Python For Data Science Cheat Sheet

plotnine

EasyShu团队-张杰出品

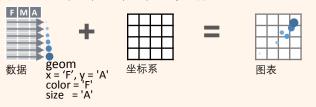
Basics

ES

Plotnine是基于图层语法开发的python数据可视化包, 主张模块间的协调与分工, 各司其职、上帝的归上帝、凯撒的归凯撒:数据集、坐标系和geom可视化数据。



为展示数据,可以通过geom把数据集的变量映射到数据可视化的属 性、比如:size(尺寸),color(颜色)和x、y轴的位置。



图表绘制的标准格式:

ggplot (data = <DATA>) + <GEOM FUNCTION> (mapping = aes(< MAPPINGS>), stat = **<STAT>**, position = **<POSITION>**) + <COORDINATE FUNCTION> + <FACET FUNCTION> + 可选 <SCALE FUNCTION> +

from plotnine import * #导入plotnine包的绘图函数 from plotnine.data import * #导入plotnine自带的数据集

base_plot=ggplot(data = mpg, aes(x = 'cty', y = 'hwy')) #使用ggplot()函 数开始绘图, 然后通过geom()函数可以添加图层。

print(base plot) #图表的显示

base_plot.save("plot.png", width = 5, height = 5) #保存绘制的图表为 5' x 5'的png格式的图片 "plot.png", 图片的保存格式也可选择为pdf。

Geoms



使用geom函数绘制数据,使用geom函数的美学 属性映射数据的变量,每个函数都返回一个图

图元

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

> a + geom_blank() (用于扩大限制)

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



a + geom_polygon(aes(group = 'date')) x, y, alpha, color, fill, group, linetype, size



b + geom_rect(aes(xmin = 'long', ymin='lat', xmax= 'long' + 1', ymax = 'lat + 1'))

xmax, xmin, ymax, ymin, alpha, color, fill, linetype, size



a + geom_ribbon(aes(ymin='unemploy - 900', ymax= 'unemploy + 900'))

x, ymax, ymin, alpha, color, fill, group, linetype, size

线条:公用的映射属性: 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 = range(0,seals.shape[0]), radius = 1))

1个变量:连续型

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c = ggplot(mpg, aes('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,



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

1个变量:离散型

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d =qqplot(mpq, aes('fl'))



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

2个变量:x-连续型, v-连续型

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e = ggplot(mpg, aes('cty', 'hwy'))



e + geom_label(aes(label = 'cty'), nudge_x = 1, nudge_y = 1) x, y, label, alpha, angle, color, family, fontface, hjust, size, vjust

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

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) x, y, label, alpha, angle, color, family, fontface, hjust, size, vjust

2个变量:x-离散型, v-连续型

ES

f=ggplot(mpg, aes('class', '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",dotsize=0.1)
x, y, alpha, color, fill, group



f + geom_violin(scale = "area")

x, y, alpha, color, fill, group, linetype, size, weight

2个变量:x-离散型,y-离散型

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g = ggplot(diamonds, aes('cut', 'color'))



g + geom_count()

x, y, alpha, color, fill, shape, size, stroke

二元连续变量分布

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h=ggplot(diamonds, aes('carat', 'price'))



h + geom_bin2d(binwidth = (0.25, 500)) x, y, alpha, color, fill, linetype, size, weight



h + geom_density2d()

x, y, alpha, colour, group, linetype, size

连续函数

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i=ggplot(economics, aes(date, unemploy))



i + geom_area()

x, y, alpha, color, fill, linetype, size



i + geom_line()

x, y, alpha, color, group, linetype, size



i + geom_step(direction = "hv")

x, y, alpha, color, group, linetype, size

误差绘制

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import pandas as pd

df = pd.DataFrame(('grp': ["A", "B"], 'fit': range(3,5), 'se' : range(0,2)})
j = ggplot(df, aes('grp', 'fit', ymin = 'fit-se', ymax = 'fit+se'))



i + geom crossbar(fatten = 2)

x, y, ymax, ymin, alpha, color, fill, group, linetype, size



j + geom_errorbar()

x, ymax, ymin, alpha, color, group, linetype, size



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

地图空间

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from geopandas import GeoDataFrame

continents = GeoDataFrame.from_file('data/lands-of-ice-and-fire/continents.shp')

westeros = continents.query('name=="Westeros"')



ggplot()+ geom_map(westeros, fill=None)

3个变量:x,y,z-连续型

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seals['z'] = np.sqrt(pow(seals['delta_long'],2) +
pow(seals['delta_lat'],2))
l=ggplot(seals, aes('long', 'lat'))

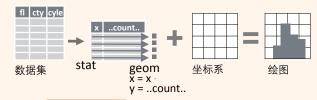


I + geom_tile(aes(fill = 'z'))

Stats

ES

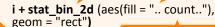
统计转换函数(stats)在数据被绘制出来之前对数据进行聚合和 其他计算。 stat_*确定了数据的计算方法。不同方法的计算会产 生不同的结果,所以一个stat()函数u必须与一个geom()函数对应 作数据的计算。



geom 绘图类型

stat()函数

geom变量映射



Stat()函数创造新的变量

c + stat_bin(binwidth = 1, origin = 10)

x, y | ...count..., ..ncount..., ..density..., ..ndensity...

c + stat_count(width = 1) x, y, | ...count..., ...prop..
c+stat_bindot(width = 1)

x, y, | ..count.., ..density.., ..scaled..
c2+stat_qq(aes(sample='hwy'))

c + stat_density(adjust = 1, kernel = "gaussian")

c2+stat_qq_line (aes(sample='hwy'))
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")

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** | ..x.., ..y..

 \mathbf{e} + \mathbf{stat} _quantile(quantiles=(0.25, 0.5, 0.75), formula = 'y ~x')

x, y | ..quantile..

e + stat_smooth(method = "lm", se=True, level=0.95**) x**, **y** | ..se.., ..x.., ..y.., ..ymin.., ..ymax..

ggplot() + stat_function() #自定义函数处理数据

e + stat_unique()

e + stat_identity()

e + stat_sum() x, y, size | ..n.., ..prop..

e + stat_summary(fun_data = "mean_cl_boot")

h + stat_summary_bin(fun_y = " mean_cl_boot ", geom = "bar")

Scales

ES

Scales 映射到图表数据点的属性,比如size, fill, color等。

最常见的scale函数

scale_*_continuous() #将连续型数值映射到可视化视觉暗示

scale_*_discrete() #将离散型数值映射到可视化视觉暗示

scale_*_identity() #将时间型数值映射到可视化视觉暗示

scale_*_manual(values = ()) #将离散型数值映射到自定义设定的可视化视觉暗示

scale_*_date(date_labels = "%m/%d"), date_breaks = "2 weeks") #将数据设定为时间型

scale_*_datetime() # 将x轴数据设定为时间型

Scales ES

X&Y坐标轴scale设定

scale_x_log10() #将x轴以log10的格式设定

scale_x_reverse() #将x坐标轴反转至y坐标轴 scale_x_sqrt() #将x轴以sqrrt的格式设定

离散型边框颜色和填充颜色的映射



n =d + geom_bar(aes(fill = 'fl'))

n + scale_fill_brewer(palette = "Blues") #palette的选择可参考: R语言的的RColorBrewer包



n + scale_fill_grey(start = 0.2, end = 0.8)

连续型边框颜色和填充颜色的映射



o = c + geom dotplot(aes(fill = '..x..'))

o + scale_fill_distiller(palette = "Blues")



o + scale_fill_gradient(low="red", high="yellow")

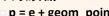


o + scale_fill_gradient2(low="red", high="blue", mid = "white", midpoint = 25)



o + scale_fill_gradientn(colors= ['red', 'blue'])

形状和尺寸的映射



p = e + geom_point(aes(shape = 'fl', size = 'cyl'))

p + scale shape() + scale size()

p + scale shape_manual(values = ('o','v',"s","h","D"))



p + scale_size_area(max size = 6)

R gaplot可以兼容三大主流可视化坐标系统(笛卡尔坐标系、极 坐标系、空间地理信息坐标系),将其整合在一套语法体系之下。 Plotnine暂时只能使用笛卡尔坐标系。



r =d + geom_bar()

r + coord_cartesian(xlim = (0, 5))

#默认为笛卡尔坐标系, 可调整xlim, ylim



r + coord_fixed(ratio = 1/2)

#设定笛卡尔坐标系的x和y轴的单位比例, 可调整ratio, xlim, ylim

r + coord flip()

#对调x和y坐标轴, 可调整xlim, ylim



r + coord trans(v= "sqrt")

#转换的笛卡尔坐标系,可根据函数调整x或者y坐标轴



r+ coord fixed()

#x和y轴同一单位的笛卡尔坐标系



r + coord_polar(theta = "x", direction=1) #极坐标系, 但是此函数plotnine包暂未实现, 如需使用, 可使用R ggplot2包

ES

s = ggplot(mpg, aes('fl', fill = 'drv'))

s + geom_bar(position = "dodge") #使柱形数据并列排布

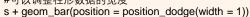
s + geom_bar(position = "stack")

#堆积柱形图, 使柱形数据堆积排列



s + geom_bar(position = "fill")

#百分比堆积柱形图,使柱形数据堆积排列 #可以调整柱形数据的宽度





e + geom_point(position = "jitter") #增加散点数据的扰动,避免重叠



e + geom_label(aes(label='cty'),position = "nudge")

#使数据标签从数据点上移开

主题

ES

plotnine包提供了各种不同图表风格的主题函数,直接调用就可以实 现,也可以在此基础上,调用theme()函数再对图表的主题元素修改 与调整。

theme 538()

theme bw()

theme classic() #R ggplot2的绘图风格. 默认

theme dark()

theme_gray()

theme_light()

theme_linedraw()

theme_matplotlib() #python matplotlib的绘图风格

theme minimal()

theme_seaborn() #python seaborn的绘图风格

theme_void()

theme xkcd() #漫画风格的绘图风格

分面

plotnine包Facets函数可以根据数据的离散型变量将图表分成若干个 子图表, 并按一定的规则排列。

t = ggplot(mpg, aes('cty', 'hwy')) + geom_point()

t + facet_grid('. ~ fl') #根据变量按列排布

t + facet_grid('year ~ .') #根据变量按行排布

t + facet_grid('year ~ fl') #根据两个变量按行列矩阵排布

_____ t + facet_wrap('~ fl') #根据变量按矩形排布

t + facet_grid('drv ~ fl', scales = "free")

#调整x和v坐标轴的取值范围

"free_x" - x坐标轴调整

"free_y" - y坐标轴调整

"fixed"-x和y坐标轴的取值范围统一

ES

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", <aes> = "New <aes> legend title") t + annotate(geom = "text", x = 8, y = 9, label = "A")

图例

n + theme(legend_position = "bottom") 图例位置的设定: "bottom", "top", "left", 或者 "right" n + scale_fill_discrete(name = "Title", labels = ("A", "B", "C", "D", "E")) #使用scale函数设定图例的名字和内容

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