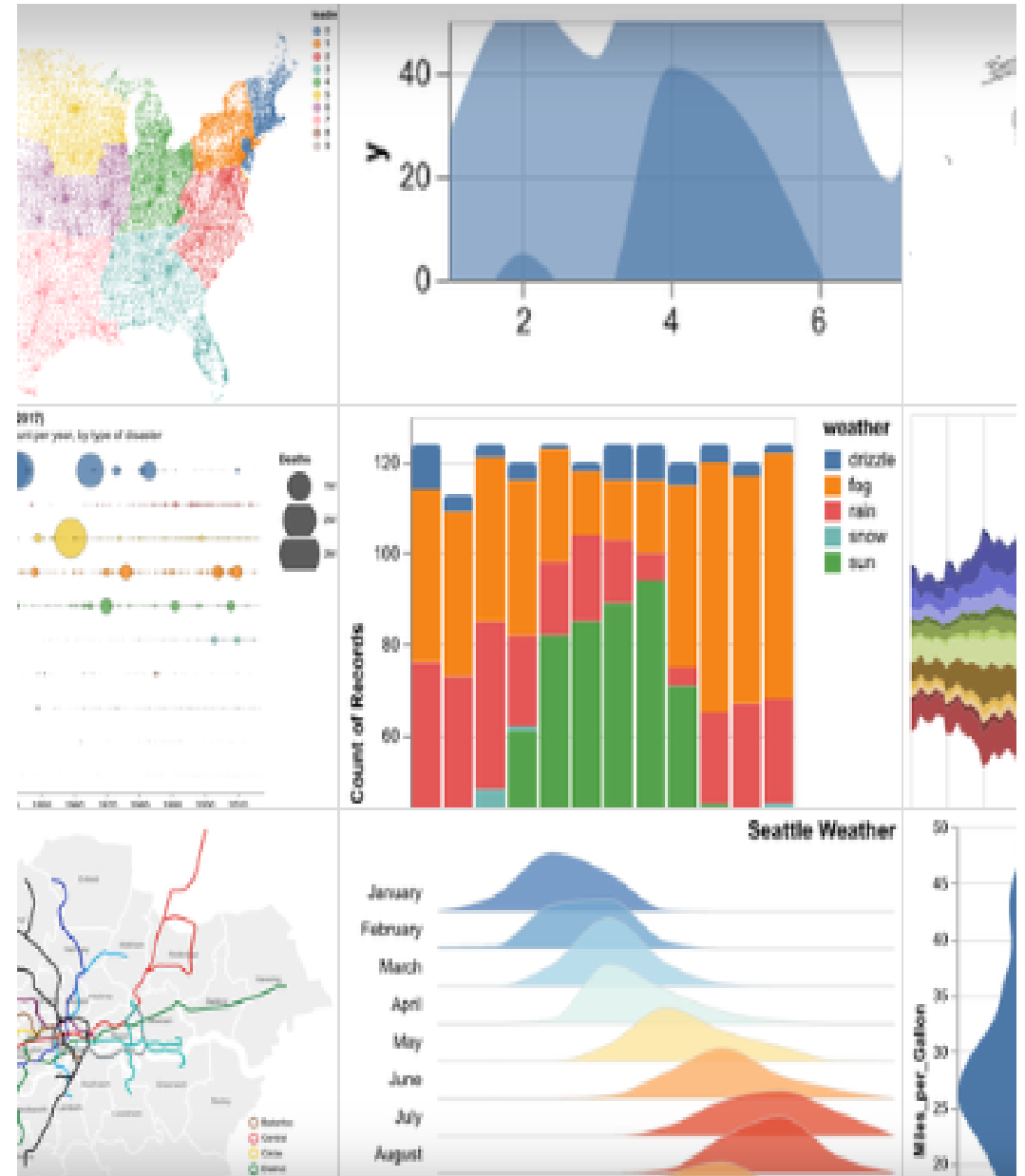


INTRO TO DATAVIZ

Altair and Python

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VISUAL VARIABLES



Data Types

- **Nominal (N)**
 - Equality relation
 - Apples, bananas, pears...
- **Ordinal (O)**
 - Ordering relation
 - Small, medium, large, darker, light...
- **Quantitative (Q)**
 - Arithmetic relations
 - 10m, 32 degrees, 2 bars...
 - **Quantitative Interval** (no reference point)
 - Dates, Location
 - Distances: A is 3 degrees hotter than B
 - **Quantitative Ratio** (reference point)
 - Length, mass
 - Proportions: A is twice as large as B

From Data to Conceptual Model

- **Data Model:** low-level representation of data and operations
- **Conceptual Model:** mental and semantic construction

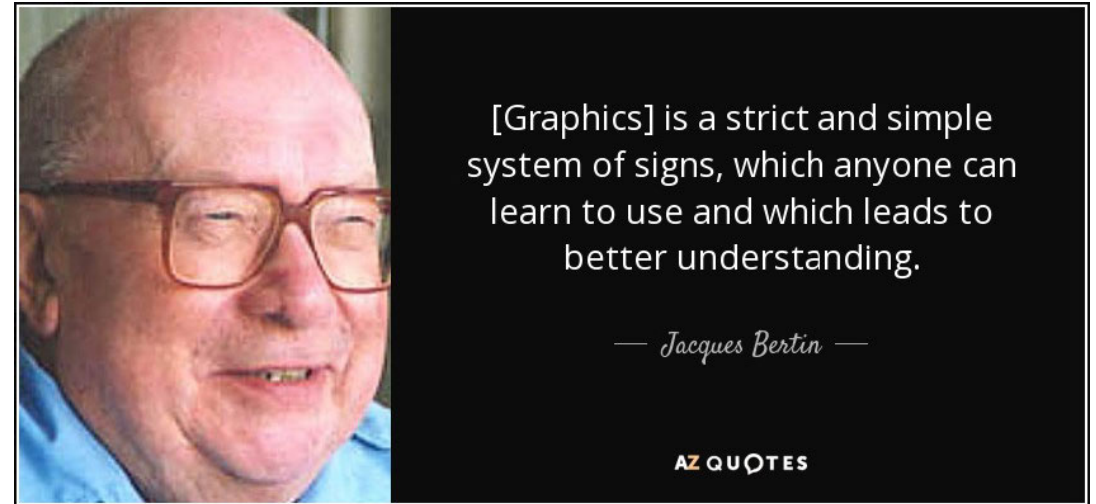
Data	Concept
1D number	Temperature
2D numbers	Geographic Coordinate
3D numbers	Spatio-temporal position

From Data to Conceptual Model

- From **data model**...
 - 70.8, 27.2, -10.2...
- ...using **conceptual model**...
 - Temperature
- ...to **data type**
 - Continuous variation
 - Warm, hot, cold
 - Burned vs not burned

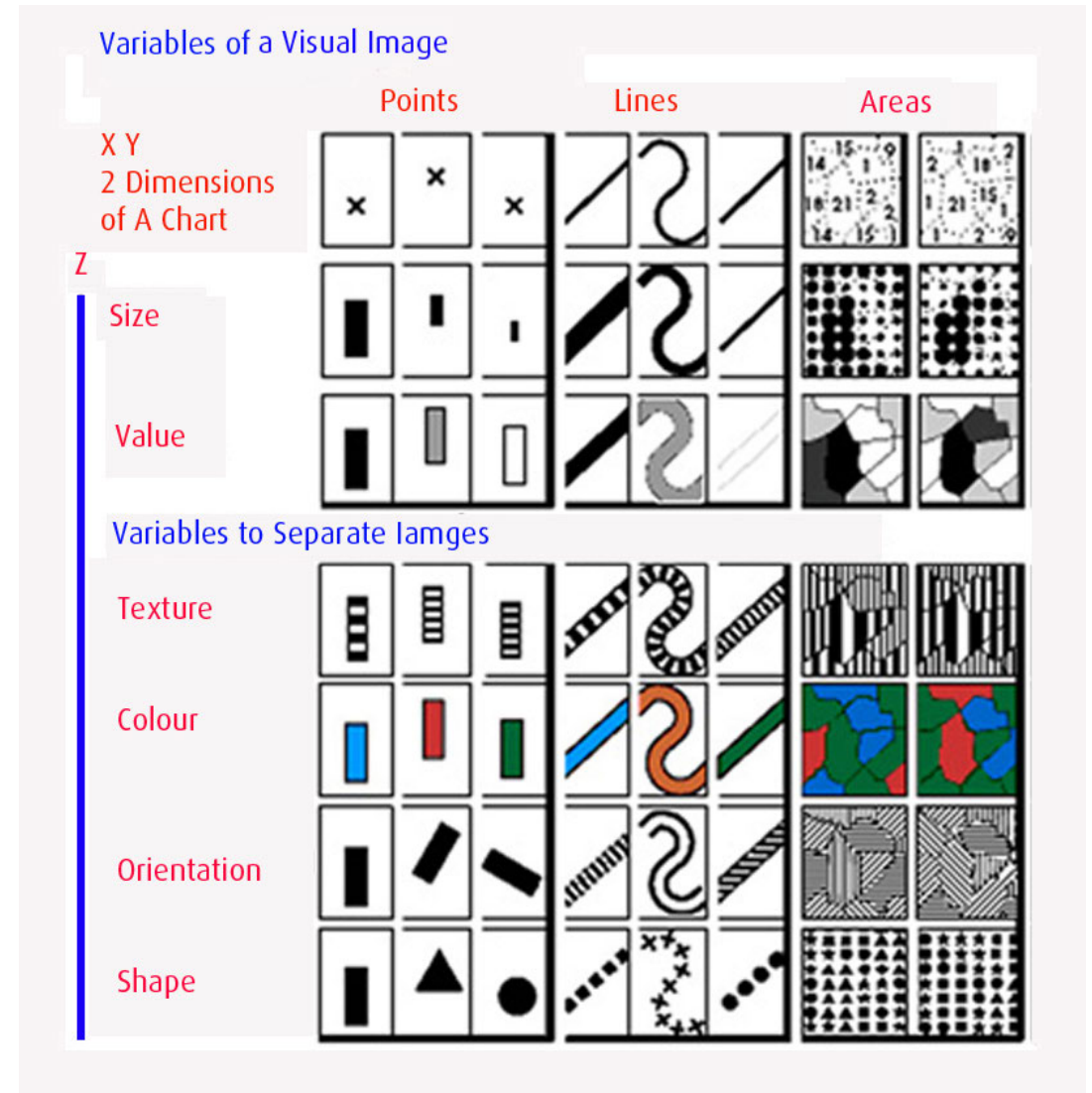
Visual Variables

- Jacques Bertin (1918-2010), cartographer
- Theoretical principles of visual encodings
- Semiology of Graphics (1967)

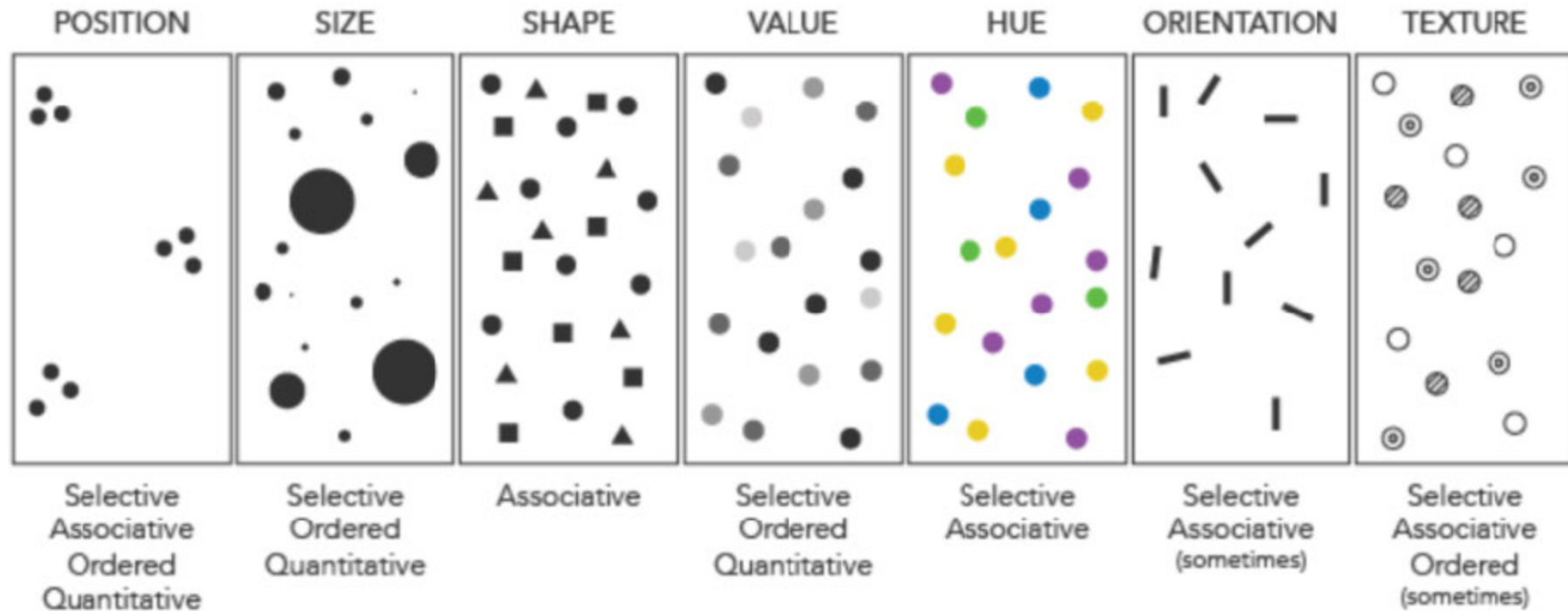


Bertin's Visual Variables

This is a visual summary of Bertin's classification of visual variables



Characteristics of Visual Variables

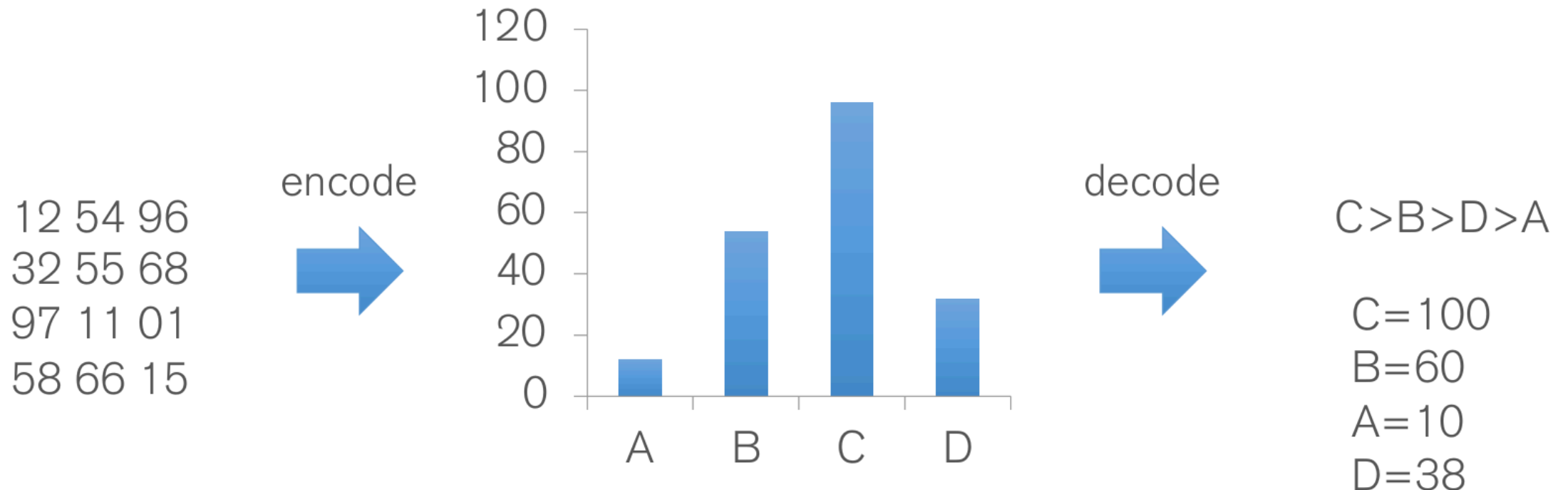


Bertin Visual Variables

	Nominal	Ordinal	Quantitative
Position	✓	✓	✓
Size	✓	✓	~
Value (intensity)	✓	✓	~
Texture	✓	~	✗
Color	✓	✗	✗
Orientation	✓	✗	✗
Shape	✓	✗	✗

Visual Encoding/Decoding

- A graph encodes a set of information as a set of graphical attributes
- The observer has to decode the graphical attributes to extract the original information



VISUALIZATION WITH VEGA-ALTAIR



Vega-Altair and Vega-Lite

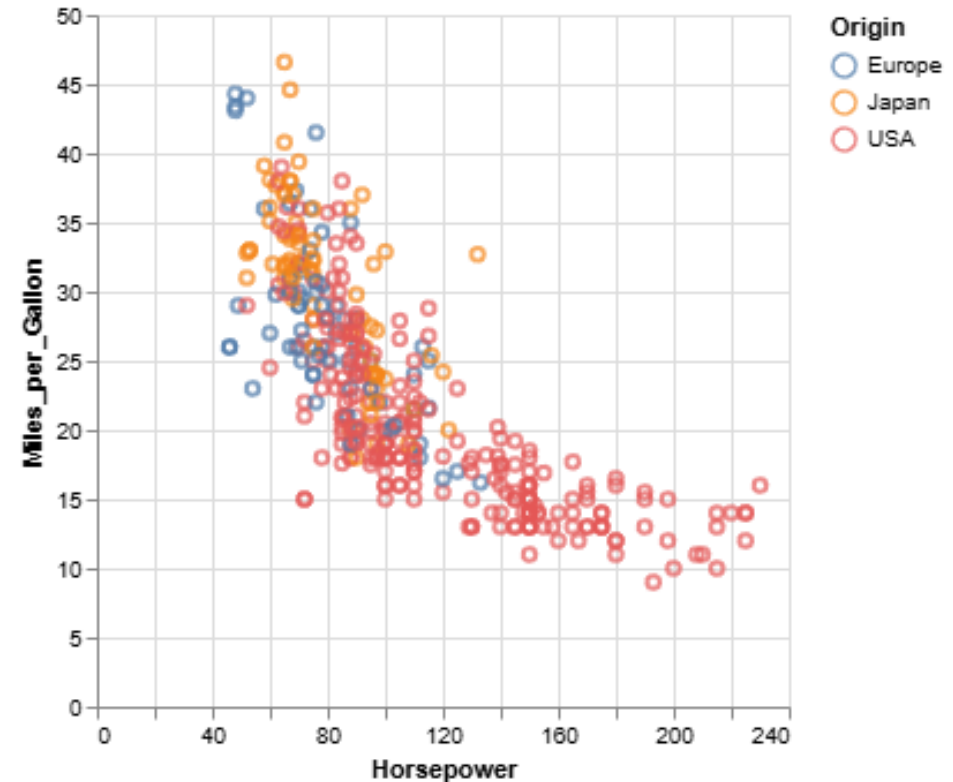
- **Vega** is a visualization grammar, a declarative format for creating, saving, and sharing interactive visualization designs.
- **Vega-Lite** is a high-level grammar of interactive graphics. It provides a concise JSON syntax for rapidly generating visualizations to support analysis.
- **Altair** is a declarative statistical visualization library for Python, based on Vega and Vega-Lite.

First Example in Altair

```
# import altair with an abbreviated alias
import altair as alt

# load a sample dataset as a pandas DataFrame
from vega_datasets import data
cars = data.cars()

# make the chart
alt.Chart(cars).mark_point().encode(
    x='Horsepower',
    y='Miles_per_Gallon',
    color='Origin',
).interactive()
```



Altair Basic Concepts

- **The Data:** The data to be visualized is passed to the Chart object as a Pandas DataFrame.
- **Encodings and Marks:** The visual appearance of the data is specified using encodings and marks.
 - Encodings map data fields to visual properties, such as position, color, size, and shape.
 - Marks are the basic building blocks of a visualization, such as points, lines, bars, and areas.
- **Data Transformation:** Altair provides a number of methods for transforming the data before it is visualized.
- **Interaction:** Altair provides a number of methods for adding interactivity to the visualization.

DATA



Altair - Specifying Data

- **Data:** Altair uses tabular data as its basic data model. Individual rows represent individual data points, and columns represent attributes of the data points.
- **Data Source:** The data to be visualized is passed to the Chart object as a `Pandas DataFrame`.
- **Data Types:** Altair supports a number of data types, including quantitative, ordinal, nominal, and temporal data. These data types are inferred from the `DataFrame` columns.
- **Data Object:** Altair provides a dedicated class to represent the data, which can be used to specify the data types and formats.

```
import altair as alt

data = alt.Data(values=[{'x': 'A', 'y': 5},
                        {'x': 'B', 'y': 3},
                        {'x': 'C', 'y': 6},
                        {'x': 'D', 'y': 7},
                        {'x': 'E', 'y': 2}])

alt.Chart(data).mark_bar().encode(
    x='x:N', # specify nominal data
    y='y:Q', # specify quantitative data
)
```


Long-form data and Wide-form data

- **Long-form data:** Each row represents a single observation, and each column represents a variable.
- **Wide-form data:** it has one row per independent variable and metadata encoded in columns and rows names.

Wide-form data

	Date	AAPL	AMZN	GOOG
0	2007-10-01	189.95	89.15	707.00
1	2007-11-01	182.22	90.56	693.00
2	2007-12-01	198.08	92.64	691.48

Long-form data

	Date	company	price
0	2007-10-01	AAPL	189.95
1	2007-11-01	AAPL	182.22
2	2007-12-01	AAPL	198.08
3	2007-10-01	AMZN	89.15
4	2007-11-01	AMZN	90.56
5	2007-12-01	AMZN	92.64
...			

ENCODINGS



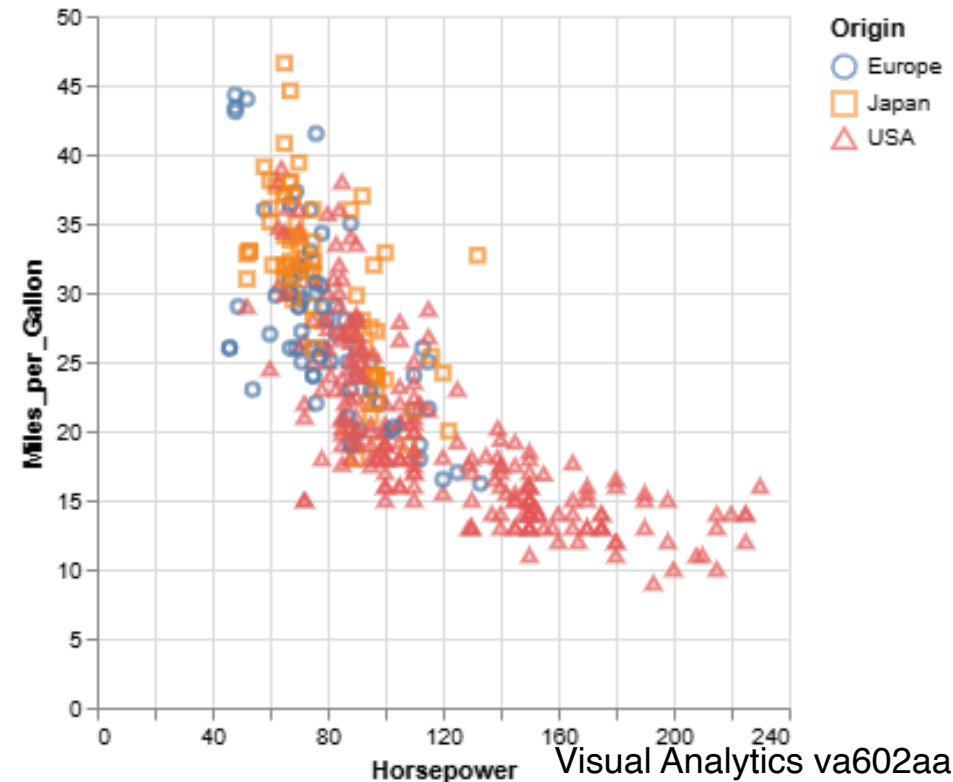
Altair - Encodings

- **Encodings:** Encodings map data fields to visual properties, such as position, color, size, and shape.
- Encodings represent the visual properties of the marks in the visualization.

```
import altair as alt
from vega_datasets import data

cars = data.cars()

alt.Chart(cars).mark_point().encode(
    x='Horsepower',
    y='Miles_per_Gallon',
    color='Origin',
    shape='Origin'
)
```



Channel Options

The **channel options** are the parameters to customize the visual mappings. The new version introduces a new style for options, with the following syntax:

Method-based Syntax

```
alt.Chart(cars).mark_point().encode(  
    alt.X('Horsepower').axis(tickMinStep=50),  
    alt.Y('Miles_per_Gallon').title('Miles per Gallon'),  
    color='Origin',  
    shape='Origin'  
)
```

Attribute-based Syntax

```
alt.Chart(cars).mark_point().encode(  
    alt.X('Horsepower', axis=alt.Axis(tickMinStep=50)),  
    alt.Y('Miles_per_Gallon', title="Miles per Gallon"),  
    color='Origin',  
    shape='Origin'  
)
```

Encodings Data Types

Altair supports five main data types for encodings:

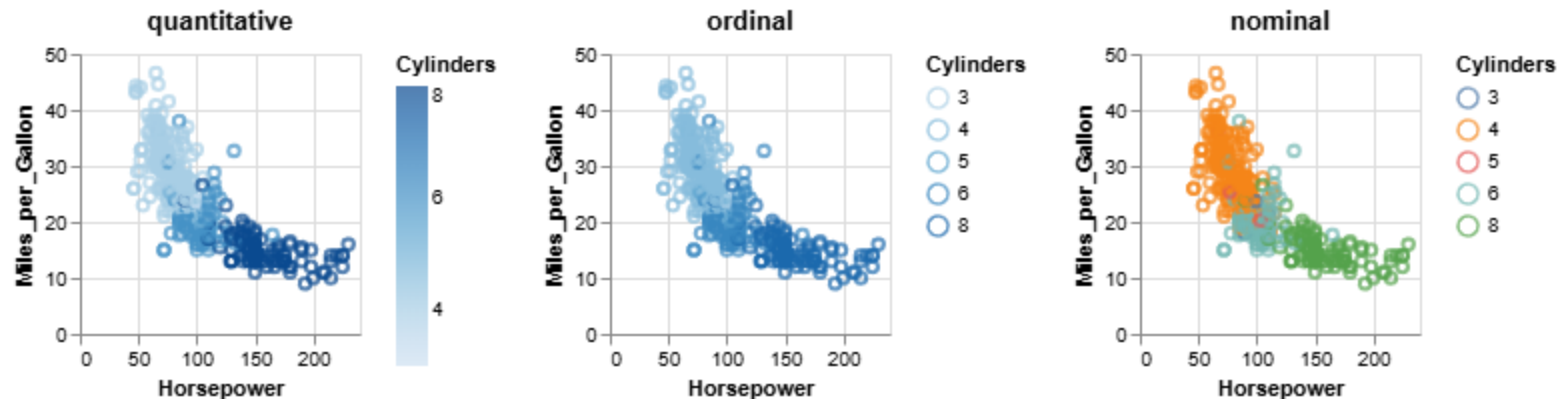
Data Type	Shorthand Code	Description
quantitative	Q	a continuous real-valued quantity
ordinal	O	a discrete ordered quantity
nominal	N	a discrete unordered category
temporal	T	a time or date value
geojson	G	a geographic shape

```
alt.Chart(cars).mark_point().encode(  
  x='Acceleration:Q',  
  y='Miles_per_Gallon:Q',  
  )
```

```
alt.Chart(cars).mark_point().encode(  
  alt.X('Acceleration', type='quantitative'),  
  alt.Y('Miles_per_Gallon', type='quantitative'),  
  alt.Color('Origin', type='nominal')  
  )
```

Choosing the correct data type (colors)

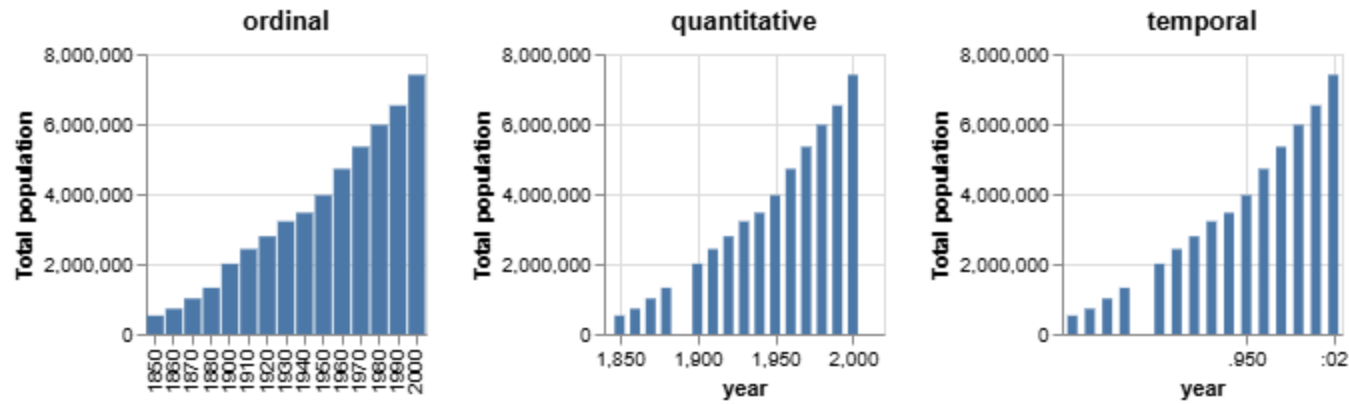
```
base = alt.Chart(cars).mark_point().encode(  
    x='Horsepower:Q',  
    y='Miles_per_Gallon:Q',  
)  
.properties(width=140, height=140)  
  
alt.hconcat(  
    base.encode(color='Cylinders:Q').properties(title='quantitative'),  
    base.encode(color='Cylinders:O').properties(title='ordinal'),  
    base.encode(color='Cylinders:N').properties(title='nominal'),  
)
```



Choosing the correct data type (axis)

```
base = alt.Chart(pop).mark_bar().encode(
    alt.Y('mean(people):Q').title('Total population')
).properties(width=140, height=140)

alt.hconcat(
    base.encode(x='year:O').properties(title='ordinal'),
    base.encode(x='year:Q').properties(title='quantitative'),
    base.encode(x='year:T').properties(title='temporal')
)
```



Encodings Shorthands

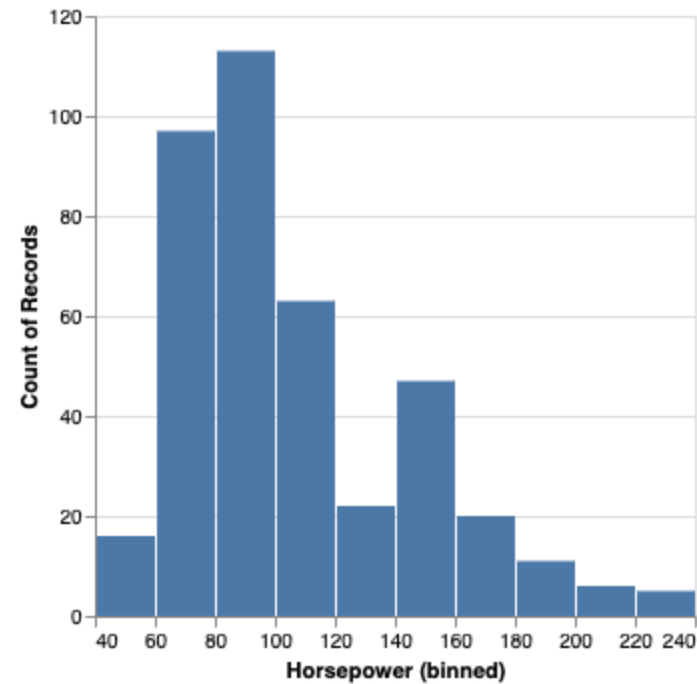
The shorthand syntax is a convenient way to specify encodings in Altair. The shorthand syntax uses a colon to separate the field name from the data type.

Shorthand Syntax	Equivalent long-form
<code>x = 'name'</code>	<code>alt.X('name')</code>
<code>x = 'name:Q'</code>	<code>alt.X('name', type='quantitative')</code>
<code>x = 'sum(name)'</code>	<code>alt.X('name', aggregate='sum')</code>
<code>x = 'sum(name):Q'</code>	<code>alt.X('name', aggregate='sum', type='quantitative')</code>
<code>x = 'count():Q'</code>	<code>alt.X(aggregate='count', type='quantitative')</code>

Binning and Aggregation - Histogram

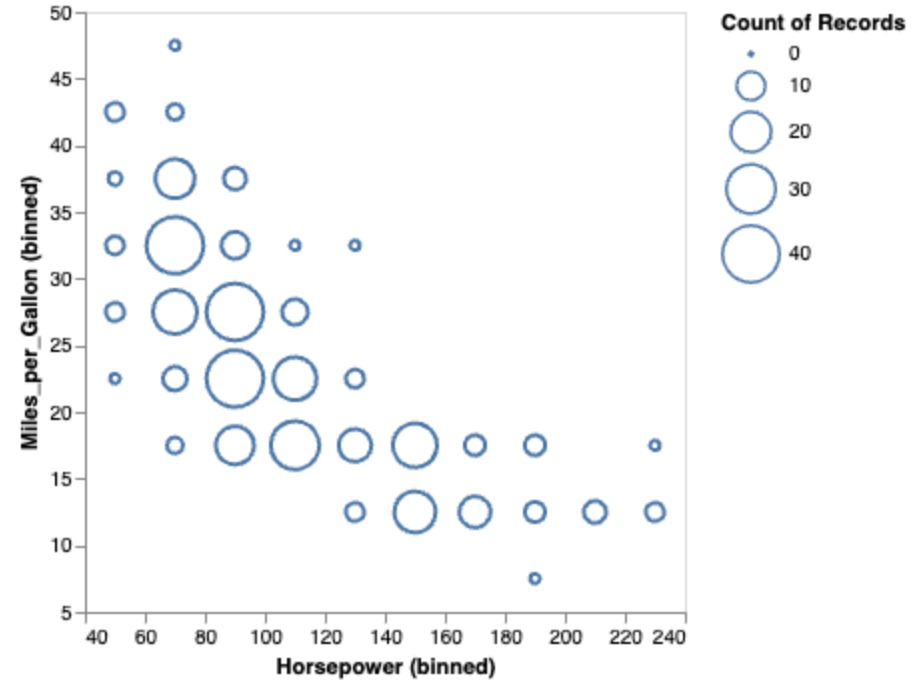
Altair's visualizations are built on the concept of the database-style grouping and aggregation. The split-apply-combine abstraction underpins many data analysis approaches.

```
alt.Chart(cars).mark_bar().encode(  
  alt.X('Horsepower').bin(),  
  y='count()'   
  # could also use alt.Y(aggregate='count', type='quantitative')  
)
```



Binning and Aggregation - 2D Histogram

```
alt.Chart(cars).mark_point().encode(  
  alt.X('Horsepower').bin(),  
  alt.Y('Miles_per_Gallon').bin(),  
  size='count()',  
)
```

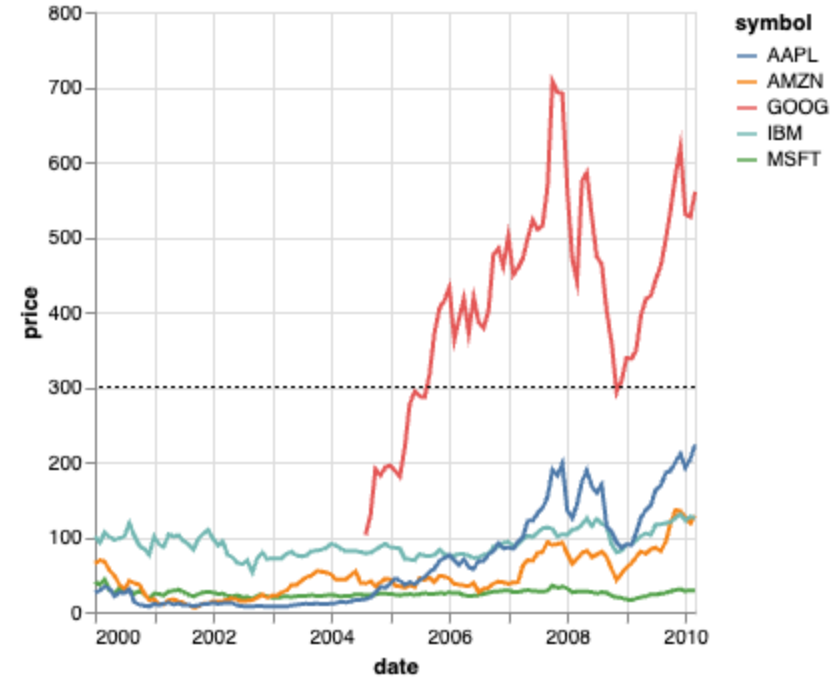


Datum

```
import altair as alt
from vega_datasets import data

source = data.stocks()
base = alt.Chart(source)
lines = base.mark_line().encode(
    x="date:T",
    y="price:Q",
    color="symbol:N"
)
rule = base.mark_rule(strokeDash=[2, 2]).encode(
    y=alt.datum(300)
)

lines + rule
```



Marks (1)

Marks are the basic building blocks of a visualization. The `mark` property is what specifies how exactly those attributes should be represented on the plot.

Mark	Method	Description
Arc	<code>mark_arc()</code>	A pie chart.
Area	<code>mark_area()</code>	A filled area plot.
Bar	<code>mark_bar()</code>	A bar plot.
Circle	<code>mark_circle()</code>	A scatter plot with filled circles.
Geoshape	<code>mark_geoshape()</code>	Visualization containing spatial data
Image	<code>mark_image()</code>	A scatter plot with image markers.
Line	<code>mark_line()</code>	A line plot.

Marks (2)

Marks are the basic building blocks of a visualization. The `mark` property is what specifies how exactly those attributes should be represented on the plot.

Mark	Method	Description
Rect	<code>mark_rect()</code>	A filled rectangle, used for heatmaps
Rule	<code>mark_rule()</code>	A vertical or horizontal line spanning the axis.
Square	<code>mark_square()</code>	A scatter plot with filled squares.
Text	<code>mark_text()</code>	A scatter plot with points represented by text.
Tick	<code>mark_tick()</code>	A vertical or horizontal tick

Composite Marks

Composite marks are a combination of multiple marks that are used to create more complex visualizations.

Mark	Method	Description
Errorband	mark_errorband()	An error band around a line
Errorbar	mark_errorbar()	An error bar around a point
Boxplot	mark_boxplot()	A box plot.

DATA TRANSFORMATION



Data Transformation

It is often necessary to transform or filter data in the process of visualizing it. In Altair you can do this one of two ways:

- Before the chart definition, using standard pandas data transformations.
- Within the chart definition, using Vega-Lite's data transformation tools.

Data Transformation Operators (1)

Tranform	Method	Description
Aggregate	transform_aggregate()	Create a new data column by aggregating an existing column.
Bin	transform_bin()	Create a new data column by binning an existing column.
Calculate	transform_calculate()	Create a new data column using an arithmetic calculation on an existing column.
Density	transform_density()	Create a new data column with the kernel density estimate of the input

Data Transformation Operators (2)

Tranform	Method	Description
Flatten	transform_flatten()	Flatten array data into columns.
Fold	transform_fold()	Convert wide-form data into long-form data (opposite of pivot).
Impute	transform_impute()	Impute missing data.
Join Aggregate	transform_joinaggregate()	Aggregate transform joined to original data.
LOESS	transform_loess()	Create a new column with LOESS

Data Transformation Operators (3)

Tranform	Method	Description
Quantile	transform_quantile()	Compute empirical quantiles of a dataset.
Regression	transform_regression()	Fit a regression model to a dataset.
Sample	transform_sample()	Random sub-sample of the rows in the dataset.
Stack	transform_stack()	Compute stacked version of values.
TimeUnit	transform_timeunit()	Discretize/group a date by a time unit (day, month, year, etc.)
Window	transform_window()	Compute a window aggregation

INTERACTION



Interaction

Altair provides a declarative grammar for specifying interactive visualizations. Interaction is based on three key concepts:

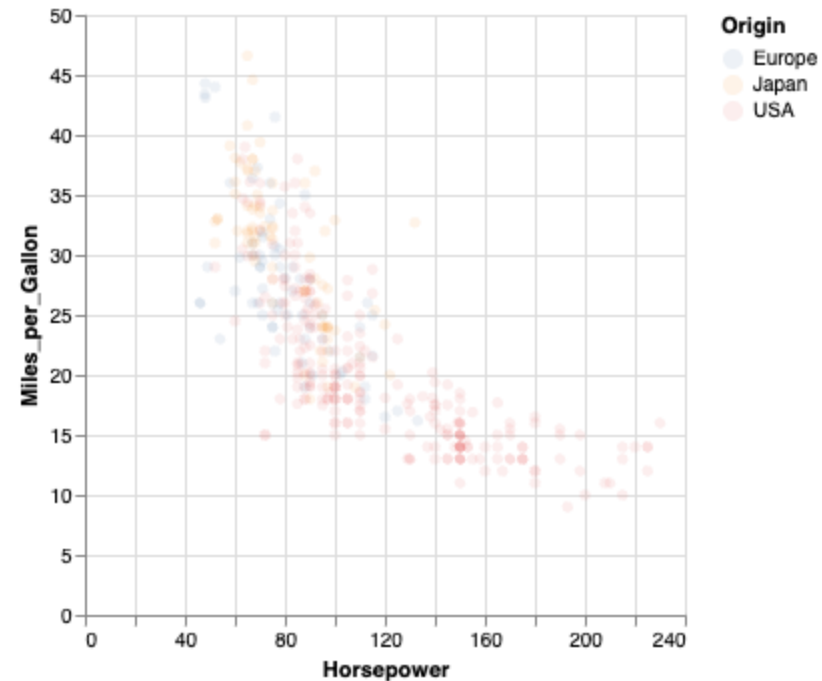
- **Parameters** are the basic building blocks in the grammar of interaction. They can either be simple variables or more complex selections that map user input (e.g., mouse clicks and drags) to data queries.
- **Conditions and filters** can respond to changes in parameter values and update chart elements based on that input.
- **Widgets** and other **chart input elements** can bind to parameters so that charts can be manipulated via drop-down menus, radio buttons, sliders, legends, etc.

Parameters - Variables

Variable parameters allow for a value to be defined once and then reused throughout the rest of the chart. This is useful for defining a value that is used in multiple places in the chart, such as a color or size.

```
op_var = alt.param(value=0.1)

alt.Chart(cars).mark_circle(opacity=op_var).encode(
    x='Horsepower:Q',
    y='Miles_per_Gallon:Q',
    color='Origin:N'
).add_params(
    op_var
)
```

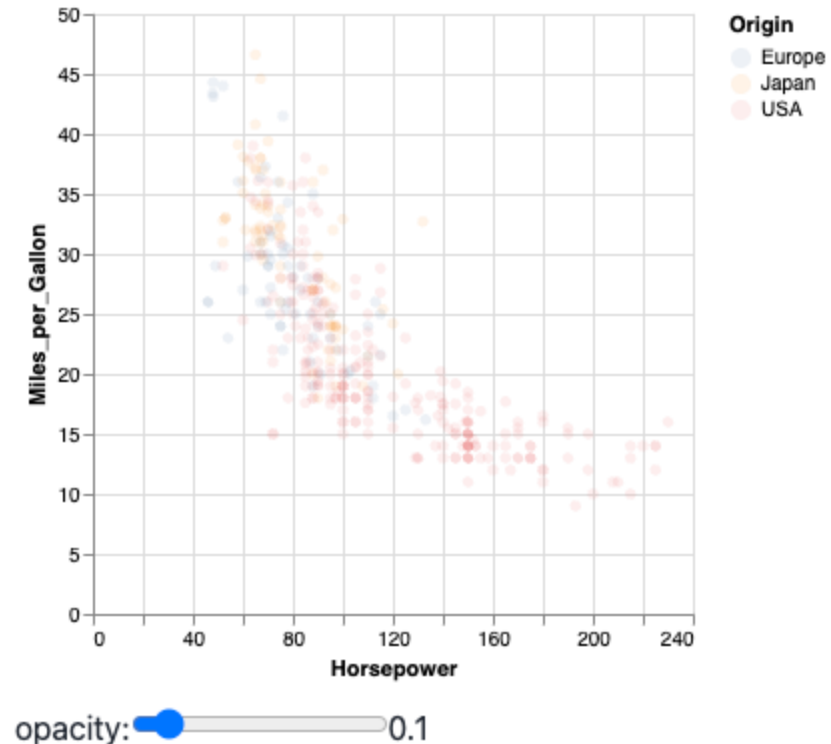


Parameters - Widgets

Widgets are interactive elements that allow users to manipulate the chart. Widgets can be bound to parameters, so that the chart updates in response to user input.

```
slider = alt.binding_range(min=0, max=1, step=0.05, name='opacity:')
op_var = alt.param(value=0.1, bind=slider)

alt.Chart(cars).mark_circle(opacity=op_var).encode(
    x='Horsepower:Q',
    y='Miles_per_Gallon:Q',
    color='Origin:N'
).add_params(
    op_var
)
```



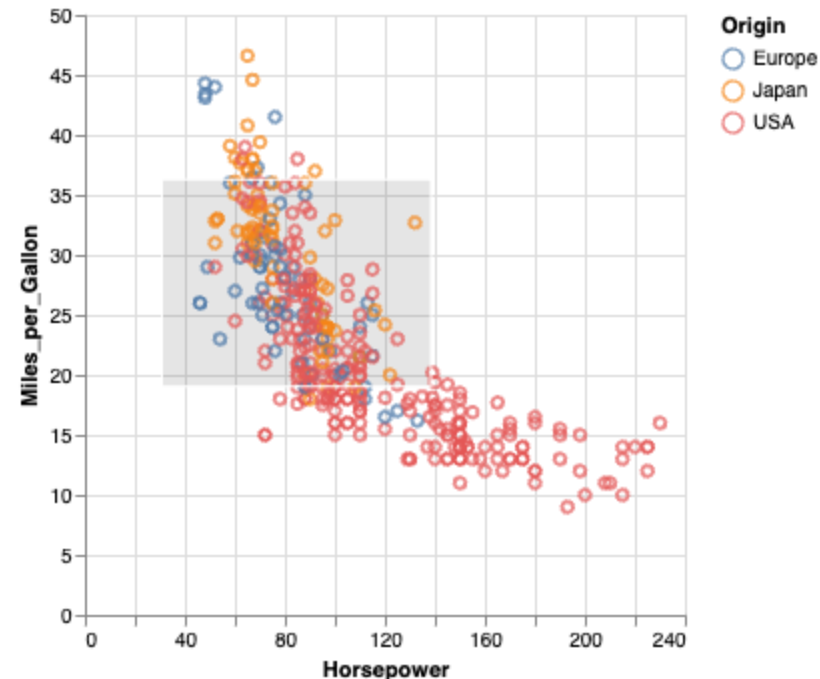
Selection and Charts Interaction

Selection parameters define data queries that are driven by interactive manipulation of the chart by the user (e.g., via mouse clicks or drags).

There are two types of selections: `selection_interval()` and `selection_point()` .

```
brush = alt.selection_interval()

alt.Chart(cars).mark_point().encode(
    x='Horsepower:Q',
    y='Miles_per_Gallon:Q',
    color='Origin:N'
).add_params(
    brush
)
```

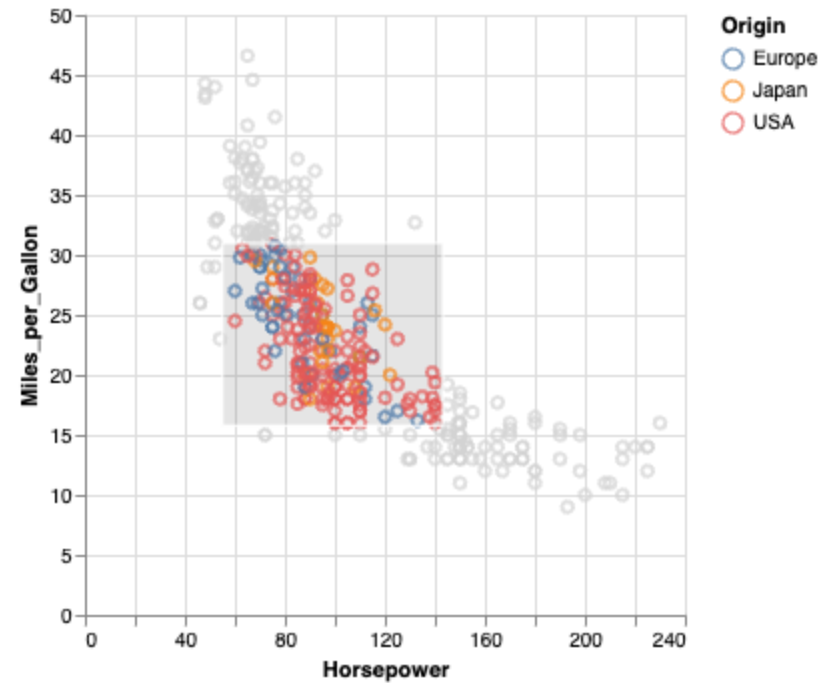


Conditions

Conditions are used to update the chart based on the value of a parameter. Conditions can be used to change the appearance of the chart, filter the data, or update the chart in other ways.

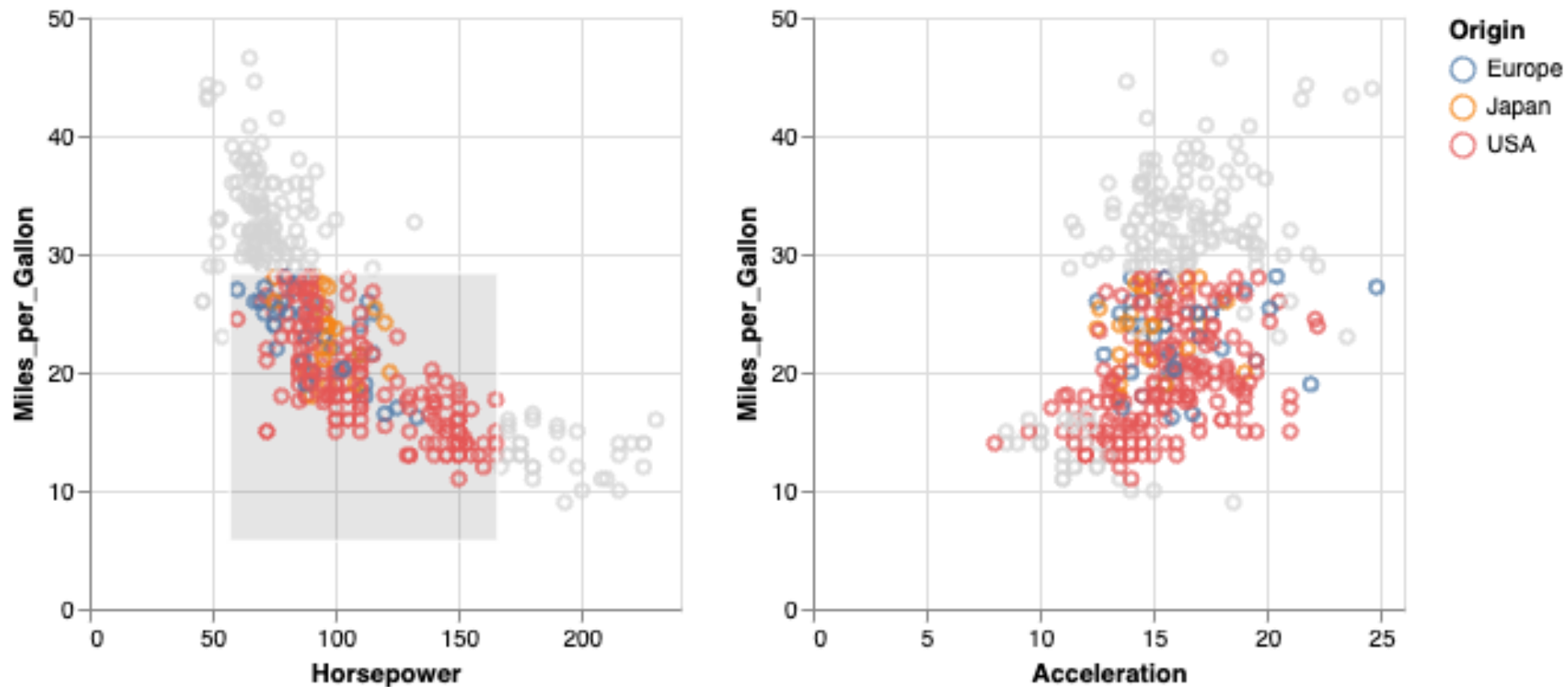
```
conditional = alt.when(brush).then("Origin:N").otherwise(alt.value("lightgray"))

alt.Chart(cars).mark_point().encode(
    x="Horsepower:Q",
    y="Miles_per_Gallon:Q",
    color=conditional,
).add_params(
    brush
)
```



Linked Conditions (1)

Conditions can be linked together to create more complex interactions among charts. This allows for more sophisticated interactions between different parts of the visualization.



Linked Conditions (2)

```
chart = alt.Chart(cars).mark_point().encode(  
    x='Horsepower:Q',  
    y='Miles_per_Gallon:Q',  
    color=alt.when(brush).then("Origin:N").otherwise(alt.value("lightgray")),  
)  
.properties(  
    width=250,  
    height=250  
)  
.add_params(  
    brush  
)  
  
chart | chart.encode(x='Acceleration:Q')
```

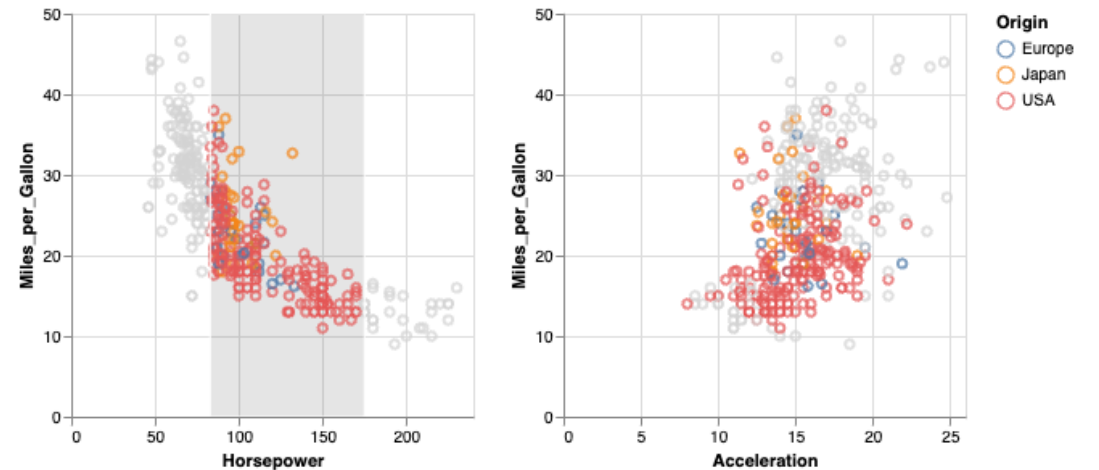
Selection Types

Each selection type can be customized with additional properties to control its behavior.

```
brush = alt.selection_interval(encodings=['x'])

chart = alt.Chart(cars).mark_point().encode(
    x='Horsepower:Q',
    y='Miles_per_Gallon:Q',
    color=alt.when(brush).then("Origin:N").otherwise(alt.value("lightgray")),
).properties(
    width=250,
    height=250
).add_params(
    brush
)

chart | chart.encode(x='Acceleration:Q')
```



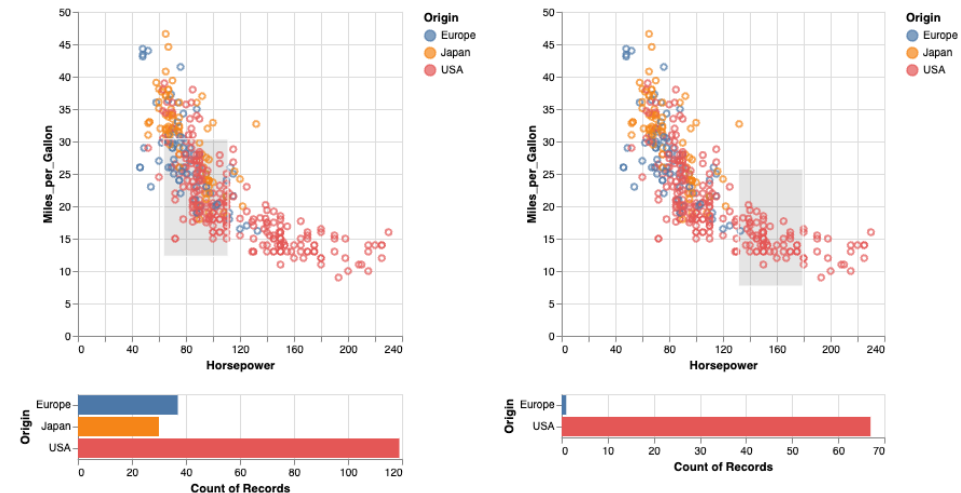
Filters

A selection parameter can be used to filter the data in the chart. This allows for interactive filtering of the data based on user input.

```
brush = alt.selection_interval()

points = alt.Chart(cars).mark_point().encode(
    x='Horsepower:Q',
    y='Miles_per_Gallon:Q',
    color='Origin:N'
).add_params(
    brush
)

bars = alt.Chart(cars).mark_bar().encode(
    x='count()',
    y='Origin:N',
    color='Origin:N'
).transform_filter(
    brush
```



Takeaways

- **Data:** Altair uses tabular data as its basic data model. The data to be visualized is passed to the Chart object as a Pandas DataFrame.
- **Encodings and Marks:** The visual appearance of the data is specified using encodings and marks. Encodings map data fields to visual properties, such as position, color, size, and shape. Marks are the basic building blocks of a visualization, such as points, lines, bars, and areas.
- **Data Transformation:** Altair provides a number of methods for transforming the data before it is visualized. These methods can be used to filter, aggregate, and sort the data.
- **Interaction:** Altair provides a number of methods for adding interactivity to the visualization. These methods can be used to add tooltips, zooming, panning, and other interactive features to the visualization.