

Stefano Mangiola

Maria Doyle

# Tidy Transcriptomics for Single-cell RNA Sequencing Analyses



# Resources for #tidytranscriptomics



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### Tidy-transcriptomics manifesto

2021-10-13·tidytranscriptomics

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### tidybulk: an R tidy framework for modular transcriptomic data analysis

[Stefano Mangiola](#), [Ramyar Molania](#), [Ruining Dong](#), [Maria A. Doyle](#) & [Anthony T. Papenfuss](#) ✉

[Genome Biology](#) **22**, Article number: 42 (2021) | [Cite this article](#)

### Interfacing Seurat with the R tidy universe 🧠

[Stefano Mangiola](#) ✉, [Maria A Doyle](#), [Anthony T Papenfuss](#) ✉

*Bioinformatics*, btab404, <https://doi.org/10.1093/bioinformatics/btab404>

# Tidy R tools

There are four basic principles to a tidy API:

- Reuse existing data structures.
- Compose simple functions with the pipe.
- Embrace functional programming.
- Design for humans.

```
# A tibble: 100 x 8
  observation variable_1 variable_2
  <glue>      <chr>      <chr>
1 observation 1 ...      ...
2 observation 2 ...      ...
3 observation 3 ...      ...
4 observation 4 ...      ...
5 observation 5 ...      ...
6 observation 6 ...      ...
7 observation 7 ...      ...
8 observation 8 ...      ...
9 observation 9 ...      ...
10 observation 10 ...      ...
# ... with 90 more rows
```

# Tidy R tools

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```
# A tibble: 100 x 8
  observation variable_1 variable_2 variable_3
  <glue>      <chr>      <chr>      <list>
1 observation 1    ...      ...      <gg>
2 observation 2    ...      ...      <gg>
3 observation 3    ...      ...      <gg>
4 observation 4    ...      ...      <gg>
5 observation 5    ...      ...      <gg>
6 observation 6    ...      ...      <gg>
7 observation 7    ...      ...      <gg>
8 observation 8    ...      ...      <gg>
9 observation 9    ...      ...      <gg>
10 observation 10  ...      ...      <gg>
# ... with 90 more rows
```

# Tidy R tools

There are four basic principles to a tidy API:

- Reuse existing data structures.
- Compose simple functions with the pipe.
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```
# A tibble: 100 x 8
  observation variable_1 variable_2 variable_3 variable_4
  <glue>      <chr>      <chr>      <list>      <list>
1 observation 1  ...      ...      <gg>      <tibble [10 x 2]>
2 observation 2  ...      ...      <gg>      <tibble [10 x 2]>
3 observation 3  ...      ...      <gg>      <tibble [10 x 2]>
4 observation 4  ...      ...      <gg>      <tibble [10 x 2]>
5 observation 5  ...      ...      <gg>      <tibble [10 x 2]>
6 observation 6  ...      ...      <gg>      <tibble [10 x 2]>
7 observation 7  ...      ...      <gg>      <tibble [10 x 2]>
8 observation 8  ...      ...      <gg>      <tibble [10 x 2]>
9 observation 9  ...      ...      <gg>      <tibble [10 x 2]>
10 observation 10 ...      ...      <gg>      <tibble [10 x 2]>
# ... with 90 more rows
```



# Tidy R tools

There are four basic principles to a tidy API:

- Reuse existing data structures.
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```
# A tibble: 100 x 8
  observation variable_1 variable_2 variable_3 variable_4 variable_5
  <glue>      <chr>      <chr>      <list>      <list>      <list>
1 observation 1  ...      ...      <gg>      <tibble [10 x 2]> <lm>
2 observation 2  ...      ...      <gg>      <tibble [10 x 2]> <lm>
3 observation 3  ...      ...      <gg>      <tibble [10 x 2]> <lm>
4 observation 4  ...      ...      <gg>      <tibble [10 x 2]> <lm>
5 observation 5  ...      ...      <gg>      <tibble [10 x 2]> <lm>
6 observation 6  ...      ...      <gg>      <tibble [10 x 2]> <lm>
7 observation 7  ...      ...      <gg>      <tibble [10 x 2]> <lm>
8 observation 8  ...      ...      <gg>      <tibble [10 x 2]> <lm>
9 observation 9  ...      ...      <gg>      <tibble [10 x 2]> <lm>
10 observation 10 ...      ...      <gg>      <tibble [10 x 2]> <lm>
# ... with 90 more rows
```

# Tidy R tools

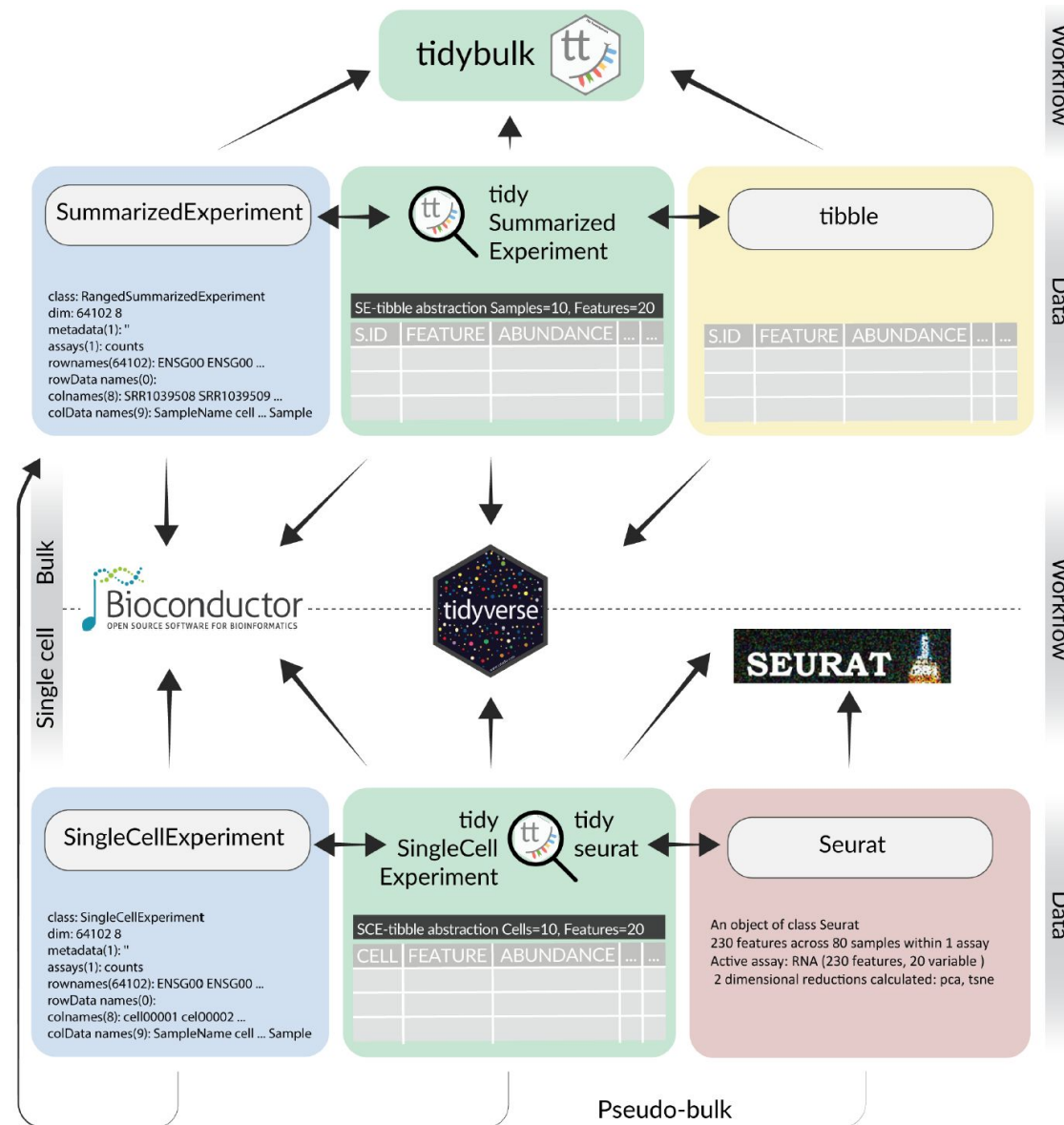
There are four basic principles to a tidy API:

- Reuse existing data structures.
- Compose simple functions with the pipe.
- Embrace functional programming.
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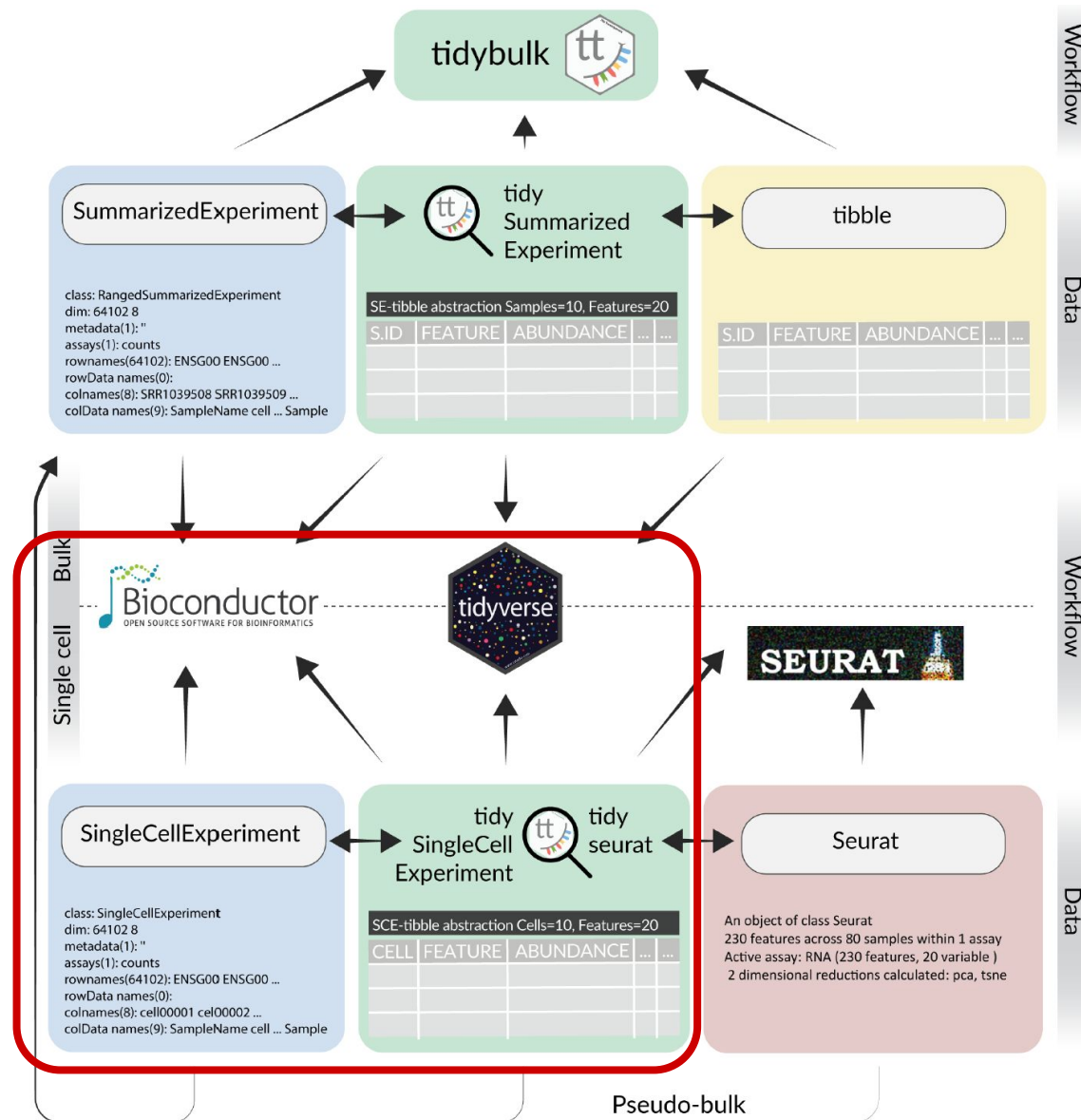
```
# A tibble: 100 x 8
  observation variable_1 variable_2 variable_3 variable_4 variable_5 variable_6 variable_7
  <glue>      <chr>      <chr>      <list>      <list>      <list>      <list>      <list>
1 observation 1  ...      ...      <gg>      <tibble [10 x 2]> <lm>      <Seurat[,80]> <SnglCl1E[,80...
2 observation 2  ...      ...      <gg>      <tibble [10 x 2]> <lm>      <Seurat[,80]> <SnglCl1E[,80...
3 observation 3  ...      ...      <gg>      <tibble [10 x 2]> <lm>      <Seurat[,80]> <SnglCl1E[,80...
4 observation 4  ...      ...      <gg>      <tibble [10 x 2]> <lm>      <Seurat[,80]> <SnglCl1E[,80...
5 observation 5  ...      ...      <gg>      <tibble [10 x 2]> <lm>      <Seurat[,80]> <SnglCl1E[,80...
6 observation 6  ...      ...      <gg>      <tibble [10 x 2]> <lm>      <Seurat[,80]> <SnglCl1E[,80...
7 observation 7  ...      ...      <gg>      <tibble [10 x 2]> <lm>      <Seurat[,80]> <SnglCl1E[,80...
8 observation 8  ...      ...      <gg>      <tibble [10 x 2]> <lm>      <Seurat[,80]> <SnglCl1E[,80...
9 observation 9  ...      ...      <gg>      <tibble [10 x 2]> <lm>      <Seurat[,80]> <SnglCl1E[,80...
10 observation 10 ...      ...      <gg>      <tibble [10 x 2]> <lm>      <Seurat[,80]> <SnglCl1E[,80...
# ... with 90 more rows
```



# The big picture



# The big picture



# Analysis infrastructure for single-cell data

## Data container



```
class: SingleCellExperiment
dim: 51958 3000
metadata(0):
assays(2): counts logcounts
rownames(51958): DDX11L1 WASH7P ... RP11-141019.1 RP11-
rowData names(0):
colnames(3000): CCAGTCACACTGGT-1 ATGAGCACATCTTC-1 ... (
colData names(7): file orig.ident ... G2M.Score ident
reducedDimNames(0):
mainExpName: NULL
altExpNames(0):
```

# Analysis infrastructure for single-cell data

## Data container



```
class: SingleCellExperiment
dim: 51958 3000
metadata(0):
assays(2): counts logcounts
rownames(51958): DDX11L1 WASH7P ... RP11-141019.1 RP11-
rowData names(0):
colnames(3000): CCAGTCACACTGGT-1 ATGAGCACATCTTC-1 ... (
colData names(7): file orig.ident ... G2M.Score ident
reducedDimNames(0):
mainExpName: NULL
altExpNames(0):
```

## Analysis

Bioconductor  
community

scraper/scater

## Manipulation

```
colData(data)
reducedDims(data, "umap")
subset(data, , class=="A")
data$info = info

data = cbind( data, cohort_info[
  match(data$sample, cohort_info$sample)
,])
subset(data, , !is.na(sample_id))
```

# Tidy data representation

## Data container

```
# A SingleCellExperiment-tibble abstraction: 80 x 15
# Features=230 | Assays=counts, logcounts
  cell      orig.ident  nCount_RNA nFeature_RNA RNA_snn_res.0.8 letter.ident groups RNA_snn_res.1  PC_1  PC_2  PC_3  PC_4  PC_5  tSNE_1  tSNE_2
  <chr>    <fct>        <dbl>      <int>    <fct>          <fct>    <chr> <fct>          <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
1 ATGCCAGAACGACT SeuratProject    70         47 0             A        g2    0             -0.774 -0.900 -0.249 0.559 0.465 0.868 -8.10
2 CATGGCCTGTGCAT SeuratProject    85         52 0             A        g1    0             -0.0260 -0.347 0.665 0.418 0.585 -7.39 -8.77
3 GAACCTGATGAACC SeuratProject    87         50 1             B        g2    0             -0.457 0.180 1.32 2.01 -0.482 -28.2 0.241
4 TGA CTGGATTCTCA SeuratProject   127         56 0             A        g2    0             -0.812 -1.38 -1.00 0.139 -1.60 16.3 -11.2
5 AGTCAGACTGCACA SeuratProject   173         53 0             A        g2    0             -0.774 -0.900 -0.249 0.559 0.465 1.91 -11.2
6 TCTGATACACGTGT SeuratProject    70         48 0             A        g1    0             -0.774 -0.900 -0.249 0.559 0.465 3.15 -9.94
7 TGGTATCTAAACAG SeuratProject    64         36 0             A        g1    0             -0.460 -1.19 -0.312 0.716 -1.65 17.9 -9.90
8 GCAGCTCTGTTTCT SeuratProject    72         45 0             A        g1    0             -0.900 -0.388 0.693 0.404 0.536 -6.49 -8.39
9 GATATAACACGCAT SeuratProject    52         36 0             A        g1    0             -0.774 -0.900 -0.249 0.559 0.465 1.33 -9.68
10 AATGTTGACAGTCA SeuratProject   100         41 0             A        g1    0             -0.488 -1.16 -0.306 0.702 -1.47 17.0 -9.43
# ... with 70 more rows
```



# Tidy analysis infrastructure

## Data container



```
# A SingleCellExperiment-tibble abstraction: 80 x 15
# Features=230 | Assays=counts, logcounts
  cell      orig.ident  nCount_RNA nFeature_RNA RNA_snn_res.0.8
  <chr>     <fct>         <dbl>      <int>    <fct>
1 ATGCCAGAACGACT SeuratProject    70        47 0
2 CATGGCCTGTGCAT SeuratProject    85        52 0
3 GAACCTGATGAACC SeuratProject    87        50 1
4 TGA CTGGATTCTCA SeuratProject   127        56 0
5 AGTCAGACTGCACA SeuratProject   173        53 0
6 TCTGATACACGTGT SeuratProject    70        48 0
7 TGGTATCTAAACAG SeuratProject    64        36 0
8 GCAGCTCTGTTTCT SeuratProject    72        45 0
9 GATATAACACGCAT SeuratProject    52        36 0
10 AATGTTGACAGTCA SeuratProject   100        41 0
# ... with 70 more rows
```

## Analysis

Bioconductor  
community

## Manipulation

```
data
data |> select(contains("UMAP"))
data |> filter(class=="A")
data |> mutate(info = info)

data |> inner_join(cohort_info, by="sample")
```

# Tidyseurat and tidySingleCellExperiment

## Data container



```
# A SingleCellExperiment-tibble abstraction: 80 x 15
# Features=230 | Assays=counts, logcounts
  cell      orig.ident  nCount_RNA nFeature_RNA RNA_snn_res.0.8
  <chr>     <fct>         <dbl>      <int>    <fct>
1 ATGCCAGAACGACT SeuratProject    70        47 0
2 CATGGCCTGTGCAT SeuratProject    85        52 0
3 GAACCTGATGAACC SeuratProject    87        50 1
4 TGA CTGGATTCTCA SeuratProject   127        56 0
5 AGTCAGACTGCACA SeuratProject   173        53 0
6 TCTGATACACGTGT SeuratProject    70        48 0
7 TGGTATCTAAACAG SeuratProject    64        36 0
8 GCAGCTCTGTTTCT SeuratProject    72        45 0
9 GATATAACACGCAT SeuratProject    52        36 0
10 AATGTTGACAGTCA SeuratProject   100        41 0
# ... with 70 more rows
```

## Analysis

Bioconductor  
community

```
data
data |> select(contains("UMAP"))
data |> filter(class=="A")
data |> mutate(info = info)

data |> inner_join(cohort_info, by="sample")
```

## Manipulation



```
# A Seurat-tibble abstraction: 80 x 15
# Features=230 | Active assay=RNA | Assays=RNA
  cell      orig.ident  nCount_RNA nFeature_RNA RNA_snn_res.0.8
  <chr>     <fct>         <dbl>      <int>    <fct>
1 ATGCCAGAACGACT SeuratProject    70        47 0
2 CATGGCCTGTGCAT SeuratProject    85        52 0
3 GAACCTGATGAACC SeuratProject    87        50 1
4 TGA CTGGATTCTCA SeuratProject   127        56 0
5 AGTCAGACTGCACA SeuratProject   173        53 0
6 TCTGATACACGTGT SeuratProject    70        48 0
7 TGGTATCTAAACAG SeuratProject    64        36 0
8 GCAGCTCTGTTTCT SeuratProject    72        45 0
9 GATATAACACGCAT SeuratProject    52        36 0
10 AATGTTGACAGTCA SeuratProject   100        41 0
# ... with 70 more rows
```

Seurat  
SeuratWrappers  
community

```
data
data |> select(contains("UMAP"))
data |> filter(class=="A")
data |> mutate(info = info)

data |> inner_join(cohort_info, by="sample")
```

# Tidy operators available

`as_tibble()`

`mutate()`

`bind_rows()`

`left_join()` `inner_join()` `*_join()`

`select()` **`distinct()`**

**`count()`** `add_count()` **`summarise()`**

**`pull()`** `slice()`

`filter()` `sample_n()` `sample_frac()`

`rename()`

`separate()` `unite()` `extract()`

`nest()` `unnest()` `map_*()`

**`pivot_longer()`**

`join_features()`

`ggplot()`

`plotly()`

# What tidy data frameworks are and what are not

**NO:** data containers

**NO:** analysis tools

**YES:** data interface

**YES:** manipulation, integration, visualisation tools

Therefore, the question “**can we go from tidyseurat to Seurat and vice versa**” is not relevant, as we never leave **Seurat**.