Chapter 7. Scales

"Scales are functions that map from an input domain to an output range."

That's Mike Bostock's definition of D3 scales, and there is no clearer way to say it.

The values in any dataset are unlikely to correspond exactly to pixel measurements for use in your visualization. Scales provide a convenient way to map those data values to new values useful for visualization purposes.

D3 scales are *functions* with parameters that you define. Once they are created, you call the scale function, pass it a data value, and it nicely returns a scaled output value. You can define and use as many scales as you like.

It might be tempting to think of a scale as something that appears visually in the final image—like a set of tick marks, indicating a progression of values. *Do not be fooled!* Those tick marks are part of an *axis*, which is a *visual representation* of a scale. A scale is a mathematical relationship, with no direct visual output. I encourage you to think of scales and axes as two different, yet related, elements.

This chapter addresses only <u>linear</u> scales, because they are most common and easiest understood. Once you understand linear scales, the others—ordinal, logarithmic, square root, and so on—will be a piece of cake.

Apples and Pixels

Imagine that the following dataset represents the number of apples sold at a roadside fruit stand each month:

```
var dataset = [ 100, 200, 300, 400, 500 ];
```

First of all, this is great news, as the stand is selling 100 additional apples each month! Business is booming. To showcase this success, you want to make a bar chart illustrating the steep upward climb of apple sales, with each data value corresponding to the height of one bar.

Until now, we've used data values directly as display values, ignoring unit differences. So if 500 apples were sold, the corresponding bar would be 500 pixels tall.

That could work, but what about next month, when 600 apples are sold? And a year later, when 1,800 apples are sold? Your audience would have to purchase ever-larger displays just to be able to see the full height of those very tall apple bars! (Mmm, apple bars!)

This is where scales come in. Because apples are not pixels (which are also not oranges), we need scales to translate between them.

Domains and Ranges

A scale's *input domain* is the range of possible input data values. Given the preceding apples data, appropriate input domains would be either 100 and 500 (the minimum and maximum values of the dataset) or 0 and 500.

A scale's *output range* is the range of possible output values, commonly used as display values in pixel units. The output range is completely up to you, as the information designer. If you decide the shortest apple bar

will be 10 pixels tall, and the tallest will be 350 pixels tall, then you could set an output range of 10 and 350.

For example, create a scale with an input domain of [100,500] and an output range of [10,350]. If you handed the low input value of 100 to that scale, it would return its lowest range value, 10. If you gave it 500, it would spit back 350. If you gave it 300, it would hand 180 back to you on a silver platter. (300 is in the center of the domain, and 180 is in the center of the range.)

We can visualize the domain and range as corresponding axes, side-by-side, displayed in Figure 7-1.

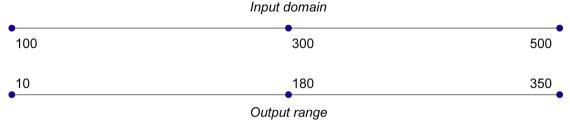


Figure 7-1. An input domain and an output range, visualized as parallel axes

One more thing: to prevent your brain from mixing up the *input domain* and *output range* terminology, I'd like to propose a little exercise. When I say "input," you say "domain." Then I say "output," and you say "range." Ready? Okay:

Input! Domain!

Output! Range!

Input! Domain!

Output! Range!

Got it? Great.

Normalization

If you're familiar with the concept of *normalization*, it might be helpful to know that, with a linear scale, that's all that is really going on here.

Normalization is the process of mapping a numeric value to a new value between 0 and 1, based on the possible minimum and maximum values. For example, with 365 days in the year, day number 310 maps to about 0.85, or 85 percent of the way through the year.

With linear scales, we are just letting D₃ handle the math of the normalization process. The input value is normalized according to the domain, and then the normalized value is scaled to the output range.

Creating a Scale

D3's scale function generators are accessed with d3.scale followed by the type of scale you want. I recommend opening up the sample code page *O1_scale_test.html* and typing each of the following into the console:

```
var scale = d3.scale.linear();
```

Congratulations! Now scale is a function to which you can pass input values. (Don't be misled by the var.

Remember that in JavaScript, variables can store functions.)

```
scale(2.5); //Returns 2.5
```

Because we haven't set a domain and a range yet, this function will map input to output on a 1:1 scale. That is, whatever we input will be returned unchanged.

We can set the scale's input domain to 100,500 by passing those values to the domain() method as an array. Note the hard brackets indicating an array:

```
scale.domain([100, 500]);
```

Set the output range in similar fashion, with range():

```
scale.range([10, 350]);
```

These steps can be done separately, as just shown, or chained together into one line of code:

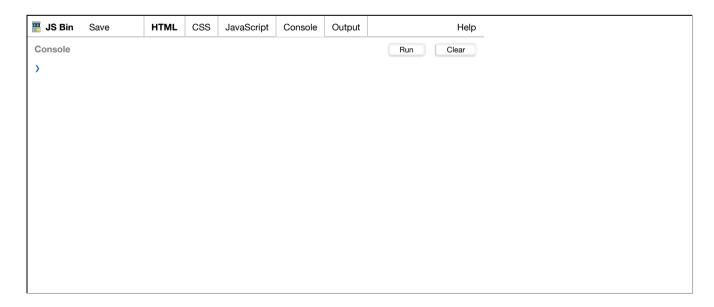
Either way, our scale is ready to use!

```
scale(100);  //Returns 10
scale(300);  //Returns 180
scale(500);  //Returns 350
```

Try It Now!

Try entering the following scale statements in the console below to see the results:

```
scale(200);scale(450.5);scale(-100);
```



Typically, you will call scale functions from within an attr() method or similar, not on their own. Let's modify our scatterplot visualization to use dynamic scales.

Scaling the Scatterplot

To revisit our dataset from the scatterplot:

You'll recall that this dataset is an array of arrays. We mapped the first value in each array onto the x-axis, and the second value onto the y-axis. Let's start with the x-axis.

Just by eyeballing the x values, it looks like they range from 5 to 480, so a reasonable input domain to specify might be 0,500, right?

Why are you giving me that look? Oh, because you want to keep your code flexible and scalable, so it will continue to work even if the data changes in the future. Very smart! Remember, if we were building a data dashboard for the roadside apple stand, we'd want our code to accommodate the enormous projected growth in apple sales. Our chart should work just as well with 5 apples sold as 5 million.

d3.min() and d3.max()

Instead of specifying fixed values for the domain, we can use the convenient array functions d3.min() and d3.max() to analyze our data set on the fly. For example, this loops through each of the x values in our arrays and returns the value of the greatest one:

```
d3.max(dataset, function(d) {
    return d[0]; //References first value in each subarray
});
```

That code will return the value 480, because 480 is the largest x value in our dataset. Let me explain how it works.

Both min() and max() work the same way, and they can take either one or two arguments. The first argument must be a reference to the array of values you want evaluated, which is dataset, in this case. If you have a simple, one-dimensional array of numeric values, like [7, 8, 4, 5, 2], then it's obvious how to compare the values against each other, and no second argument is needed. For example:

```
var simpleDataset = [7, 8, 4, 5, 2];
d3.max(simpleDataset); // Returns 8
```

The max() function simply loops through each value in the array, and identifies the largest one.

But our dataset is not just an array of numbers; it is an array of arrays. Calling d3.max(dataset) might produce unexpected results:

To tell max() which *specific* values we want compared, we must include a second argument, an *accessor* function:

```
d3.max(dataset, function(d) {
    return d[0];
});
```

The accessor function is an anonymous function to which max() hands off each value in the data array, one at a time, as d. The accessor function specifies *how to access* the value to be used for the comparison. In this case, our data array is dataset, and we want to compare only the x values, which are the first values in each subarray, meaning in position [0]. So our accessor function looks like this:

```
function(d) {
    return d[0]; //Return the first value in each subarray
}
```

Note that this looks suspiciously similar to the syntax we used when generating our scatterplot circles, which also used anonymous functions to retrieve and return values:

```
.attr("cx", function(d) {
    return d[0];
})
.attr("cy", function(d) {
    return d[1];
})
```

This is a common D3 pattern. Soon you will be very comfortable with all manner of anonymous functions crawling all over your code.

Setting Up Dynamic Scales

Putting together what we've covered, let's create the scale function for our x-axis:

First, notice that I named it xScale. Of course, you can name your scales whatever you want, but a name like xScale helps me remember what this function does.

Second, notice that both the domain and range are specified as two-value arrays in hard brackets.

Third, notice that I set the low end of the input domain to o. (Alternatively, you could use min() to calculate a dynamic value.) The upper end of the domain is set to the maximum value in dataset (which is currently 480, but could change in the future).

Finally, observe that the output range is set to 0 and w, the SVG's width.

We'll use very similar code to create the scale function for the y-axis:

```
.range([0, h]);
```

Note that the max() function here references d[1], the y value of each subarray. Also, the upper end of range() is set to h instead of w.

The scale functions are in place! Now all we need to do is use them.

Incorporating Scaled Values

})

Revisiting our scatterplot code, we now simply modify the original line where we created a circle for each data value:

```
.attr("cx", function(d) {
    return d[0]; //Returns original value bound from dataset
})
to return a scaled value (instead of the original value):
.attr("cx", function(d) {
    return xScale(d[0]); //Returns scaled value
})
Likewise, for the y-axis, this:
.attr("cy", function(d) {
    return d[1];
})
is modified as:
.attr("cy", function(d) {
    return yScale(d[1]);
})
For good measure, let's make the same change where we set the coordinates for the text labels, so these
lines:
.attr("x", function(d) {
    return d[0];
})
.attr("y", function(d) {
    return d[1];
})
become this:
.attr("x", function(d) {
    return xScale(d[0]);
})
.attr("y", function(d) {
    return yScale(d[1]);
```

And there we are!

Check out the working code in *o2_scaled_plot.html*. Visually, the result in <u>Figure 7-2</u> is disappointingly similar to our original scatterplot! Yet we are making more progress than might be apparent.

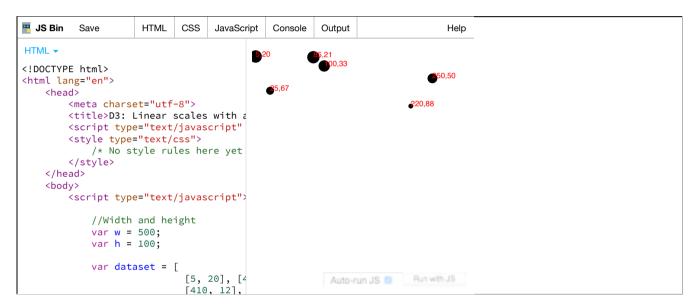


Figure 7-2. Scatterplot using x and y scales

Try It Now!

Experiment with the code for the scatterplot below. Adjust the w and h values to see the effect on the scale. Do the same with the dataset values.

Open in new window



Refining the Plot

You might have noticed that smaller y values are at the top of the plot, and the larger y values are toward the bottom. Now that we're using D3 scales, it's super easy to reverse that, so greater values are higher up, as you would expect. It's just a matter of changing the output range of yScale from:

```
.range([0, h]);
to:
.range([h, 0]);
```

See *o3_scaled_plot_inverted.html* for the code that results in <u>Figure 7-3</u>.



Figure 7-3. Scatterplot with y scale inverted

Yes, now a smaller input to yScale will produce a larger output value, thereby pushing those circles and

text elements down, closer to the base of the image. I know, it's almost too easy!

Yet some elements are getting cut off. Let's introduce a padding variable:

```
var padding = 20;
```

Then we'll incorporate the padding amount when setting the range of both scales. This will help push our elements in, away from the edges of the SVG, to prevent them from being clipped.

The range for xScale was range([0, w]), but now it's:

```
.range([padding, w - padding]);
```

The range for yScale was range([h, 0]), but now it's:

```
.range([h - padding, padding]);
```

This should provide us with 20 pixels of extra room on the left, right, top, and bottom edges of the SVG. And it does (see Figure 7-4)!



Figure 7-4. Scatterplot with padding

But the text labels on the far right are still getting cut off, so I'll double the amount of xScale's padding on the right side by multiplying by two to achieve the result shown in <u>Figure 7-5</u>:

```
.range([padding, w - padding * 2]);

20,88 330,95 480,90

25,67 550,50 975,44
```

Figure 7-5. Scatterplot with more padding

The way I've introduced padding here is simple, but not elegant. Eventually, you'll want more control over how much padding is on each side of your charts (top, right, bottom, left), and it's useful to standardize how you specify those values across projects. Although I haven't used Mike Bostock's margin convention for the code samples in this book, I recommend taking a look to see if it could work for you.

Better! Reference the code so far in *04_scaled_plot_padding.html*. But there's one more change I'd like to make. Instead of setting the radius of each circle as the square root of its y value (which was a bit of a hack, and not visually useful), why not create another custom scale?

Then, setting the radius looks like this:

```
.attr("r", function(d) {
    return rScale(d[1]);
```

});

This is exciting because we are guaranteeing that our radius values will *always* fall within the range of 2,5. (Or *almost* always; see reference to clamp() later.) So data values of 0 (the minimum input) will get circles of radius 2 (or a diameter of 4 pixels). The very largest data value will get a circle of radius 5 (diameter of 10 pixels).

Voila: Figure 7-6 shows our first scale used for a visual property other than an axis value. (See o5_scaled_plot_radii.html.)



Figure 7-6. Scatterplot with scaled radii

Finally, just in case the power of scales hasn't yet blown your mind, I'd like to add one more array to the dataset: [600, 150].

Boom! Check out *o6_scaled_plot_big.html*. Notice how all the old points in <u>Figure 7-7</u> maintained their relative positions but have migrated closer together, down and to the left, to accommodate the newcomer in the top-right corner.



Figure 7-7. Scatterplot with big numbers added

And now, one final revelation: we can now very easily change the size of our SVG, and *everything scales accordingly*. In <u>Figure 7-8</u>, I've increased the value of h from 100 to 300 and made *no other changes*.

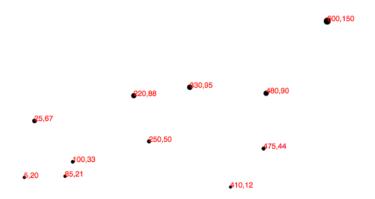


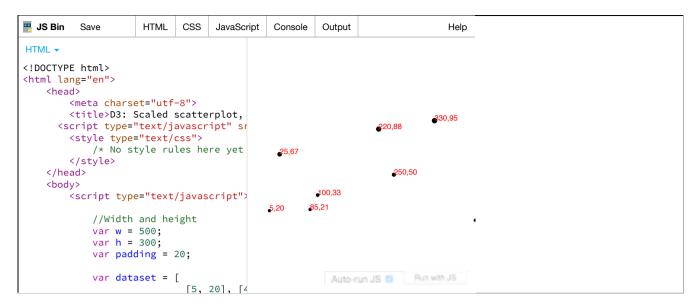
Figure 7-8. Large, scaled scatterplot

Boom, again! See *o7_scaled_plot_large.html*. Hopefully, you are seeing this and realizing: no more late nights tweaking your code because the client decided the graphic should be 800 pixels wide instead of 600. Yes, you will get more sleep because of me (and D3's brilliant built-in methods). Being well-rested is a competitive advantage. You can thank me later.

Try It Now!

Experiment with the code for the final scatterplot below. Adjust the code at left and see the chart output at right.

Open in new window



Other Methods

d3.scale.linear() has several other handy methods that deserve a brief mention here:

nice()

This tells the scale to take whatever input domain that you gave to range() and expand both ends to the nearest round value. From the D3 wiki: "For example, for a domain of [0.20147987687960267, 0.996679553296417], the nice domain is [0.2, 1]." This is useful for normal people, who are not computers and find it hard to read numbers like 0.20147987687960267.

rangeRound()

Use rangeRound() in place of range(), and all values output by the scale will be rounded to the nearest whole number. This is useful if you want shapes to have exact pixel values, to avoid the fuzzy edges that could arise with antialiasing.

clamp()

By default, a linear scale *can* return values outside of the specified range. For example, if given a value outside of its expected input domain, a scale will return a number also outside of the output range. Calling clamp(true) on a scale, however, forces all output values to be within the specified range. This means excessive values will be rounded to the range's low or high value (whichever is nearest).

To use any of these special methods, just tack them onto the chain in which you define the original scale function. For example, to use nice():

Other Scales

In addition to <u>linear scales</u> (discussed earlier), D3 has several other built-in scale methods:

sgrt

A square root scale.

<u>pow</u>

A power scale (good for the gym, er, I mean, useful when working with exponential series of values, as in "to the power of" some exponent).

log

A logarithmic scale.

quantize

A linear scale with discrete values for its output range, for when you want to sort data into "buckets." guantile

Similar to quantize, but with discrete values for its input domain (when you already have "buckets").

ordinal

Ordinal scales use nonquantitative values (like category names) for output; perfect for comparing apples and oranges.

d3.scale.category10(), d3.scale.category20(), d3.scale.category20b(), and
d3.scale.category20c()

Handy preset ordinal scales that output either 10 or 20 categorical colors.

d3.time.scale()

A scale method for date and time values, with special handling of ticks for dates.

Now that you have mastered the power of scales, it's time to express them visually as, yes, axes!

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