

Topological Equivariant Artist Model

April 18, 2021

Hannah Aizenman

Advisor: Dr. Michael Grossberg

Committee: Dr. Robert Haralick, Dr. Lev Manovich, Dr. Huy Vo

External Member: Dr. Marcus Hanwell

Visualizations are structure preserving maps

DATE	LATITUDE	LONGITUDE	TAVG
2021-01-01	42.1	-77.1	27.86
2021-01-01	41.5	-73.9	29.48
2021-01-01	43.8	-73.7	24.08
2021-01-01	43.35	-73.6167	25.52
2021-01-01	43.1111	-76.1039	33.8

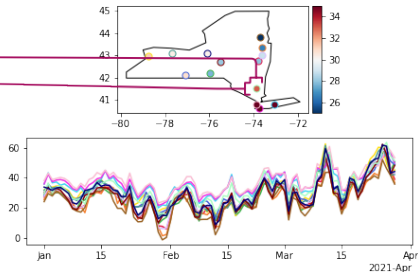
Visualizations are structure preserving maps

DATE	LATITUDE	LONGITUDE	TEMP	NAME
2021-01-01	42.1	-77.1	27.86	GANG MILLS NEW YORK
2021-01-01	41.5	-73.9	29.48	STONYKILL NEW YORK
2021-01-01	43.8	-73.7	24.08	SCHROON LAKE NEW YORK
2021-01-01	43.35	-73.6167	25.52	GLENS FALLS AP
2021-01-01	43.1111	-76.1039	33.80	SYRACUSE HANCOCK INTL AP
2021-01-01	43.1167	-77.6767	31.64	ROCHESTER GTR INTL AP
2021-01-01	40.6386	-73.7622	35.96	NEW YORK JFK INTL AP
2021-01-01	42.1997	-75.985	28.76	BINGHAMTON
2021-01-01	40.7939	-73.1017	35.78	ISLIP LI MACARTHUR AP
2021-01-01	43.0078	-73.6511	27.68	SARA NEW YORK

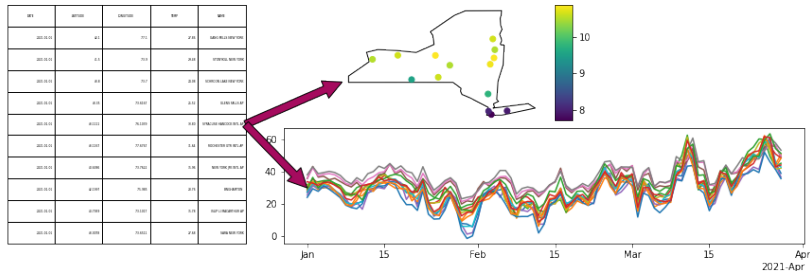


Visualizations are structure preserving maps

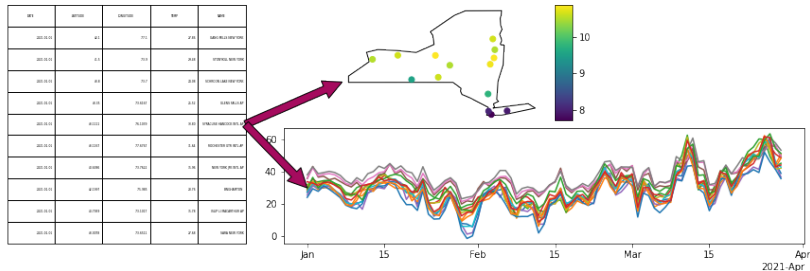
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Visualizations are structure preserving maps



Visualizations are structure preserving maps



equivariance properties of data and visual encoding match

continuity connectivity of data and visual encoding match

Domain specific libraries assume data structure[HA06]

DATE	LATITUDE	LONGITUDE	TEMP	NAME
2021-01-01	42.7431	-73.8692	27.14	ALBANY AP
2021-01-01	42.7	-73.5	30.02	SHEPARDINE NEW YORK
2021-01-01	43.8	-73.7	28.08	SCHROON LAKE NEW YORK
2021-01-01	43.0070	-73.4511	27.48	SARA NEW YORK
2021-01-01	43.1187	-77.8787	31.64	ROCHESTER GTR INTL AP
2021-01-01	43.7784	-73.8803	38.12	NEW YORK LAGUARDIA AP
2021-01-01	40.6386	-75.7622	35.98	NEW YORK JFK INTL AP
2021-01-01	43.1111	-76.1039	32.80	SYRACUSE Hancock INTL AP
2021-01-01	40.7938	-73.1017	35.78	SLIP LI MACARTHUR AP
2021-01-01	40.38	-73.8287	25.52	GLENS FALLS AP

ggplot[Wic16]

Vega[SWH14]

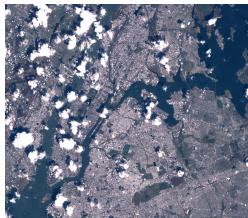
Altair[Van+18]

Tableau [STH02]

[Han06; MHS07]

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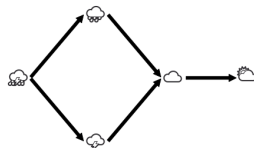
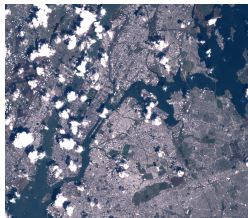
ImageJ[SRE12]

ImagePlot[Stu21]

Napari[Sof+21]

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ggplot[Wic16]
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ImageJ[SRE12]
ImagePlot[Stu21]
Napari[Sof+21]

Gephi[BHJ09]
Graphviz[EI+02]
Networkx[HSS08]

General purpose libraries can't[TM04]

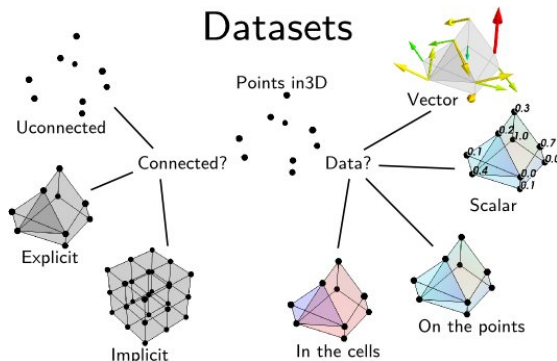


Figure: Data Representation, MayaVi 4.7.2 docs[Dat]

- 1 Matplotlib[Hun07] → Seaborn[Wt20], xarray [HH17]
- 2 D3 [BOH11]
- 3 VTK [**geveci2012vtk**; Han+15](MayaVi[RV11]) → Titan[BJ09], ParaView[AGL05]

Best practices in visualization design

Expressiveness structure preserving mappings[Mac86]

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Graphical Integrity graphs show **only** the data[Tuf01]

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Expressiveness structure preserving mappings[Mac86]

Graphical Integrity graphs show **only** the data[Tuf01]

Naturalness easier to understand when properties match[Nor93]

Contributions

Topological continuity

Equivariant monoid action

Artist Matplotlib $artist : data \rightarrow graphic$

Model

Topological Equivariant Artist Model

An artist \mathcal{A} is an equivariant map

$$\mathcal{A} : \mathcal{E} \rightarrow \mathcal{H}$$

from data \mathcal{E} space to graphic \mathcal{H} space.

Model data as a fiber bundle [BB92; BP89]

A fiber bundle is a tuple (E, K, π, F) defined by the map π

$$F \hookrightarrow E \xrightarrow{\pi} K$$

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total space E topology

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fiber space F schema

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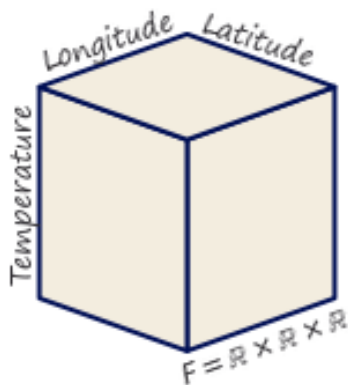
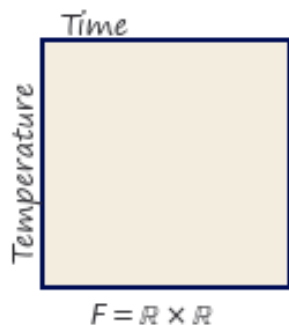
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total space E topology

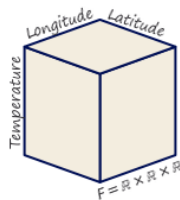
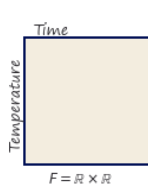
fiber space F schema

base space K continuity

Encode variable types in a schema like fiber [Spi10; Spi]

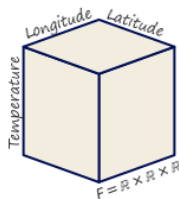
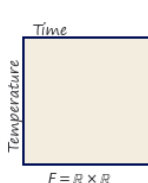


Monoids are the structure of the components of F



$$F = F_0 \times \dots \times F_i \times \dots \times F_n$$

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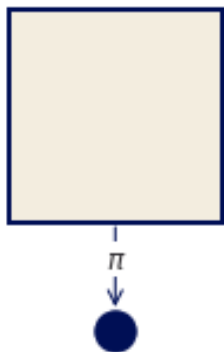
$$F = F_0 \times \dots \times F_i \times \dots \times F_n$$

Monoid actions M_i (e.g. rotation, partial ordering) define the structure on F_i

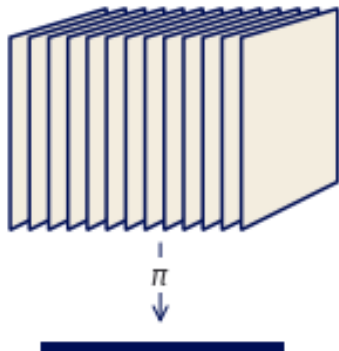
$$\bullet : M_i \times F_i \rightarrow F_i$$

where \bullet is associative and has an identity action.

K is an indexing space

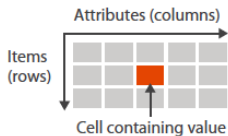


K

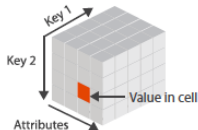


K is the space of keys into data in E [Mun14]

→ Tables



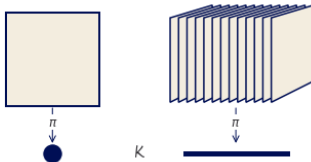
→ Multidimensional Table



→ Geometry (Spatial)



Figure: Figure 2.8 in Munzner's Visualization Analysis and Design[Mun14]

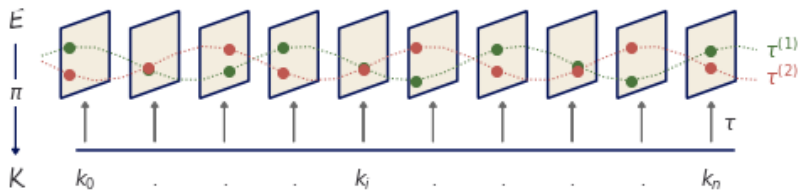


Data are sections τ on E

For any fiber bundle, there exists a map

$$\begin{array}{ccc} F & \hookrightarrow & E \\ & \pi \downarrow & \uparrow \tau \\ & K & \end{array}$$

s.t. $\pi(\tau(k)) = k$. $\Gamma(E)$ is the set of all global sections.



Graphic Bundle (H, S, π, D)

$$\begin{array}{ccc} F & \hookrightarrow & E \\ & & \pi \downarrow \uparrow \tau \\ & & K \end{array} \qquad \begin{array}{ccc} D & \hookrightarrow & H \\ & & \pi \downarrow \uparrow \rho \\ & & S \end{array}$$

Graphic Bundle (H, S, π, D)

Continuity is preserved via the many s to one k map $\xi: S \rightarrow K$

$$\begin{array}{ccc} F \hookrightarrow E & & D \hookrightarrow H \\ \pi \downarrow \uparrow \tau & & \pi \downarrow \uparrow \rho \\ K \xleftarrow{\xi} & S \end{array}$$

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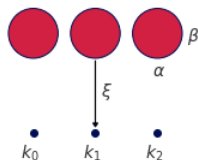
D is a proxy for the target display, for example $(x, y, r, g, b) \in D$

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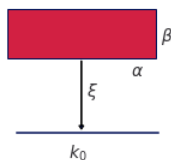
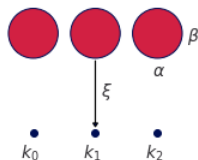


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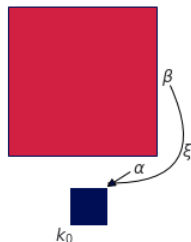
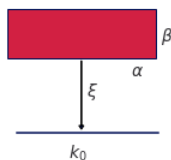
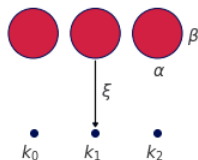


Graphic Bundle (H, S, π, D)

Continuity is preserved via the many s to one k map $\xi : S \rightarrow K$

$$\begin{array}{ccc}
 F & \hookrightarrow & E \\
 & \pi \downarrow \nearrow \tau & \\
 & K & \\
 & \xi \longleftarrow & S \\
 & \pi \downarrow \nearrow \rho & \\
 & H & \\
 D & \hookrightarrow &
 \end{array}$$

D is a proxy for the target display, for example $(x, y, r, g, b) \in D$



Visual bundle (V, K, π, P)

$$\mathcal{A} : \mathcal{E} \rightarrow \mathcal{H}$$

Visual bundle (V, K, π, P)

$$\mathcal{A} : \mathcal{E} \rightarrow \mathcal{H}$$

$$\begin{array}{ccccccc} E' & \xrightarrow{\nu} & V & \xleftarrow{\xi^*} & \xi^* V & \xrightarrow{Q} & H \\ & \searrow \pi & \downarrow \pi & & \downarrow \xi^* \pi & & \swarrow \pi \\ & & K & \xleftarrow{\xi} & S & & \end{array}$$

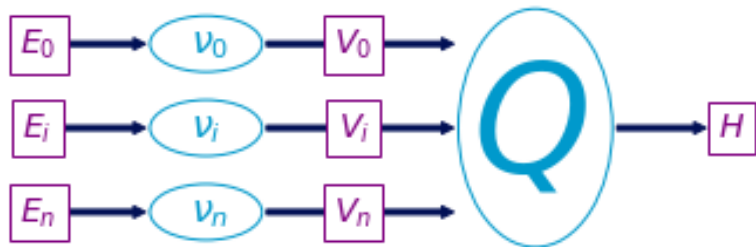
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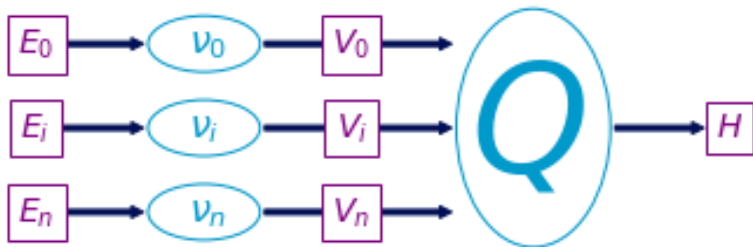
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$$A : \mathcal{O}(E) \rightarrow \mathcal{O}(H)$$

Visualization Assembly Function

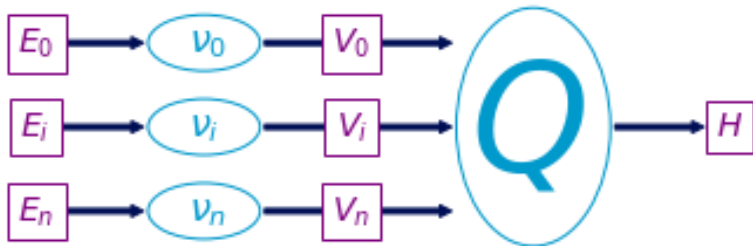


Visualization Assembly Function



$$\{v_0, \dots, v_n\} : \{\tau_0, \dots, \tau_n\} \mapsto \{\mu_0, \dots, \mu_n\}$$

Visualization Assembly Function



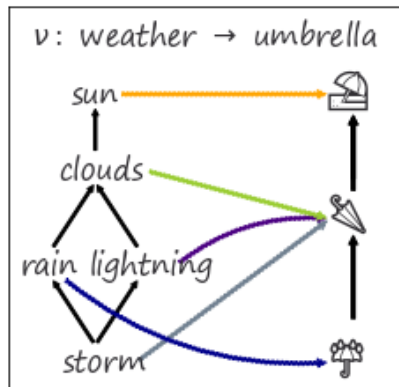
$$\{v_0, \dots, v_n\} : \{\tau_0, \dots, \tau_n\} \mapsto \{\mu_0, \dots, \mu_n\}$$

$$Q = v \circ \tau$$

Group Equivariance: Stevens' Scales [Ste46]

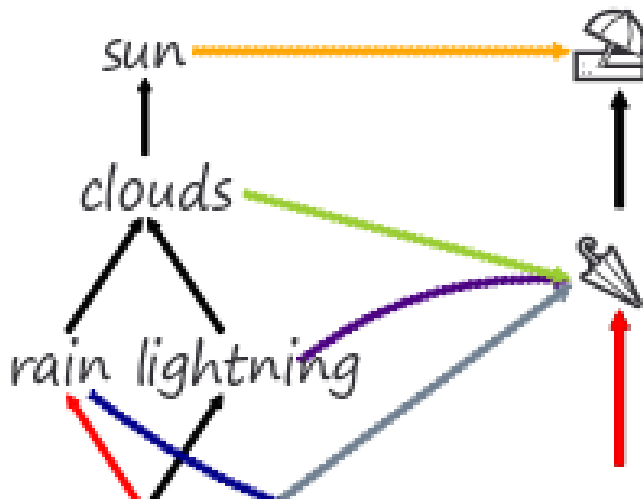
scale	group	constraint
nominal	permutation	if $r_1 \neq r_2$ then $v(r_1) \neq v(r_2)$
ordinal	monotonic	if $r_1 \leq r_2$ then $v(r_1) \leq v(r_2)$
interval	translation	$v(x + c) = v(x) + c$
ratio	scaling	$v(xc) = v(x) * c$

Monoid Equivariance: Partial Orders

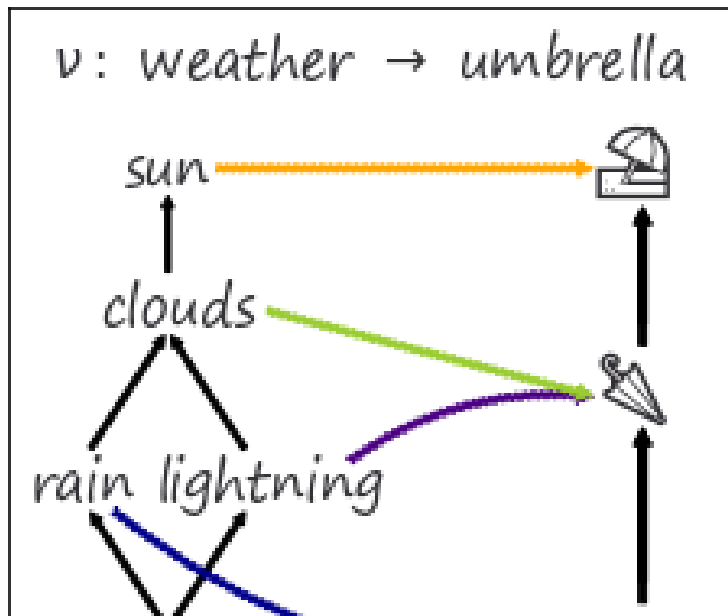


Monoid Equivariance: Partial Orders

~~x~~: weather \rightarrow umbrella



Monoid Equivariance: Partial Orders

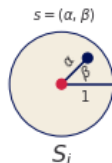
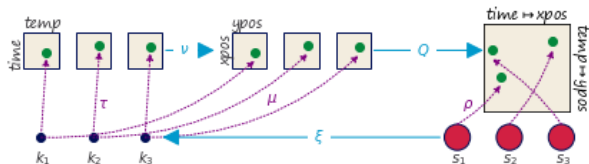


Visualization Equivariance

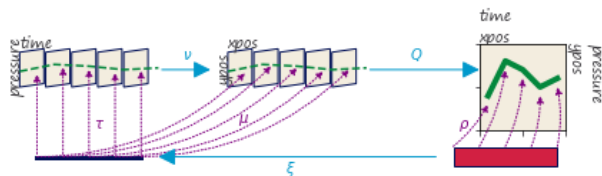


$$Q(\mu) = Q(\mu') \implies Q(m \circ \mu) = Q(m \circ \mu')$$

Scatter: $Q(xpos, ypos)(\alpha, \beta)$



Line: $Q(xpos, \hat{n}_1, ypos, \hat{n}_2)(\alpha, \beta)$



Line: $Q(xpos, \hat{n}_1, ypos, \hat{n}_2)(\alpha, \beta)$

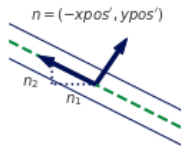
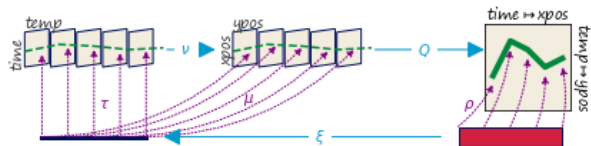
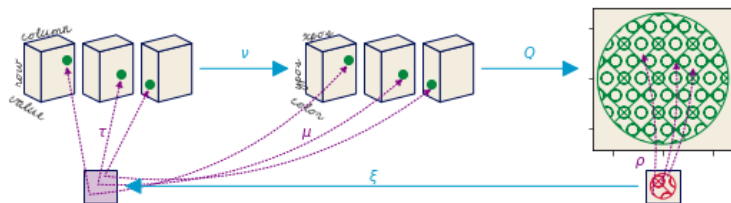
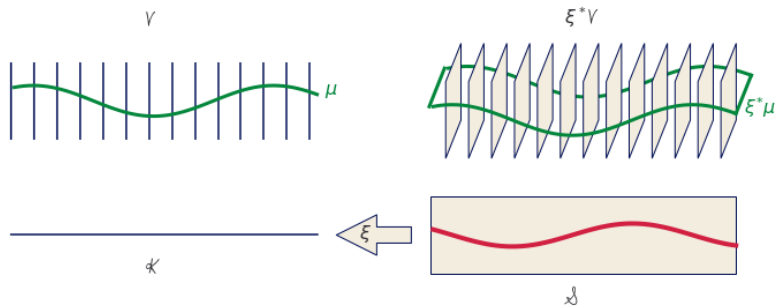


Image $Q(xpos, ypos, color)$

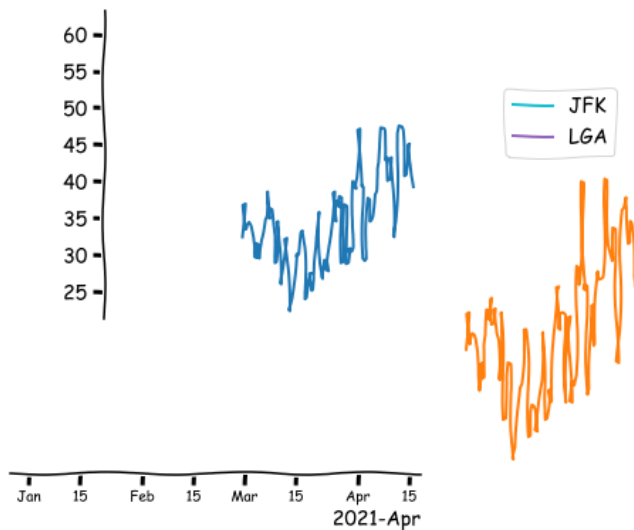


Build Q over K : \hat{Q}



$$\hat{Q}(\mu(k))(s) := Q((\xi^*\mu)(s))$$

Composition of artists $+ := \sqcup E_i$



TEAM driven rearchitecture of Matplotlib

- complex visualizations

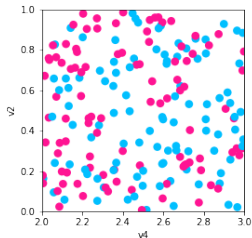
TEAM driven rearchitecture of Matplotlib

- complex visualizations
- structure preserving maps from data to visual
 - data and graphics have equivalent continuity
 - properties are equivariant under monoid actions

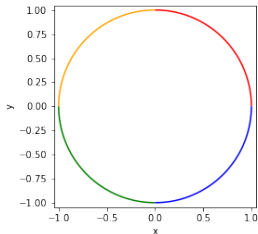
TEAM driven rearchitecture of Matplotlib

- complex visualizations
- structure preserving maps from data to visual
 - data and graphics have equivalent continuity
 - properties are equivariant under monoid actions
- fiber bundles are an abstraction
 - topologically complex heterogeneous data
 - target display spaces

How do we make things?



```
1 fig, ax = plt.subplots()
2 artist = Point(data, transforms)
3 ax.add_artist(artist)
```



```
1 fig, ax = plt.subplots()
2 artist = Line(data, transforms)
3 ax.add_artist(artist)
```

```
1 cmap = color.Categorical({'true':'deeppink', 'false':'deepskyblue'})
2 transforms = {'x': {'name': 'v4', 'encoder': lambda x: x},
3               'y': {'name': 'v2', 'encoder': lambda x: x},
4               'facecolors': {'name': 'v3', 'encoder': cmap},
5               's': {'name': None ,
6                   'encoder': lambda _: itertools.repeat(.02)}}
```

- `lambda x: x` is identity ν
- `{'name':None}` map into P without corresponding τ
- `color.Categorical` is custom ν

```
1 class ArtistClass(matplotlib.artist.Artist):
2     def __init__(self, E, V, *args, **kwargs):
3         # set properties that are specific to the artist
4         # stash the input E and V
5         super().__init__(*args, **kwargs)
6
7     def qhat(self, **args):
8         # set the properties of the graphic
9
10    def draw(self, renderer):
11        # returns tau, indexed on fiber then key
12        tau = self.E.view(self.axes)
13        # visual channel encoding applied fiberwise
14        visual = {p_i: nu_i(tau_i)
15                  for p_i, nu_i, tau_i
16                  in zip(self.V, tau)}
17        self.qhat(**visual)
18        # pass configurations off to the renderer
19        super().draw(renderer)
```

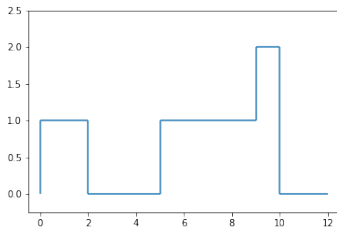
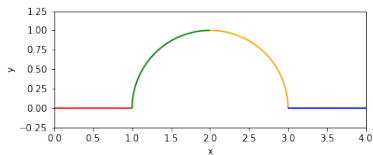


```
1 class Point(mcollections.Collection):
2     def assemble(self, x, y, s, facecolors='C0' ):
3         # construct geometries of the circle glyphs in visual coordinates
4         # set attributes of glyphs
5
6 class Line(mcollections.LineCollection):
7     def assemble(self, x, y, color='C0'):
8         # assemble line marks as set of segments
```

Continuity

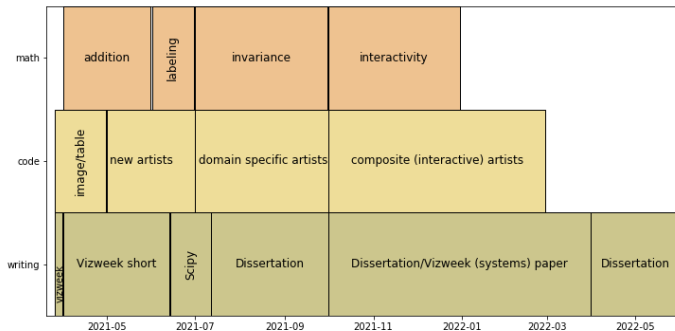
```
1 class PointData:
2     # Fiberbundle is consistent across all sections
3     FB = FiberBundle({'tables': ['vertex']},
4                       {'v1': float, 'v2': str, 'v3': float})
5     def tau(self, k):
6         return # tau evaluated at one point k
7
8 class LineData:
9     FB = FiberBundle({'tables': ['edge']},
10                      {'x': float, 'y': float, 'color': mtypes.Color()})
11     def tau(self, k):
12         return # tau evaluated on interval k
```

Same Artist, Different E



-
- 1 `LineData(FB, edge_table, vertex_table, connect=True)`
 - 2 `LineData(FB, edge_table, vertex_table, num_samples=2, connect=False)`
-

Proposed Work



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Mathematical Models of Visualization

algebraic process data and viz transforms are symmetric [KS14]

$$\begin{array}{ccccc} E & \xrightarrow{v} & V & \xrightarrow{Q} & H \\ m \downarrow & & & & \downarrow \varphi(m) \\ E & \xrightarrow{v} & V & \xrightarrow{Q} & H \end{array}$$

language APT and GoG: syntax, semantics, and grammar
[Mac86; Mac87; Wil05]

functional dependencies relationship between components [SSS09]

category theory *understanding* = *read* \circ *render* [VFR13]

Fiber is all possible values a variable can be [Spi10; Spi]

Given a space of all possible values \mathbb{U}

$$\begin{array}{ccc} \mathbb{U}_\sigma & \longrightarrow & \mathbb{U} \\ \pi_\sigma \downarrow & & \downarrow \pi \\ \mathcal{C} & \xrightarrow{\sigma} & \mathbf{DT} \end{array}$$

a fiber component is the restricted space $\mathbb{U}_{\sigma(c)}$.

$$F = \mathbb{U}_{\sigma(c)} = \mathbb{U}_T$$

DT data types of the variables in the dataset

\mathbb{U} disjoint union of all values of type $T \in \mathbf{DT}$

\mathcal{C} variable names, $c \in \mathcal{C}$

\mathbb{U}_σ \mathbb{U} restricted to the data type of a named variable

Monoid actions

A monoid M is a set with

associative binary operator $*$: $M \times M \rightarrow M$

identity element $e \in M$ such that $e * a = a * e = a$ for all $a \in M$.

left monoid action

A set F with an action [nLa21] $\bullet : M \times F \rightarrow F$ with the properties:

associativity for all $f, g \in M$ and $x \in F$, $f \bullet (g \bullet x) = (f * g) \bullet x$

identity for all $x \in F$, $e \in M$, $e \bullet x = x$

Keeping track of sections with sheafs

Restriction maps of a sheaf describe how local $\iota^*\tau$ can be glued into larger sections [Ghr14; Ghr18]

$$\begin{array}{ccc} \iota^*E & \xhookrightarrow{\iota^*} & E \\ \pi \downarrow \uparrow \iota^*\tau & & \pi \downarrow \uparrow \tau \\ U & \xhookrightarrow{\iota} & K \end{array}$$

The inclusion map $\iota : U \rightarrow K$ pulls E over U such that the pulled back $\iota^*\tau$ only contains records over $U \subset K$.

Rendering: Define a Pixel

Given a pixel

$$p = [y_{top}, y_{bottom}, x_{right}, x_{left}]$$

the inverse map of the bounding box

$$S_p = \rho_{xy}^{-1}(p)$$

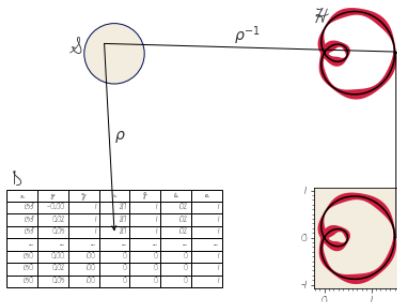
is a region $S_p \subset S$ such that

$$r_p = \iint_{S_p} \rho_r(s) ds^2 \quad (1)$$

$$g_p = \iint_{S_p} \rho_g(s) ds^2 \quad (2)$$

$$b_p = \iint_{S_p} \rho_b(s) ds^2 \quad (3)$$

yields the color of the pixel.



$$\mathcal{A} : \mathcal{E} \rightarrow \mathcal{E}$$

The topological artist is a sheaf map

$$A : \mathcal{O}(E) \rightarrow \mathcal{O}(H)$$

that carries homomorphism of monoid actions $\varphi : M \rightarrow M'$ [Ceg19]

$$A(m \cdot r) = \varphi(m) \cdot A(r)$$

Visual Channel Encoders

We define the visual transformers \mathbf{v} on components of the data bundle τ_i

$$\{\mathbf{v}_0, \dots, \mathbf{v}_n\} : \{\tau_0, \dots, \tau_n\} \mapsto \{\mu_0, \dots, \mu_n\}$$

as the set of equivariant maps with the constraint

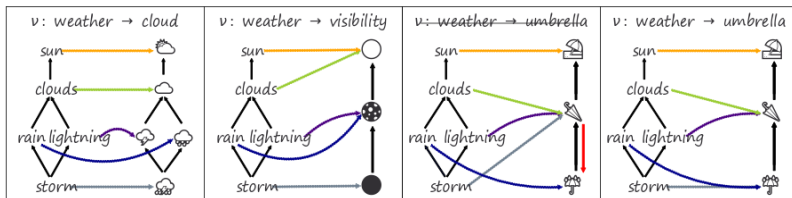
$$\mathbf{v}_i(m_r(E_i)) = \varphi(m_r)(\mathbf{v}_i(E_i))$$

where $\varphi : M \rightarrow M'$ carries a homomorphism of monoid actions.

P Components

ν_i	μ_i	$\text{codomain}(\nu_i) \subset P_i$
position	x, y, z, theta, r	\mathbb{R}
size	linewidth, markersize	\mathbb{R}^+
shape	markerstyle	$\{f_0, \dots, f_n\}$
color	color, facecolor, markerfacecolor, edgecolor	\mathbb{R}^4
texture	hatch	\mathbb{N}^{10}
	linestyle	$(\mathbb{R}, \mathbb{R}^{+n, n\%2=0})$

Monoid Equivariance: Partial Orders



The glyph is the graphic generated by $Q(S_j)$ where the path connected components $J \subset K$ are defined

$$J = \{j \in K \text{ s. t. } \exists \gamma \text{ s.t. } \gamma(0) = k \text{ and } \gamma(1) = j\}$$

such that the path γ from k to j is a continuous function from the interval $[0,1]$ and S_j is the region

$$H \xrightleftharpoons[\rho(S_j)]{} S_j \xrightleftharpoons[\xi^{-1}(J)]{\xi(s)} J_k$$

such that the glyph is differentiable, in keeping with Ziemkiewicz and Kosara's description of a glyph[ZK09].

Artist Equivalence class

When artists share a base space

$$K_2 \hookrightarrow K_1$$

a composition operator can be defined such that the the artists can be considered to be acting on different components of the same section.

Complex γ

```
1 class Categorical:
2     def __init__(self, mapping):
3         # check that the conversion is to valid colors
4         assert(mcolors.is_color_like(color) for color in mapping.values())
5         self._mapping = mapping
6
7     def __call__(self, value):
8         # convert value to a color
9         return [mcolors.to_rgba(self._mapping[v]) for v in values]
```

That we can test for action equivariance

```
1 def test_nominal(values, encoder):
2     m1 = list(zip(values, encoder(values)))
3     random.shuffle(values)
4     m2 = list(zip(values, encoder(values)))
5     assert sorted(m1) == sorted(m2)
```

Artist

```
1 class ArtistClass(matplotlib.artist.Artist):
2     def __init__(self, data, transforms, *args, **kwargs):
3         # properties that are specific to the graphic
4         self.data = data
5         self.transforms = transforms
6         super().__init__(*args, **kwargs)
7
8     def assemble(self, **args):
9         # set the properties of the graphic
10
11    def draw(self, renderer):
12        # returns K, indexed on fiber then key
13        view = self.data.view(self.axes)
14        # visual channel encoding applied fiberwise
15        visual = {p: t['encoder'](view[t['name']])
16                  for p, t in self.transforms.items()}
17        self.assemble(**visual)
18        # pass configurations off to the renderer
19        super().draw(renderer)
```

Artists: Scatter & Line

```
1 class Point(mcollections.Collection):
2     def assemble(self, x, y, s, facecolors='C0' ):
3         # construct geometries of the circle glyphs in visual coordinates
4         self._paths = [mpath.Path.circle(center=(xi,yi), radius=si)
5                         for (xi, yi, si) in zip(x, y, s)]
6         # set attributes of glyphs, these are vectorized
7         # circles and facecolors are lists of the same size
8         self.set_facecolors(facecolors)
9
10    class Line(mcollections.LineCollection):
11        def assemble(self, x, y, color='C0'):
12            #assemble line marks as set of segments
13            segments = [np.vstack((vx, vy)).T for vx, vy in zip(x, y)]
14            self.set_segments(segments)
15            self.set_color(color)
```

```
1 def view(self, axes):
2     table = defaultdict(list)
3     for k in self.keys:
4         table['index'].append(k)
5         for (name, value) in zip(self.FB.fiber.keys(), self.tau(k)[1]):
6             table[name].append(value)
7     return table
```

VertexSimplex (name, value), value is scalar

EdgeSimplex (name, value), value is $[x_0, \dots, x_n]$

Fiber Bundle

```
1 @dataclass
2 class FiberBundle:
3     """
4     Attributes
5     -----
6     K: {'tables': []}
7     F: {variable name: type}
8     """
9     K: dict
10    F: dict
```

GraphLine Data Model

```
1 class GraphLine:
2     def __init__(self, FB, edge_table, vertex_table, num_samples=1000,
3                 connect=False):
4         #set args as attributes and generate distance
5         if connect: # test connectivity if edges are continuous
6             assert edge_table.keys() == self.FB.F.keys()
7             assert is_continuous(vertex_table)
8
9     def tau(self, k):
10        # evaluates functions defined in edge table
11        return(k, (self.edges[c][k](self.distances)
12                  for c in self.FB.F.keys()))
13
14    def view(self, axes):
15        # walk the edge_vertex table to return the edge function
16        table = defaultdict(list)
17        for (i, (start, end)) in sorted(zip(self.ids, self.vertices),
18                                       key=lambda v:v[1][0]):
19            table['index'].append(i)
20            # same as view for line, returns nested list
21            for (name, value) in zip(self.FB.F.keys(), self.tau(i)[1]):
22                table[name].append(value)
23        return table
```