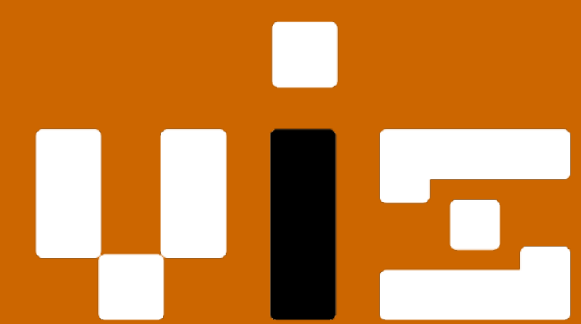


Towards *Effective* Interaction

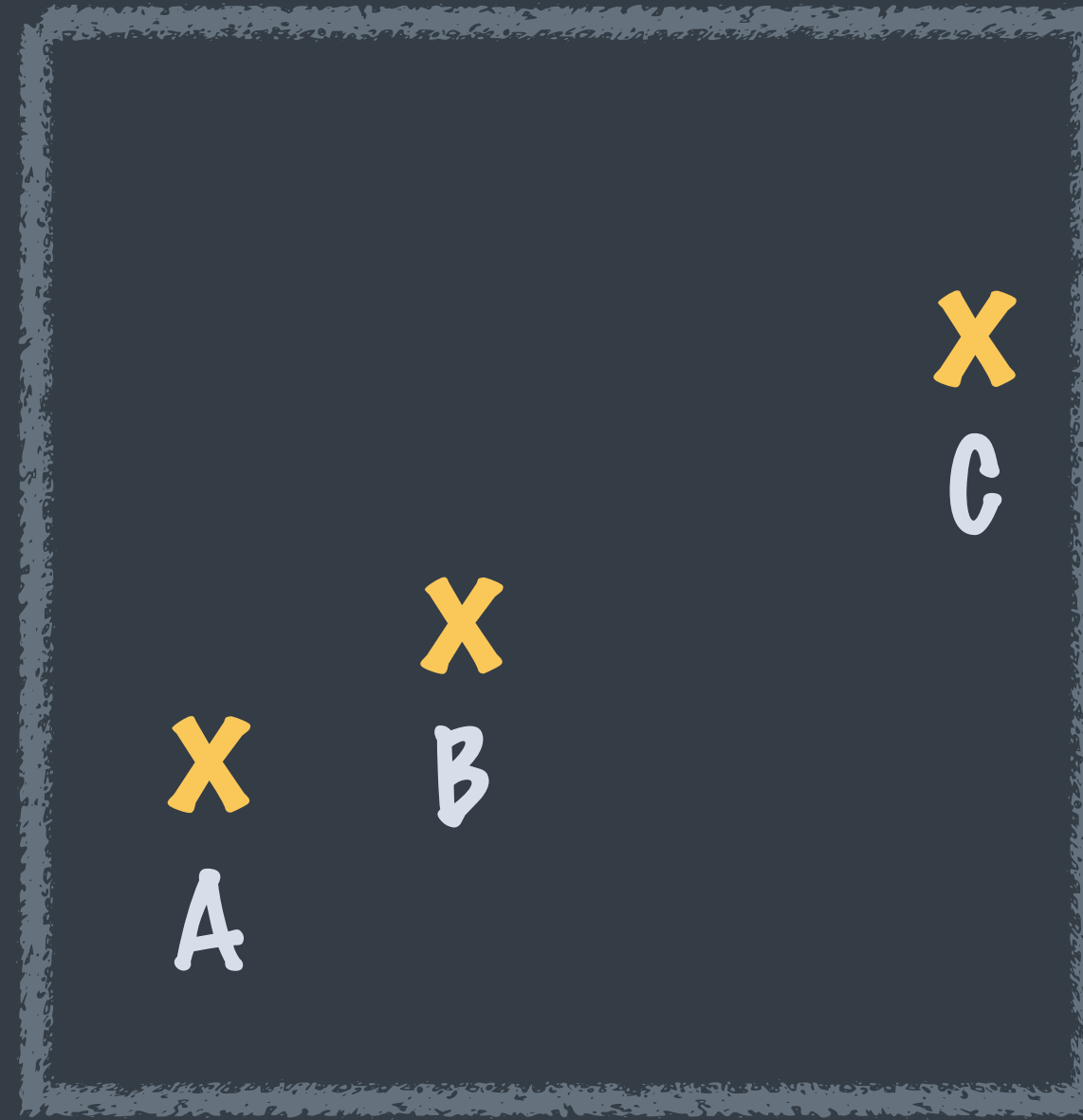
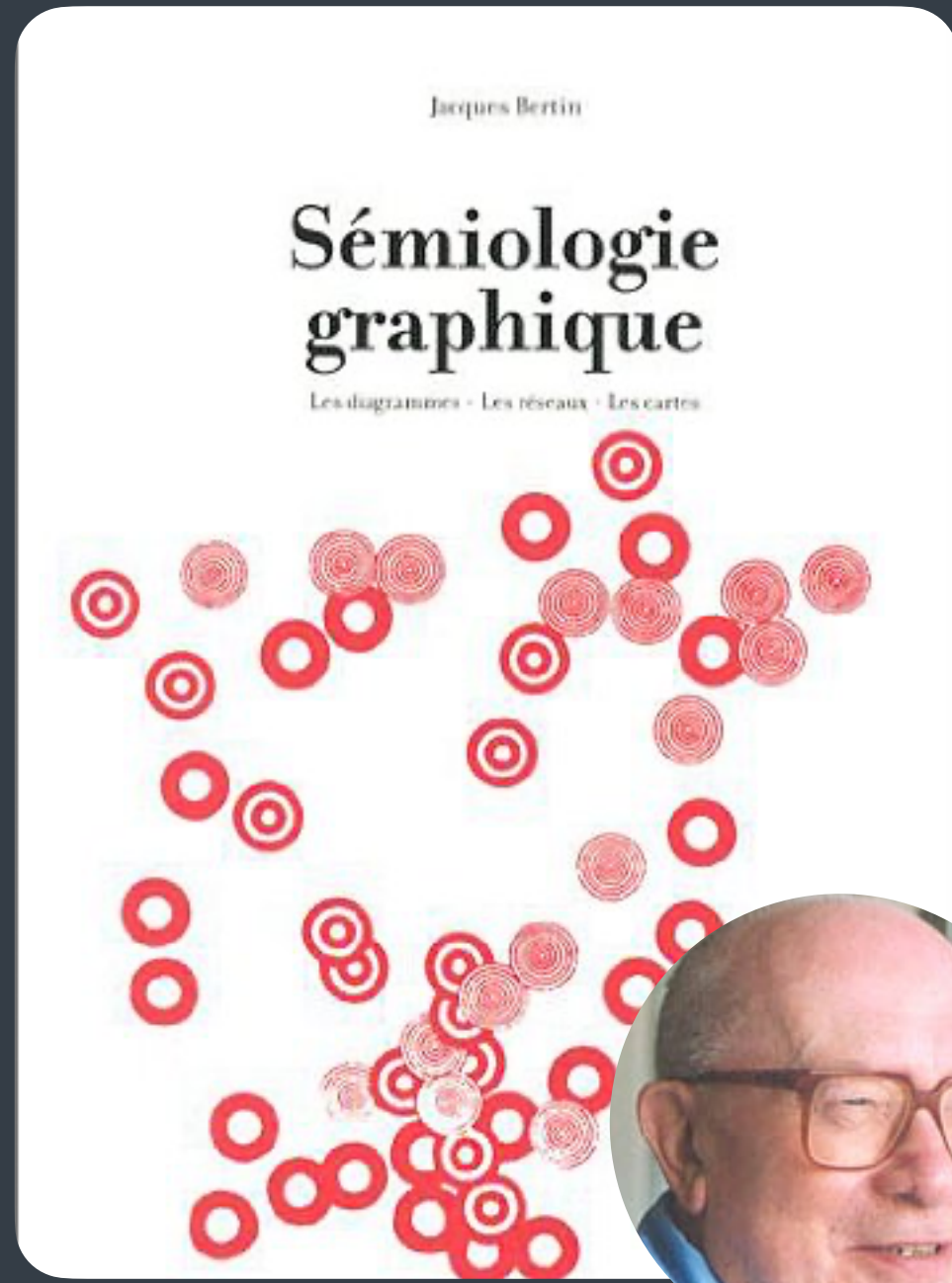
With Data Visualization

Arvind Satyanarayan
@arvindsatya1

MIT Visualization Group
@mitvis • vis.csail.mit.edu



+ friends



Visual Variables

1. A, B, C are distinguishable.
2. B is between A and C.
3. BC is twice as long as AB.

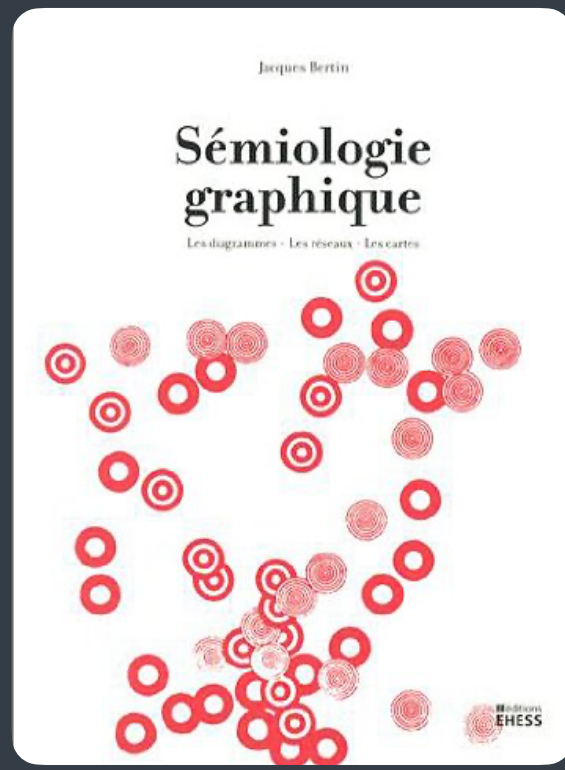
LES VARIABLES DE L'IMAGE

XY	dimensions du plan								
	Taille								
Z	Valeur								

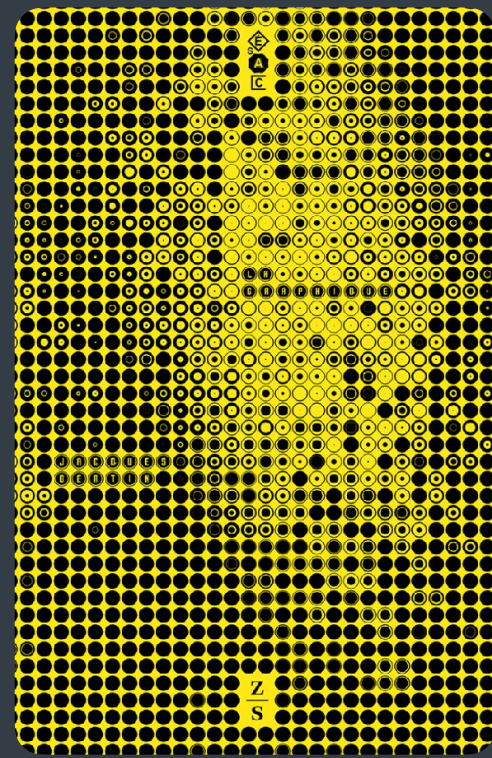
LES VARIABLES DE SÉPARATION

Grain								
Couleur								
Orientation								
Forme								
	Points	Lignes	Zones					

Marks



1967



1977

Automating the Design of Graphical Presentations of Relational Information
JOCK MACKINLAY
Stanford University

1984

1987



1999

2002



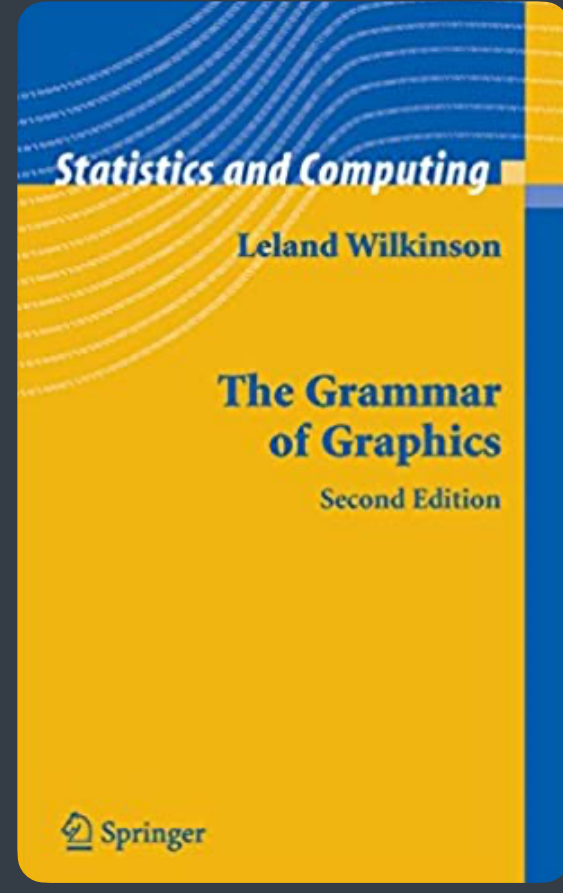
2010

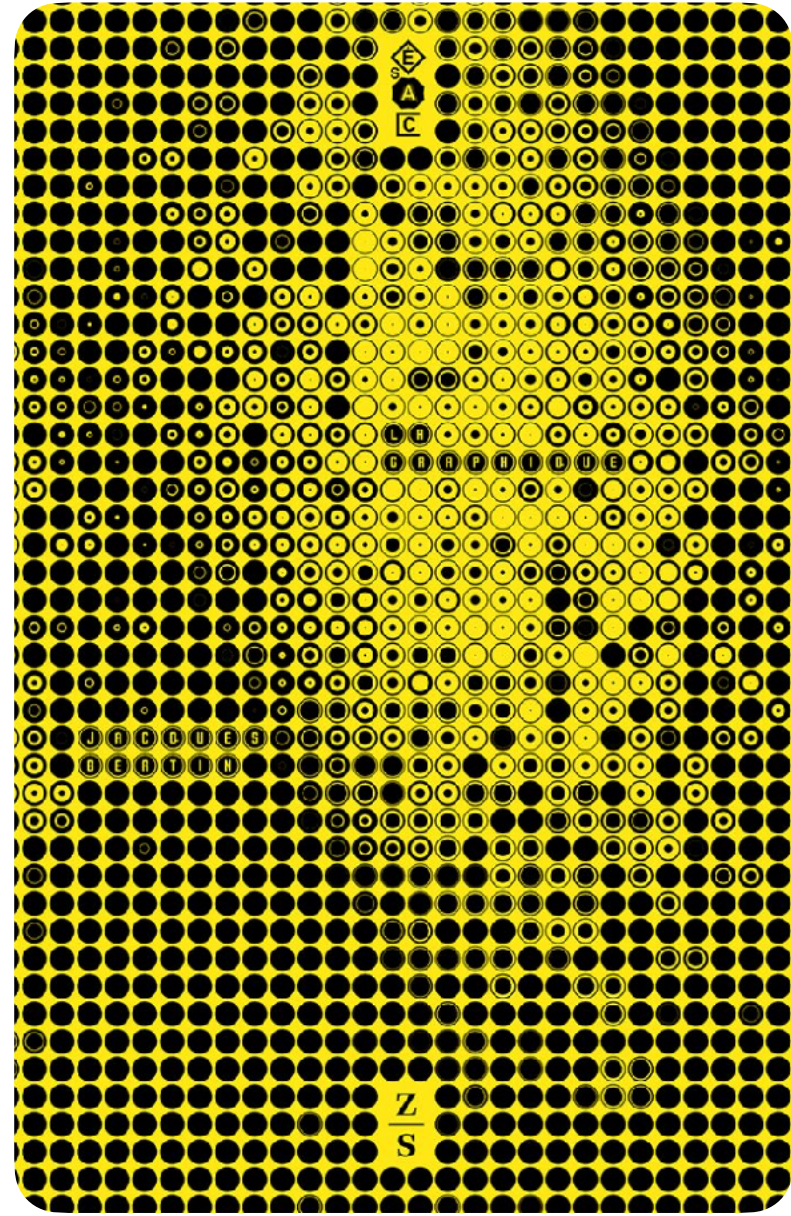
2011

2013

2016

Graphical Perception: Theory, Experimentation, and Application to the Development of Graphical Methods
WILLIAM S. CLEVELAND and ROBERT MCGILL*



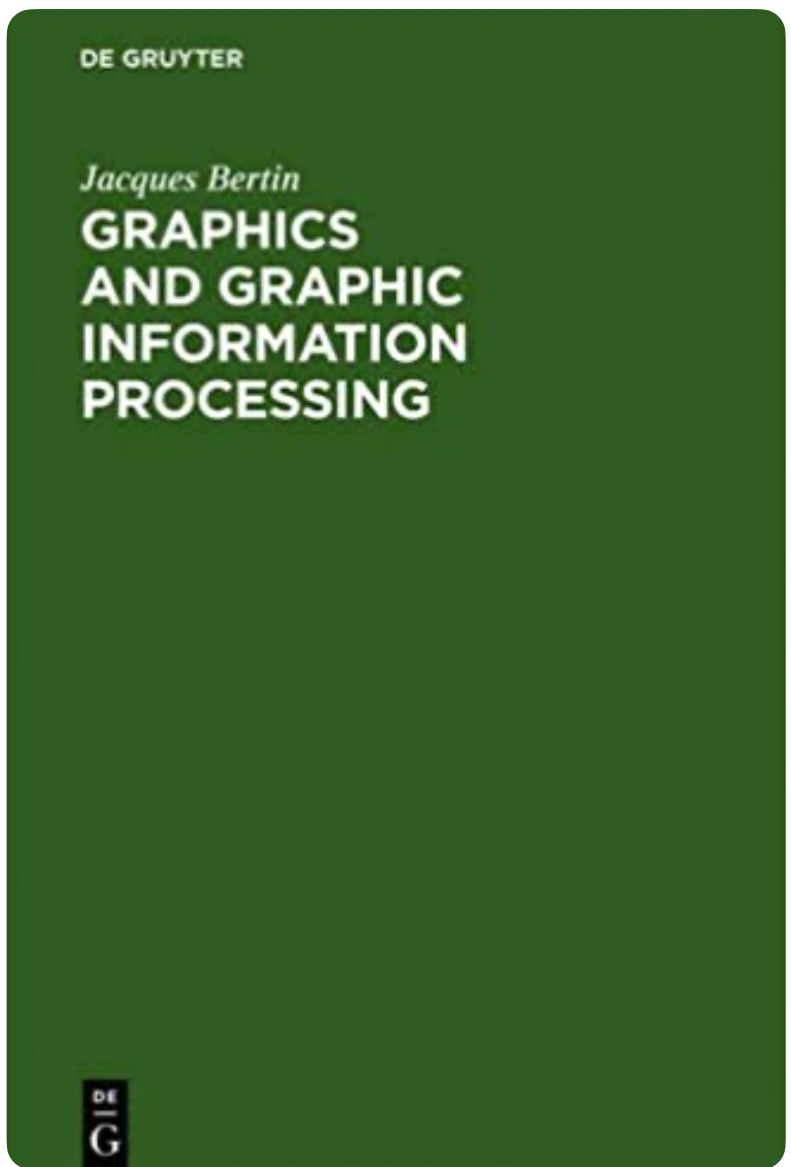
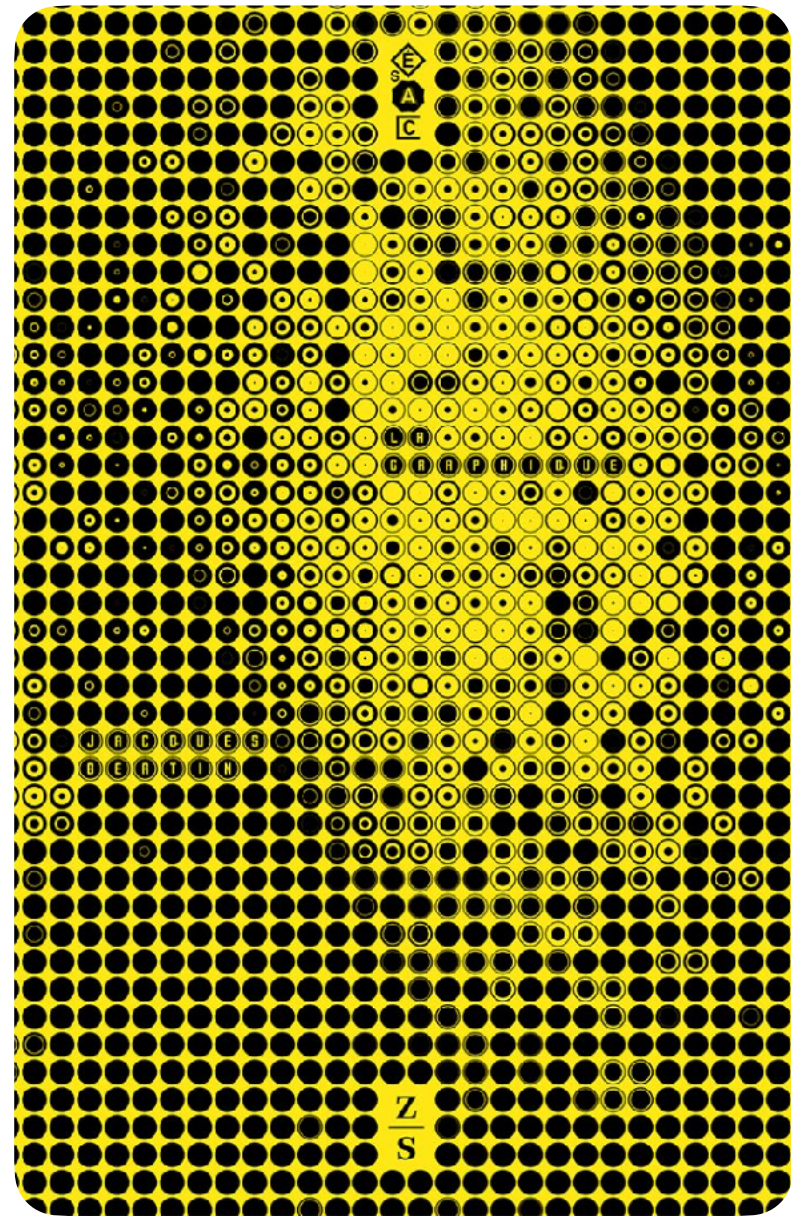


DE GRUYTER

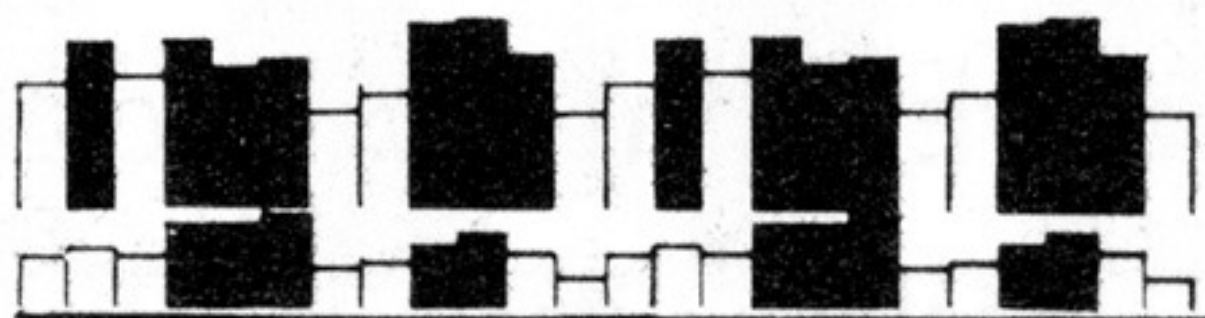
Jacques Bertin
**GRAPHICS
 AND GRAPHIC
 INFORMATION
 PROCESSING**

GIP

J	F	M	A	M	J	J	A	S	O	N	D		
26	21	26	28	20	20	20	20	20	40	15	40	1	% CLIENTELE FEMALE
69	70	77	71	37	36	39	39	55	60	68	72	2	% —" — LOCAL
7	6	3	6	23	14	19	14	9	6	8	8	3	% —" — U.S.A.
0	0	0	0	8	6	6	4	2	12	0	0	4	% —" — SOUTH AMERICA
20	15	14	15	23	27	22	30	27	19	19	17	5	% —" — EUROPE
1	0	0	8	6	4	6	4	2	1	0	1	6	% —" — M.EAST, AFRICA
3	10	6	0	3	13	8	9	5	2	5	2	7	% —" — ASIA
78	80	85	86	85	87	70	76	87	85	87	80	8	% BUSINESSMEN
22	20	15	14	15	13	30	24	13	15	13	20	9	% TOURISTS
70	70	75	74	69	68	74	75	68	68	64	75	10	% DIRECT RESERVATIONS
20	18	19	17	27	27	19	19	26	27	21	15	11	% AGENCY —" —
10	12	6	9	4	5	7	6	6	5	15	10	12	% AIR CREWS
2	2	4	2	2	1	1	2	2	4	2	5	13	% CLIENTS UNDER 20 YEARS
25	27	37	35	25	25	27	28	24	30	24	30	14	% —" — 20-35 —" —
48	49	42	48	54	55	53	51	55	46	55	43	15	% —" — 35-55 —" —
25	22	17	15	19	19	19	19	19	20	19	22	16	% —" — MORE THAN 55 —" —
163	167	166	174	152	155	145	170	157	174	165	156	17	PRICE OF ROOMS
1.65	1.71	1.65	1.91	1.90	2.	1.54	1.60	1.73	1.82	1.66	1.44	18	LENGTH OF STAY
67	82	70	83	74	77	56	62	90	92	78	55	19	% OCCUPANCY
			X	X	X			X	X	X	X	20	CONVENTIONS

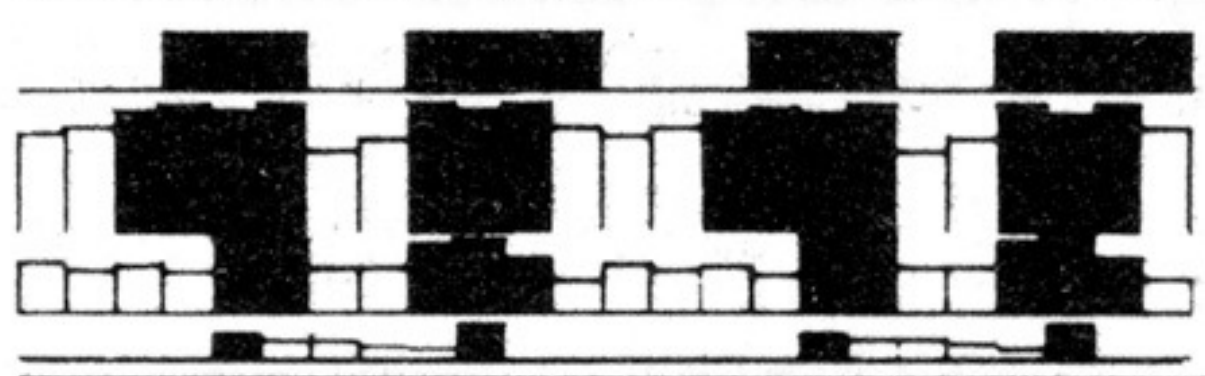


J F M A M J J A S O N D J F M A M J J A S O N D



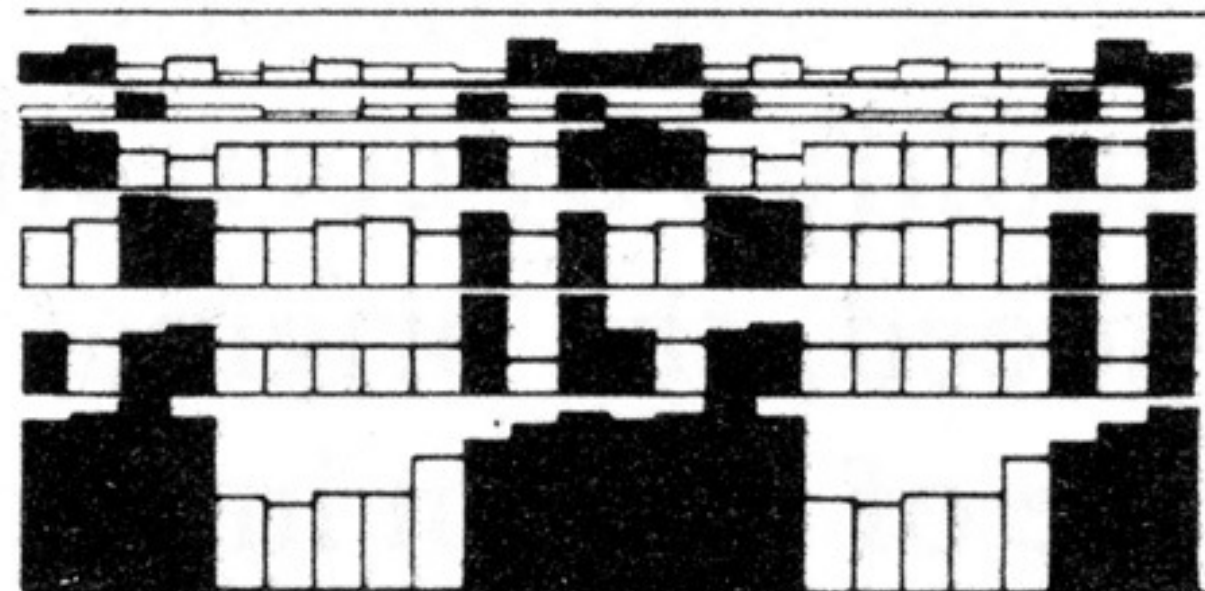
18 % OCCUPANCY
18 LENGTH OF STAY

ACTIVE AND SLOW PERIODS



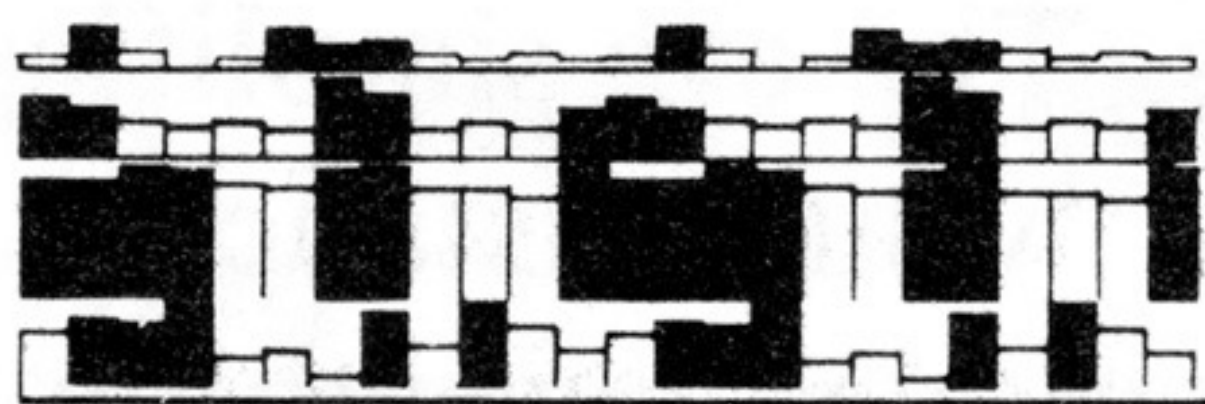
20 CONVENTIONS
8 BUSINESSMEN
11 AGENCY RESERVATIONS
4 SOUTH AMERICA

DISCOVERY FACTORS



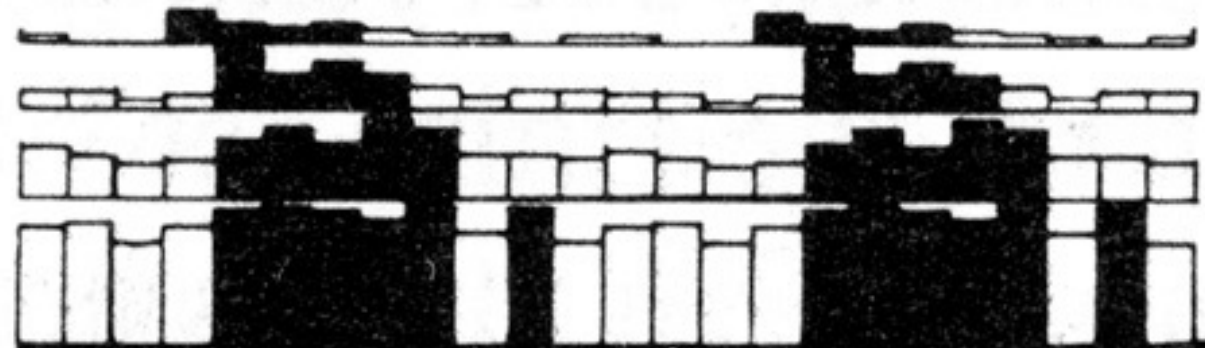
18 AIR CREWS
18 CLIENTS UNDER 20 YEARS
18 CLIENTS MORE THAN 55 YEARS
14 CLIENTS FROM 20-35 YEARS
1 FEMALE CLIENTELE
2 LOCAL CLIENTELE

RECOVERY FACTORS
WINTER



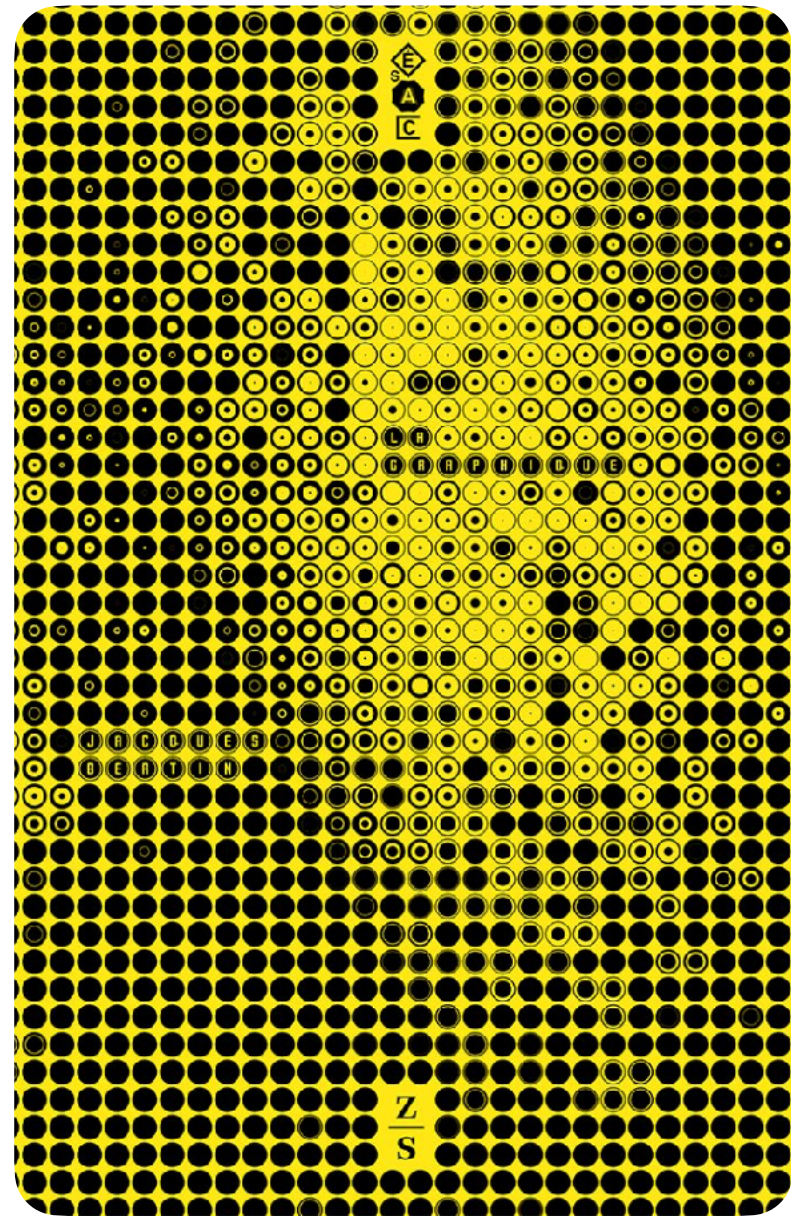
7 ASIA
9 TOURISTS
10 DIRECT RESERVATION
17 PRICE OF ROOMS

WINTER-SUMMER



6 MIDDLE EAST, AFRICA
3 U. S. A.
5 EUROPE
15 CLIENTS FROM 35-55 YEARS

SUMMER

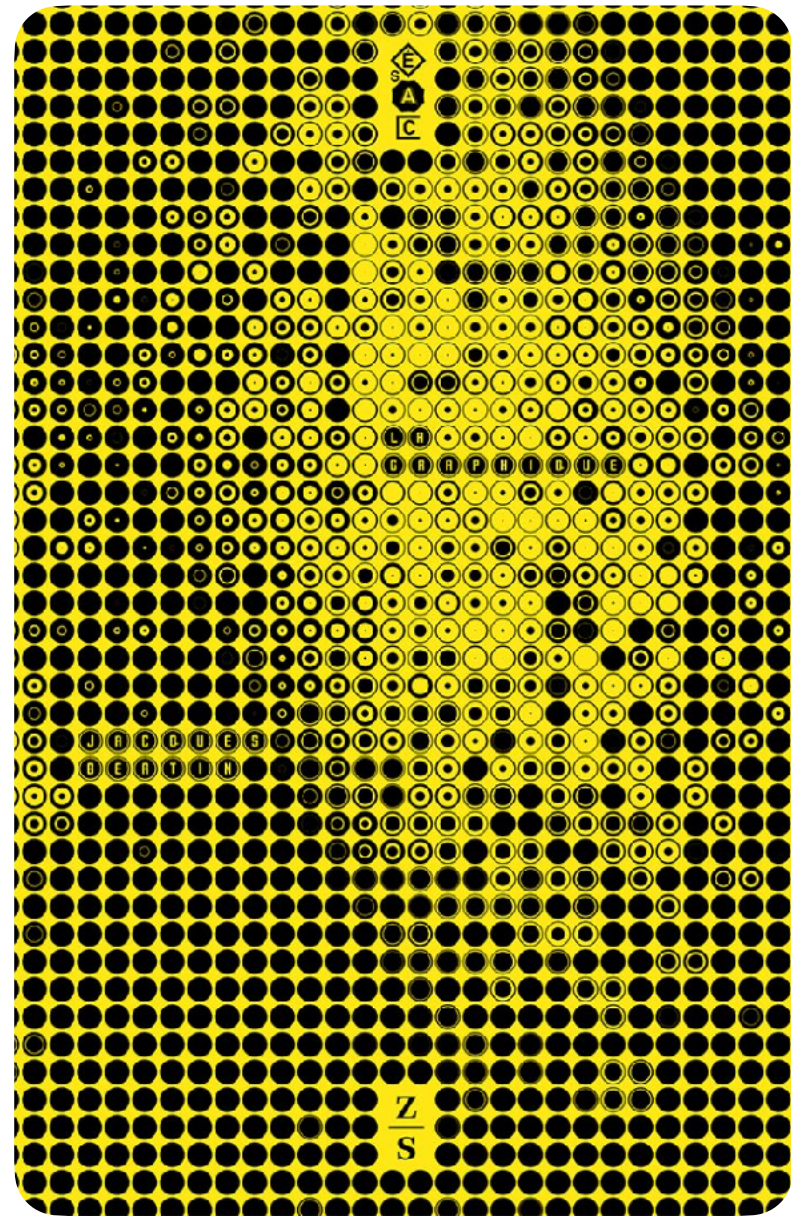


DE GRUYTER

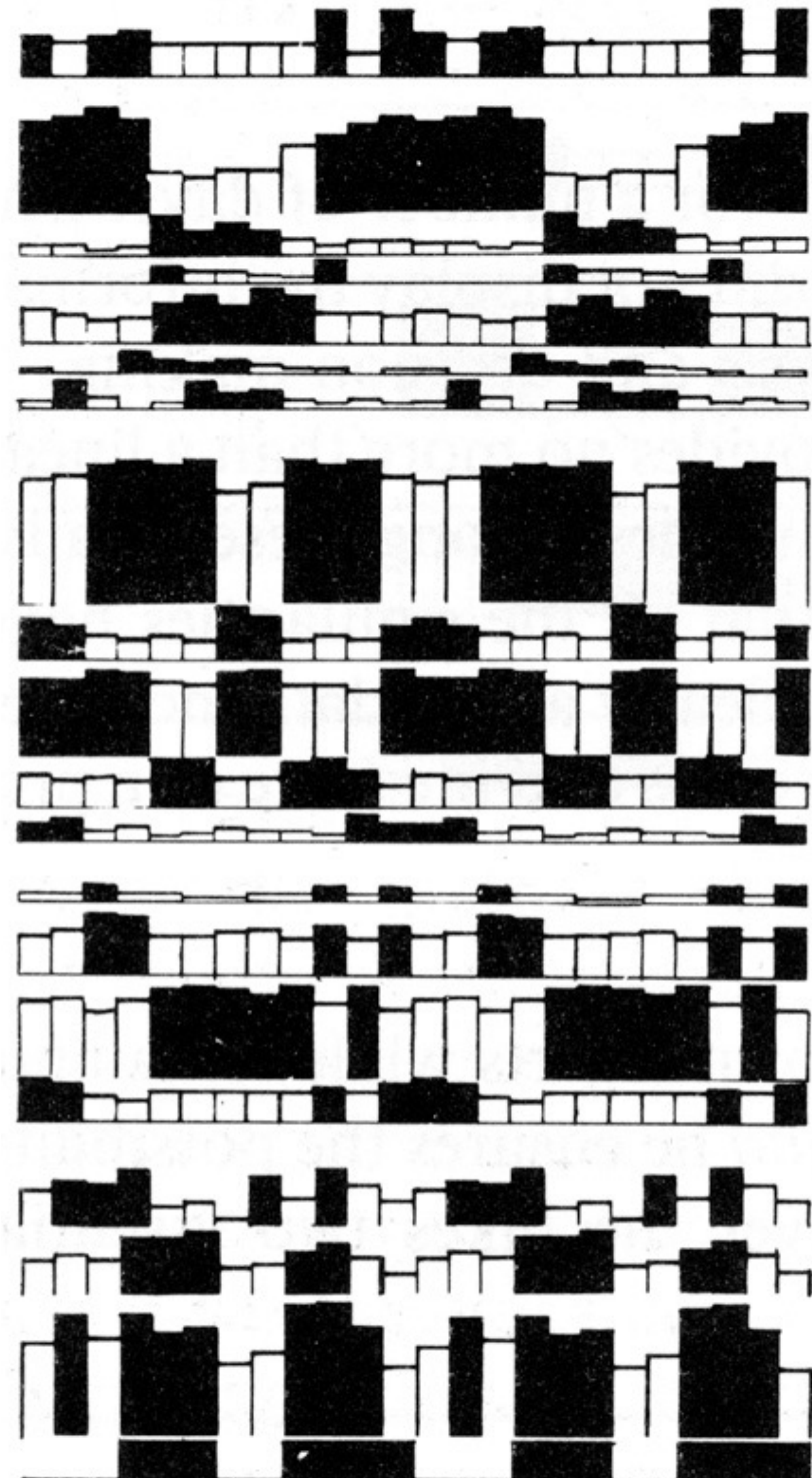
Jacques Bertin
**GRAPHICS
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 PROCESSING**

GIP

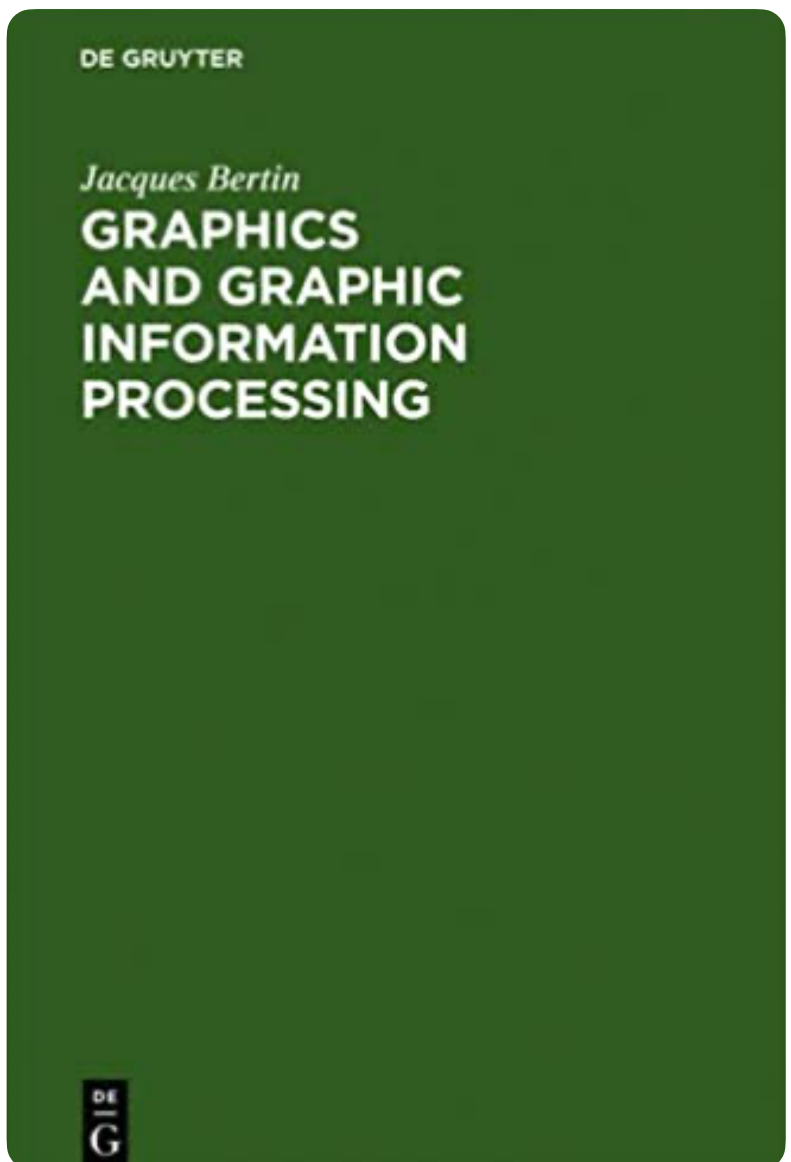
J	F	M	A	M	J	J	A	S	O	N	D		
26	21	26	28	20	20	20	20	20	40	15	40	1	% CLIENTELE FEMALE
69	70	77	71	37	36	39	39	55	60	68	72	2	% —"—— LOCAL
7	6	3	6	23	14	19	14	9	6	8	8	3	% —"—— U.S.A.
0	0	0	0	8	6	6	4	2	12	0	0	4	% —"—— SOUTH AMERICA
20	15	14	15	23	27	22	30	27	19	19	17	5	% —"—— EUROPE
1	0	0	8	6	4	6	4	2	1	0	1	6	% —"—— M.EAST, AFRICA
3	10	6	0	3	13	8	9	5	2	5	2	7	% —"—— ASIA
78	80	85	86	85	87	70	76	87	85	87	80	8	% BUSINESSMEN
22	20	15	14	15	13	30	24	13	15	13	20	9	% TOURISTS
70	70	75	74	69	68	74	75	68	68	64	75	10	% DIRECT RESERVATIONS
20	18	19	17	27	27	19	19	26	27	21	15	11	% AGENCY ———"——
10	12	6	9	4	5	7	6	6	5	15	10	12	% AIR CREWS
2	2	4	2	2	1	1	2	2	4	2	5	13	% CLIENTS UNDER 20 YEARS
25	27	37	35	25	25	27	28	24	30	24	30	14	% —"—— 20-35 —"——
48	49	42	48	54	55	53	51	55	46	55	43	15	% —"—— 35-55 —"——
25	22	17	15	19	19	19	19	19	20	19	22	16	% —"—— MORE THAN 55 —"——
163	167	166	174	152	155	145	170	157	174	165	156	17	PRICE OF ROOMS
1.65	1.71	1.65	1.91	1.90	2.	1.54	1.60	1.73	1.82	1.66	1.44	18	LENGTH OF STAY
67	82	70	83	74	77	56	62	90	92	78	55	19	% OCCUPANCY
			X	X	X			X	X	X	X	20	CONVENTIONS



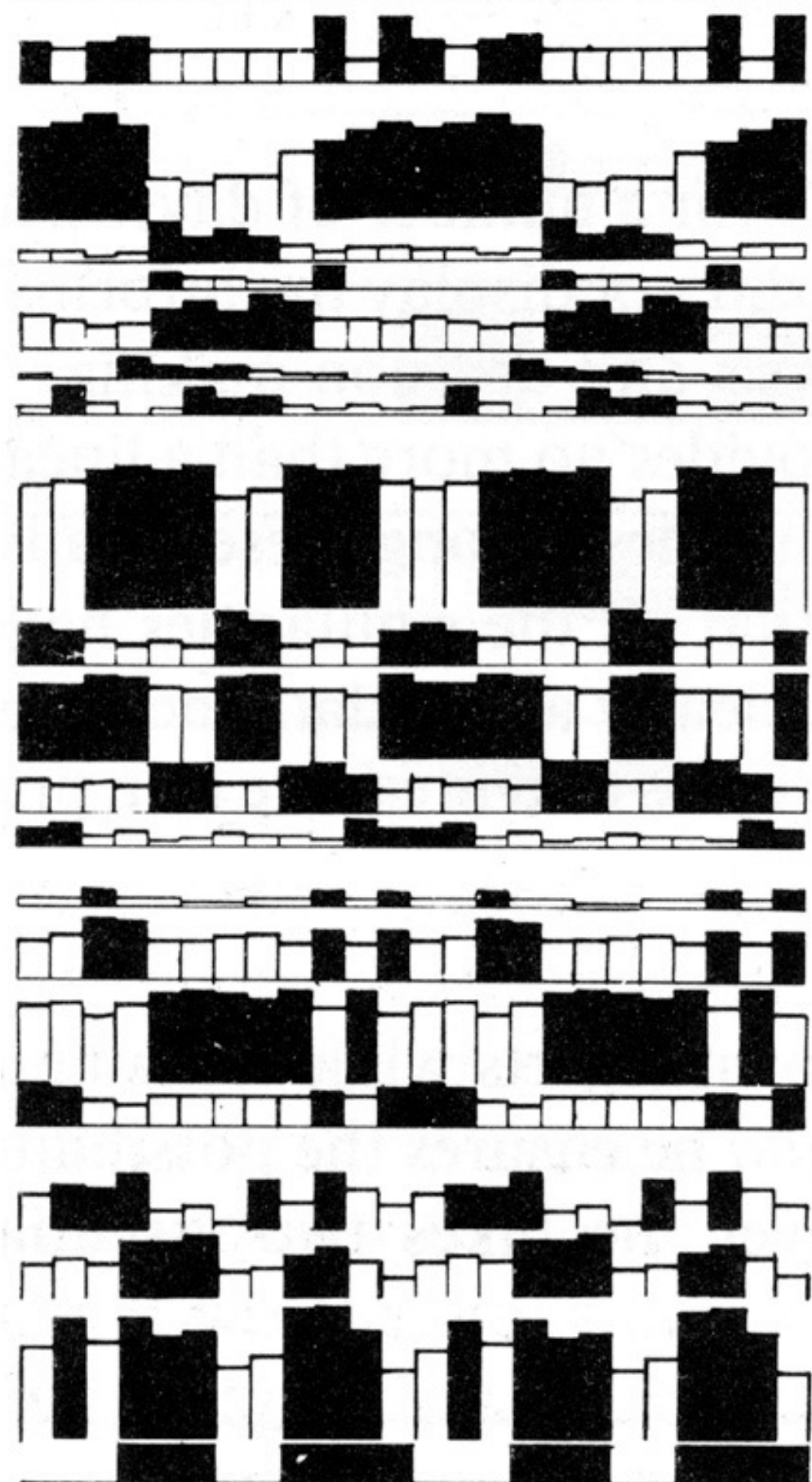
J FMAMJ JASOND J FMAMJ JASOND



1	% CLIENTELE FEMALE
2	% —"—— LOCAL
3	% —"—— U.S.A.
4	% —"—— SOUTH AMERICA
5	% —"—— EUROPE
6	% —"—— M.EAST, AFRICA
7	% —"—— ASIA
8	% BUSINESSMEN
9	% TOURISTS
10	% DIRECT RESERVATIONS
11	% AGENCY ———"———
12	% AIR CREWS
13	% CLIENTS UNDER 20 YEARS
14	% —"——— 20-35 —"—
15	% —"——— 35-55 —"—
16	% —"——— MORE THAN 55 —"—
17	PRICE OF ROOMS
18	LENGTH OF STAY
19	% OCCUPANCY
20	CONVENTIONS

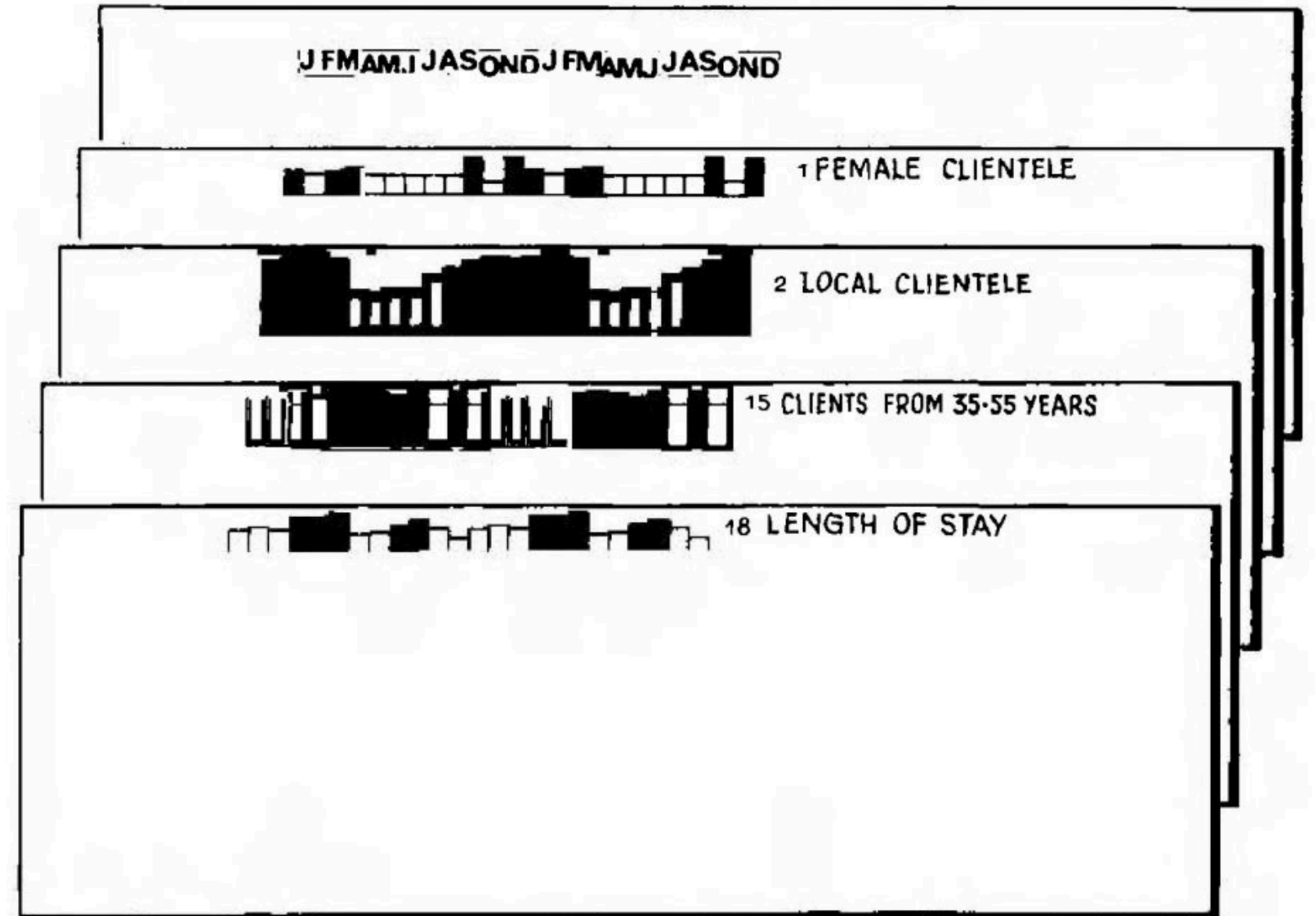


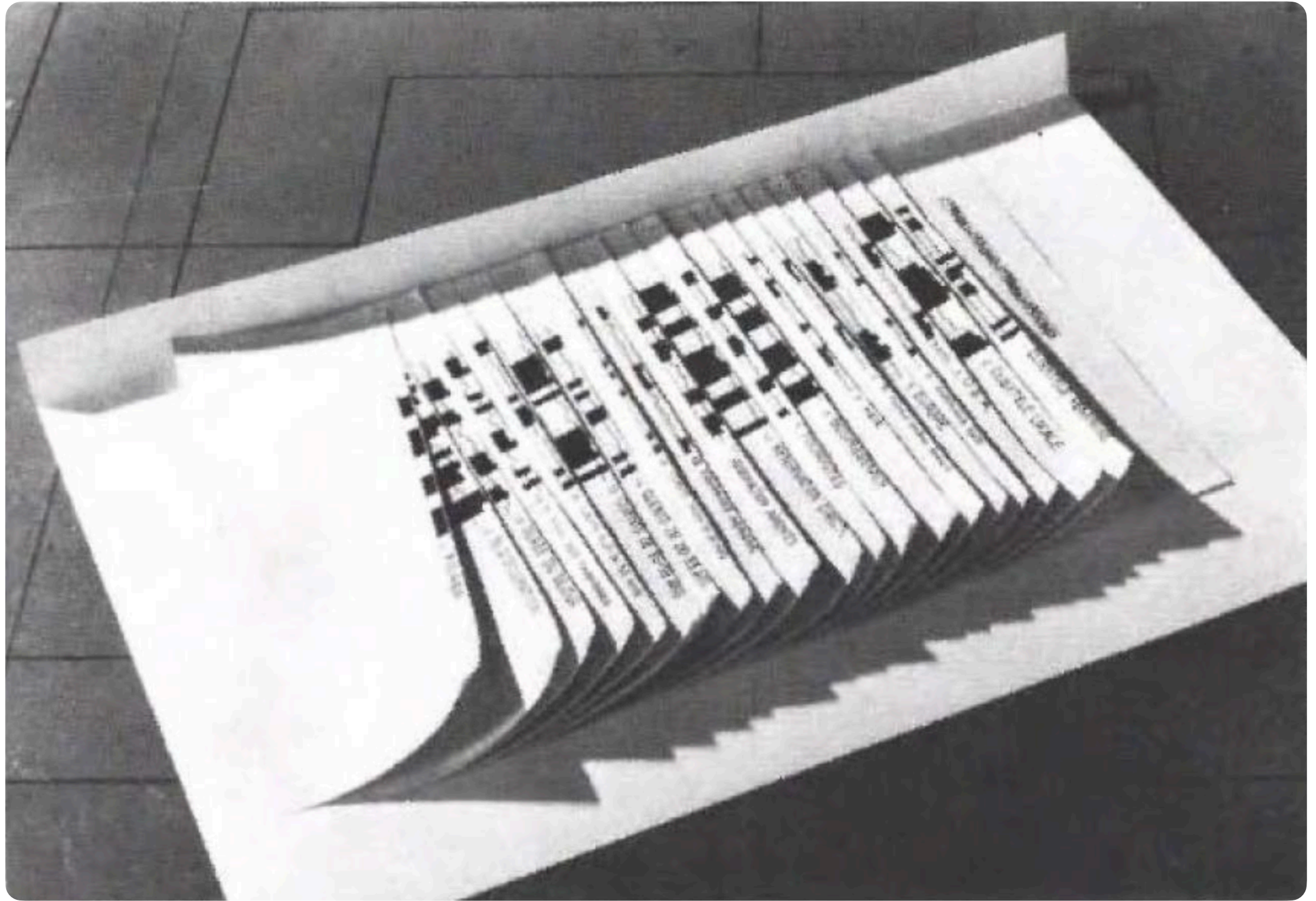
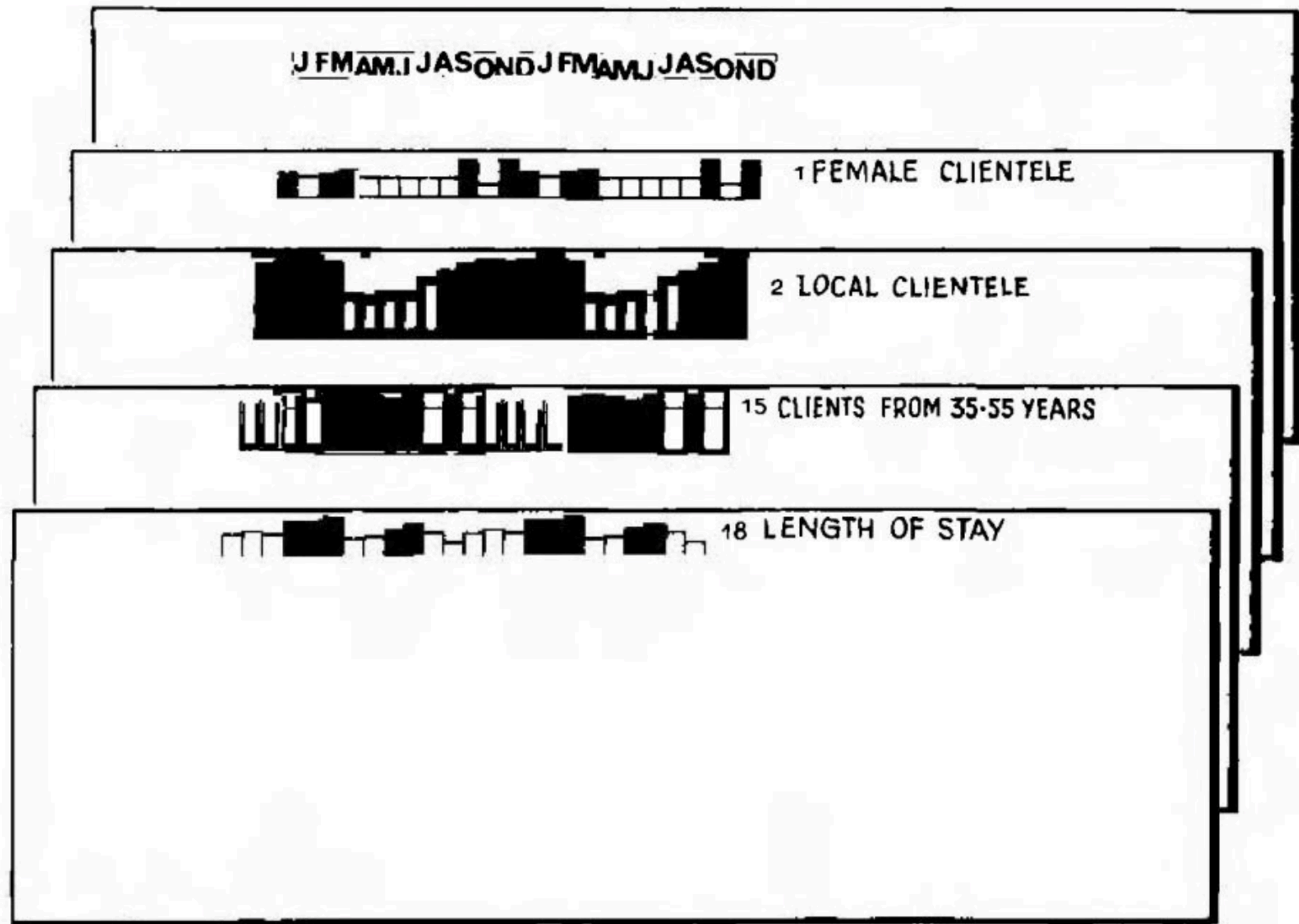
J FMAMJ JASONDJ FMAMJ JASONDJ



1	% CLIENTELE FEMALE
2	% —//— LOCAL
3	% —//— U.S.A.
4	% —//— SOUTH AMERICA
5	% —//— EUROPE
6	% —//— M.EAST, AFRICA
7	% —//— ASIA
8	% BUSINESSMEN
9	% TOURISTS
10	% DIRECT RESERVATIONS
11	% AGENCY —//—
12	% AIR CREWS
13	% CLIENTS UNDER 20 YEARS
14	% —//— 20-35 —//—
15	% —//— 35-55 —//—
16	% —//— MORE THAN 55 —//—
17	PRICE OF ROOMS
18	LENGTH OF STAY
19	% OCCUPANCY
20	CONVENTIONS

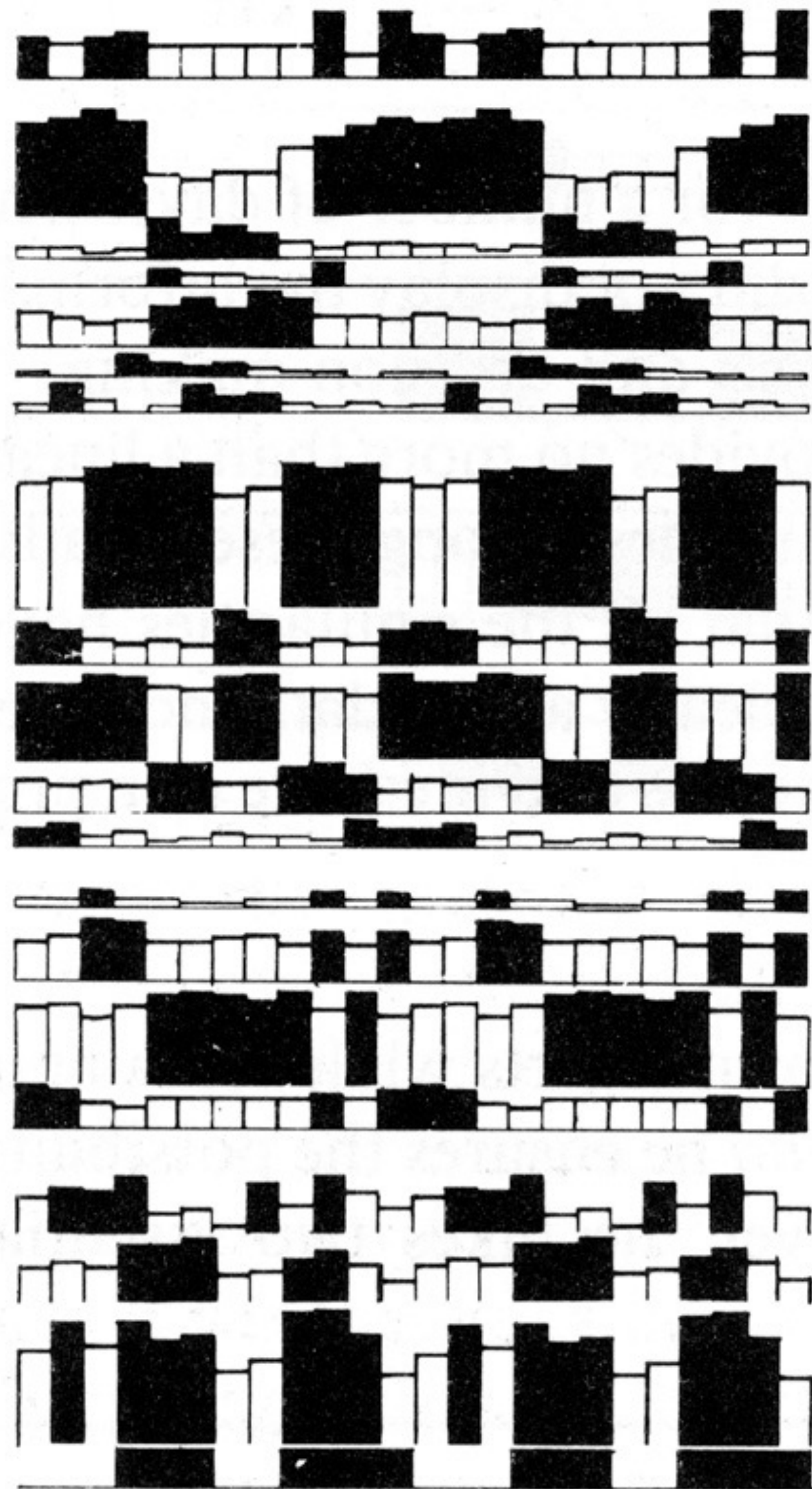
J FMAMJ JASONDJ FMAMJ JASONDJ





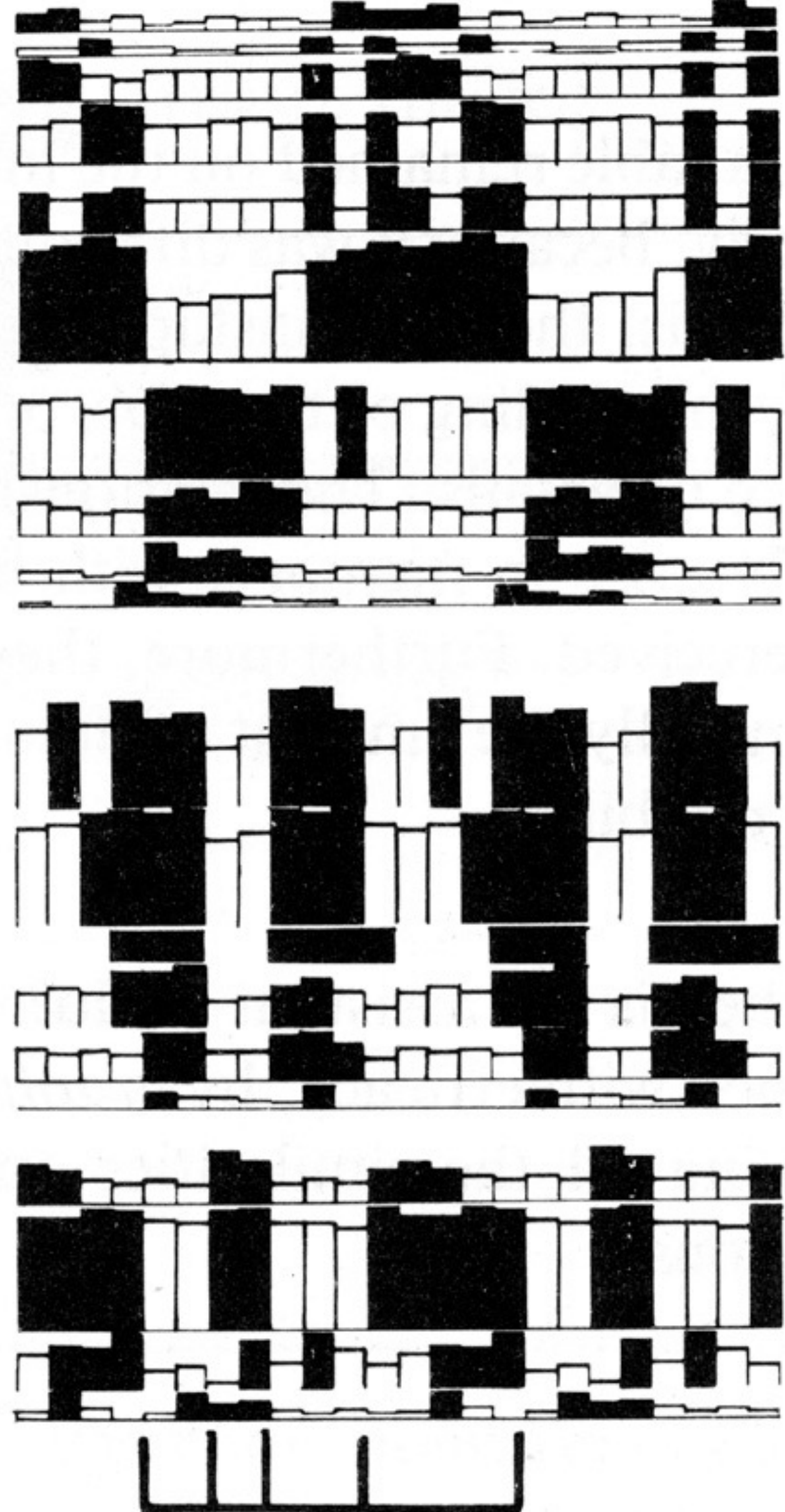
1

J FMAMJJASONDJ FMAMJJASOND

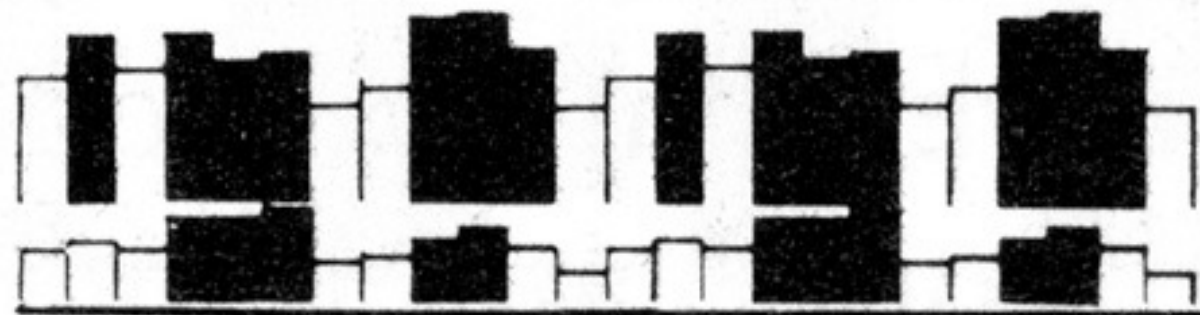


2

J FMAMJJASONDJ FMAMJJASOND



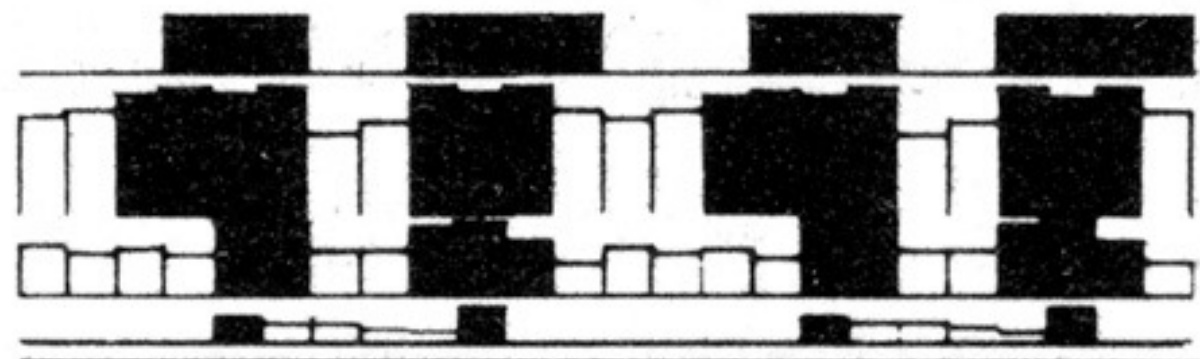
J F M A M J J A S O N D J F M A M J J A S O N D



18 % OCCUPANCY

18 LENGTH OF STAY

ACTIVE AND SLOW PERIODS



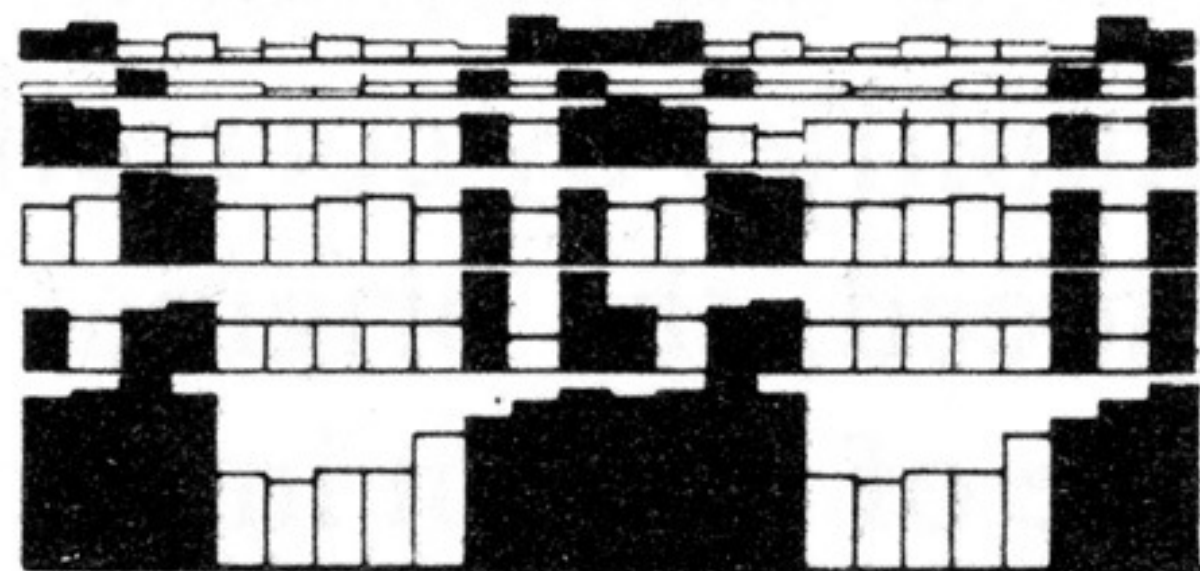
20 CONVENTIONS

8 BUSINESSMEN

11 AGENCY RESERVATIONS

4 SOUTH AMERICA

DISCOVERY FACTORS



18 AIR CREWS

18 CLIENTS UNDER 20 YEARS

18 CLIENTS MORE THAN 55 YEARS

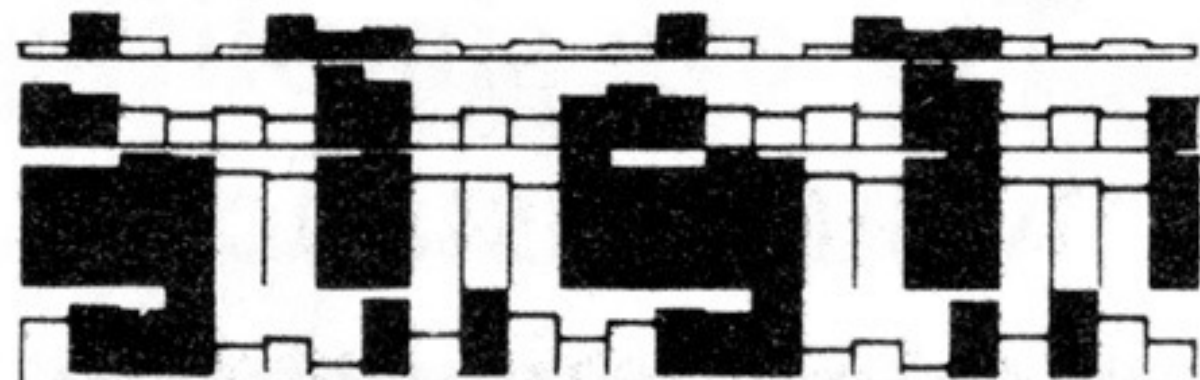
14 CLIENTS FROM 20-35 YEARS

1 FEMALE CLIENTELE

2 LOCAL CLIENTELE

RECOVERY FACTORS

WINTER



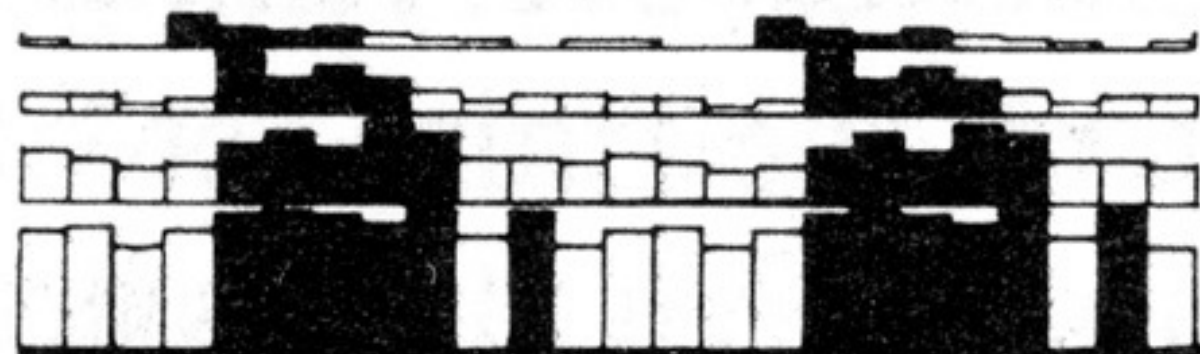
7 ASIA

9 TOURISTS

10 DIRECT RESERVATION

17 PRICE OF ROOMS

WINTER-SUMMER



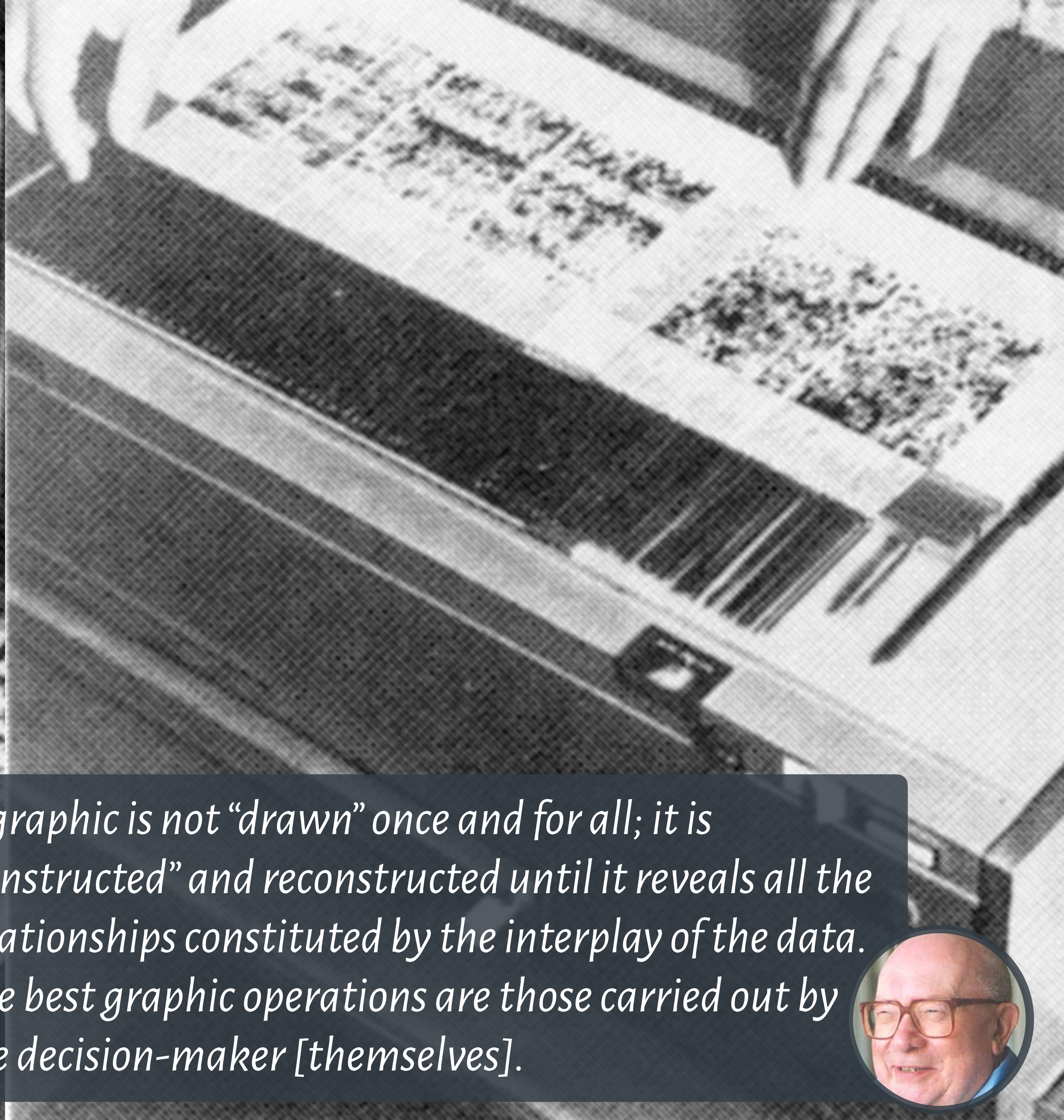
6 MIDDLE EAST, AFRICA

3 U.S.A.

5 EUROPE

15 CLIENTS FROM 35-55 YEARS

SUMMER



A graphic is not “drawn” once and for all; it is “constructed” and reconstructed until it reveals all the relationships constituted by the interplay of the data. The best graphic operations are those carried out by the decision-maker [themselves].

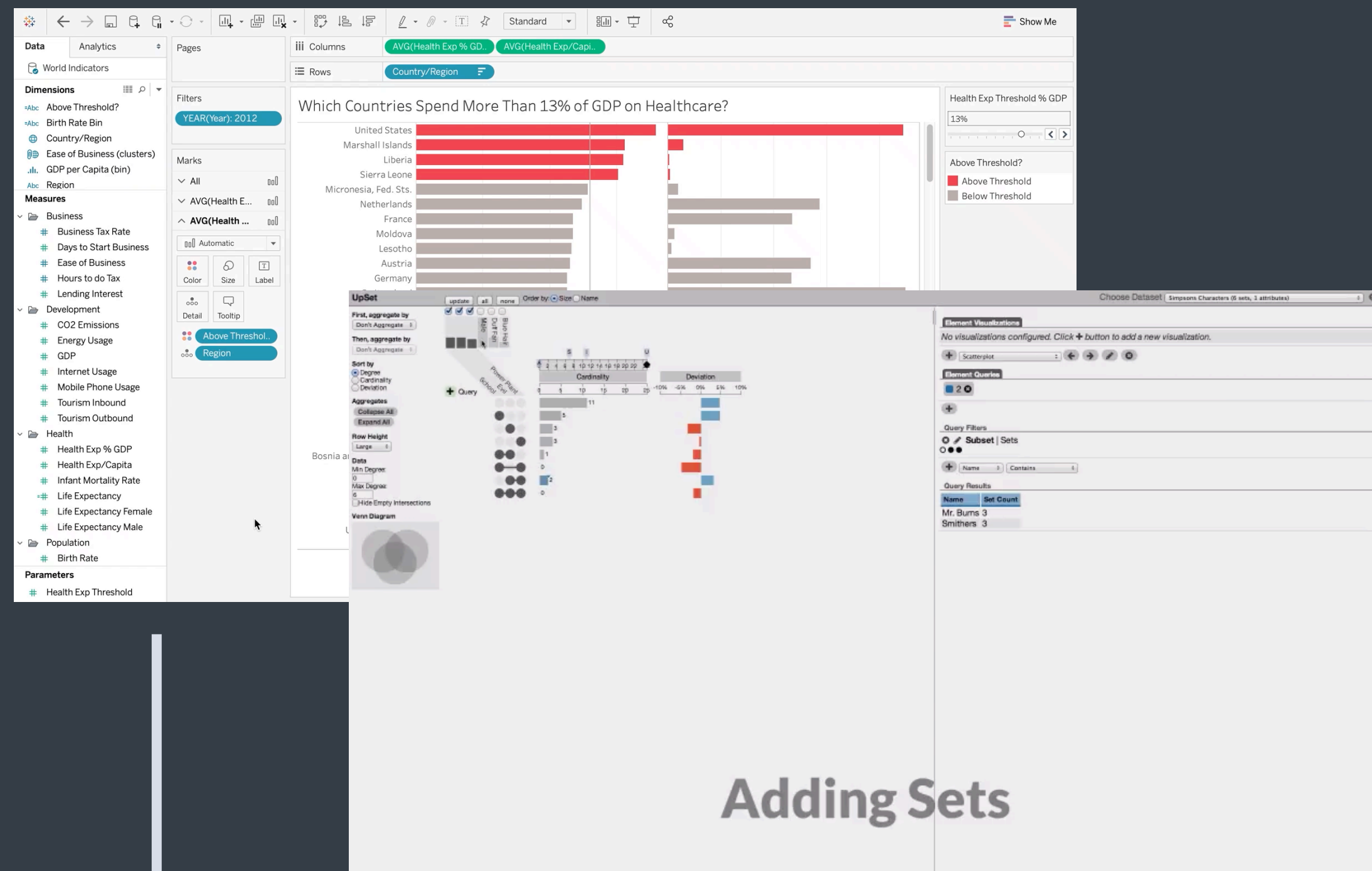


“Visual representations alone cannot satisfy analytical needs. Interaction techniques are required to support the dialogue between the analyst and the data.”

[Thomas & Cook, 2005]

“It is through the interactive manipulation of a visual interface – the analytic discourse – that knowledge is constructed, tested, refined, and shared.”

[Pike et al., 2009]



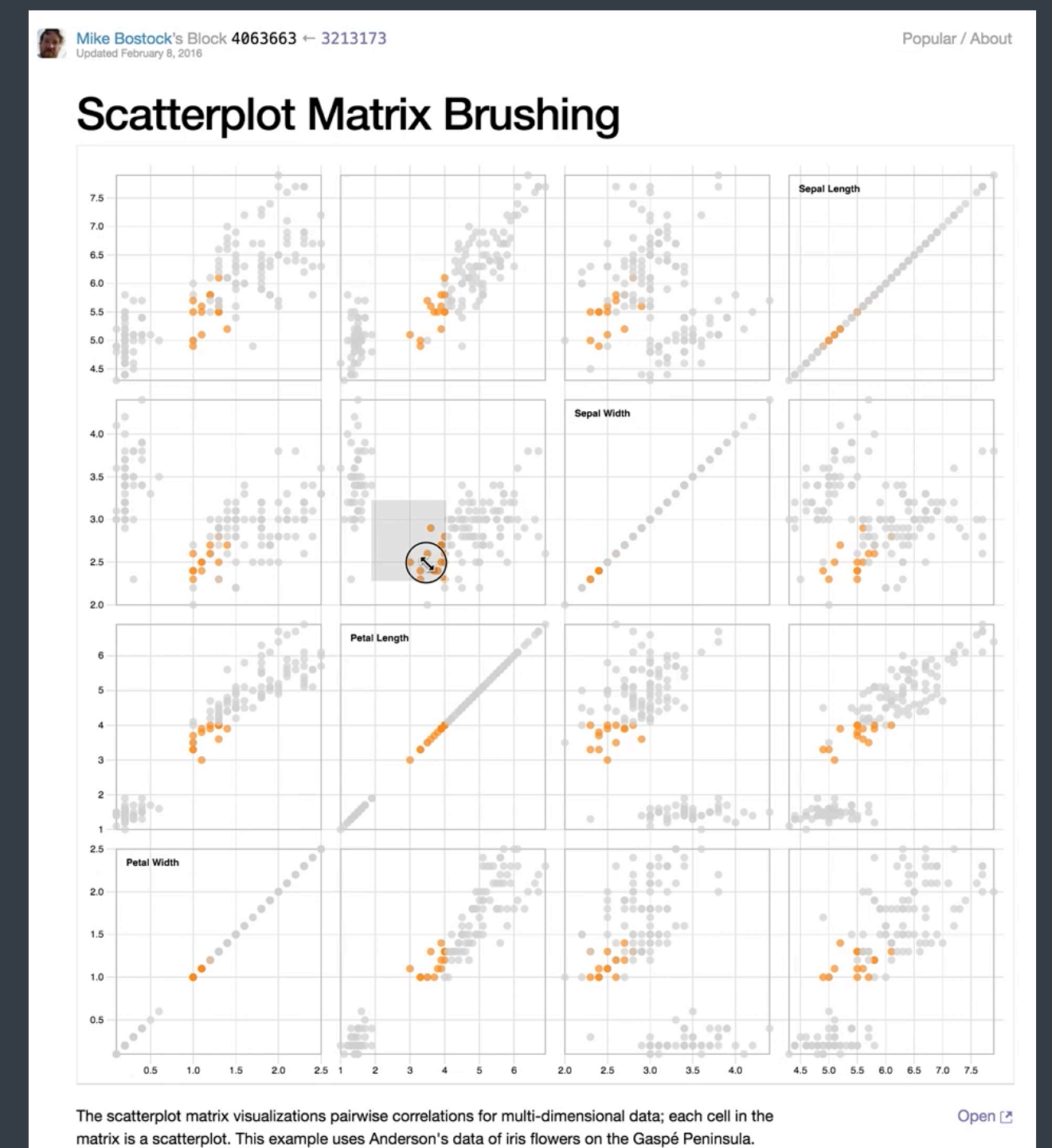
Adding Sets

Fixed palette of general-purpose techniques

Custom techniques in purpose-built tools

???

Programming custom techniques



The scatterplot matrix visualizations pairwise correlations for multi-dimensional data; each cell in the matrix is a scatterplot. This example uses Anderson's data of iris flowers on the Gaspé Peninsula.

Vega-Lite: A Grammar of Interactive Graphics

with



```
{
  "data": {"url": "data/gapminder.json"},
  "mark": "circle",
  "encoding": {
    "x": {"field": "fertility", "type": "Q"},
    "y": {"field": "life_expectancy", "type": "Q"},
    "color": {"field": "region", "type": "N"},
    "size": {"field": "population", "type": "Q"},
  }
}
```




```
{
  "data": {"url": "data/gapminder.json"},
  "mark": "circle",
  "encoding": {
    "x": {"field": "fertility", "type": "Q"},
    "y": {"field": "life_expectancy", "type": "Q"},
    "color": {"field": "region", "type": "N"},
    "size": {"field": "population", "type": "Q"},
  }
}
```

Altair (Python)

```
alt.Chart("data/gapminder.json")
  .mark_circle()
  .encode(
    x='fertility:Q',
    y='life_expectancy:Q',
    color='region:N',
    size='population:Q'
  )
```

vegalite (R)

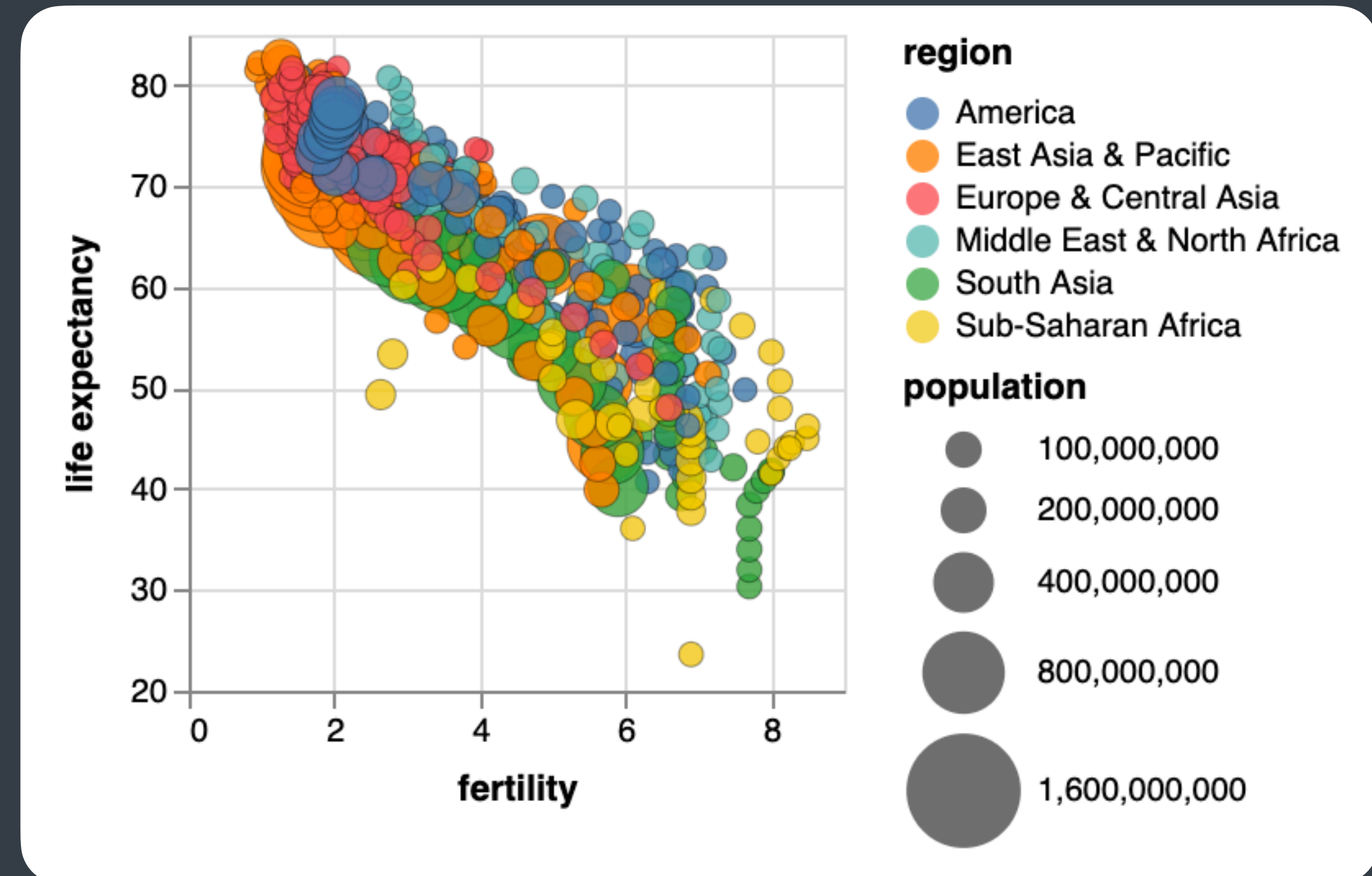
```
vegalite() %>%
  add_data("data/gapminder.json") %>%
  encode_x("fertility", "Q") %>%
  encode_y("life_expectancy", "Q") %>%
  encode_color("region", "N") %>%
  encode_size("population", "Q") %>%
  mark_circle()
```


Vega-Lite: A Grammar of Interactive Graphics

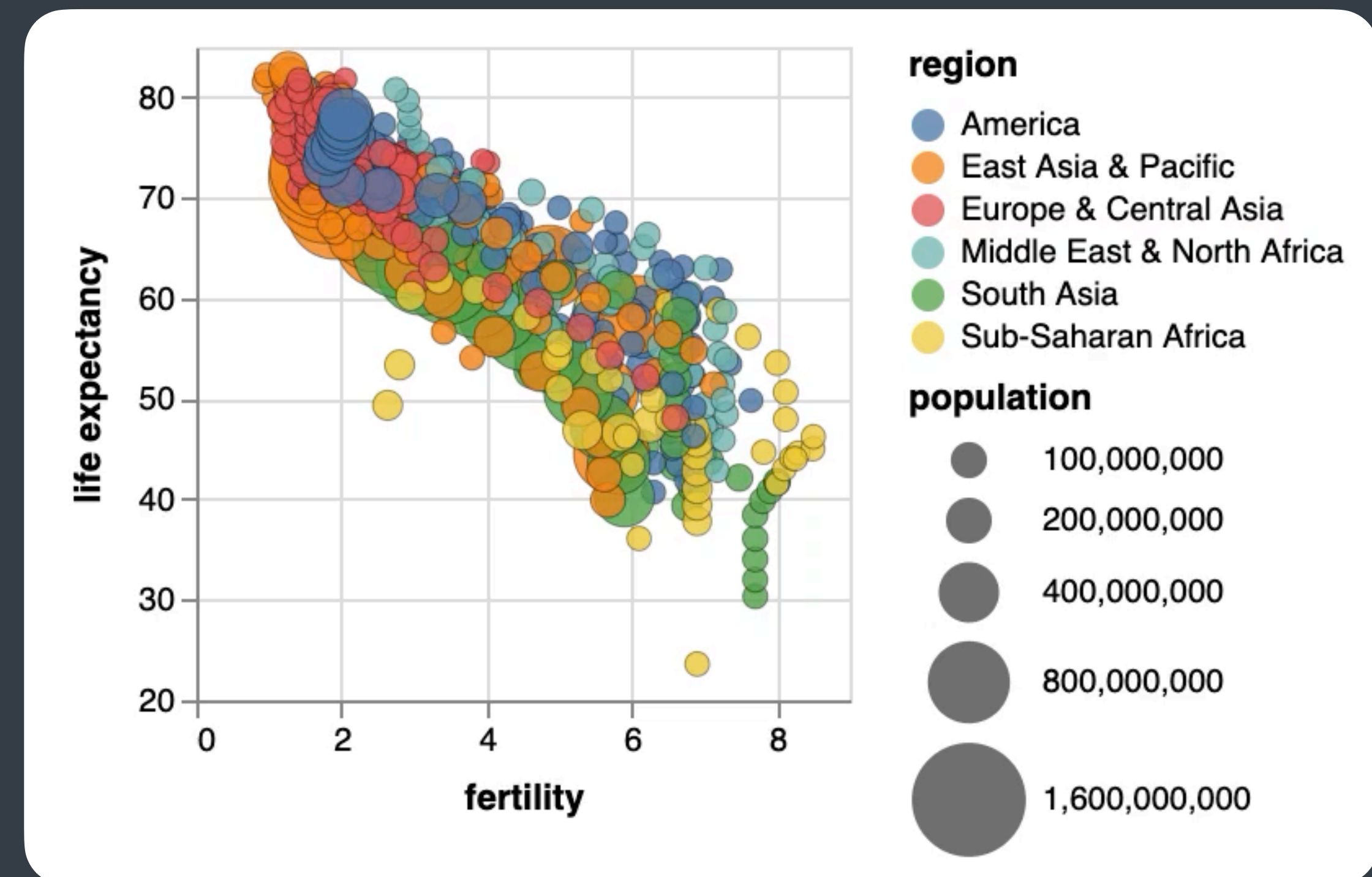
with



```
{  
  "data": {"url": "data/gapminder.json"},  
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  "encoding": {  
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    "y": {"field": "life_expectancy", "type": "Q"},  
    "color": {"field": "region", "type": "N"},  
    "size": {"field": "population", "type": "Q"},  
  }  
}
```



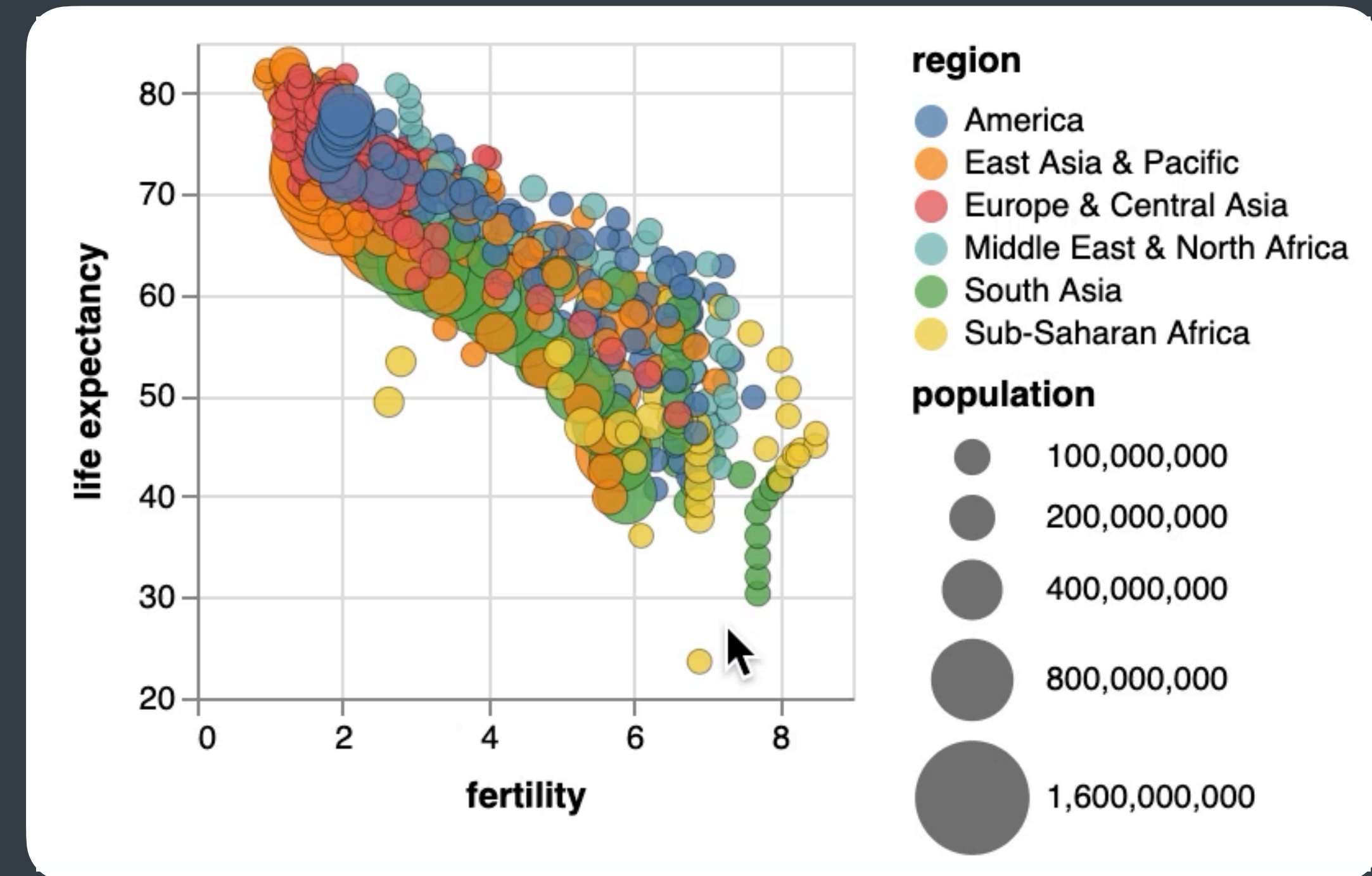

```
{
  "data": {"url": "data/gapminder.json"},
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  "encoding": {
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    "y": {"field": "life_expectancy", "type": "Q"},
    "color": {"field": "region", "type": "N"},
    "size": {"field": "population", "type": "Q"},
    "tooltip": [
      {"field": "country"},
      {"field": "year"}
    ]
  }
}
```




```

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  "encoding": {
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    "y": {"field": "life_expectancy", "type": "Q"},
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  },
  "tooltip": [
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    {"field": "year"}
  ],
}

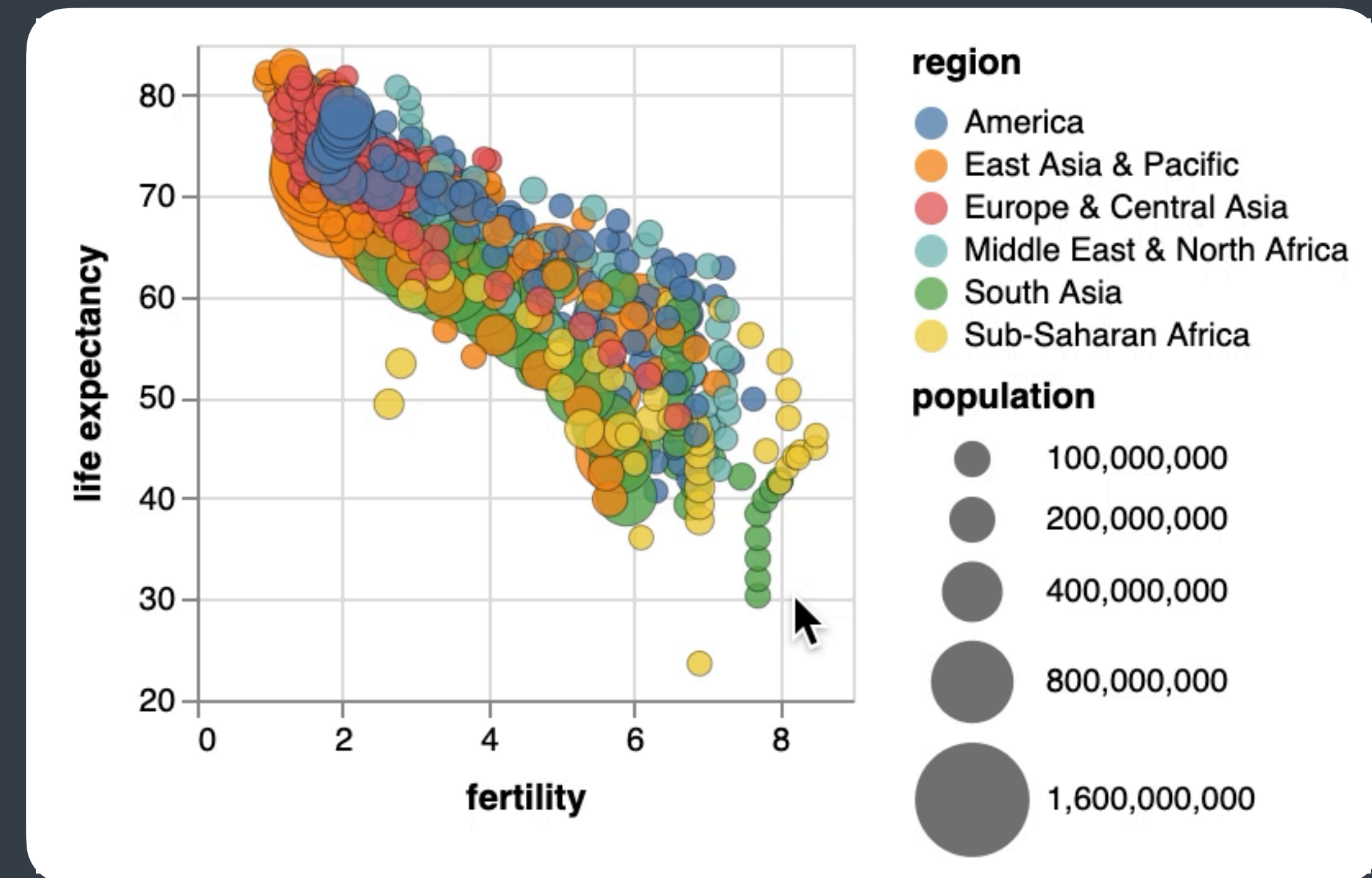
```




```

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  },
  "encoding": {
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    "color": {"field": "region", "type": "N"},
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  },
  "tooltip": [
    {"field": "country"},
    {"field": "year"}
  ],
}

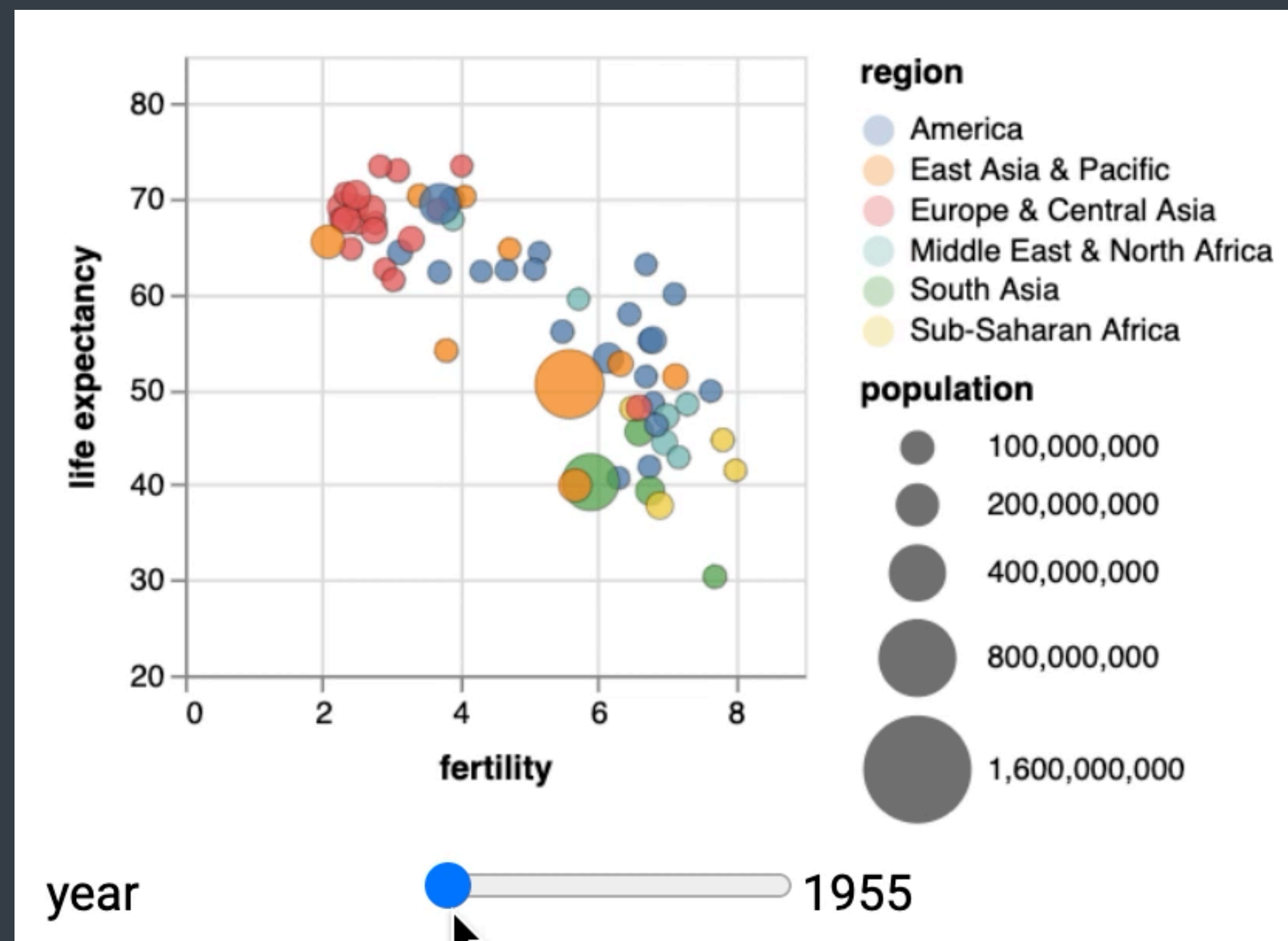
```




```

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  "encoding": {
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    "size": {"field": "population", "type": "Q"},
    "tooltip": [
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      {"field": "year"}
    ],
  },
}

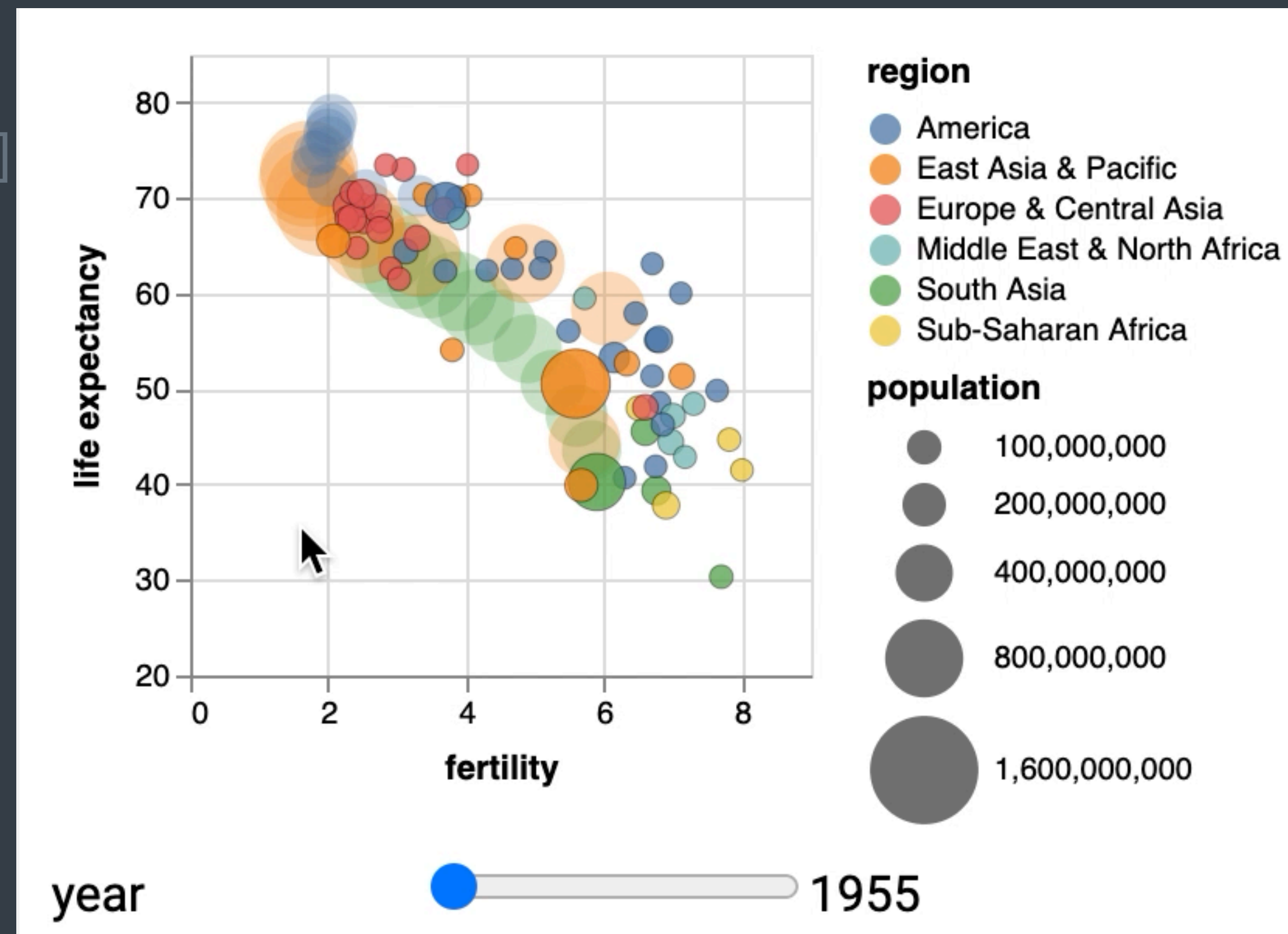
```




```

{
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      }
    }
  },
  "encoding": {...}
}, {
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  "encoding": {
    ...,
    "opacity": {
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      "value": 0.25
    }
  },
}
}]

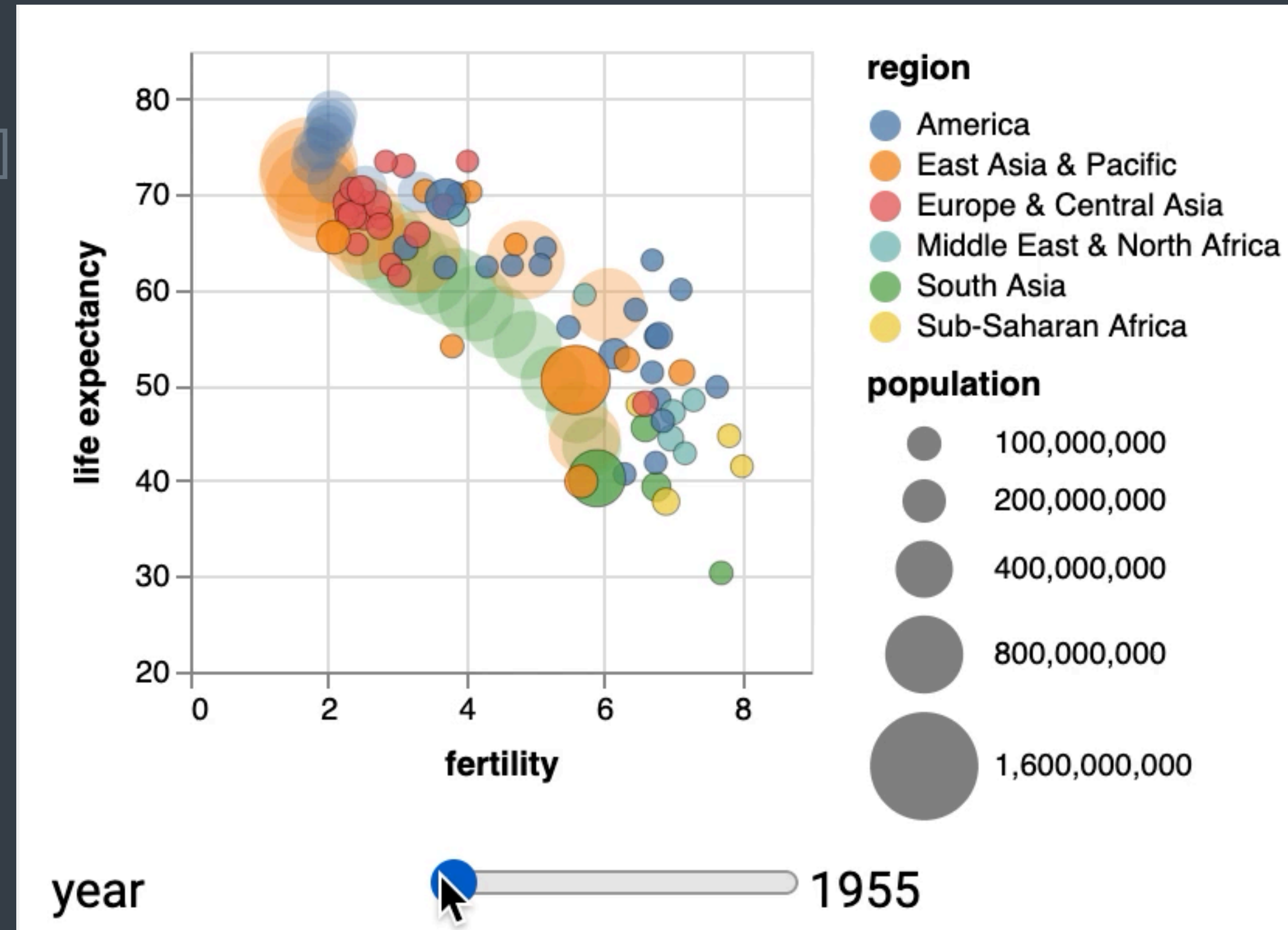
```




```

{
  "data": {"url": "data/gapminder.json"},
  "layer": [{
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    "mark": "circle",
    "selection": {
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      "yr": {
        "type": "single", "fields": ["year"]
        "bind": {"input": "range", ...}
      },
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  }, {
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    "encoding": {
      ...
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        "value": "grey"
      }
    }
  }
}

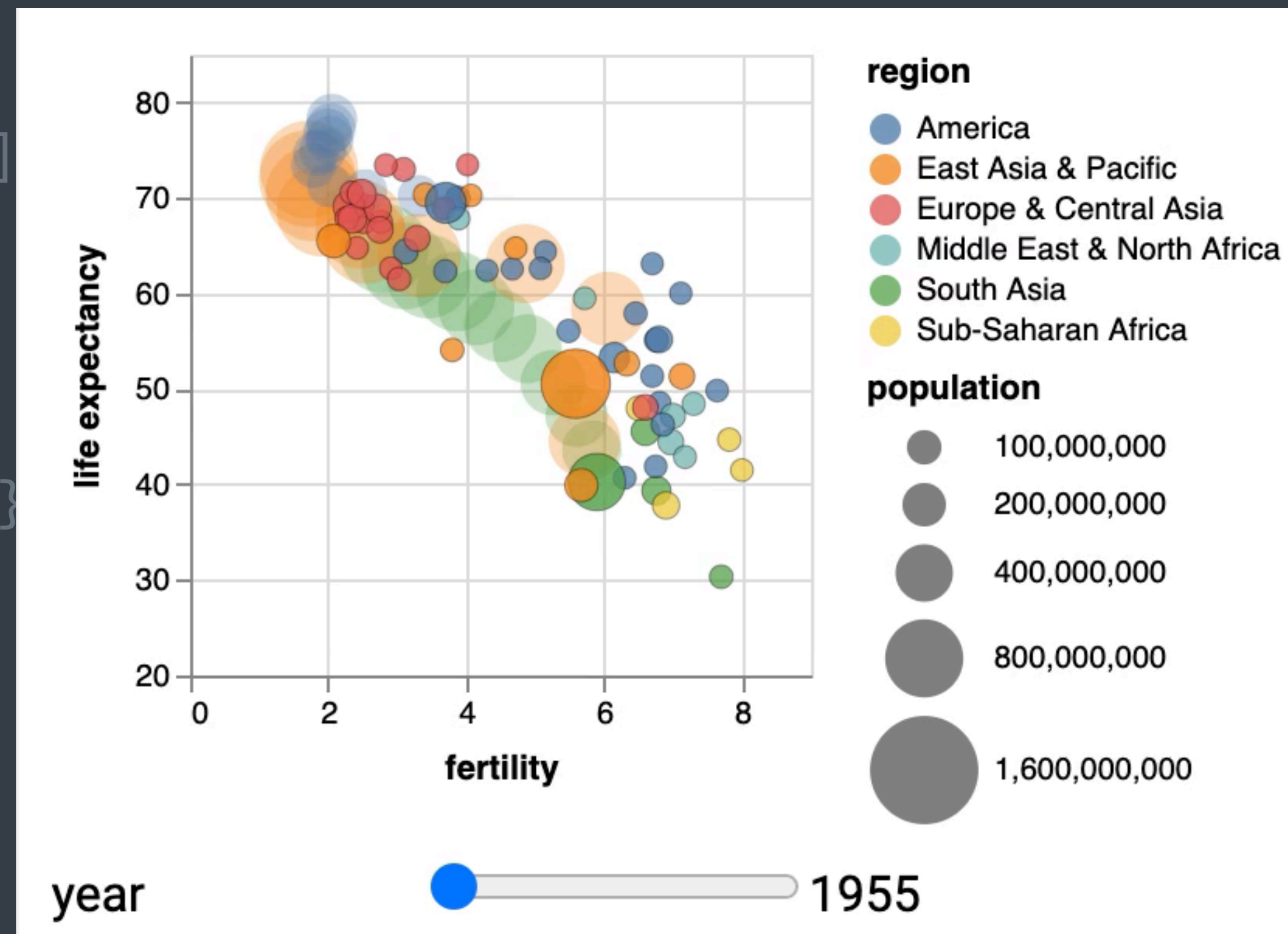
```



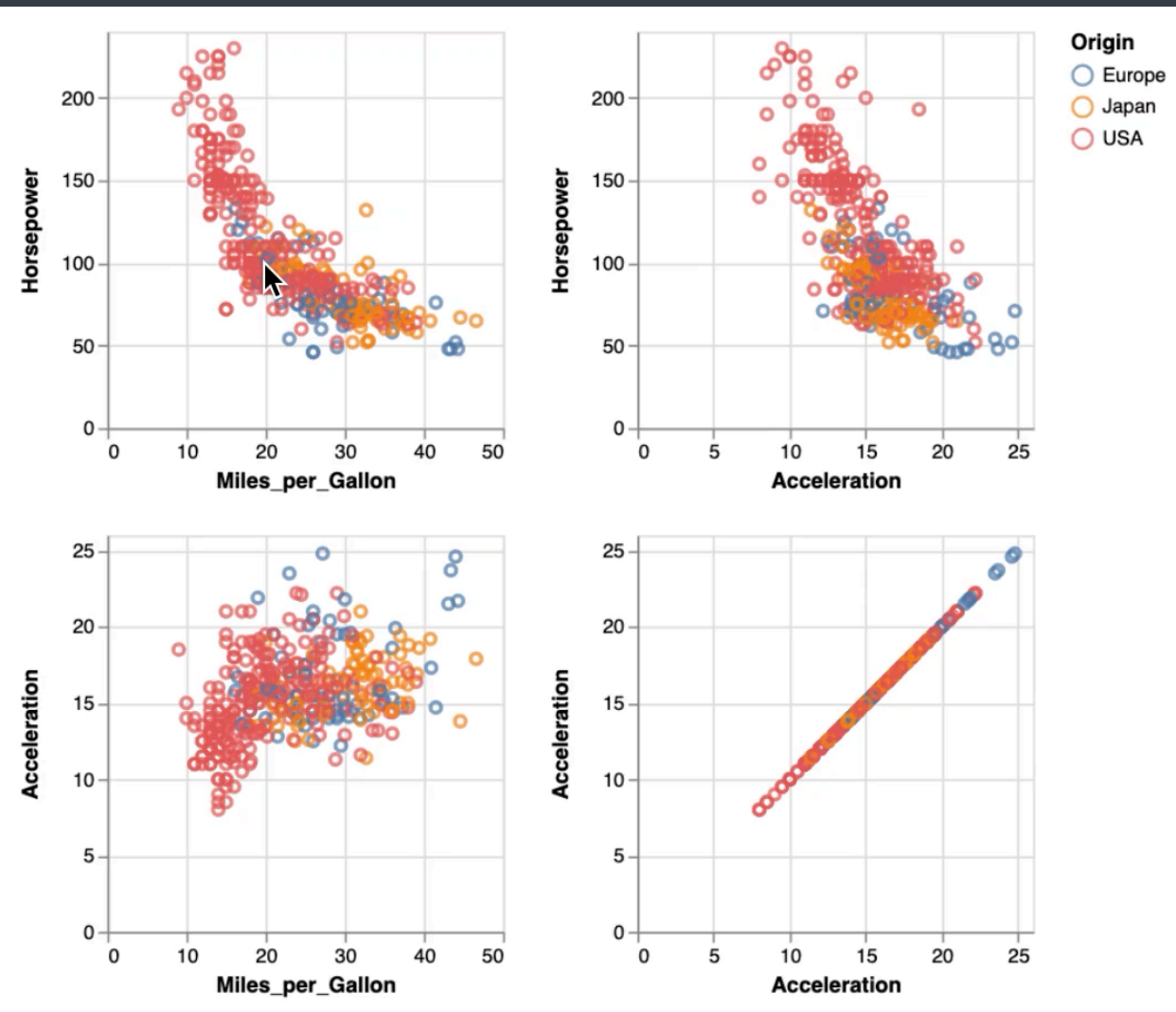

```

{
  "data": {"url": "data/gapminder.json"},
  "layer": [{
    "transform": [{"filter": {"selection": "yr"}}],
    "mark": "circle",
    "selection": {
      "cnty": {"type": "multi", "fields": ["country"]},
      "yr": {
        "type": "single", "fields": ["year"]
        "bind": {"input": "range", ...}
      },
      "brush": {"type": "interval", "bind": "scales"}
    },
    "encoding": {...}
  }, {
    "mark": "circle",
    "encoding": {
      ...
      "color": {
        "condition": {"selection": "brush", "field": "region", "type": "N"},
        "value": "grey"
      }
    }
  }
}

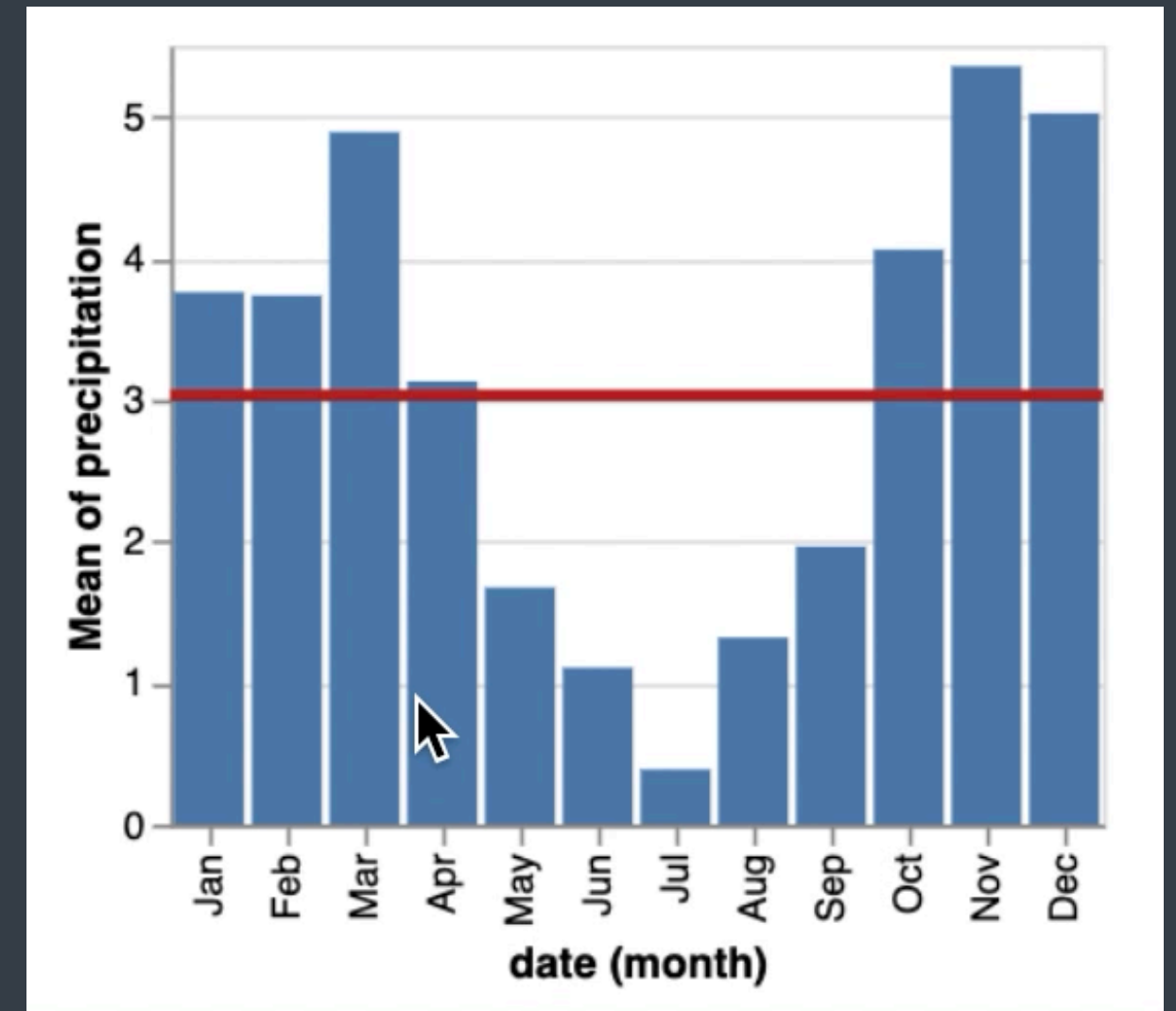
```



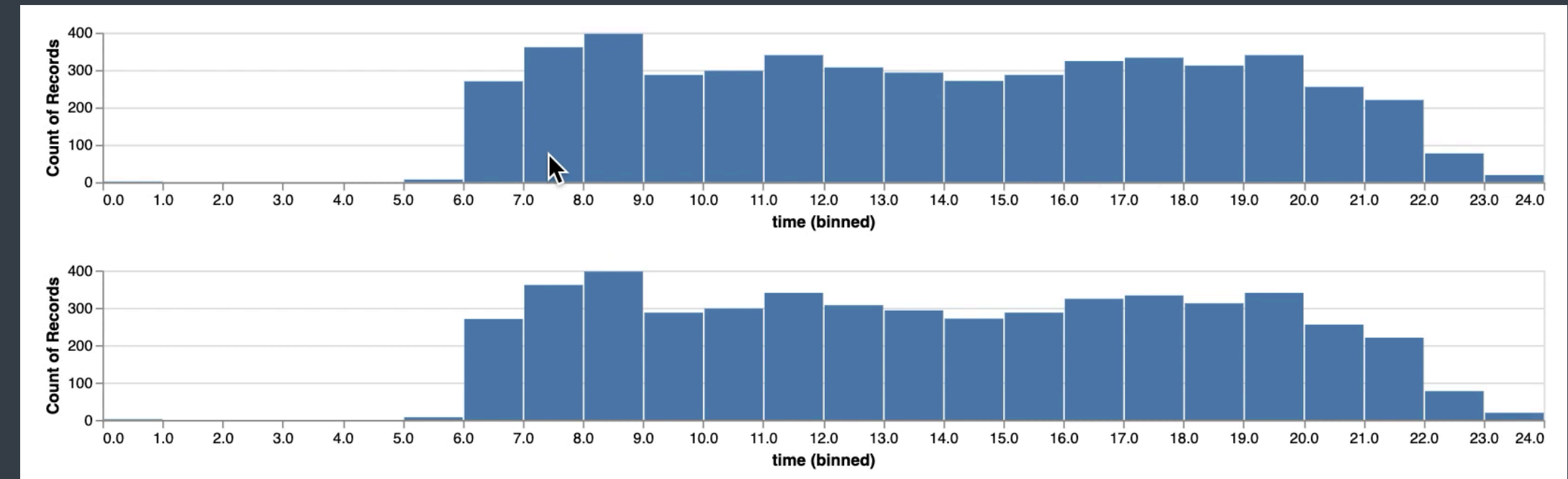
Union Multiple Brushes



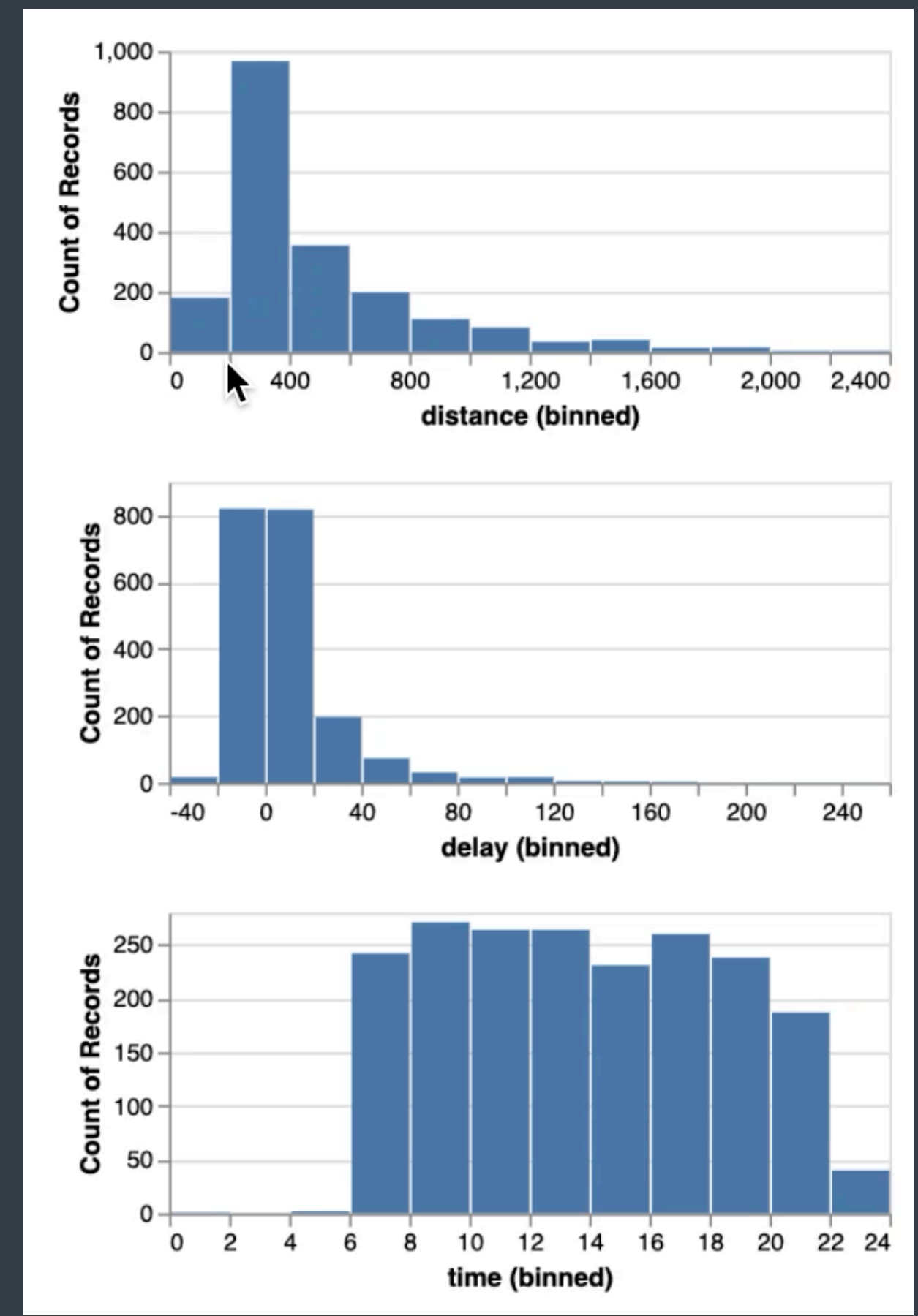
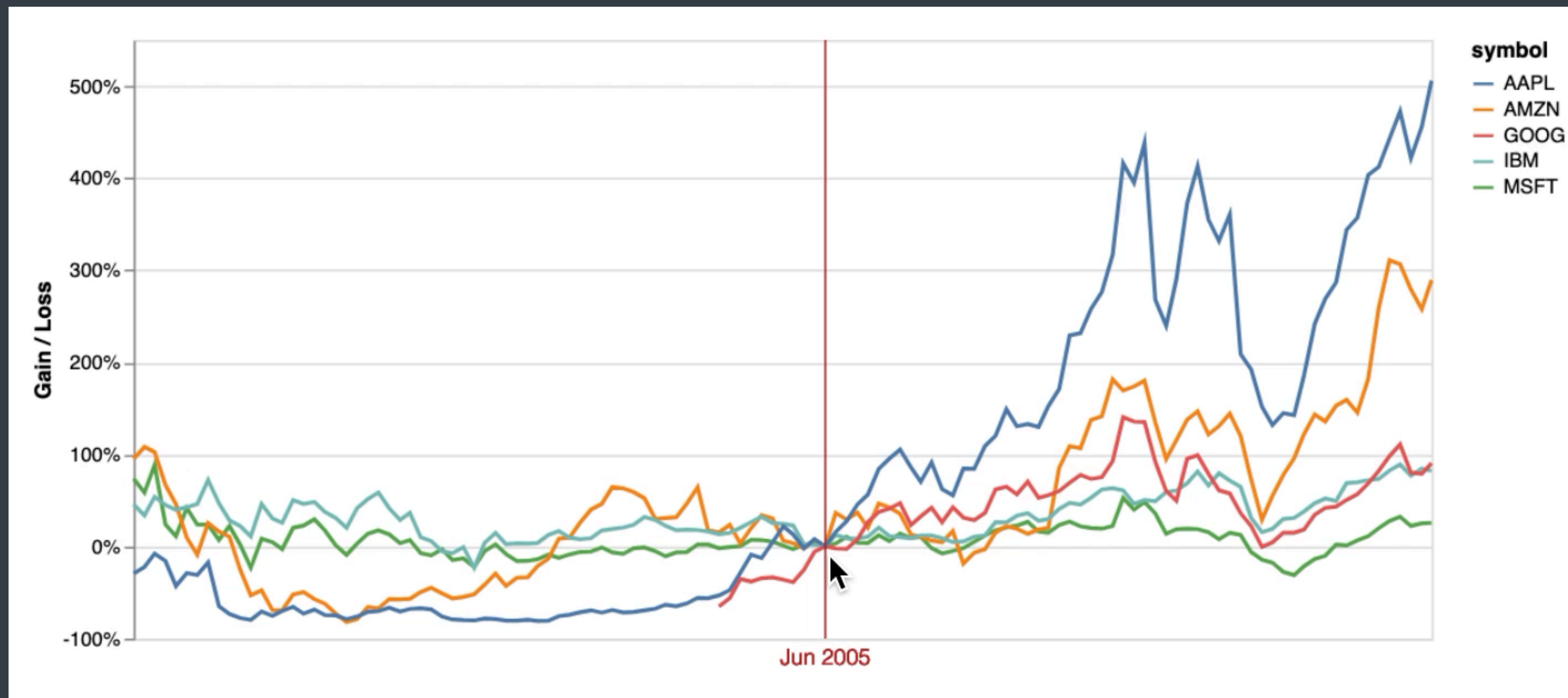
Interactive Aggregation



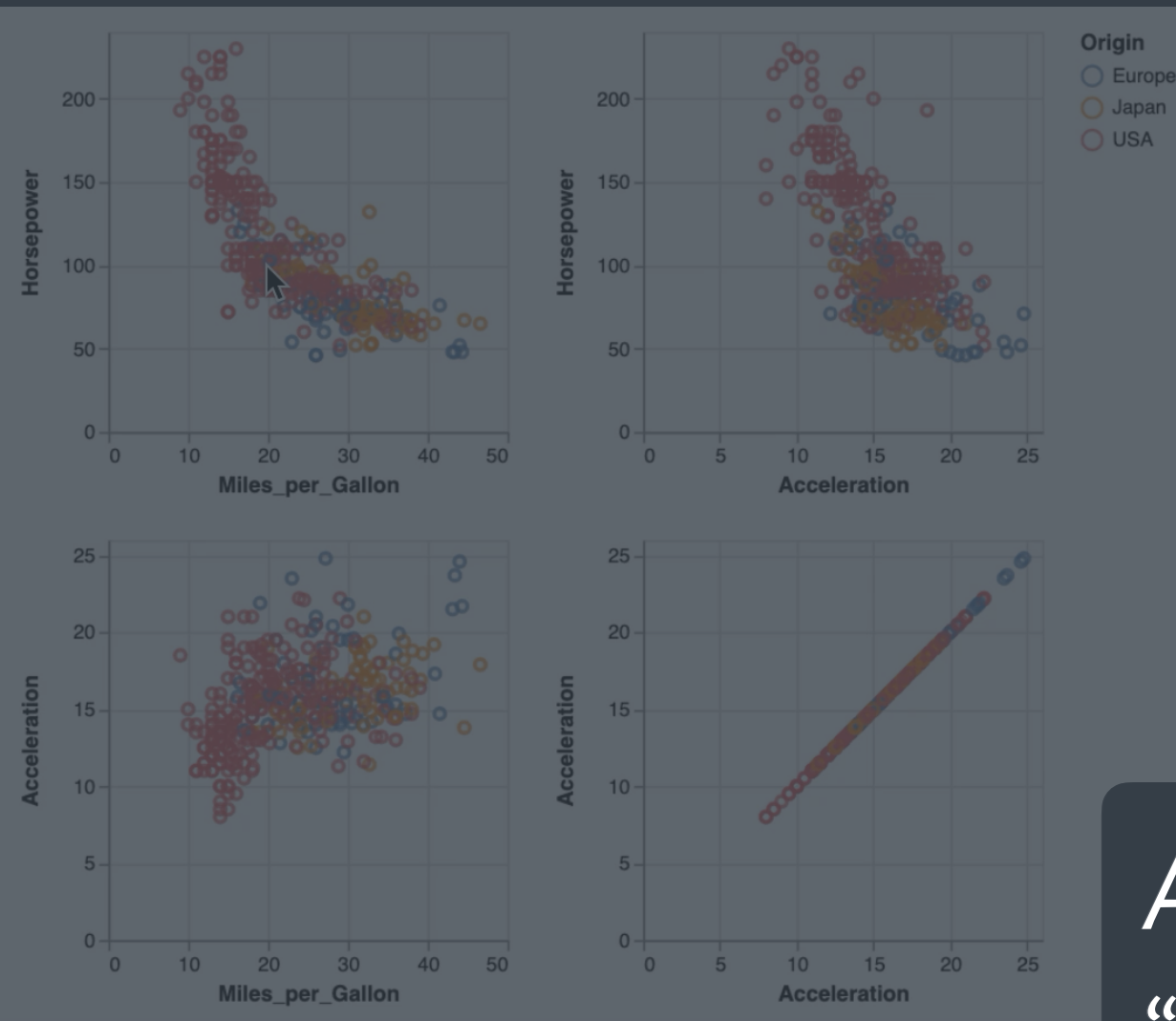
Overview+Detail Interactive Binning



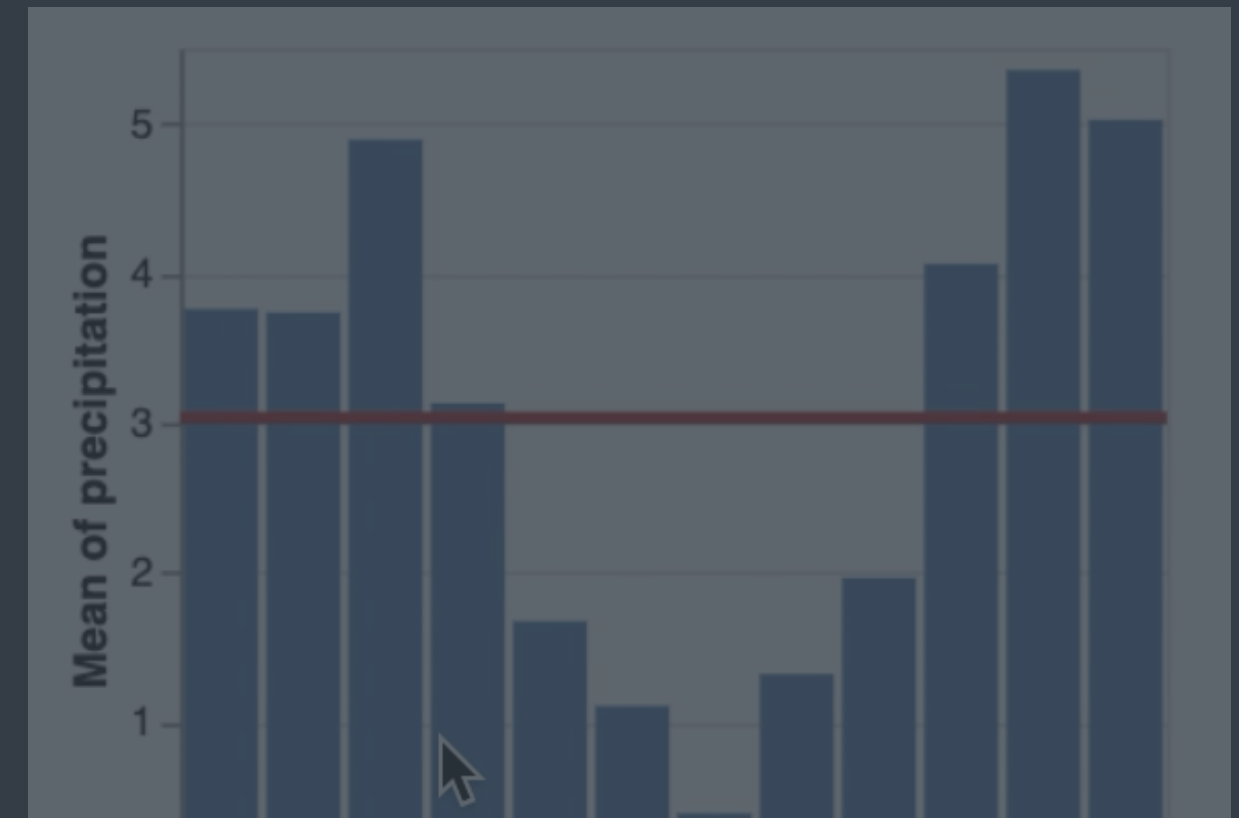
Interactive Re-Normalization



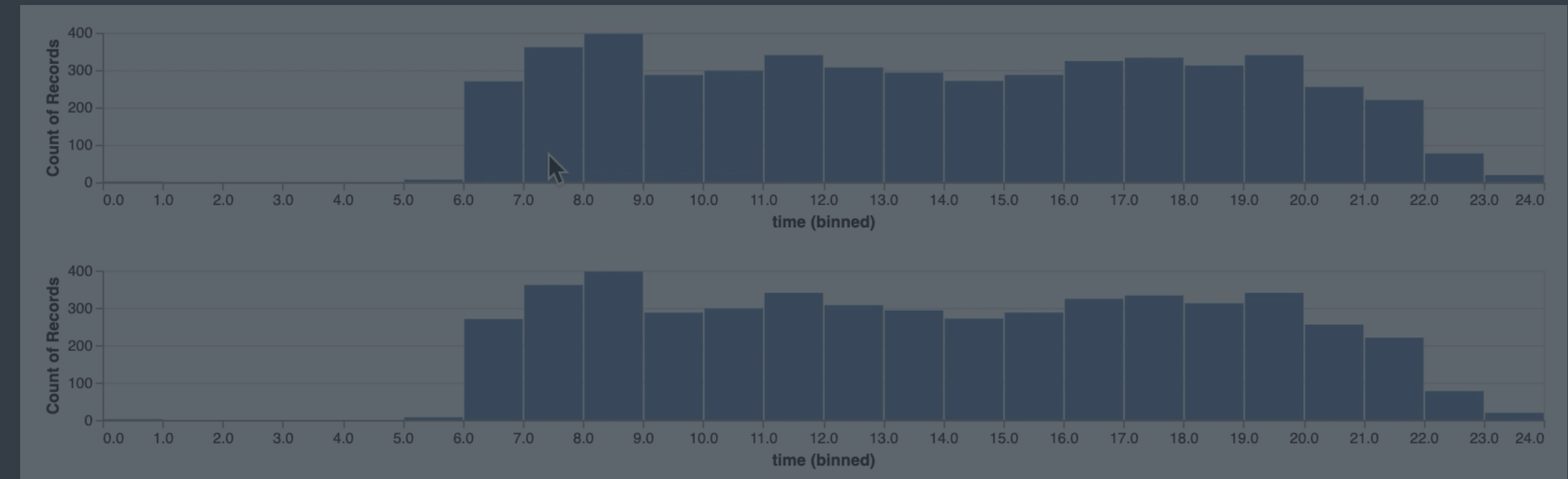
Union Multiple Brushes



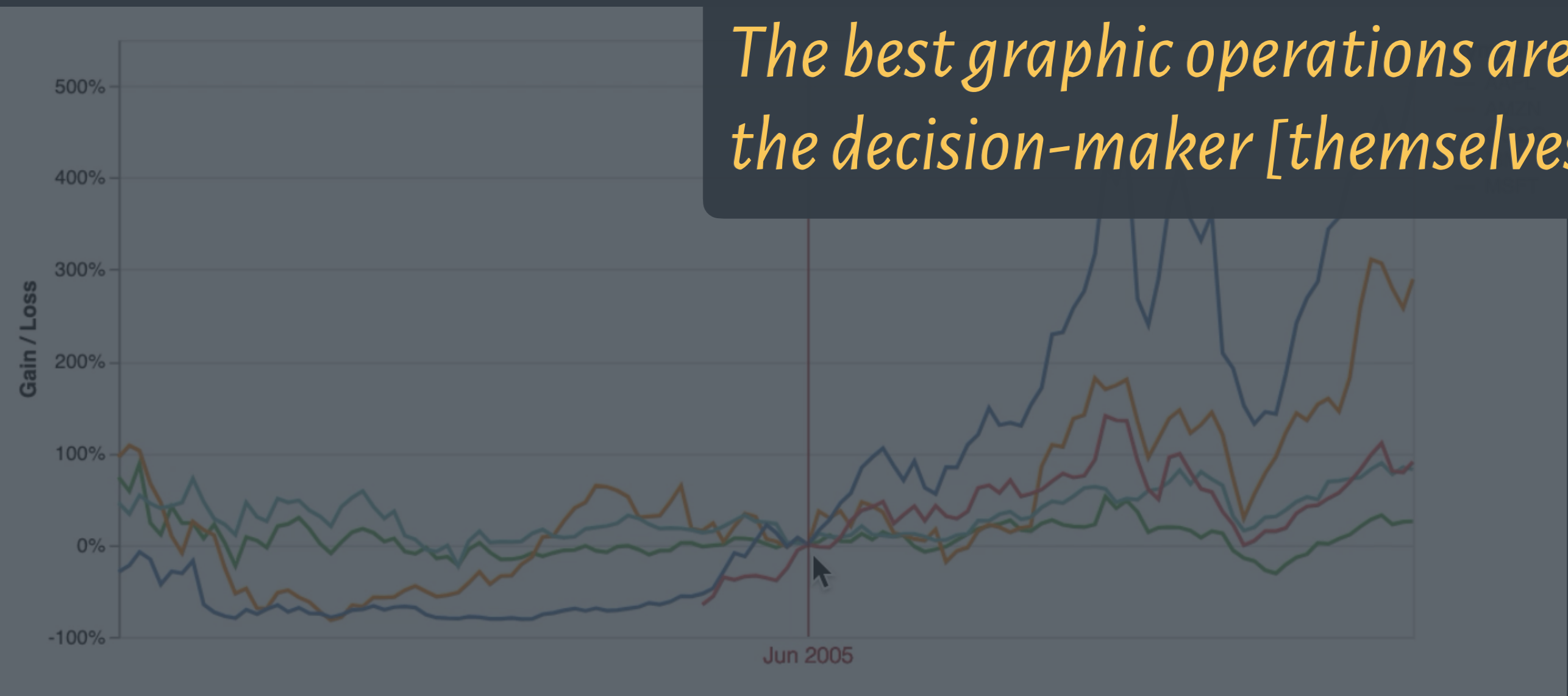
Interactive Aggregation



Overview+Detail Interactive Binning



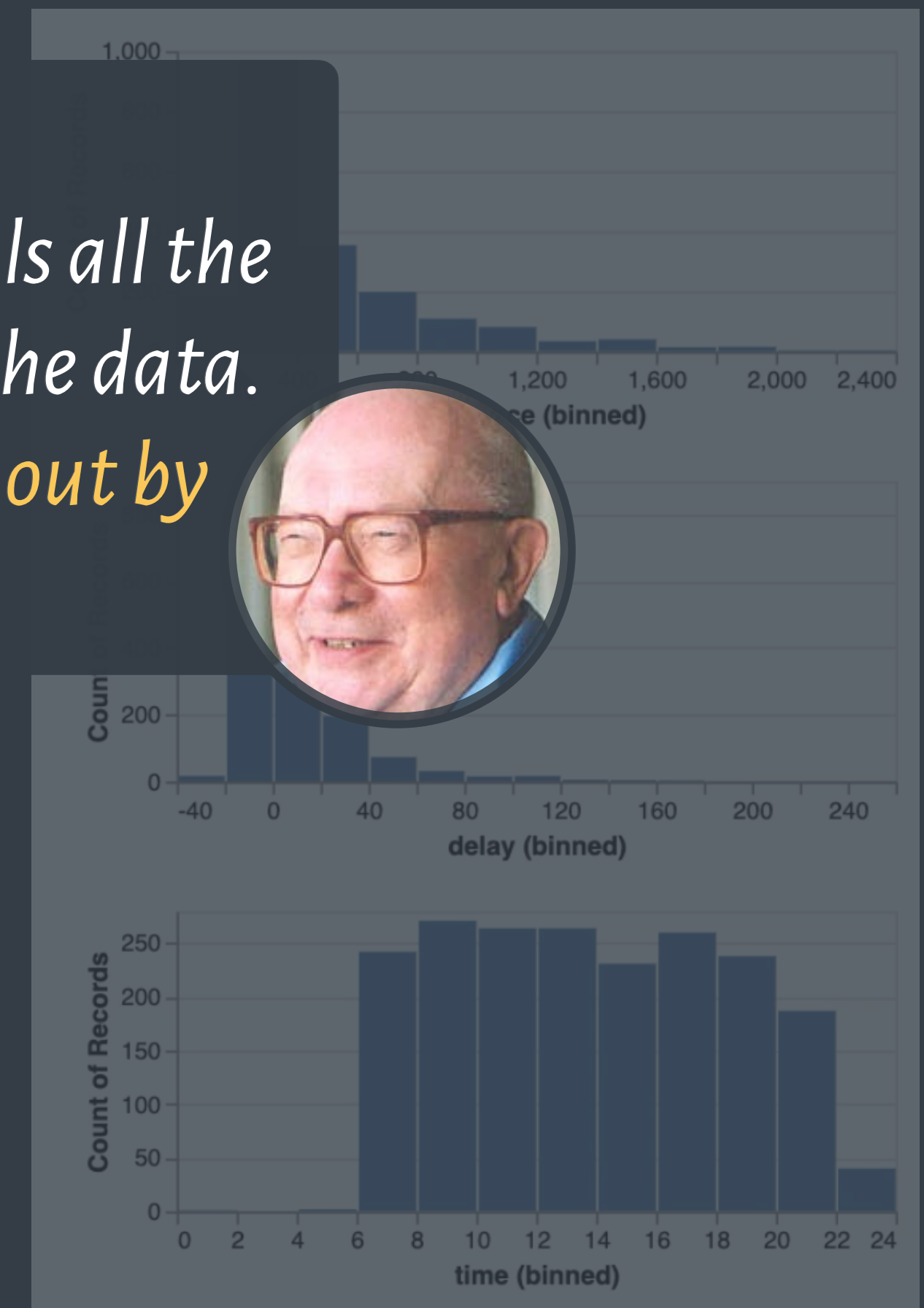
Interactive F



A graphic is not “drawn” once and for all; it is “constructed” and reconstructed until it reveals all the relationships constituted by the interplay of the data. The best graphic operations are those carried out by the decision-maker [themselves].



Interactive Cr



A *Layout* Gap

Visualizations interleaved with code makes it difficult to coordinate multiple views.

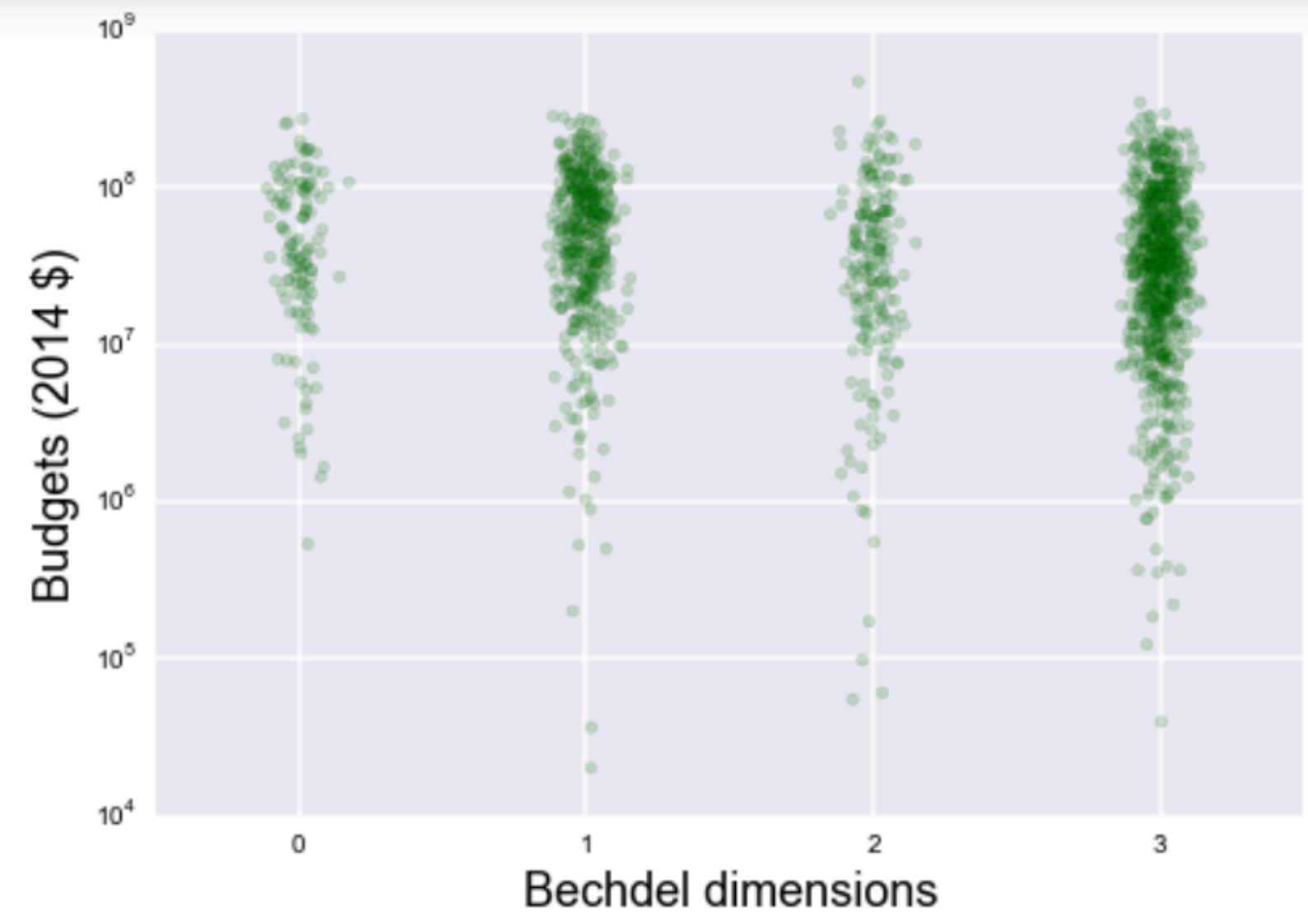
A *Semantic* Gap

Visualizations need to be manually specified despite the provenance expressed by dataframe transformations.

Interactive results are siloed, and not available for further analysis in code.

A *Temporal* Gap

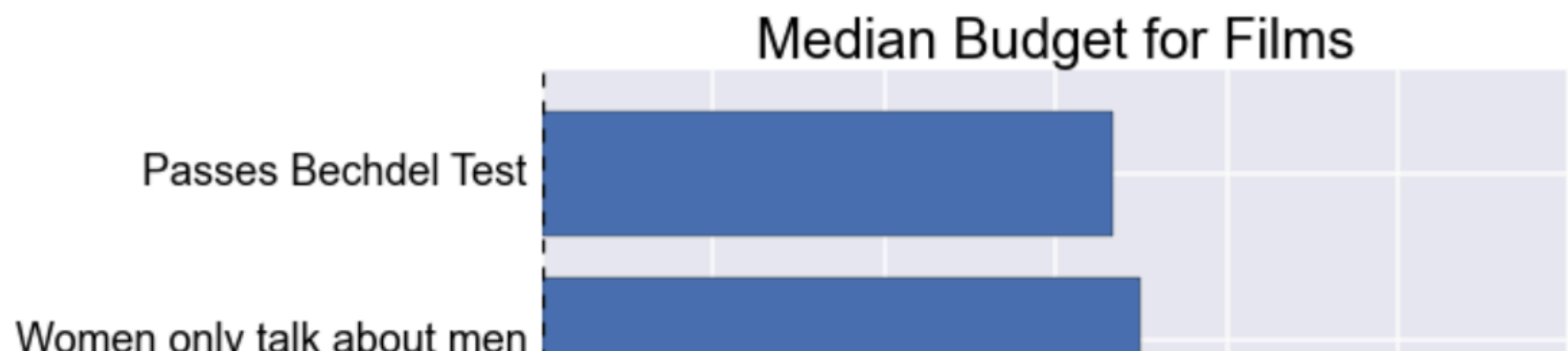
Cell execution is persistent, but interaction is transient.



We can also visualize the median budgets with a bar chart to see that movies featuring two women who don't talk to each other appear to have much larger budgets than the rest. Movies that pass the Bechdel test also appear to have slightly smaller budgets than movies that don't pass.

```
In [43]: df['Adj_Budget'].groupby(df['rating']).agg(np.median).plot(kind='barh')
plt.xticks(arange(0e7,7e7,1e7),arange(0,70,10),fontsize=15)
plt.yticks(plt.yticks()[0],["Fewer than two women",
                             "Women don't talk to each other",
                             "Women only talk about men",
                             "Passes Bechdel Test"],
           fontsize=18)
plt.xlabel('Budget (2014 $millions)',fontsize=18)
plt.title('Median Budget for Films',fontsize=24)
plt.ylabel('')
```

Out[43]: <matplotlib.text.Text at 0x10c9fa490>



B2: Bridging Code & Interactive Visualization in Notebooks

with



The screenshot displays the Jupyter Notebook interface. At the top, the header includes the Jupyter logo, the text "jupyter Untitled2", and a status message "Last Checkpoint: a few seconds ago (unsaved changes)". On the right side of the header, there is a Python logo and a "Logout" button. Below the header is a menu bar with options: File, Edit, View, Insert, Cell, Kernel, Widgets, and Help. To the right of the menu bar, it shows "Trusted" and "Python 3". Below the menu bar is a toolbar with icons for saving, adding a new cell, deleting a cell, copying, pasting, undo, redo, and running code. A dropdown menu is currently set to "Code". The main area of the notebook contains a single code cell with the prompt "In []:" and a cursor pointing to the start of the cell.

B2: Bridging Code & Interactive Visualization in Notebooks

with



The screenshot shows a Jupyter Notebook interface. At the top, the Jupyter logo and 'Untitled2' are visible, along with a 'Logout' button and a Python 3 kernel indicator. A menu bar includes File, Edit, View, Insert, Cell, Kernel, Widgets, and Help. Below the menu bar is a toolbar with icons for file operations and execution. The main area contains two code cells. The first cell has the following code:

```
In [1]: from midas import B2
        # init the env
        b2 = B2()
```

The second cell is empty. To the right of the code cells is a sidebar with a light blue background containing the following text:

Load Data

To load data, use `.from_file("path/to/your_data.csv")`, the columns will show up to the right.

Load Charts

For any dataframes, you can visualize it here with `.vis()`.

Making Interactions

All the loaded charts by default are augmented with interactivity by default. You can select a subset of the data by shift click with a bar chart, or shift-drag to draw the brush, with a scatter plot or line chart.

Recording/Restoring Interactions

By default, your interactions are executed via a "log" in a code cell to the left. You can look at the code to get a sense of what you have interacted with. You can also execute the code by uncommenting the relevant selections.

Toggle and Resize Panes

To toggle this pane (restoring the traditional notebook view), click on `Toggle midas`, from the menu bar on the top. Similarly, you can click `Toggle Column Shelf` to just hide the pane to the right.

To resize, you can also drag the left edge of the the main (blue)pane---the resizer will be highlighted with a darker shade of blue when you hover over.

Column Pane

Columns of loaded data will show here. You can click on the x symbol to the right to hide the column from view.

See Distributions

You can click on the column to get the distribution of the data.

B2: Bridging Code & Interactive Visualization in Notebooks

with



jupyter Untitled2 Last Checkpoint: a few seconds ago (unsaved changes) Python 3

File Edit View Insert Cell Kernel Widgets Help Trusted | Python 3

Toggle Toggle B2 charts Toggle Columns

```
# init the env
b2 = B2()

In [2]: df = b2.from_file("./data/fire.csv")
df

Out[2]:
```

SourceType	Name	Year	DiscoveryDate	DiscoveryTime	Caus
NONFED	nan	1992	2.44865e+06	1540	Arso
NONFED	nan	1998	2.45091e+06	nan	Debris Burnin
NONFED	nan	1995	2.44983e+06	nan	Childre
FED	MINE	2011	2.45575e+06	1800	Lightnin
NONFED	2013039	2013	2.45648e+06	nan	Equipment Us
NONFED	nan	1994	2.44944e+06	1030	Missing/Undefine
NONFED	nan	2015	2.45706e+06	1352	Debris Burnin
NONFED	nan	2000	2.4517e+06	nan	Equipment Us
FED	WAVES AGAIN	2011	2.45575e+06	2144	Campfir
NONFED	MERCED COUNTY	1993	2.44912e+06	nan	Equipment Us

... (9990 rows omitted)

```
In [4]: Year_dist = df.groupby('Year')
Year_dist.vis()
```

In []:

df

- SourceType
- Name
- Year
- DiscoveryDate
- DiscoveryTime
- Cause
- ContDate
- ContDoy
- ContTime
- Size
- SizeClass
- Lat
- Lon
- Owner
- State
- County

B2: Bridging Code & Interactive Visualization in Notebooks

with



jupyter Untitled2 Last Checkpoint: a minute ago (unsaved changes) Python 3

File Edit View Insert Cell Kernel Widgets Help Trusted | Python 3

Toggle charts Toggle Columns

NONFED	nan	1992	2.44865e+06	1540	Arso
NONFED	nan	1998	2.45091e+06	nan	Debris Burnin
NONFED	nan	1995	2.44983e+06	nan	Childre
FED	MINE	2011	2.45575e+06	1800	Lightnin
NONFED	2013039	2013	2.45648e+06	nan	Equipment Us
NONFED	nan	1994	2.44944e+06	1030	Missing/Undefine
NONFED	nan	2015	2.45706e+06	1352	Debris Burnin
NONFED	nan	2000	2.4517e+06	nan	Equipment Us
FED	WAVES AGAIN	2011	2.45575e+06	2144	Campfir
NONFED	MERCED COUNTY	1993	2.44912e+06	nan	Equipment Us

... (9990 rows omitted)

```
In [4]: Year_dist = df.group('Year')
Year_dist.vis()
```

```
In [7]: # 12:54 PM
State_df_dist = df.group('State')
State_df_dist.vis()
```

```
In [8]: # 12:54 PM
Cause_df_dist = df.group('Cause')
Cause_df_dist.vis()
```

```
In [ ]:
```

Year_dist

State_df_dist

Cause_df_dist

df

- SourceType
- Name
- Year
- DiscoveryDate
- DiscoveryTime
- Cause
- ContDate
- ContDoy
- ContTime
- Size
- SizeClass
- Lat
- Lon
- Owner
- State
- County

B2: Bridging Code & Interactive Visualization in Notebooks

with



jupyter Untitled2 Last Checkpoint: a minute ago (unsaved changes) Python 3

File Edit View Insert Cell Kernel Widgets Help Trusted Python 3

Toggle charts Toggle Columns

NONFED	nan	1992	2.44865e+06	1540	Arso
NONFED	nan	1998	2.45091e+06	nan	Debris Burnin
NONFED	nan	1995	2.44983e+06	nan	Childre
FED	MINE	2011	2.45575e+06	1800	Lightnin
NONFED	2013039	2013	2.45648e+06	nan	Equipment Us
NONFED	nan	1994	2.44944e+06	1030	Missing/Undefine
NONFED	nan	2015	2.45706e+06	1352	Debris Burnin
NONFED	nan	2000	2.4517e+06	nan	Equipment Us
FED	WAVES AGAIN	2011	2.45575e+06	2144	Campfir
NONFED	MERCED COUNTY	1993	2.44912e+06	nan	Equipment Us

... (9990 rows omitted)

```
In [4]: Year_dist = df.group('Year')
Year_dist.vis()
```

```
In [7]: # 12:54 PM
State_df_dist = df.group('State')
State_df_dist.vis()
```

```
In [8]: # 12:54 PM
Cause_df_dist = df.group('Cause')
Cause_df_dist.vis()
```

```
In [15]: # 12:55 PM
b2.sel({'State_df_dist': {'State': ['NC', 'TX', 'GA']}})
b2.sel({'State_df_dist': {'State': ['NC', 'TX', 'GA', '']}})
```

Year_dist

State_df_dist

Cause_df_dist

df

- SourceType
- Name
- Year
- DiscoveryDate
- DiscoveryTime
- Cause
- ContDate
- ContDoy
- ContTime
- Size
- SizeClass
- Lat
- Lon
- Owner
- State
- County

B2: Bridging Code & Interactive Visualization in Notebooks

with



jupyter Untitled2 Last Checkpoint: a minute ago (unsaved changes) Python 3

File Edit View Insert Cell Kernel Widgets Help Trusted | Python 3

NONFED	2013039	2013	2.45648e+06	nan	Equipment Us
NONFED	nan	1994	2.44944e+06	1030	Missing/Undefined
NONFED	nan	2015	2.45706e+06	1352	Debris Burnin
NONFED	nan	2000	2.4517e+06	nan	Equipment Us
FED	WAVES AGAIN	2011	2.45575e+06	2144	Campfir
NONFED	MERCED COUNTY	1993	2.44912e+06	nan	Equipment Us

... (9990 rows omitted)

```
In [4]: Year_dist = df.group('Year')
Year_dist.vis()
```

```
In [7]: # 12:54 PM
State_df_dist = df.group('State')
State_df_dist.vis()
```

```
In [8]: # 12:54 PM
Cause_df_dist = df.group('Cause')
Cause_df_dist.vis()
```

```
In [18]: # 12:55 PM
# b2.sel([{"State_df_dist": {"State": ["CA"]}}])
# b2.sel([{"State_df_dist": {"State": ["GA"]}}])
# b2.sel([{"State_df_dist": {"State": ["TX"]}}])
# b2.sel([{"State_df_dist": {"State": ["NC"]}}])
# b2.sel([{"State_df_dist": {"State": ["NC", "TX"]}}])
# b2.sel([{"State_df_dist": {"State": ["NC", "TX", "GA"]}}])
b2.sel([{"State_df_dist": {"State": ["NC", "TX", "GA", "
```

Year_dist

State_df_dist

Cause_df_dist

df

- Source Type
- Name
- Year
- Discovery Date
- Discovery Time
- Cause
- Cont Date
- Cont Doy
- Cont Time
- Size
- Size Class
- Lat
- Lon
- Owner
- State
- County

B2: Bridging Code & Interactive Visualization in Notebooks

with



jupyter Untitled2 Last Checkpoint: 17 minutes ago (autosaved) Python 3 Logout

File Edit View Insert Cell Kernel Widgets Help

Trusted | Python 3

Toggle charts Toggle Columns

NONFED	nan	2000	2.4517e+06	nan	Equipment Us
FED	WAVES AGAIN	2011	2.45575e+06	2144	Campfir
NONFED	MERCED COUNTY	1993	2.44912e+06	nan	Equipment Us

... (9990 rows omitted)

```
In [4]: Year_dist = df.groupby('Year')
Year_dist.vis()
```

```
In [7]: # 12:54 PM
State_df_dist = df.groupby('State')
State_df_dist.vis()
```

```
In [8]: # 12:54 PM
Cause_df_dist = df.groupby('Cause')
Cause_df_dist.vis()
```

```
In [35]: # 12:55 PM
# b2.sel({"State_df_dist": {"State": ["CA"]}})
# b2.sel({"State_df_dist": {"State": ["GA"]}})
# b2.sel({"State_df_dist": {"State": ["TX"]}})
# b2.sel({"State_df_dist": {"State": ["NC"]}})
# b2.sel({"State_df_dist": {"State": ["NC", "TX"]}})
# b2.sel({"State_df_dist": {"State": ["NC", "TX", "GA"]}})
b2.sel({"State_df_dist": {"State": ["NC", "TX", "GA", "
```

```
In [37]: dist": {"Year": [1998, 1999, 2000, 2001, 2002, 2003]}}})
st": {"Year": [2004, 2005, 2006, 2007, 2008, 2009]}}})
```

```
In [ ]:
```

Year_dist

State_df_dist

Cause_df_dist

df

- SourceType
- Name
- Year
- DiscoveryDate
- DiscoveryTime
- Cause
- ContDate
- ContDoy
- ContTime
- Size
- SizeClass
- Lat
- Lon
- Owner
- State
- County

B2: Bridging Code & Interactive Visualization in Notebooks

with



jupyter Untitled2 Last Checkpoint: 22 minutes ago (unsaved changes) Python 3 Logout

File Edit View Insert Cell Kernel Widgets Help

Trusted | Python 3

Toggle charts Toggle Columns

Children	22
Debris Burning	286
Equipment Use	143
Fireworks	2
Lightning	56
Miscellaneous	219
Missing/Undefined	77
Powerline	2

... (2 rows omitted)

```
In [40]: locs = df.where('State', b2.are.contained_in(['NC', 'TX']))
         locs.plot_heatmap(zoom_start=3, radius=6)
```

Out[40]:

Leaflet | Data by © OpenStreetMap, under ODbL.

```
In [42]: # 01:16 PM
         # b2.sel({"State_df_dist": {"State": ["NC", "TX", "GA"],
         # b2.sel({"State_df_dist": {"State": ["NC", "TX", "GA", "
```

Year_dist

Year	count
1992	400
1993	300
1994	400
1995	350
1996	400
1997	350
1998	350
1999	500
2000	500
2001	450
2002	400
2003	350
2004	350
2005	450
2006	600
2007	550
2008	450
2009	400
2010	450
2011	500
2012	400
2013	350
2014	350
2015	400

State_df_dist

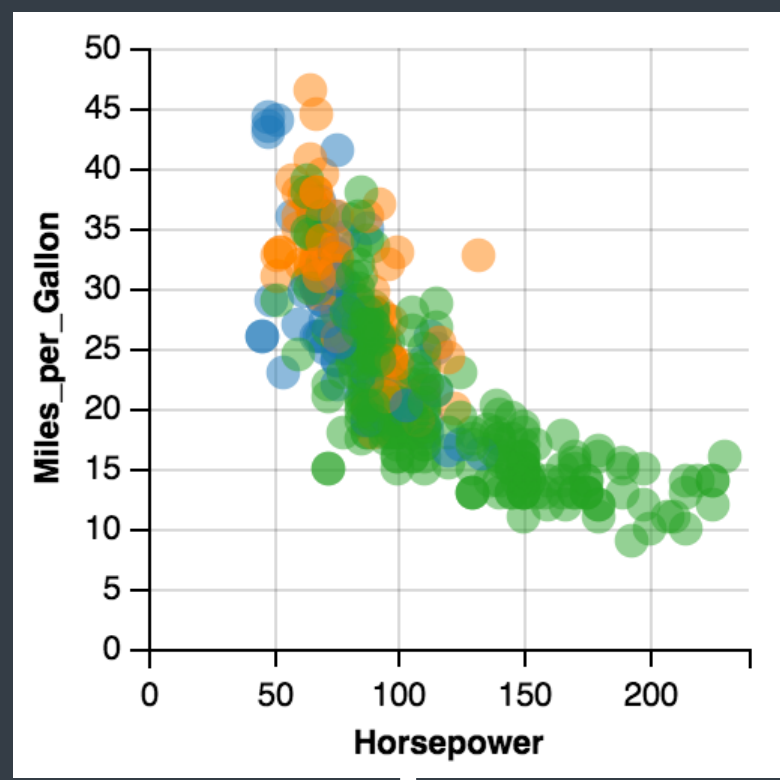
State	count
CA	100
GA	900
TX	750
NC	550
FL	450
NY	450
SC	400
MS	400
AZ	400
AL	350
OR	350
OK	250
MT	250
MN	250
ID	250
NM	250
WA	250
UT	250
CO	250
SD	250
TN	250
WI	250
LA	250
AR	250
NJ	250
KY	250
VA	250

Cause_df_dist

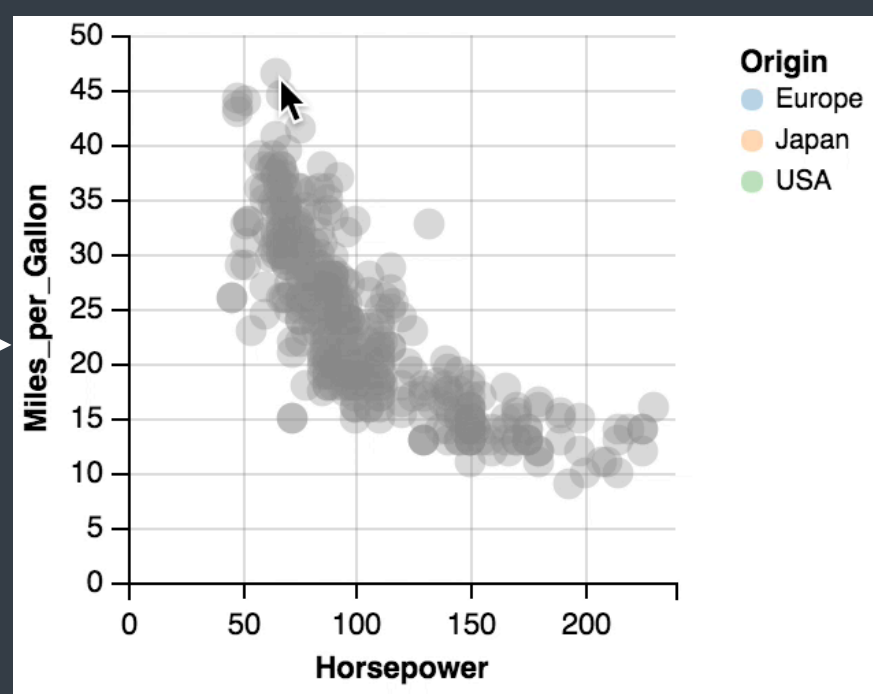
Cause	count
Debris Burning	2200
Miscellaneous	1800
Lightning	1500
Arson	1500
Missing/Undefined	1000
Equipment Use	800
Campfire	400
Children	300
Smoking	300
Railroad	200
Powerline	100
Fireworks	100
Structure	100

df

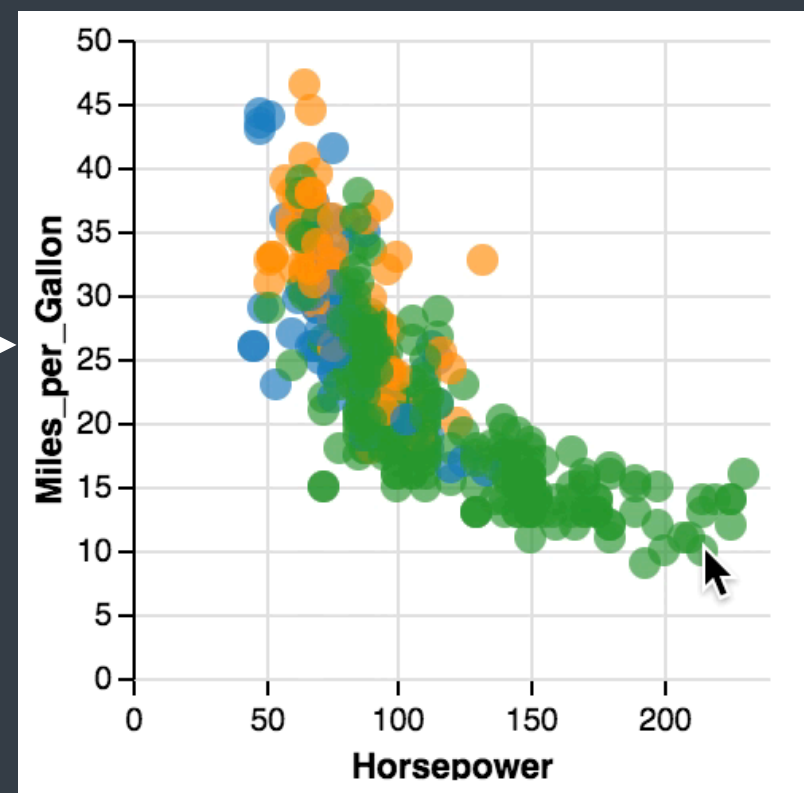
- SourceType
- Name
- Year
- DiscoveryDate
- DiscoveryTime
- Cause
- ContDate
- ContDoy
- ContTime
- Size
- SizeClass
- Lat
- Lon
- Owner
- State
- County



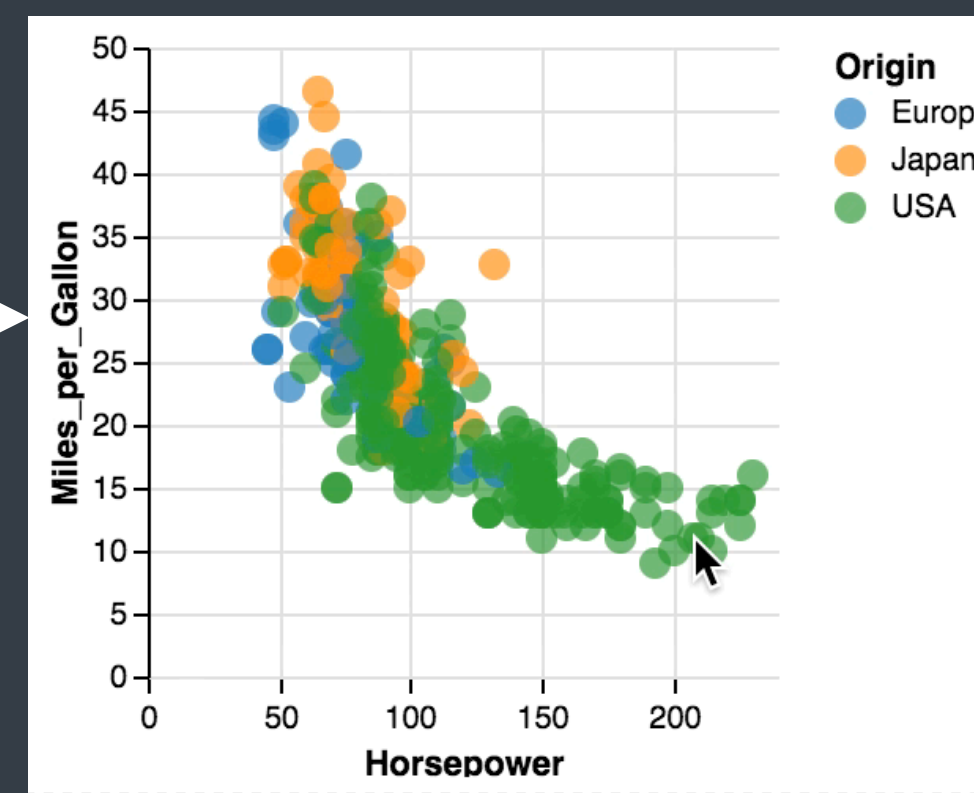
Single



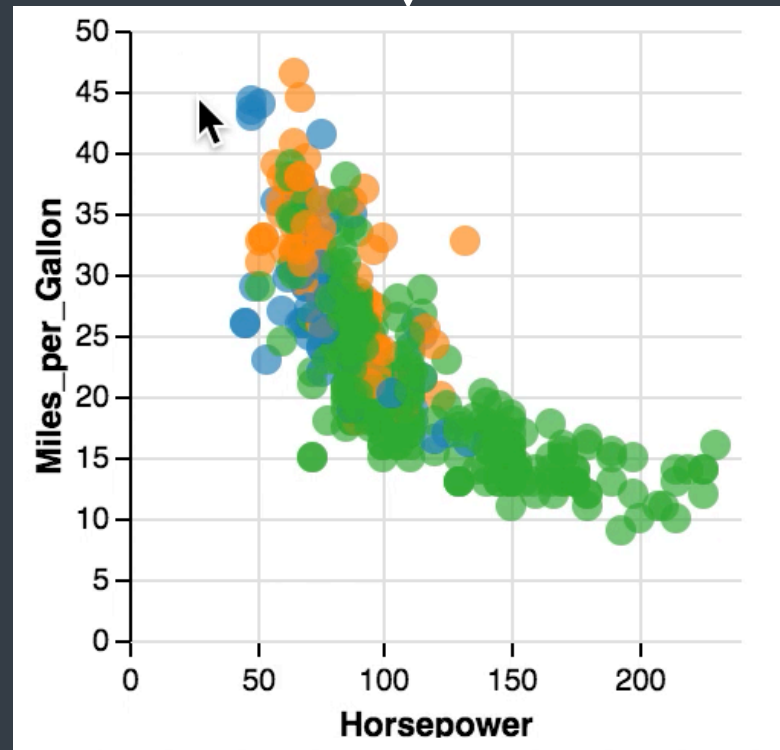
Project



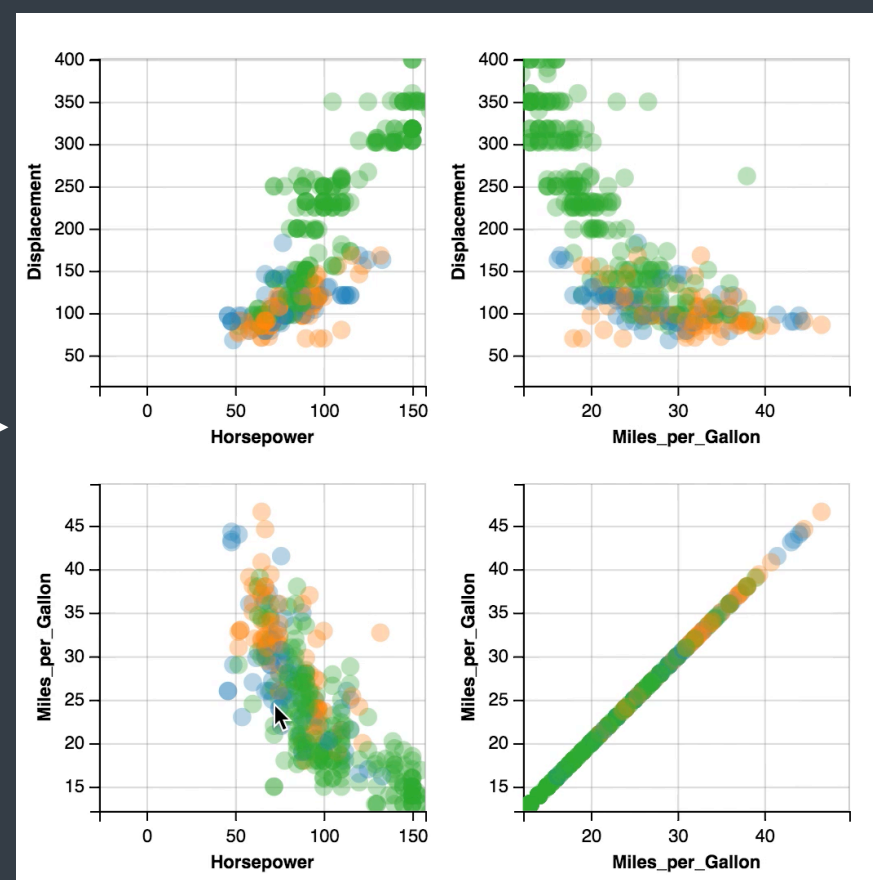
Bind



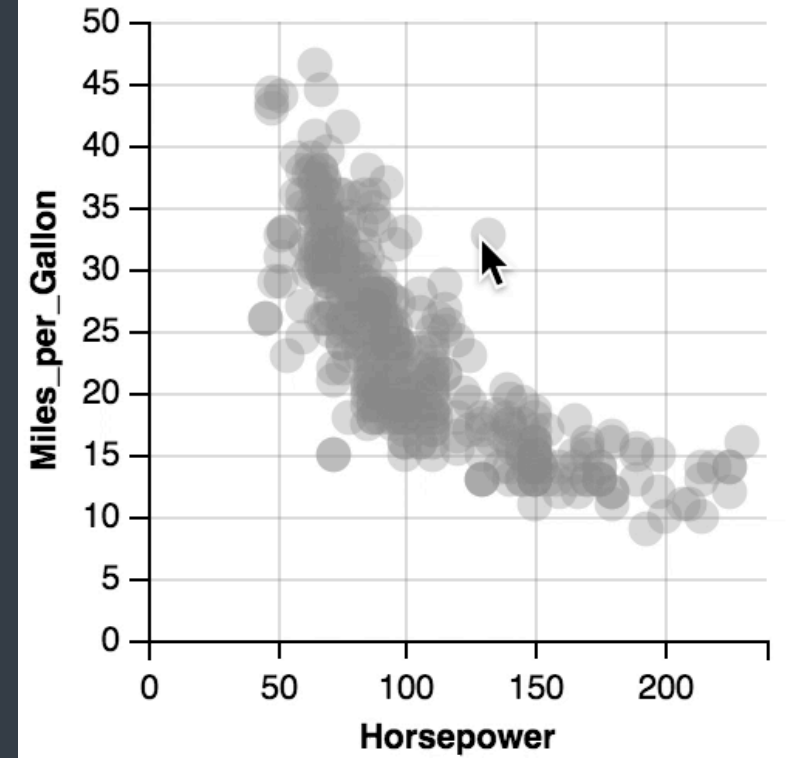
Interval



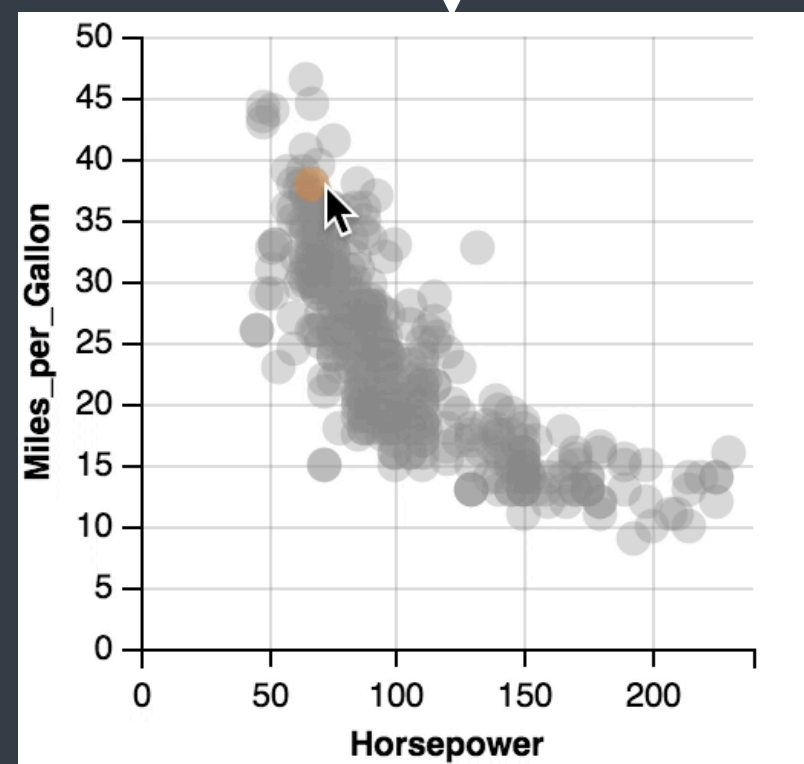
Repeat



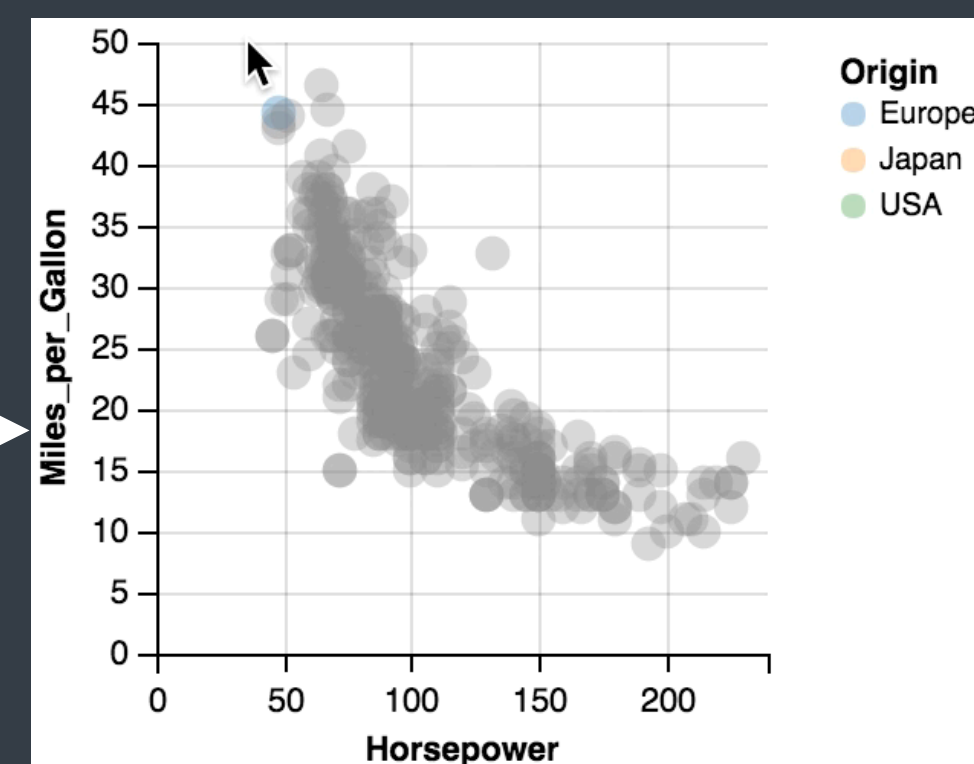
Multi



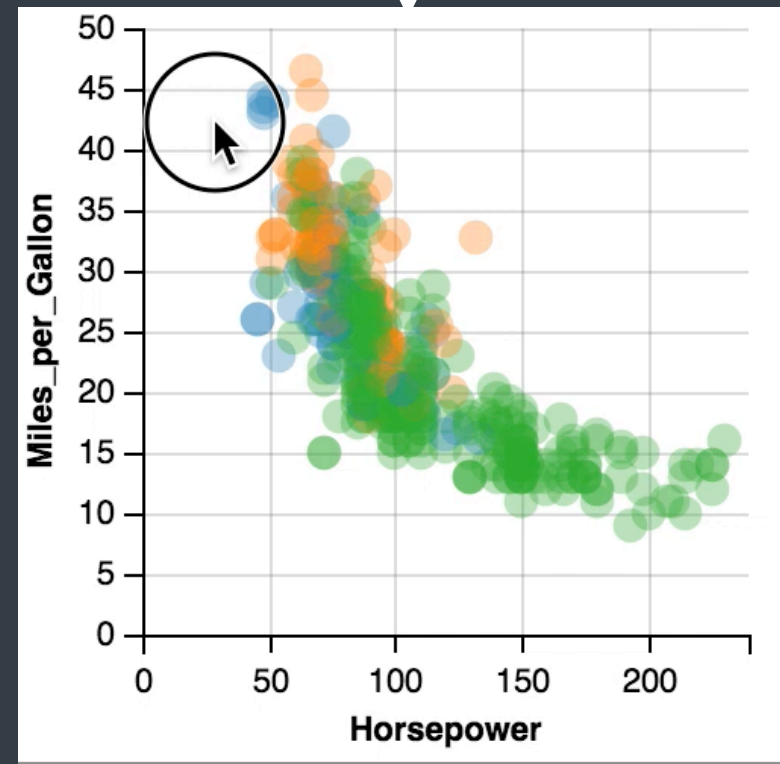
Hover



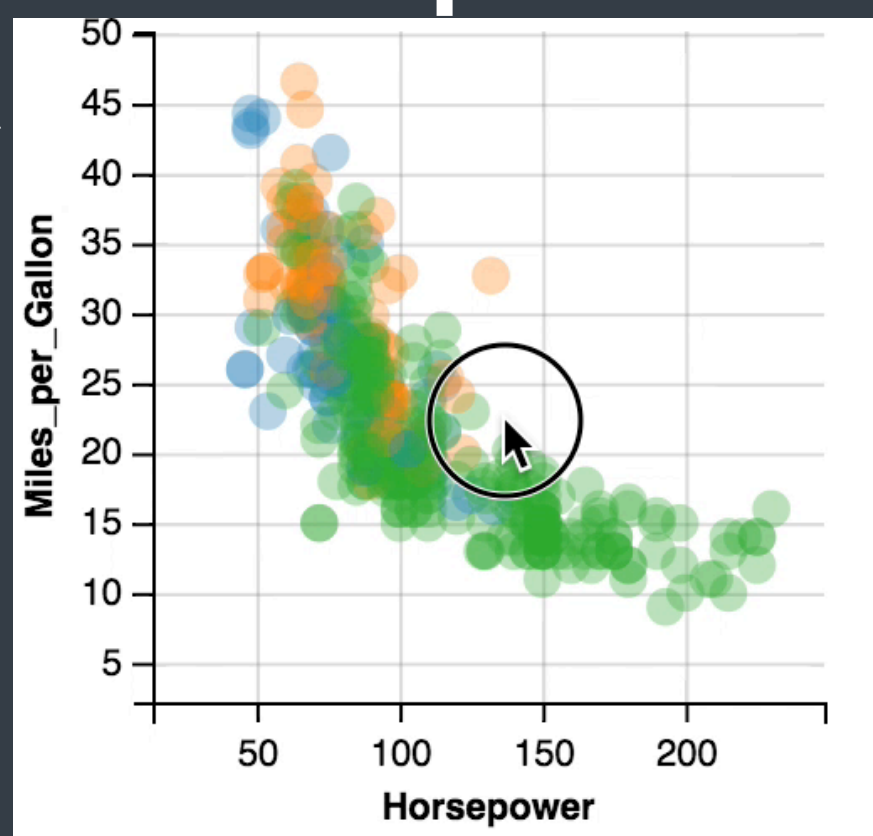
Nearest



Project



Bind



Repeat

Systematically Enumerate the Design Space

Lyra: Interactive Visualization by Demonstration

with



GROUPS +

GROUP 1

PIPELINE None

X POSITION

Left 0

Set to group width

Width 640

Y POSITION

Top 0

Set to group height

Height 360

FILL

Color transparent

Opacity 1

STROKE

Color

Width 0

SCALES

DATA PIPELINES World + New

Birth_Rate	0.011	0.019	0.019	Region	Africa	Africa	Africa
Life_Exp...	77	78	77	Infant_M...	0.013	0.012	0.014
Life_Exp...	70	68	73	Year	135433...	135433...	135433...
Country/...	Mauritius	Seychelles	Tunisia	Life_Exp...	74	73	75
GDP_per...	8862	11689	4197				

1-20 of 207

RECT

SYMBOL

TEXT

LINE

AREA

EXPORT

Lyra: Interactive Visualization by Demonstration

with



GROUPS

Edit Scene

Group 1

GUIDES

MARKS

Symbol 1

INTERACTIONS

WIDGETS

SCALES

SYMBOL 1

PIPELINE: None

POSITION

X: 100

Y: 100

GEOMETRY

Shape: circle

Size: 200

FILL

Color: #4682b4

Opacity: 1

STROKE

Color: #000000

Width: 0

DATA PIPELINES

World + New

Birth_Rate	0.011	0.019	0.019	Region	Africa	Africa	Africa
# Life_Expectancy_Female				Infant_M...	0.013	0.012	0.014
Life_Exp...	70	68	73	Year	135433...	135433...	135433...
Country/...	Mauritius	Seychelles	Tunisia	Life_Exp...	74	73	75
GDP per...	8862	11689	4197				
# Life_Expectancy_Female							

RECT

SYMBOL

TEXT

LINE

AREA

EXPORT

Lyra: Interactive Visualization by Demonstration

with



GROUPS

Edit Scene

Group 1

GUIDES

MARKS

Symbol 1

INTERACTIONS

WIDGETS

SCALES

SYMBOL 1

PIPELINE: None

POSITION

X: 100

Y: 100

GEOMETRY

Shape: circle

Size: 200

FILL

Color: #4682b4

Opacity: 1

STROKE

Color: #000000

Width: 0

DATA PIPELINES

World + New

Birth_Rate	0.011	0.019	0.019	Region	Africa	Africa	Africa
# Life_Expectancy_Female				Infant_M...	0.013	0.012	0.014
Life_Exp...	70	68	73	Year	135433...	135433...	135433...
Country/...	Mauritius	Seychelles	Tunisia	Life_Exp...	74	73	75
GDP_per...	8862	11689	4197				

1-20 of 207

Life_Expectancy_Female

RECT

SYMBOL

TEXT

LINE

AREA

EXPORT

Lyra: Interactive Visualization by Demonstration

with



GROUPS +

Edit Scene

▼ Group 1

GUIDES

X Axis

Y Axis

Color Legend

MARKS

Symbol 1

INTERACTIONS +

WIDGETS

SYMBOL 1

PIPELINE World

POSITION

X Life_Exp...

Y Birth_Rate

GEOMETRY

Shape circle

Size 200

FILL

Color color Region

Opacity 1

STROKE

Color #000000

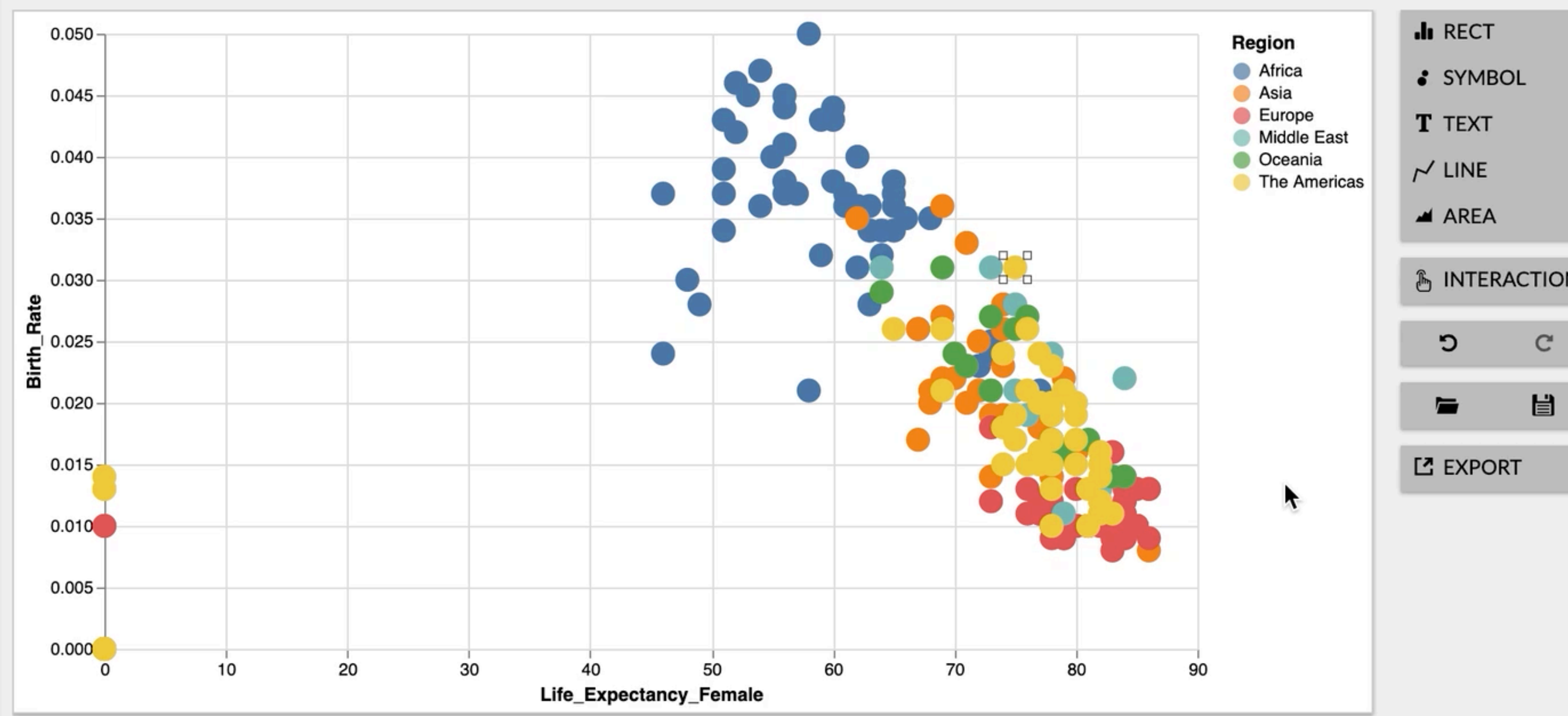
Width 0

SCALES

DATA PIPELINES World + New

Birth_Rate	0.011	0.019	0.019	Region	Africa	Africa	Africa
Life_Exp...	77	78	77	Infant_M...	0.013	0.012	0.014
Life_Exp...	70	68	73	Year	135433...	135433...	135433...
Country/...	Mauritius	Seychelles	Tunisia	Life_Exp...	74	73	75
GDP_per...	8862	11689	4197				

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Lyra: Interactive Visualization by Demonstration

with



GROUPS +

Edit Scene

▼ Group 1

GUIDES

X Axis

Y Axis

Color Legend

MARKS

Symbol 1

INTERACTIONS +

Interaction 1

WIDGETS

SCALES

INTERACTION 1

INPUT EVENT

Mouse: Drag

Keyboard: None

SELECTIONS

Brush

Brush (y-axis)

Brush (x-axis)

APPLICATIONS

Color

Opacity

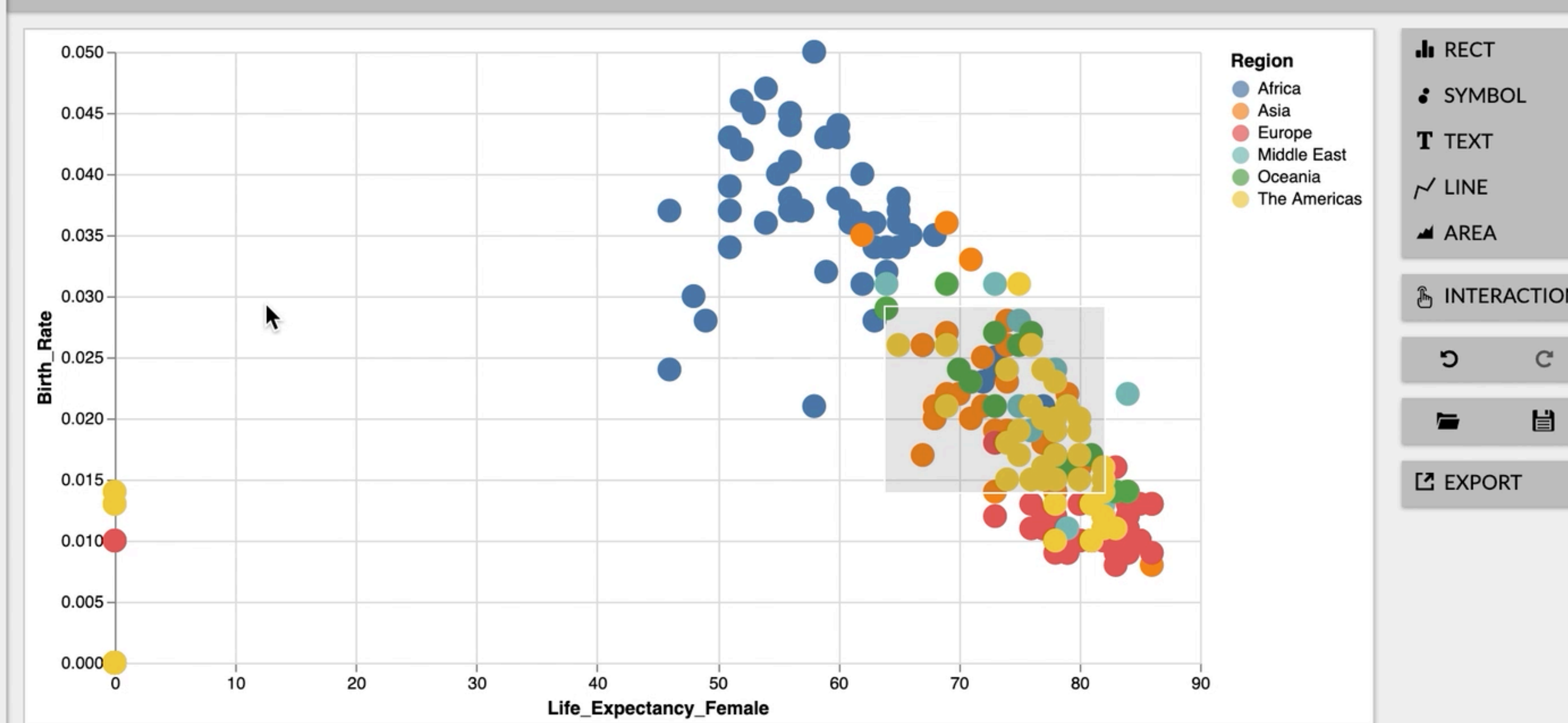
Size

Pan and zoom

DATA PIPELINES World + New

Birth_Rate	0.011	0.019	0.019	Region	Africa	Africa	Africa
Life_Exp...	77	78	77	Infant_M...	0.013	0.012	0.014
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1-20 of 207



Towards *Effective* Interaction With Data Visualization

Empirically Derive Effectiveness "Rankings"

Use an interaction grammar to enumerate alternate techniques for a given analytic task. Test with human subjects.

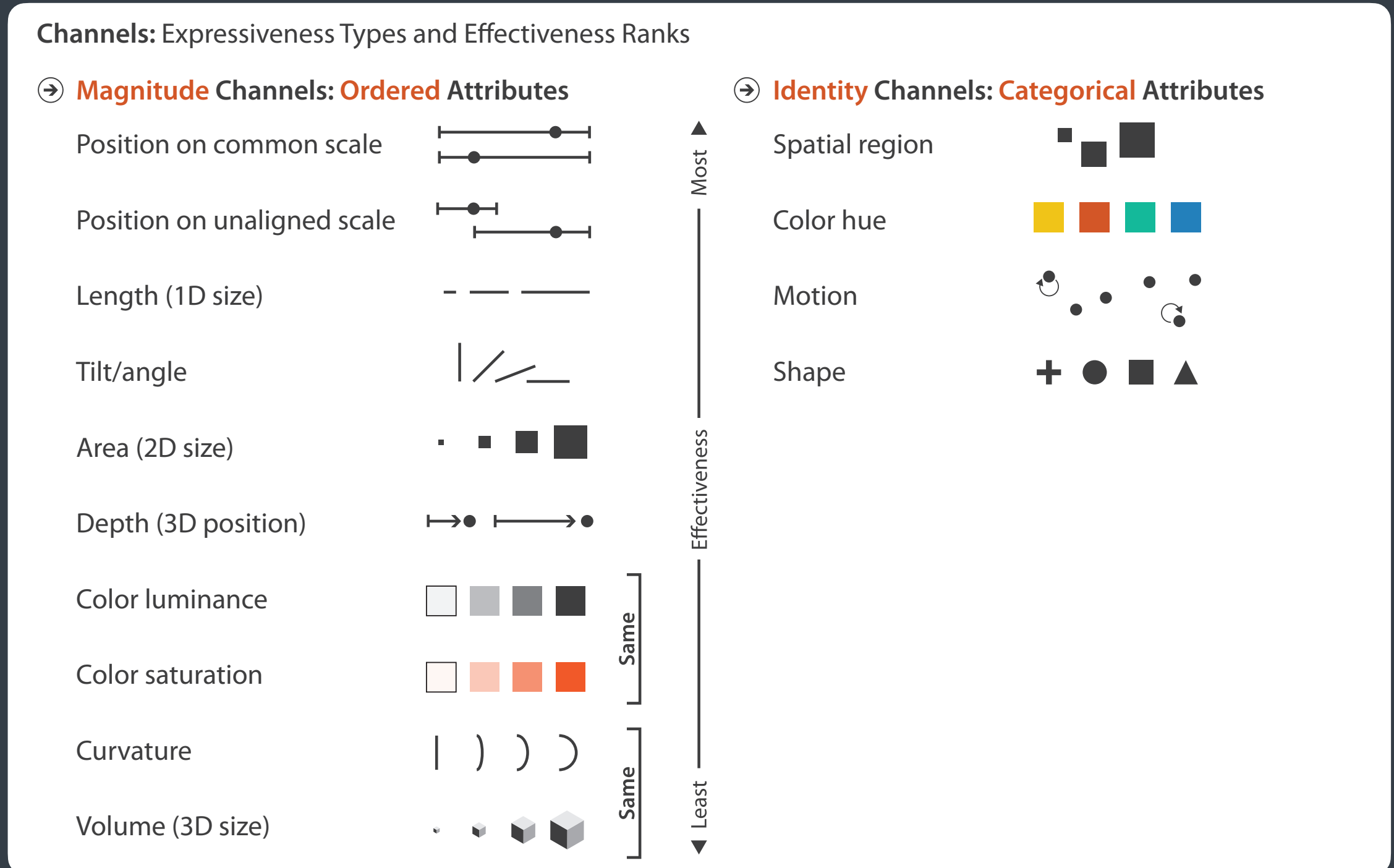
A key challenge: measuring effectiveness beyond time/error?

Interaction Recommender Systems

Synthesize appropriate interactive visualizations using effectiveness rankings.

Mine interaction histories recorded as semantically-meaningful *selections*. Accelerate interactive analysis, suggest unexplored paths.

A key challenge: how to infer user intent from context?



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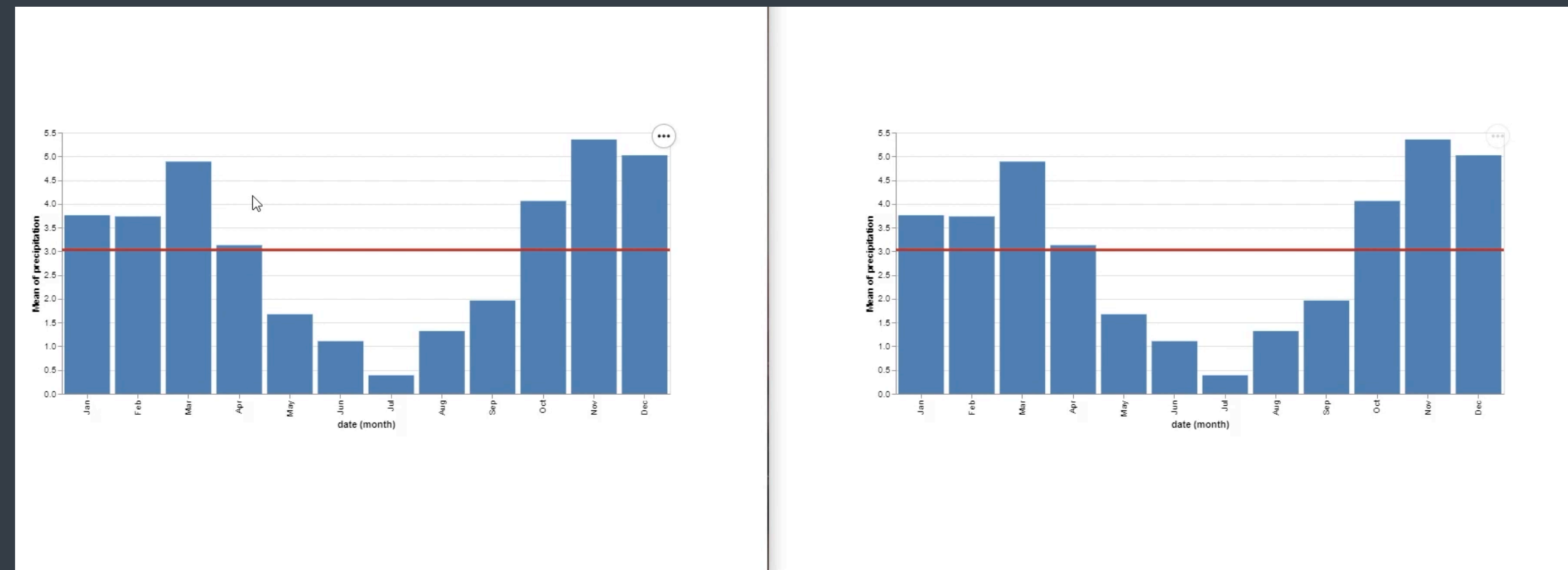
Mine interaction histories recorded as semantically-meaningful *selections*. Accelerate interactive analysis, suggest unexplored paths.

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Supporting (A)Synchronous Collaboration

Establishing common ground and shared awareness.

A key challenge: interactive visualization beyond the desktop?

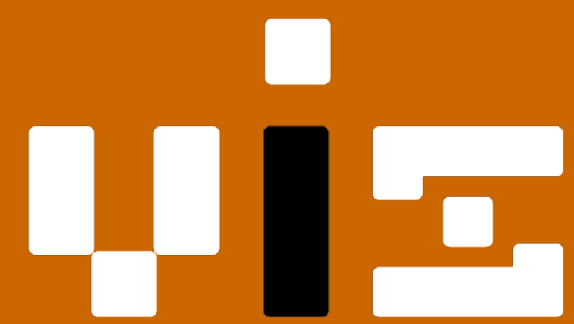


Towards *Effective* Interaction

With Data Visualization

Arvind Satyanarayan
@arvindsatya1

MIT Visualization Group
@mitvis • vis.csail.mit.edu



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