

# Restructuring Data

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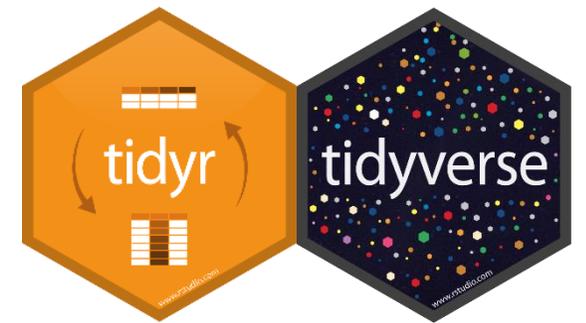
V2020-07

# Wide Format

Gene	WT_1	WT_2	WT_3	KO_1	KO_2	KO_3
ABC1	8.86	4.18	8.90	4.00	14.52	13.39
DEF1	29.60	41.22	36.15	11.18	16.68	1.64

- Compact
- Easy to read
- Shows linkage for genes
- No explicit genotype or replicate
- Values spread out over multiple rows and columns
- Not extensible to more metadata

# Long Format



Gene	Genotype	Replicate	Value
ABC1	WT	1	8.86
ABC1	WT	2	4.18
ABC1	WT	3	8.90
ABC1	KO	1	4.00
ABC1	KO	2	14.52
ABC1	KO	3	13.39
DEF1	WT	1	29.60
DEF1	WT	2	41.22
DEF1	WT	3	36.15
DEF1	KO	1	11.18
DEF1	KO	2	16.68
DEF1	KO	3	1.64

- More verbose (repeated values)
- Explicit genotype and replicate
- All values in a single column
- Extensible to more metadata



# Converting to "Tidy" format

```
# A tibble: 3 x 8
  Gene      Chr  Start      End  WT_1  WT_2  KO_1  KO_2
  <chr> <dbl>  <dbl>  <dbl> <dbl> <dbl> <dbl> <dbl>
1 Gnai3     2  163898  167465  9.39  10.9  33.5  81.9
2 Pbsn      5  4888573 4891351  91.7  59.6  45.3  82.3
3 Cdc45     7 1250084 1262669  69.2  36.1  54.4  38.1
```

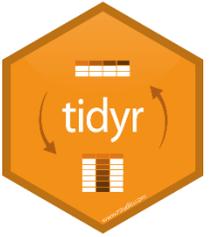
- Put all measures into a single column
- Add a 'genotype' and 'replicate' column
- Duplicate the gene information as required



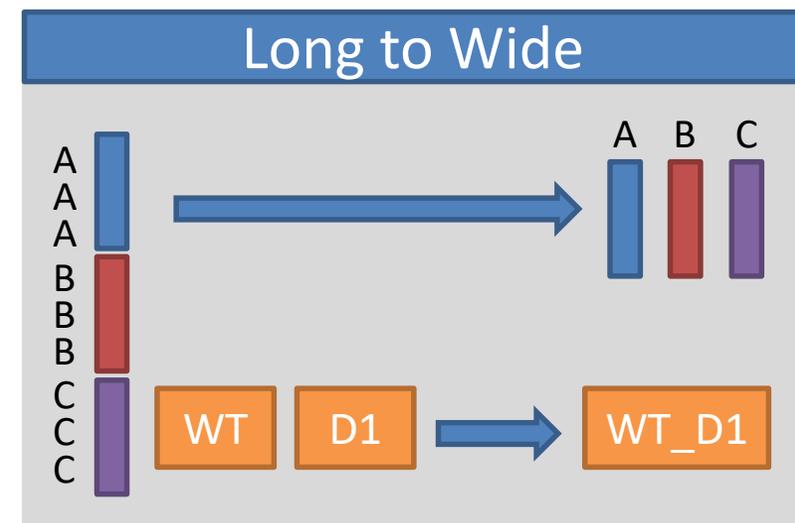
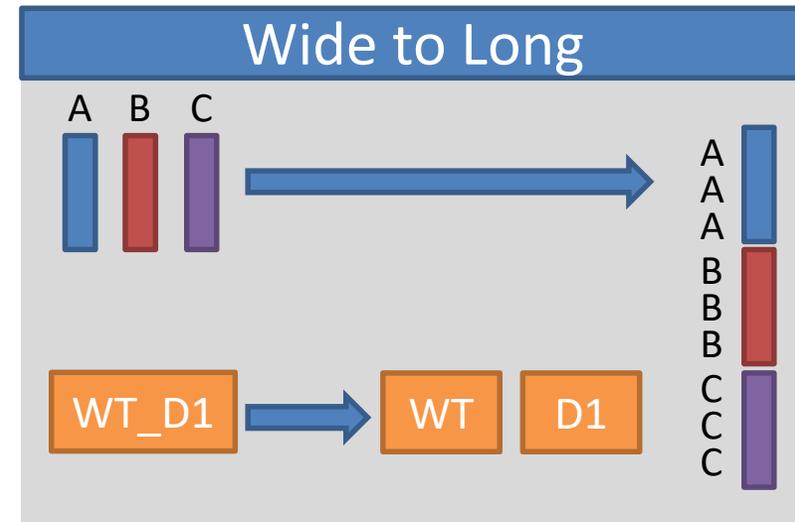
# Converting to "Tidy" format

```
# A tibble: 12 x 7
  Gene      Chr  Start      End genotype replicate value
  <chr> <dbl>  <dbl>    <dbl> <chr>      <int>  <dbl>
1 Gnai3     2  163898  167465 WT           1    9.39
2 Pbsn      5 4888573 4891351 WT           1   91.7
3 Cdc45     7 1250084 1262669 WT           1   69.2
4 Gnai3     2  163898  167465 WT           2   10.9
5 Pbsn      5 4888573 4891351 WT           2   59.6
6 Cdc45     7 1250084 1262669 WT           2   36.1
7 Gnai3     2  163898  167465 KO            1   33.5
8 Pbsn      5 4888573 4891351 KO            1   45.3
9 Cdc45     7 1250084 1262669 KO            1   54.4
10 Gnai3    2  163898  167465 KO            2   81.9
11 Pbsn     5 4888573 4891351 KO            2   82.3
12 Cdc45    7 1250084 1262669 KO            2   38.1
```

# Tidying operations



- **pivot\_longer**
  - Takes multiple columns of the same type and puts them into a pair of key-value columns
- **separate**
  - Splits a delimited column into multiple columns
- **pivot\_wider**
  - Takes a key-value column pair and spreads them out to multiple columns of the same type
- **unite**
  - Combines multiple columns into one



# Converting to "Tidy" format



```
non.normalised %>%
```

```
  pivot_longer(
    cols=WT_1:KO_2,
    names_to="sample",
    values_to="value"
  ) %>%
```

```
  separate(
    col=sample,
    into=c("genotype", "replicate"),
    convert = TRUE,
    sep="_"
  )
```

```
# A tibble: 3 x 8
  Gene      Chr  Start      End  WT_1  WT_2  KO_1  KO_2
  <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
1 Gnai3     2 163898 167465  9.39  10.9  33.5  81.9
2 Pbsn      5 4888573 4891351 91.7  59.6  45.3  82.3
3 Cdc45     7 1250084 1262669 69.2  36.1  54.4  38.1
```

```
# A tibble: 12 x 7
  Gene      Chr  Start      End genotype replicate value
  <chr> <dbl> <dbl> <dbl> <chr> <int> <dbl>
1 Gnai3     2 163898 167465 WT      1  9.39
2 Pbsn      5 4888573 4891351 WT      1 91.7
3 Cdc45     7 1250084 1262669 WT      1 69.2
```



**convert=TRUE** makes separate re-detect the type of the column, so replicate becomes a numeric value



# Pivoting Examples

```
> pivot.data
# A tibble: 4 x 3
  gene      WT      KO
  <chr> <dbl> <dbl>
1 ABC1  18608  7831
2 DEF1  31988  55502
3 GHI1   7647  93299
4 JKL1  96002  47945
```

- Log transform all of the values
- Pivot longer
  - Which columns are we pivoting?
  - What do we want to call the new column of names?
  - What do we want to call the new column of values?

```
pivot.data %>%
  pivot_longer(
    cols=WT:KO,
    names_to = "Condition",
    values_to = "Count"
  ) -> pivot.long
```

```
# A tibble: 8 x 3
  gene      Condition Count
  <chr> <chr>      <dbl>
1 ABC1   WT         18608
2 ABC1   KO          7831
3 DEF1   WT         31988
4 DEF1   KO         55502
5 GHI1   WT          7647
6 GHI1   KO         93299
7 JKL1   WT         96002
8 JKL1   KO         47945
```



# Pivoting Examples

```
> pivot.long
# A tibble: 8 x 3
  gene Condition Count
  <chr> <chr>     <dbl>
1 ABC1  WT         14.2
2 ABC1  KO         12.9
3 DEF1  WT         15.0
4 DEF1  KO         15.8
5 GHI1  WT         12.9
6 GHI1  KO         16.5
7 JKL1  WT         16.6
8 JKL1  KO         15.5
```

```
pivot.long %>%
  pivot_wider(
    names_from = Condition,
    values_from = Count
  )
```

```
# A tibble: 4 x 3
  gene      WT      KO
  <chr> <dbl> <dbl>
1 ABC1  14.2  12.9
2 DEF1  15.0  15.8
3 GHI1  12.9  16.5
4 JKL1  16.6  15.5
```

- Plot WT vs KO
- Pivot wider
  - Which column of names?
  - Which column of values?

# Exercise

## Restructuring data into 'tidy' format

We have provided 3 files which need to be restructured into tidy format. These all need turning from wide format to long format so use a combination of `pivot_wider` (and if needed, `separate`) to restructure them.

If you have time – take the restructured data and use `pivot_wider` (and if needed, `unite`) to put them back into wide format

# Grouping and Summarising

# Grouping and Summarising Workflow



1. Load a tibble with repeated values in one or more columns
2. Use `group_by` to select all of the categorical columns you want to combine to define your groups
3. Run `summarise` saying how you want to combine the quantitative values
4. Run `ungroup` to remove any remaining group information

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# Grouping and Summarising

```
> group.data
```

```
# A tibble: 8 x 5
```

```
  Sample Genotype Sex   Height Length
  <dbl>  <chr>   <chr>  <dbl>  <dbl>
1     1     WT     M      15     200
2     2     WT     F      13     185
3     3     WT     F      14     221
4     4     WT     M      18     265
5     5     KO     M      26     120
6     6     KO     F      22     165
7     7     KO     F      19     143
8     8     KO     M      27     110
```

- Want to get the average Height and Length for each combination of sex and genotype

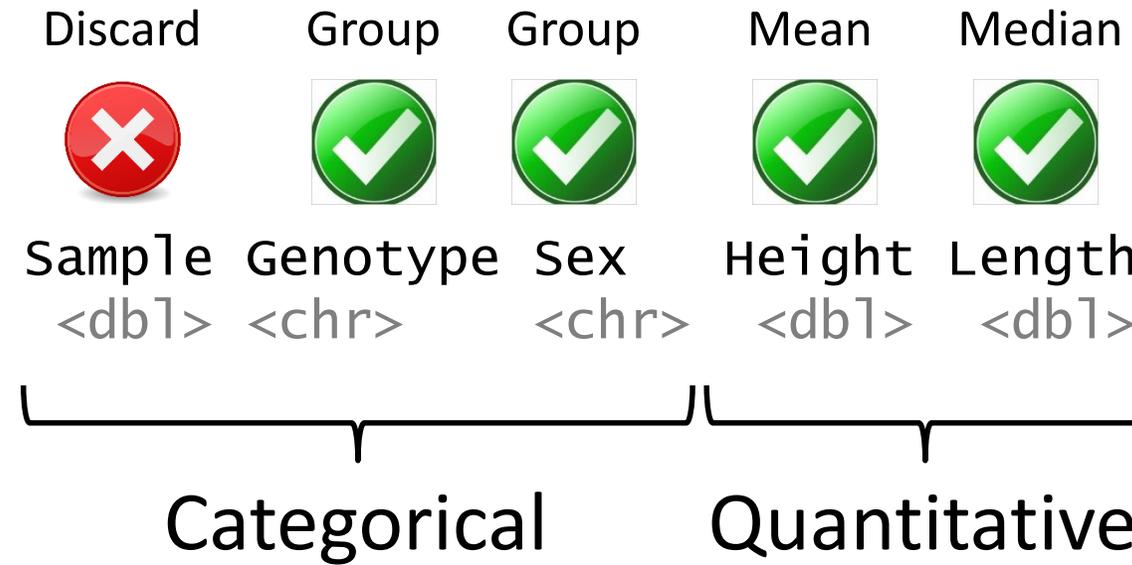
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# Grouping and Summarising



- Want to get the average Height and Length for each combination of sex and genotype



# Grouping and Summarising

```
group.data %>% group_by(Genotype, Sex)
```

```
# A tibble: 8 x 5
```

```
# Groups:   Genotype, Sex [4]
```

	Sample	Genotype	Sex	Height	Length
	<dbl>	<chr>	<chr>	<dbl>	<dbl>
1	1	WT	M	15	200
2	2	WT	F	13	185
3	3	WT	F	14	221
4	4	WT	M	18	265
5	5	KO	M	26	120
6	6	KO	F	22	165
7	7	KO	F	19	143
8	8	KO	M	27	110

Discard



Sample  
<dbl>

Group



Genotype  
<chr>

Group



Sex  
<chr>

Mean



Height  
<dbl>

Median



Length  
<dbl>

# Grouping and Summarising Workflow



1. Load a tibble with repeated values in one or more columns
2. Use `group_by` to select all of the categorical columns you want to combine to define your groups
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# Grouping and Summarising

```
group.data %>%  
  group_by(Genotype, Sex) %>%  
  count()
```

```
# A tibble: 4 x 3  
# Groups:   Genotype, Sex [4]  
  Genotype Sex      n  
  <chr>    <chr> <int>  
1 KO      F        2  
2 KO      M        2  
3 WT      F        2  
4 WT      M        2
```

Discard	Group	Group	Mean	Median
sample	Genotype	Sex	Height	Length
<dbl>	<chr>	<chr>	<dbl>	<dbl>



# Grouping and Summarising

```
group.data %>%  
  group_by(Genotype, Sex) %>%  
  summarise(  
    Height2=mean(Height),  
    Length=median(Length)  
  )
```

```
# A tibble: 4 x 4  
# Groups:   Genotype [2]  
  Genotype Sex    Height2 Length  
  <chr>    <chr>    <dbl>  <dbl>  
1 KO      F         20.5   154  
2 KO      M         26.5   115  
3 WT      F         13.5   203  
4 WT      M         16.5   232.
```

Discard	Group	Group	Mean	Median
Sample	Genotype	Sex	Height	Length
<dbl>	<chr>	<chr>	<dbl>	<dbl>



If you want the count of values as part of a summarised result use the `n()` function

# Grouping and Summarising Workflow



1. Load a tibble with repeated values in one or more columns
2. Use `group_by` to select all of the categorical columns you want to combine to define your groups
3. Run `summarise` saying how you want to combine the quantitative values
4. Run `ungroup` to remove any remaining group information



# Ungrouping

- A summarise operation removes the the last level of grouping (“Sex” in our worked example)
- Other levels of grouping (“Genotype”) remain annotated on the data, so you could do an additional summarisation if needed
- If you’re not going to use them it’s a good idea to use `ungroup` to remove remaining groups so they don’t interfere with other operations

# Grouping affects lots of operations

Find the tallest member of each Sex



```
group.data %>%  
  arrange(desc(Height)) %>%  
  group_by(Sex) %>%  
  slice(1)
```

```
# A tibble: 2 x 5  
# Groups:   Sex [2]  
  Sample Genotype Sex    Height Length  
  <dbl> <chr>   <chr> <dbl> <dbl>  
1     6 KO     F      22    165  
2     8 KO     M      27    110
```

# Exercise

## Grouping and Summarising

In your restructured tidy2 data find the mean and sem (sem is  $\text{mean}/\sqrt{n}$ ) for groups A to E. You'll need to group by sample and then summarise to calculate the two values. You get the number of observations using the `n()` function

If you have time – use the summarised values to plot a barplot of the data. Use `geom_col()` and `geom_errorbar()`. You'll need to set the `ymin` and `ymax` aesthetics for `geom_errorbar` – these will be the `mean-sem` and `mean+sem`