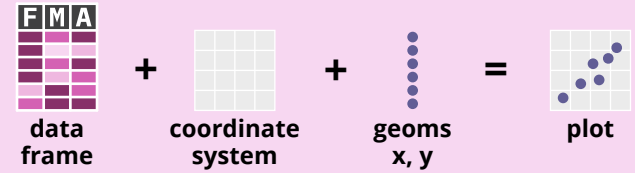


Data visualization with Plotnine # cheatsheet

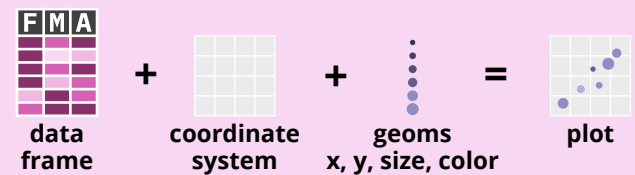


Basics

Plotnine is based on the **grammar of graphics**, the idea that you can build every plot from the same components: a **data frame**, a **coordinate system**, and **geoms**—visual marks that represent data points.



To display values, map columns in the data to visual properties of the geom (**aesthetics**) like **size**, **color**, and **x** and **y** locations.



Complete the template below to build a plot:

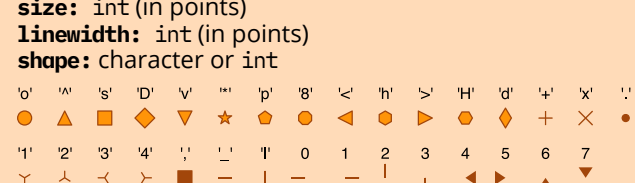
```
from plotnine import *
from plotnine.data import *

(
  ggplot(data=data_frame)
  + geom_function(
    mapping=aes(**mappings),
    stat=stat,
    position=position)
  + coord_function()
  + facet_function()
  + scale_function()
  + theme_function()
  + theme(**settings)
)
```

```
For example:
p = ggplot(mpg, aes("cty", "hwy")) + geom_point()
p.save("plot.png", width=6, height=4, dpi=200)
```

Aesthetics

color and **fill**: color name or "#RRGGBB"
linetype: str one of ["solid", "dashed", "dashdot", "dotted"] or tuple of the form (offset, (on, off, on, off, ...))
size: int (in points)
linewidth: int (in points)
shape: character or int



Geoms

Use a geom function to represent data points, use the geom's aesthetic properties to represent variables. Each function returns a layer.

Graphical primitives

```
a = ggplot(economics, aes("date", "unemploy"))
b = ggplot(seals, aes(x="long", y="lat"))

a + geom_blank() # Ensure limits include values across all plots

a + geom_path(lineend="butt", linejoin="round", linemitre=1) # x, y, alpha, color, group, linetype, size

a + geom_polygon(aes(alpha=50)) # color, fill, group, subgroup, linetype, size

b + geom_rect(aes(xmin="long", ymin="lat", xmax="long+1", ymax="lat+1")) # xmin, xmax, ymin, ymax, alpha, color, fill, linetype, size

a + geom_ribbon(aes(ymin="unemploy-900", ymax="unemploy+900")) # x, ymin, ymax, alpha, color, fill, group, linetype, size
```

Line segments

```
Common aesthetics: x, y, alpha, color, linetype, size

b + geom_abline(aes(intercept=0, slope=1))
b + geom_hline(aes(yintercept="lat"))
b + geom_vline(aes(xintercept="long"))

b + geom_segment(aes(yend="lat+1", xend="long+1"))
b + geom_spoke(aes(angle=1:1155, radius=1))
```

One continuous variable

```
c = ggplot(mpg, aes(x="hwy")); c2 = ggplot(mpg)

c + geom_area(stat="bin") # x, y, alpha, color, fill, linetype, size

c + geom_density(kernel="gaussian") # x, y, alpha, color, fill, group, linetype, size, weight

c + geom_dotplot() # x, y, alpha, color, fill

c + geom_freqpoly() # x, y, alpha, color, group, linetype, size

c + geom_histogram(binwidth=5) # x, y, alpha, color, fill, linetype, size, weight

c2 + geom_qq(aes(sample="hwy")) # x, y, alpha, color, fill, linetype, size, weight
```

One discrete variable

```
d = ggplot(mpg, aes("fl"))

d + geom_bar() # x, alpha, color, fill, linetype, size, weight
```

Two continuous variables

```
e = ggplot(mpg, aes(x="cty", y="hwy"))

e + geom_label(aes(label="cty"), nudge_x=1, nudge_y=1) # alpha, angle, color, family, fontface, hjust, lineheight, size, vjust

e + geom_point() # x, y, alpha, color, fill, shape, size, stroke

e + geom_quantile() # x, y, alpha, color, group, linetype, size, weight

e + geom_rug(sides="bl") # x, y, alpha, color, linetype, size

e + geom_smooth(method="lm") # x, y, alpha, color, fill, group, linetype, size, weight

e + geom_text(aes(label="cty"), nudge_x=1, nudge_y=1) # alpha, angle, color, family, fontface, hjust, lineheight, size, vjust
```

One discrete, one continuous variable

```
f = ggplot(mpg, aes(x="class", y="hwy"))

f + geom_col() # x, y, alpha, color, fill, group, linetype, size

f + geom_boxplot() # x, y, lower, middle, upper, ymax, ymin, alpha, color, fill, group, linetype, shape, size, weight

f + geom_dotplot(binaxis="y", stackdir="center") # x, y, alpha, color, fill, group

f + geom_violin(scale="area") # x, y, alpha, color, fill, group, linetype, size, weight
```

Two discrete variables

```
g = ggplot(diamonds, aes(x="cut", y="color"))

g + geom_count() # x, y, alpha, color, fill, shape, size, stroke

e + geom_jitter(height=2, width=2) # x, y, alpha, color, fill, shape, size
```

Three continuous variables

```
seals = pl.DataFrame(seals).with_columns(z=(pl.col("delta_long")**2 + pl.col("delta_lat")**2).sqrt())
l = ggplot(seals, aes(long, lat))

l + geom_contour(aes(z="z")) # x, y, z, alpha, color, group, linetype, size, weight

l + geom_contour_filled(aes(fill="z")) # x, y, alpha, color, fill, group, linetype, size, subgroup
```

Continuous function

```
h = ggplot(economics, aes("date", "unemploy"))

h + geom_area() # x, y, alpha, color, fill, linetype, size

h + geom_line() # x, y, alpha, color, group, linetype, size

h + geom_step(direction="hv") # x, y, alpha, color, group, linetype, size
```

Continuous bivariate distribution

```
i = ggplot(diamonds, aes("carat", "price"))

i + geom_bin_2d(binwidth=[0.25, 500]) # x, y, alpha, color, fill, linetype, size, weight

i + geom_density_2d() # x, y, alpha, color, group, linetype, size
```

Visualizing error

```
df = pl.DataFrame({"grp": ["A", "B", "C"], "fit": [4, 5, 6], "se": [1, 2, 1]}); j = ggplot(df, aes("grp", "fit", ymin="fit-se", ymax="fit+se"))

j + geom_crossbar(fatten=2) # x, y, ymax, ymin, alpha, color, fill, group, linetype, size

j + geom_errorbar(); j + geom_errorbarh() # x, ymax, ymin, alpha, color, group, linetype, size, width

j + geom_linerange() # x, ymin, ymax, alpha, color, group, linetype, size

j + geom_pointrange() # x, y, ymin, ymax, alpha, color, fill, group, linetype, shape, size
```

Maps

```
import geopandas as gp
import geodatasets as gd

np = gp.read_file(gd.get_path("geoda.nepal"))

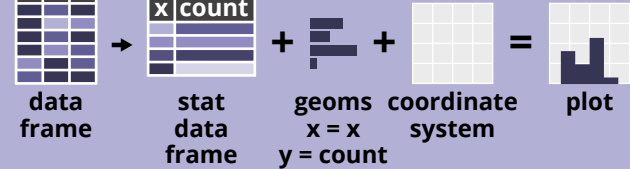
ggplot(np) + geom_map(aes(fill="population"))
```



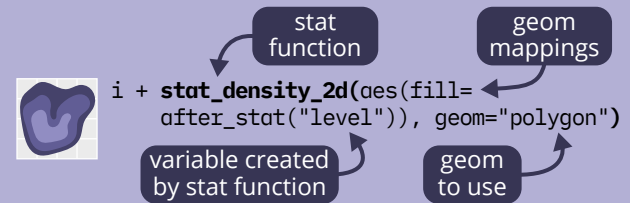
Stats

An alternative way to build a layer.

A stat builds new variables to plot (e.g., count, prop).



Visualize a stat by changing the default stat of a geom function, `geom_bar(stat="count")` or by using a stat function, `stat_count(geom="bar")`, which calls a default geom to make a layer (equivalent to a function). Use `after_stat(name)` syntax to map the stat variable name to an aesthetic.



```

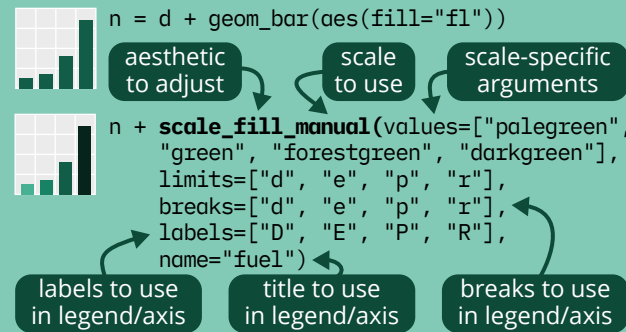
c + stat_bin(binwidth=1, boundary=10) # x, y | count, ncount, density, ndensity
c + stat_count(width=1) # x, y | count, prop
c + stat_density(adjust=1, kernel="gaussian") # x, y | count, density, scaled
.....
e + stat_bin_2d(bins=30, drop=True) # x, y, fill | count, density
e + stat_density_2d(contour=True, n=100) # x, y, color, size | level
e + stat_ellipse(level=0.95, segments=51, type="t")
.....
l + stat_contour(aes(z="z")) # x, y, z, order | level
l + stat_summary_hex(aes(z="z"), bins=30, fun=max) # x, y, z, fill | value
l + stat_summary_2d(aes(z="z"), bins=30, fun=mean) # x, y, z, fill | value
f + stat_boxplot(coef=1.5) # x, y | lower, middle, upper, width, ymin, ymax
f + stat_ydensity(kernel="gaussian", scale="area") # x, y | density, scaled, count, n, violinwidth, width
.....
e + stat_ecdf(n=40) # x, y
e + stat_quantile(quantiles=(0.1, 0.9), formula="y ~ np.log(x)") # x, y | quantile
e + stat_smooth(method="lm", formula="y ~ x", se=True, level=0.95) # x, y | se, ymin, ymax
.....
import scipy.stats as stats

ggplot() + lims(x=(-5, 5)) +
  stat_function(fun=stats.norm.pdf, n=20, geom="point") # x | y
ggplot() + stat_qq(aes(sample=range(100))) # x, y | sample, theoretical
e + stat_sum() # x, y, size | n, prop
e + stat_summary(fun_data="mean_cl_boot")
i + stat_summary_bin(fun="mean", geom="bar")
e + stat_identity()
e + stat_unique()

```

Scales

Override default mappings. Scales map data values to the visual values of an aesthetic. To change a mapping, add a new scale.



General purpose scales

Use with most aesthetics

```

scale_*_continuous() # Map continuous values
scale_*_discrete() # Map discrete values
scale_*_binned() # Map continuous values to bins
scale_*_identity() # Use data values literally
scale_*_manual(values=[]) # Map discrete values to manually chosen visual ones
scale_*_date(date_labels="%d/%m"), date_breaks="2 weeks" # Treat data values as dates
scale_*_datetime() # Treat values as date times

```

X and Y location scales

Use with x or y aesthetics (x shown here)

```

scale_x_log10() # Plot on log10 scale
scale_x_reverse() # Reverse direction of axis
scale_x_sqrt() # Plot on square root scale

```

Color and fill scales, discrete

```

n + scale_fill_brewer(palette="Blues")
n + scale_fill_grey(start=0.2, end=0.8, na_value="red")

```

Color and fill scales, continuous

```

o = c + geom_dotplot(aes(fill="x"))
o + scale_fill_distiller(palette="Blues")
o + scale_fill_gradient(low="red", high="yellow")
o + scale_fill_gradient2(low="red", high="blue", mid="white", midpoint=25)
o + scale_fill_gradientn(colors=["green", "purple", "papayawhip"])

```

Shape and size scales

```

p = e + geom_point(aes(shape="fl", size="cyl"))
p + scale_shape() + scale_size()
p + scale_shape_manual(values=["o", "^", "s", "D", "v"]) # see page 1 for shapes
p + scale_radius(range=range(1, 7))
p + scale_size_area(max_size=6)

```

Coordinate Systems

```

r = d + geom_bar()
r + coord_cartesian(xlim=[0, 5]) # xlim, ylim. The default cartesian coordinate system
r + coord_fixed(ratio=1/2) # ratio, xlim, ylim. Cartesian coordinates with fixed aspect ratio between x and y units
r + coord_flip() # Flip cartesian coordinates by switching x and y aesthetic mappings.
r + coord_trans(y="sqrt") # x, y, xlim, ylim. Transformed cartesian coordinates.

```

Position Adjustments

Position adjustments determine how to arrange geoms that would otherwise occupy the same space.

```

s = ggplot(mpg, aes("fl", fill="drv"))
s + geom_bar(position="dodge") # Arrange elements side by side.
s + geom_bar(position="fill") # Stack elements on top of one another, normalize height.
e + geom_point(position="jitter") # Add random noise to X and Y position of each element to avoid overplotting.
e + geom_label(position="nudge") # Nudge labels away from points.
s + geom_bar(position="stack") # Stack elements on top of one another.

```

Each position adjustment can be recast as a function with manual width and height arguments: `s + geom_bar(position=position_dodge(width=1))`

Themes

```

r + theme_bw() # White background with grid
r + theme_gray() # Grey background (default)
r + theme_dark() # Dark for contrast
r + theme_classic()
r + theme_light()
r + theme_linedraw()
r + theme_minimal()
r + theme_void() # Empty theme

r + theme() # Customize aspects of the theme such as axis, legend, panel, and facet properties.
r + theme(plot_title_position="plot")
r + theme(panel_background=element_rect(fill="blue"))

```

Faceting

Facets divide a plot into subplots based on the values of one or more discrete variables.

```

t = ggplot(mpg, aes("cty", "hwy")) + geom_point()
t + facet_grid(cols="fl") # Facet into columns based on fl
t + facet_grid("year") # Facet into rows based on year
t + facet_grid("year", "fl") # Facet into both rows and columns
t + facet_wrap("fl", ncol=4) # Wrap facets into a rectangular layout

```

Set **scales** to let axis limits vary across facets: `t + facet_grid("drv", "fl", scales="free")`

x and y axis limits adjust to individual facets: `"free_x" # x axis limits adjust`, `"free_y" # y axis limits adjust`

Set **labeller** to adjust facet label: `t + facet_grid(cols="fl", labeller="label_both")`

fl: c	fl: d	fl: e	fl: p	fl: r
-------	-------	-------	-------	-------

Labels and Legends

Use **labs()** to label the elements of your plot.

```

t + labs(x="New x axis label", y="New y axis label", title="Add a title above the plot", subtitle="Add a subtitle below title", caption="Add a caption below plot", tag="Add a tag to the plot", aes="New aes legend title")

```

`t + annotate(geom="text", x=8, y=9, label="A")` # Places a geom with manually selected aesthetics

`p + guides(x=guide_axis(n_dodge=2))` # Avoid crowded or overlapping labels with `n_dodge` or `angle`

`n + guides(fill="none")` # Set legend type for each aesthetic: `colorbar`, `legend`, or `none` (no legend)

`n + theme(legend_position="bottom")` # Place legend at "bottom", "top", "left", or "right"

`n + scale_fill_discrete(name="Title", labels=["A", "B", "C", "D", "E"])` # Set legend title and labels with a scale function

Zooming

Without clipping (preferred): `t + coord_cartesian(xlim=[10, 100], ylim=[0, 20])`

With clipping (removes unseen data points): `t + lims(x=(10, 100), y=(0, 20)) # option 1`, `t + scale_x_continuous(limits=[10, 100]) + scale_y_continuous(limits=[0, 20]) # option 2`

