma</td> <td>925</td> </tr> <tr> <td>7</td> <td>MGH60</td> <td>Oligodendroglioma</td> <td>430</td> </tr> <tr> <td>8</td> <td>MGH64</td> <td>Astrocytoma</td> <td>795</td> </tr> <tr> <td colspan="3">Show data files</td> <td>Download</td> </tr> </tbody> </table>						Sample name	Attributes				Subtype	# Cells	1	MGH36	Oligodendroglioma	788	2	MGH42	Astrocytoma	682	3	MGH45	Astrocytoma	608	4	MGH53	Oligodendroglioma	861	5	MGH54	Oligodendroglioma	1225	6	MGH56	Astrocytoma	925	7	MGH60	Oligodendroglioma	430	8	MGH64	Astrocytoma	795	Show data files			Download
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Figure 21. Sample data table listing the name and the number of cells for each sample

The sample table is pre-populated with two sample attributes: # Cells and Subtype. Sample attributes can be added and edited manually by clicking *Manage* in the *Sample attributes* menu on the left. If a new attribute is added, click *Assign values* to assign samples to different groups. Alternatively, you can use the *Assign values from a file* option to assign sample attributes using a tab-delimited text file. For more information about sample attributes, see [here](#).

For this tutorial, we do not need to edit or change any sample attributes.

Filtering cells in single cell RNA-Seq data

With samples imported and annotated, we can begin analysis.

- Click **Analyses** to switch to the *Analyses* tab

For now, the *Analyses* tab has only a single node, *Single cell counts*. As you perform the analysis, additional nodes representing tasks and new data will be created, forming a visual representation of your analysis pipeline.

- Click on the **Single cell counts** node

A context-sensitive menu will appear on the right-hand side of the pipeline (Figure 9). This menu includes tasks that can be performed on the selected counts data node.

An important step in analyzing single cell RNA-Seq data is to filter out low-quality cells. A few examples of low-quality cells are doublets, cells damaged during cell isolation, or cells with too few counts to be analyzed.

- Expand the **QA/QC** section of the task menu

- Click on **Single cell QA/QC** (Figure 6)

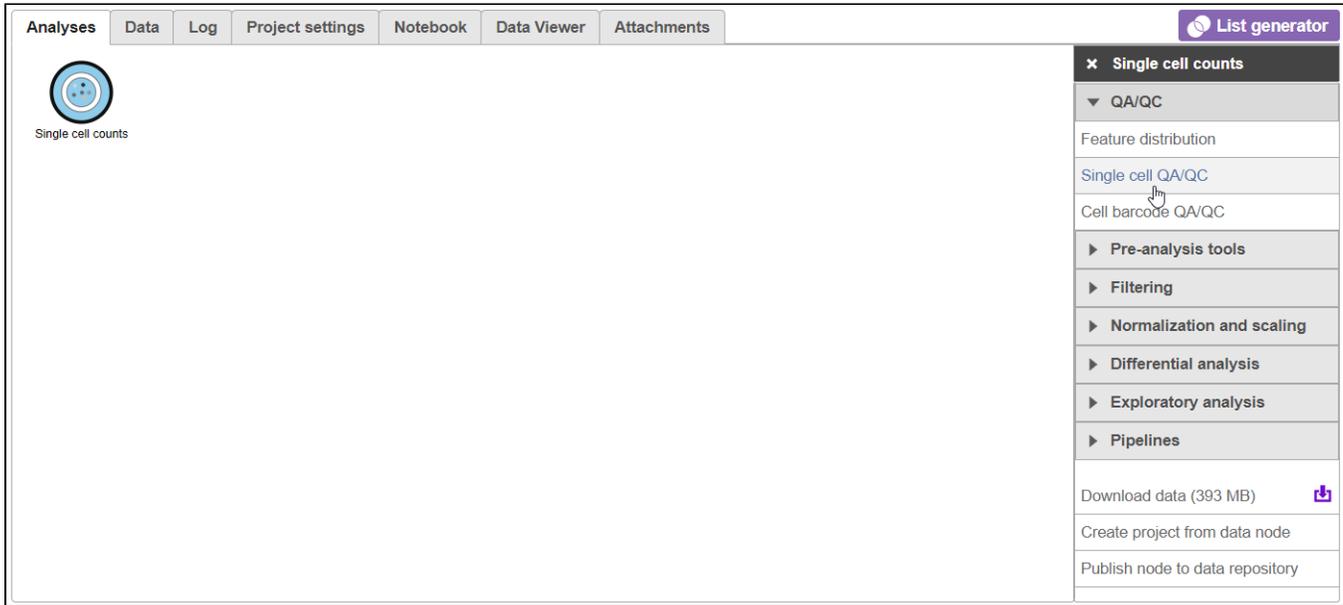


Figure 22. Selecting the *Single cell QA/QC* task from the task menu

A task node, *Single cell QA/QC*, is produced. Initially, the node will be semi-transparent to indicate that it has been queued, but not completed. A progress bar will appear on the *Single cell QA/QC* task node to indicate that the task is running.

- Click the **Single cell QA/QC** node once it finishes running
- Click **Task report** on the task menu (Figure 7)

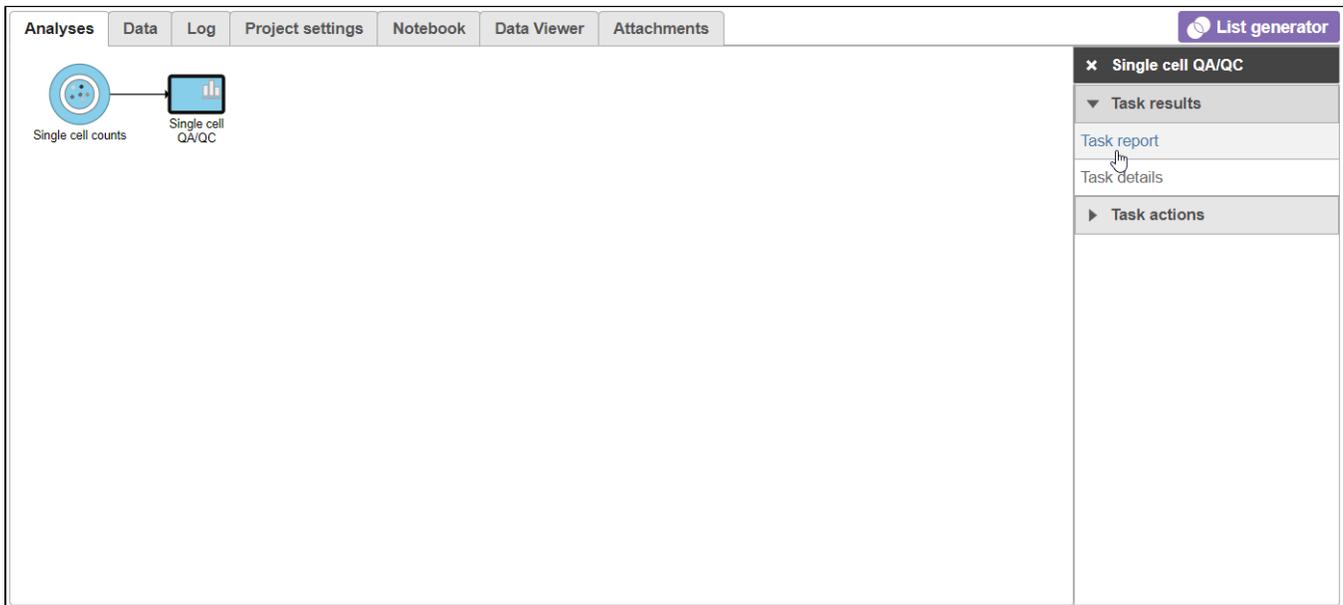


Figure 23. Selecting the task report for any task node opens a report with any tables or charts the task produced

The *Single cell QA/QC* report opens in a new data viewer session. There are interactive violin plots showing the most commonly used quality metrics for each cell from all samples combined (Figure 8). For this data set, there are two relevant plots: the total count per cell and the number of detected genes per cell. Each point on the plots is a cell and the violins illustrate the distribution of values for the y-axis metric. Typically, there is a third plot showing the percentage of mitochondrial counts per cell, but mitochondrial transcripts were not included in the data set by the study authors, so this plot is not informative for this data set.

- Remove the % mitochondrial counts and the extra text box in the bottom right by clicking **Remove plot** in the top right corner of each plot (Figure 8).

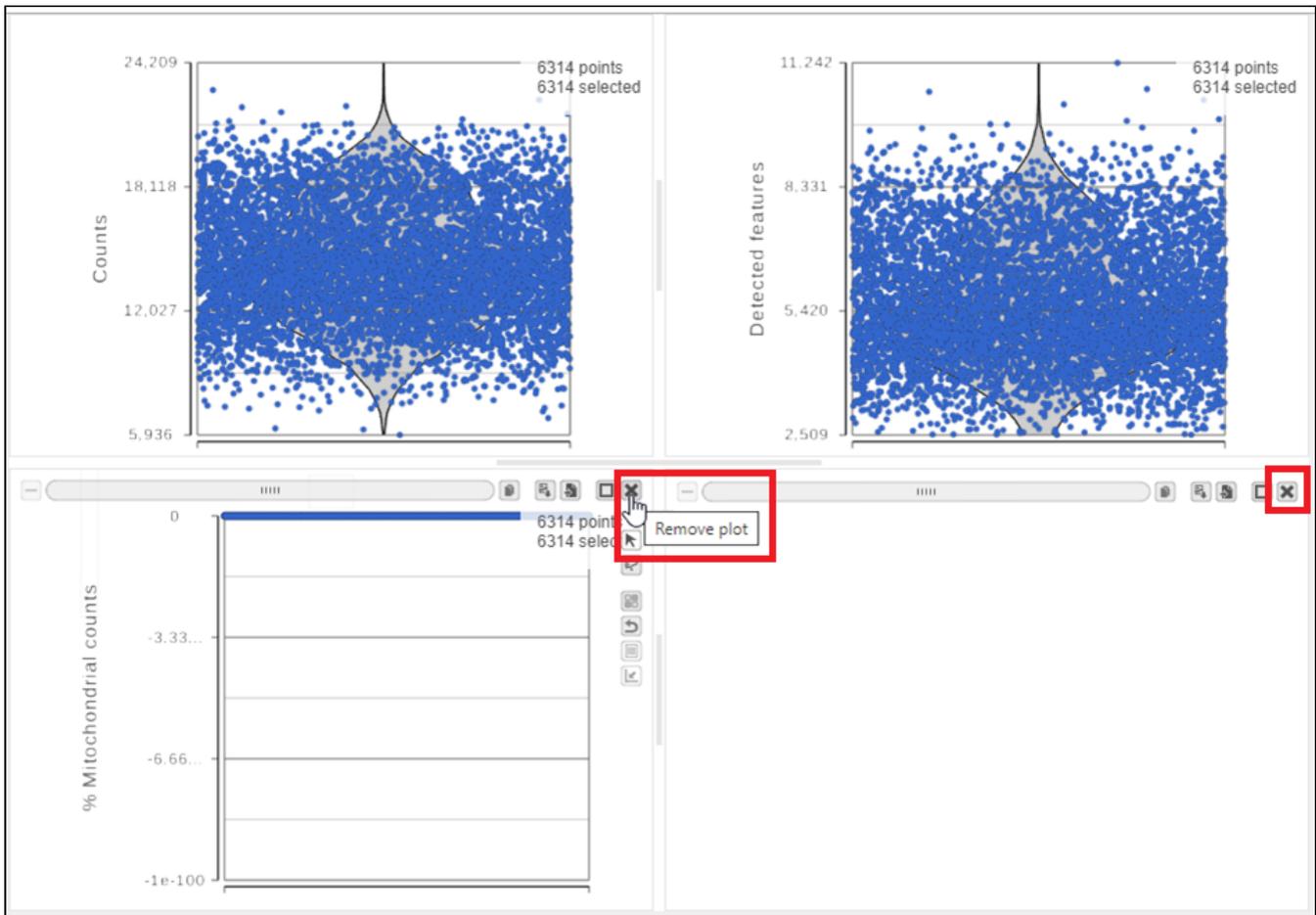


Figure 24. Each cell is shown as a point on the plot. Remove the % mitochondrial counts and empty text box using the X icons

The plots are highly customizable and can be used to explore the quality of cells in different samples.

- Click on **Single cell counts** in the **Get Data** icon on the left (Figure 9)
- Click and drag the **Sample name** attribute onto the *Counts plot* and drop it onto the *X-axis*
- Repeat this for the *Detected genes* plot

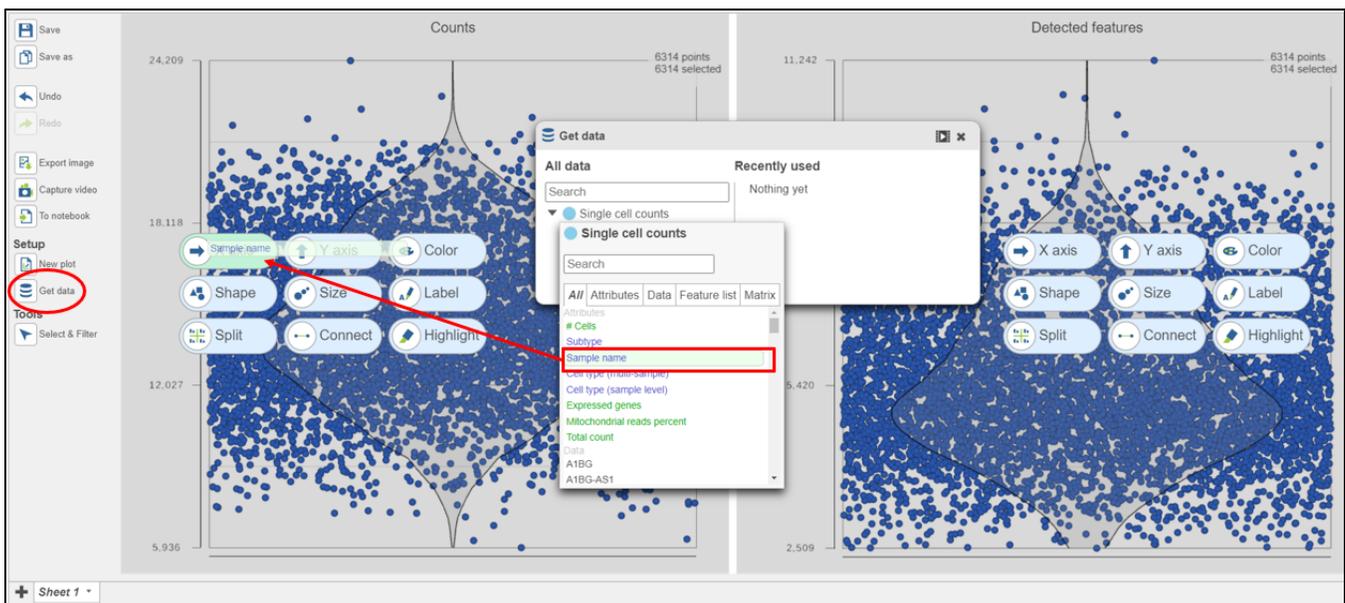


Figure 25. Click and drag the Sample name attribute onto the X-axis for each plot

The cells are now separated into different samples along the x-axis (Figure 10)

- Hold Control and left-click to select both plots
- Open the **Style** icon on the left under *Configure*
- Under *Color*, use the slider to reduce the **Opacity**
- Open the **Axis** icon on the left
- Adjust the **X-rotation** on the plots to **90**

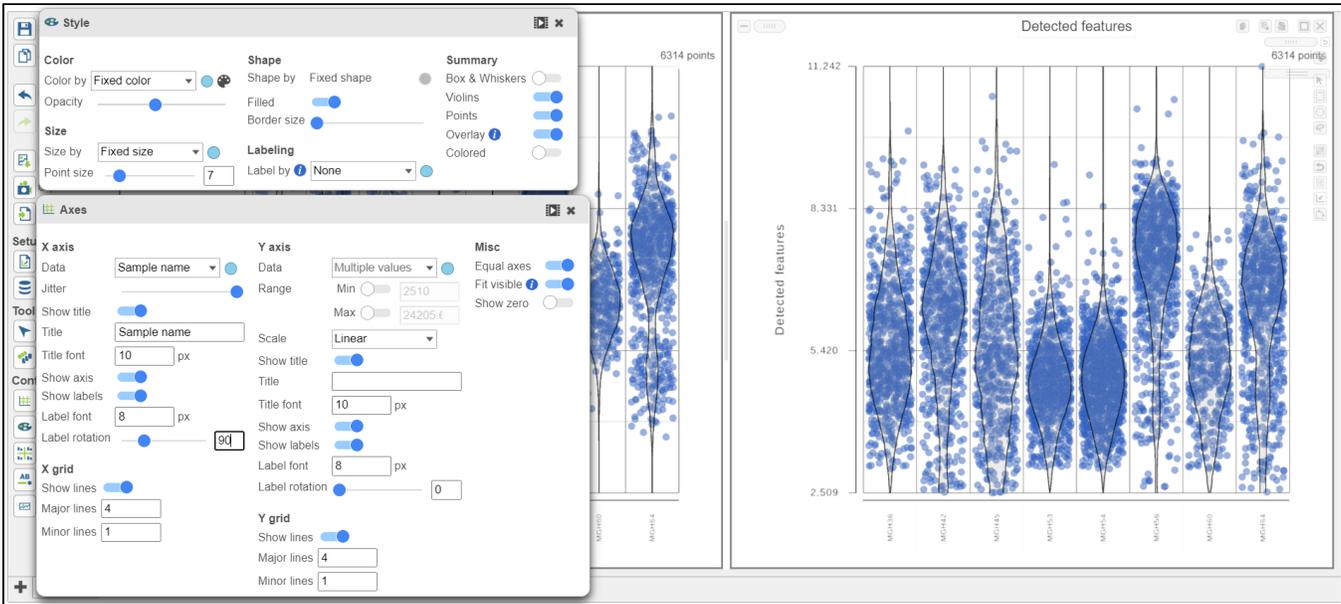


Figure 26. Counts and detected genes plots can be customized to compare cells from different samples

Note how both plots were modified at the same time.

Cells can be selected by setting thresholds using the **Select & Filter** tool. Here, we will select cells based on the total count

- Open **Select & Filter** under *Tools* on the left
- Under *Criteria*, Click **Pin histogram** to see the distribution of counts
- Set the *Counts* thresholds to **8000** and **20500**

Selected cells will be in blue and deselected cells will be dimmed (Figure 11).

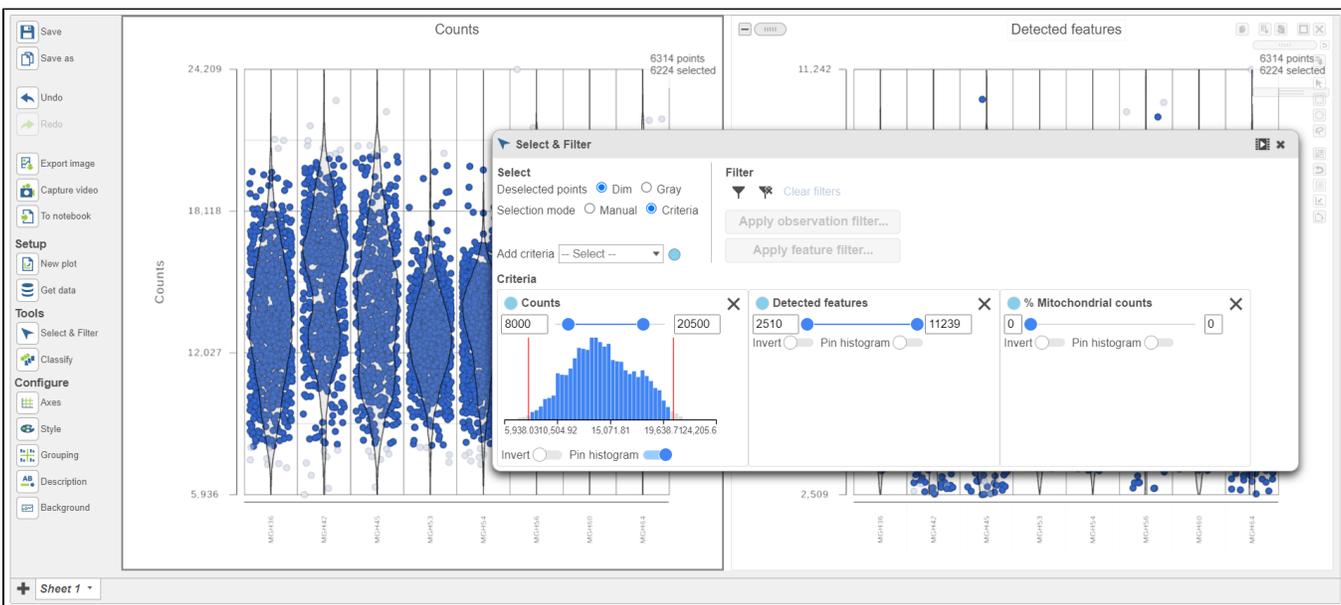


Figure 27. Previewing a filter using the Single cell QA/QC violin plots

Because this data set was already filtered by the study authors to include only high-quality cells, this count filter is sufficient.



- Click  under *Filter* to include the selected cells
- Click **Apply observation filter**
- Click the **Single cell counts** data node in the pipeline preview (Figure 12)
- Click **Select**

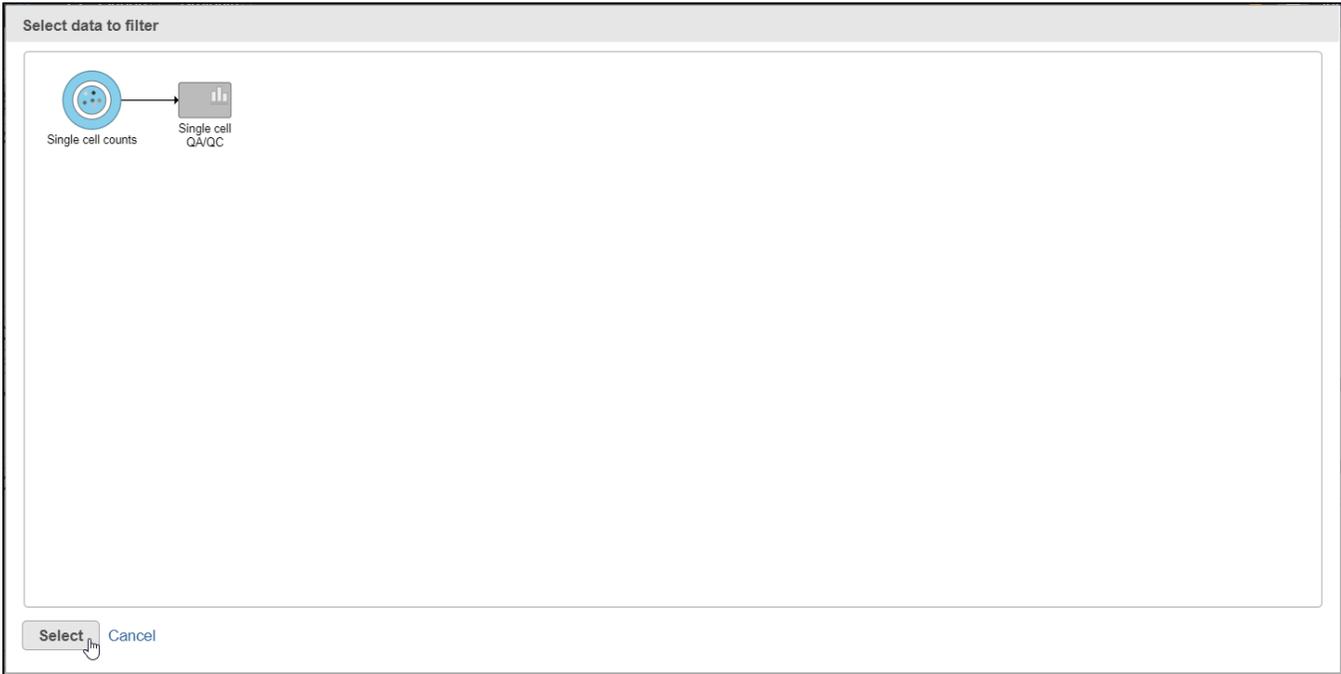


Figure 28. After the Apply filter button is selected, you will be presented with a preview of your pipeline. You need to select the appropriate data node to apply the filtering to

A new task, *Filter counts*, is added to the *Analyses* tab. This task produces a new *Filter counts* data node (Figure 13).

- Click on the **Glioma (multi-sample)** project name at the top to go back to the *Analyses* tab
- Your browser may warn you that any unsaved changes to the data viewer session will be lost. Ignore this message and proceed to the *Analyses* ta

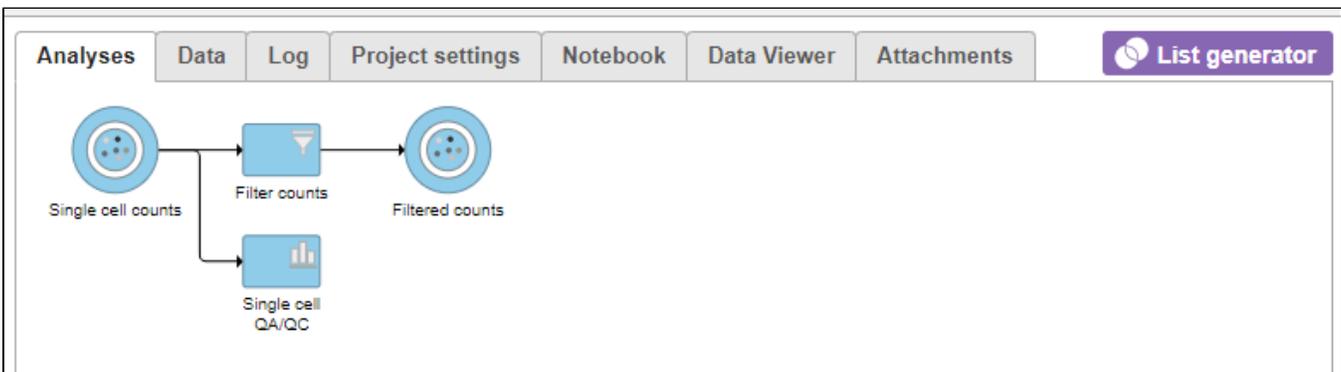


Figure 29. Applying a cell quality filter

Most tasks can be queued up on data nodes that have not yet been generated, so you can wait for filtering step to complete, or proceed to the next section.

Filtering genes in single cell RNA-Seq data

A common task in bulk and single-cell RNA-Seq analysis is to filter the data to include only informative genes. Because there is no gold standard for what makes a gene informative or not, ideal gene filtering criteria depends on your experimental design and research question. Thus, Partek Flow has a wide variety of flexible filtering options.

- Click the **Filter counts** node produced by the *Filter counts* task
- Click **Filtering** in the task menu
- Click **Filter features** (Figure 14)

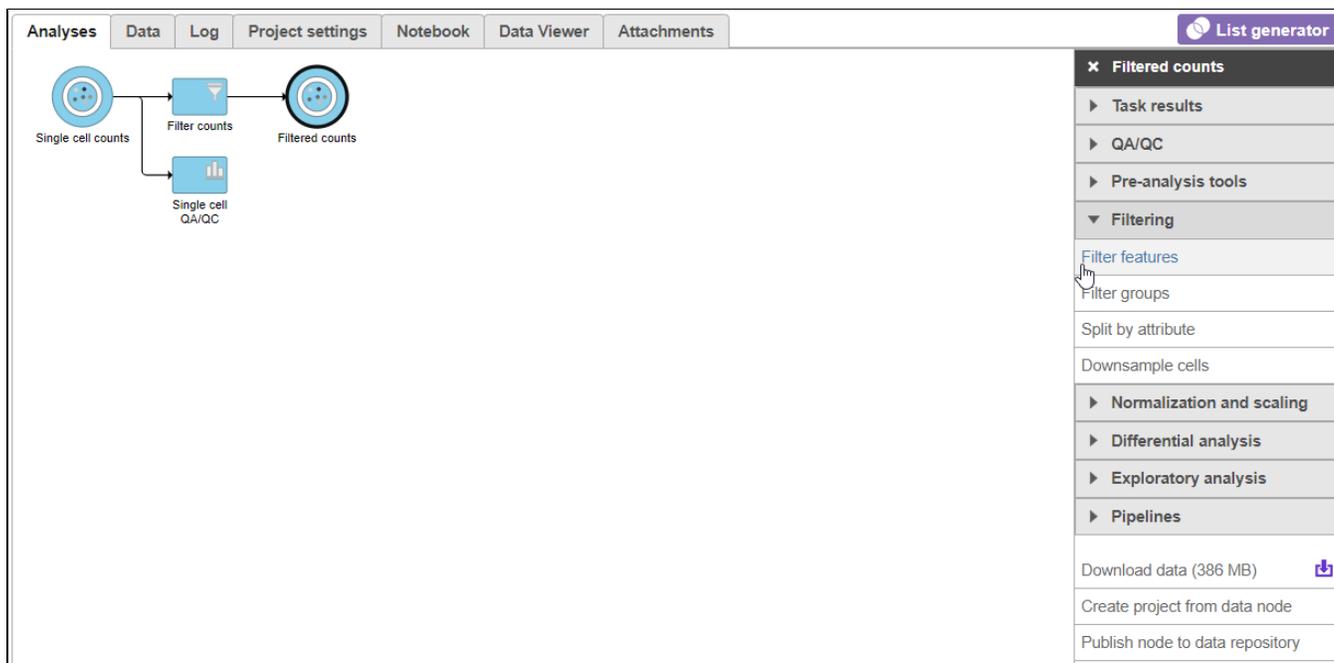


Figure 30. Invoking Filter features

There are four categories of filter available - noise reduction, statistics based, feature metadata, and feature list (Figure 15).

For a detailed explanation of options, please see the [documentation](#) for this task.

Noise reduction filter i

Exclude features where maximum <= 1.0

Statistics based filter i

Filter features by Counts Percentiles

Keep the top 100.0 features with highest variance

Feature metadata filter i

Filter features by

include
Chromosome
in
Terms
OR

AND

Feature list filter

Filter criteria Include Exclude

Case sensitive

Feature identifier i

- Gene symbol (Values: A1BG, A1BG-AS1, A1CF, A2M, A2M-AS1, A2ML1...)
- gene_id (Values: ENSG00000121410, ENSG00000268895, ENSG00000148584,...)
- gene_name (Values: A1BG, A1BG-AS1, A1CF, A2M, A2M-AS1, A2ML1...)
- transcript_id (Values: ENST00000600966, ENST00000595302, ENST00000414883,...)

Back

Finish

Figure 31. Viewing the filtering options

The noise reduction filter allows you to exclude genes considered background noise based on a variety of criteria. The statistics based filter is useful for focusing on a certain number or percentile of genes based on a variety of metrics, such as variance. The feature list filter allows you to filter your data set to include or exclude particular genes.

We will use a noise reduction filter to exclude genes that are not expressed by any cell in the data set but were included in the matrix file.

- Click the **Noise reduction filter** checkbox
- Set the *Noise reduction filter* to **Exclude features where value <= 0 in 99% of cells** using the drop-down menus and text boxes (Figure 16)
- Click **Finish** to apply the filter

For a detailed explanation of options, please see the [documentation](#) for this task.

Noise reduction filter i

Exclude features where value <= 0 in at least 99 % of the cells

Statistics based filter i

Filter features by Counts Percentiles

Keep the top 100.0 features with highest variance

Feature metadata filter i

Filter features by

include Chromosome in Terms OR

AND

Feature list filter

Filter criteria Include Exclude

Case sensitive

Feature identifier i

- Gene symbol (Values: A1BG, A1BG-AS1, A1CF, A2M, A2M-AS1, A2ML1...)
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- transcript_id (Values: ENST00000600966, ENST00000595302, ENST00000414883,...)

Back
Finish

Figure 32. Configuring a noise reduction filter to exclude genes not expressed in the data set

This produces a *Filtered counts* data node. This will be the starting point for the next stage of analysis - identifying cell types in the data using the interactive t-SNE plot.

Normalizing single cell RNA-Seq data

We are omitting normalization in this tutorial because the data has already been normalized.

The tutorial data set is taken from a published study and has already been normalized using TPM (Transcripts per million), which normalizes for the length of feature and total reads, and transformed as $\log_2(\text{TPM}/10+1)$. This normalization and transformation scheme can be performed in Partek Flow, along with other commonly used RNA-Seq data normalization methods.

For more information on normalizing data in Partek Flow, please see the [Normalization](#) section of the user manual.

« [Single Cell RNA-Seq Analysis \(Multiple Samples\)](#) [Classify cells from multiple samples using t-SNE](#) »

Additional Assistance

If you need additional assistance, please visit [our support page](#) to submit a help ticket or find phone numbers for regional support.



★

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