# Aesthetics in ggraph

Your name here

# Two types of aesthetics

starwars

There are two main types of "aesthetics" in ggraph: changes that apply to all of the nodes/edges, and changes that differ based on some attribute of the nodes/edges.

For this tutorial, we'll be visualizing data from a 1990 paper on the French financial elite (ffe) in the networkdata package.

Let's load the data as a tbl\_graph

```
library(networkdata)

Attaching package: 'networkdata'

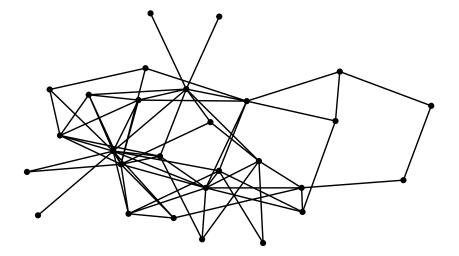
The following object is masked from 'package:dplyr':
```

```
G <- ffe_friends |> as_tbl_graph()
```

Now, let's see what it looks like it our basic ggraph plot.

```
G |>
    ggraph() +
    geom_edge_fan() +
    geom_node_point()
```

Using "stress" as default layout



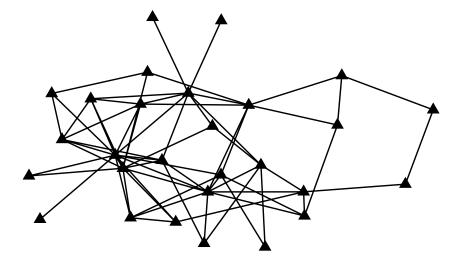
# **General aesthetics**

We've already learned about the first kind of aesthetics—general aesthetics that apply to the whole graph. Things like changing the size of the nodes or the color of the edges.

For example, this changes the shape of the nodes to triangles and makes them a bit bigger.

```
G |>
    ggraph() +
    geom_edge_fan() +
    geom_node_point(shape='triangle', size = 3)
```

Using "stress" as default layout



#### **Exercise**

Change the width of the edges to .5 and their color to "purple"

# Your code here

# "Mapping" aesthetics

The second type of aesthetics maps some aspect of the data to some aspect of the visualization. For example, we might color nodes differently based on their major or size them differently based on their centrality.

Let's start by figuring out what the attributes of the nodes are in this data.

G

```
# A tbl_graph: 28 nodes and 66 edges
#
# An undirected simple graph with 1 component
#
# Node Data: 28 x 26 (active)
```

	birthdate	birthplace	cabinet	clubs	eliteprom	elitevote	ena	enayear	
	<dbl></dbl>								
1	1943	2	1	2	1	24	1	1970	
2	1935	2	2	3	1	52	2	NA	
3	1924	1	2	1	1	45	2	NA	
4	1933	2	2	0	1	28	1	1960	
5	1931	1	1	3	1	45	1	1957	
6	1934	2	1	3	0	6	1	1962	
7	1932	1	2	1	1	48	2	NA	
8	1931	1	2	3	1	34	2	NA	
9	1911	2	2	0	0	13	2	NA	
10	1935	1	2	3	1	23	2	NA	
# i 18 more rows									
# i 18 more variables: fathers.lev <dbl>, finance.min <dbl>, igyear <dbl>,</dbl></dbl></dbl>									
<pre># inspec.gen <dbl>, masons <dbl>, na <lgl>, normal.sch <dbl>, party <dbl>,</dbl></dbl></lgl></dbl></dbl></pre>									
<pre># polytech <dbl>, polyyear <dbl>, prestige <dbl>, religion <dbl>,</dbl></dbl></dbl></dbl></pre>									
<pre># sciencepoly <dbl>, socialreg <dbl>, topboards <dbl>, university <dbl>,</dbl></dbl></dbl></dbl></pre>									
<pre># vertex.names <chr>, zipcode <dbl></dbl></chr></pre>									
#			_						
# Edge Data: 66 x 2									
from to									
<int> <int></int></int>									
1	2 6	3							
2	3 6	3							
3	1 7	7							
# i 63 more rows									

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This is rich data! There are 26 different variables, some of which are clear (e.g, birthdate) and some of which are confusing (e.g., topboards or igyear). Let's choose one and visualize it.

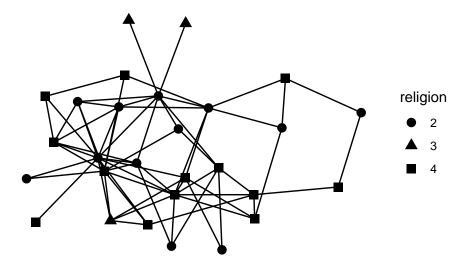
Usually, you would want to be guided by a question about the data which would inform your visualizations. Let's pretend like we really want to know whether people form friendships across religions. Let's visualize different religion as different shapes.

To do that, we "map" the shape of the nodes to the religion variable. The syntax to do this is to put the thing we want to change, then =, then the variable (column name) we want to use to change it. So, for changing shape to religion, it's shape = religion. We put that whole thing in aes (which stands for mapping aesthetic), and put it into the geom\_node\_point function.

```
G |>
mutate(religion = as_factor(religion)) |>
```

```
ggraph() +
geom_edge_fan() +
geom_node_point(aes(shape=religion), size = 3)
```

Using "stress" as default layout



Note the line mutate(religion = as\_factor(religion)) Try commenting out that line and running the code again.

You should get an error that is something like A continuous variable can not be mapped to shape. This is saying that in this data religion is a number and it doesn't make much sense to map a number to a shape, because usually numbers are ordered while shapes are unordered. So, we have to add the mutate step to change the religion variable type using as\_factor which changes how R sees it from a number to a factor—in other words, a category (which is what religion really is, anyway!).

## **Exercise**

Instead of mapping their shape to religion, map the color of the nodes to religion.

#### # Your code here

There are a number of node aesthetics that we can map variables to. The most common are: shape, size, and color. Let's do one more exercise to practice these.

#### **Exercise**

Go back to your code from when we learned about mutate and figure out how to calculate degree centrality for our network; save the centrality in a column called degree and change the size of nodes based on the degree.

# Your code here

## **Edge** aesthetics

We can also map the edge aesthetics to variables from our edgelist spreadsheet. The most common approaches are to map the type of the edge to color and the weight of the edge to the width, color, or alpha of the edge.

Looking at our data, the edges don't have weights. For your exercise, we'll put together a few skills from the last few labs.

#### **Exercise**

Load the bkfrab dataset (also from networkdata, so load it in the same way as the ffe\_friends data above). This records how often pairs of subjects were seen talking to each other in a fraternity. Map the alpha of the edge to the weight in the edges dataframe.

Hints: Remember to use as\_tbl\_graph to change the network type, and to activate the edges dataframe.

# Your code here

# **Challenge Exercise**

When we look at the bkfrab data, there are a lot of nodes with edge weight 10. Filter to just the edges whose weight is greater than 10, and map the color of the edges to their weight.

Note that this will leave a lot of nodes without any edges. The default 'stress' layout puts them along the bottom, which looks kind of odd. You may want to use a layout like 'fr' instead.

# # Your code here

You now have all of the key skills to load in, mutate and filter networks, and to do some pretty cool visualizations with them. From here on out, we'll mostly be reusing these skills in new ways.