

cs6630 | October 2 2014

FILTERING & AGGREGATION

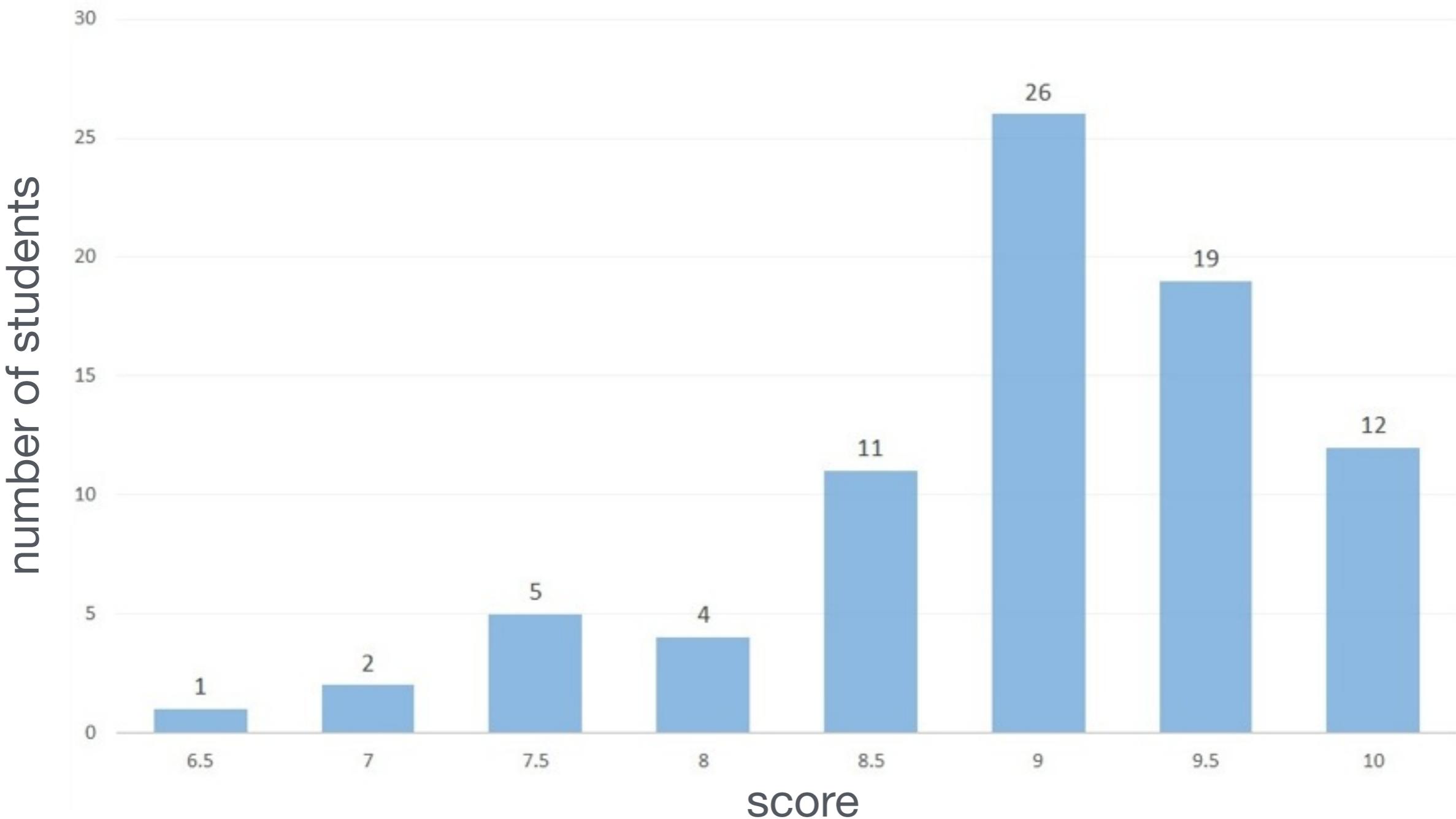
Miriah Meyer
University of Utah



administrivia . . .

- exam on Tuesday
- data exploration grades are in

data exploration assignment



last time . . .

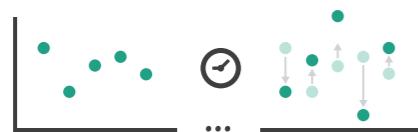
FOCUS + CONTEXT

carefully pick what to show

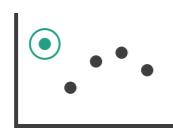
hint at what you are not showing

Manipulate

→ Change



→ Select



→ Navigate

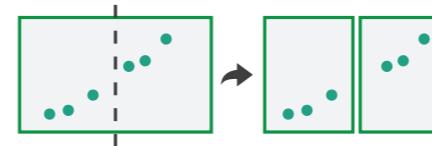


Facet

→ Juxtapose



→ Partition



→ Superimpose



Reduce

→ Embed

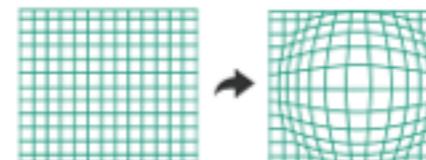
→ Elide Data



→ Superimpose Layer



→ Distort Geometry



what is elision?

what is elision?

focus items shown in detail, other items
summarized for context

superimpose

superimpose

focus layer limited to a local region of view, instead of stretching across the entire view

distort

distort

use geometric distortion of the contextual regions to make room for the details in the focus region(s)

distortion concerns

- unsuitable for relative spatial judgements**
- overhead of tracking distortion**
- visual communication of distortion**
 - gridlines, shading
- target acquisition problem**
 - lens displacing items away from screen location
- mixed results compared to separate views and temporal navigation**
- fisheye follow-up: concern with enthusiasm over distortion**
 - what is being shown: selective filtering
 - how it is being shown: distortion as one possibility

today . . .

-filtering

-aggregation

Reducing Items and Attributes

④ Filter

→ Items

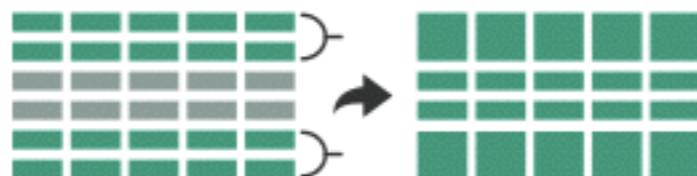


→ Attributes

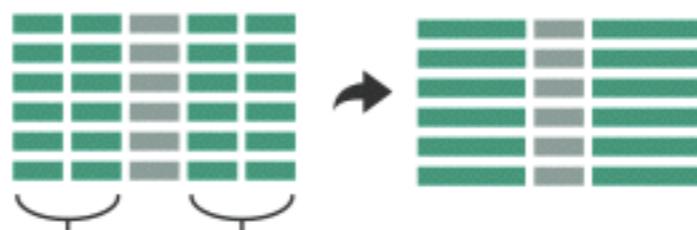


⑤ Aggregate

→ Items



→ Attributes



why reduce?

filter vs aggregation

filter

elements are eliminated

→ Items



→ Attributes



filter

elements are eliminated

→ Items



→ Attributes



dynamic queries

filter

elements are eliminated

→ Items



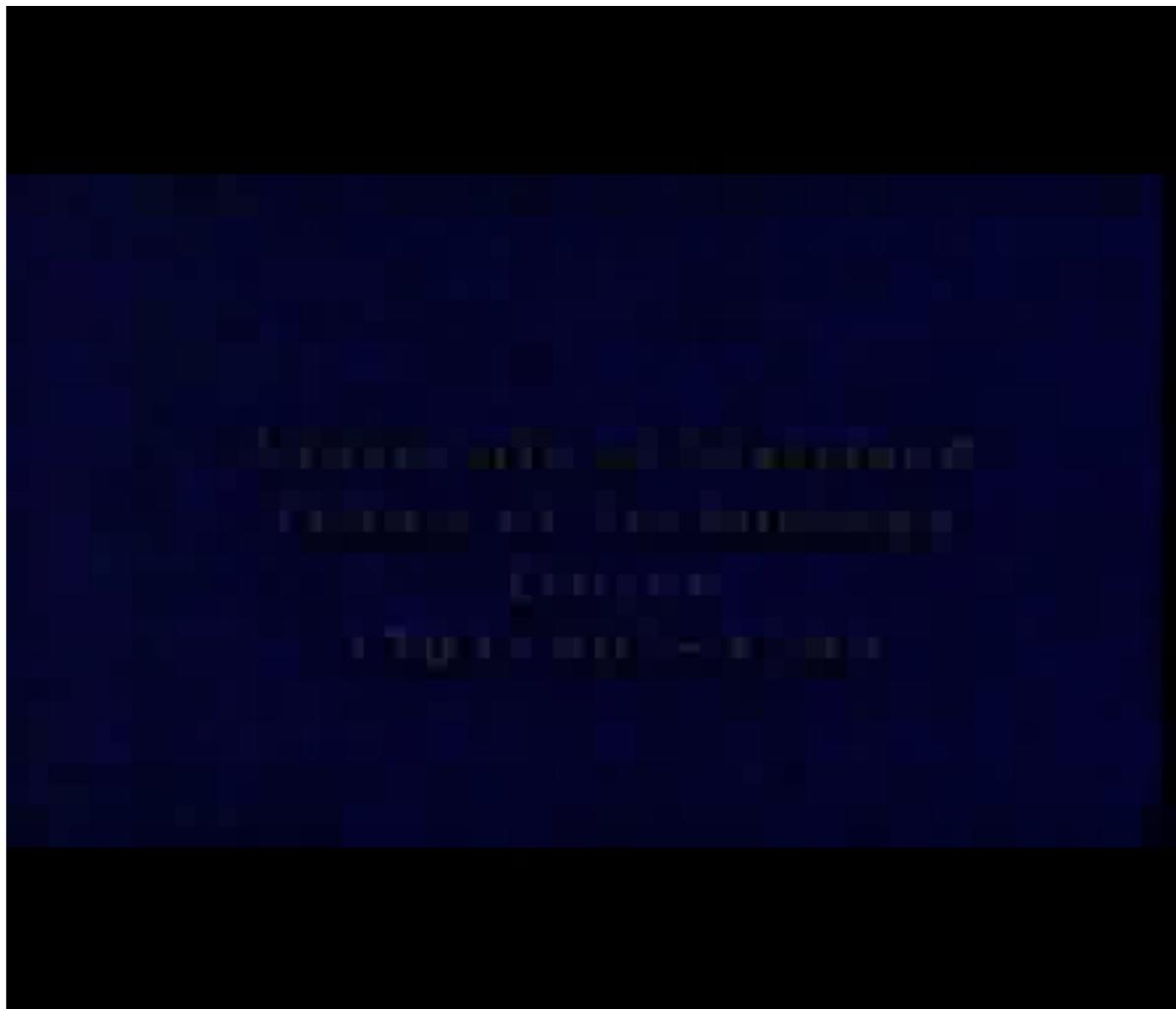
→ Attributes



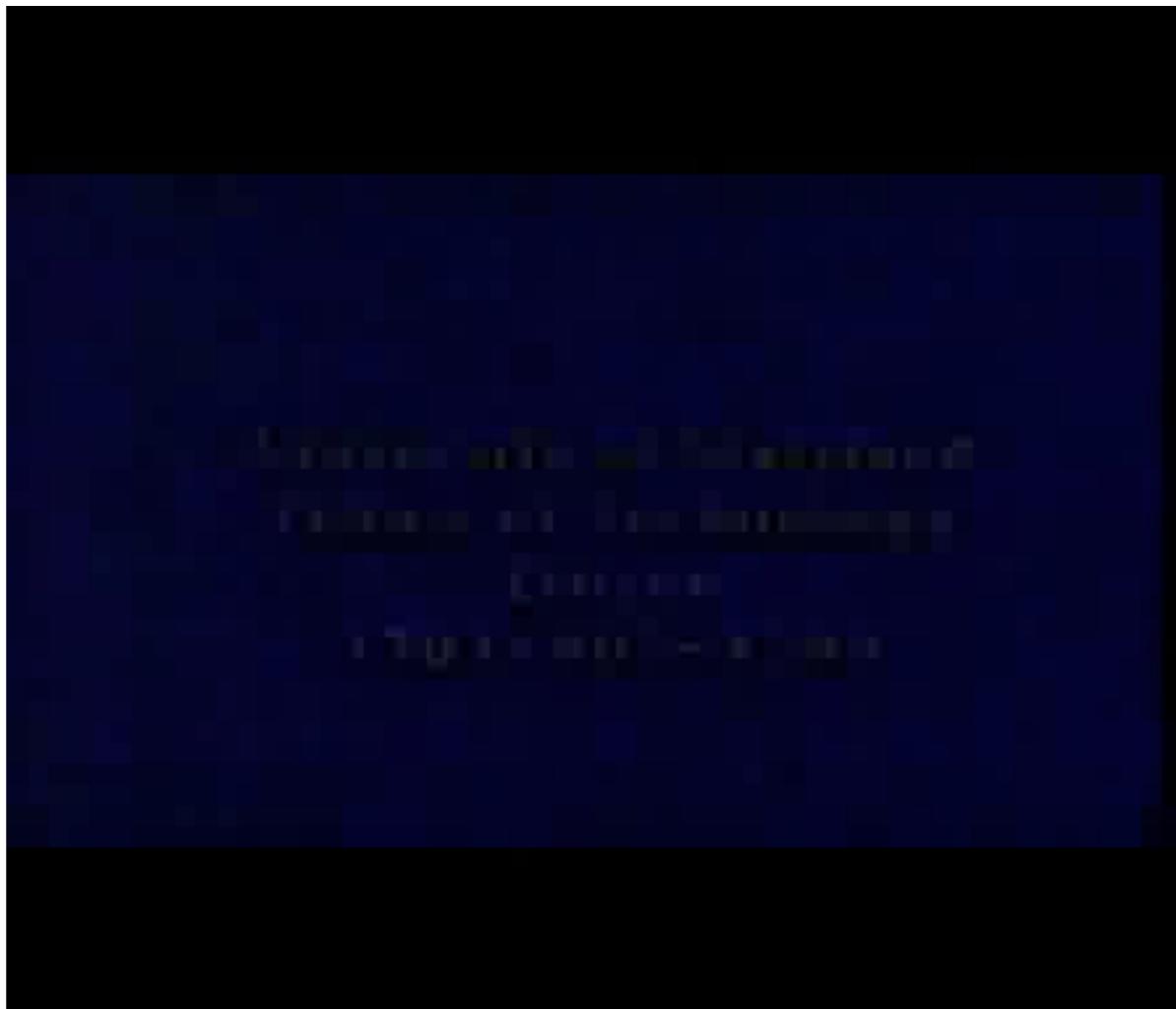
dynamic queries

coupling between encoding and interaction so that user can immediately see the results of an action

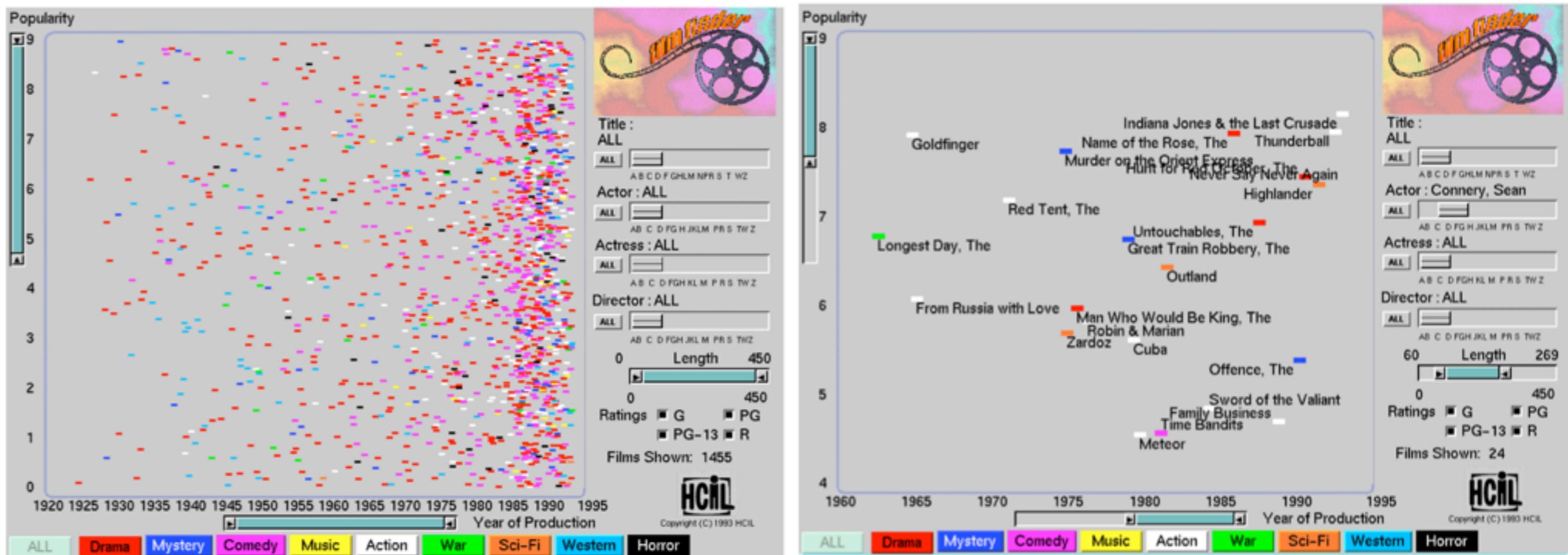
ITEM FILTERING

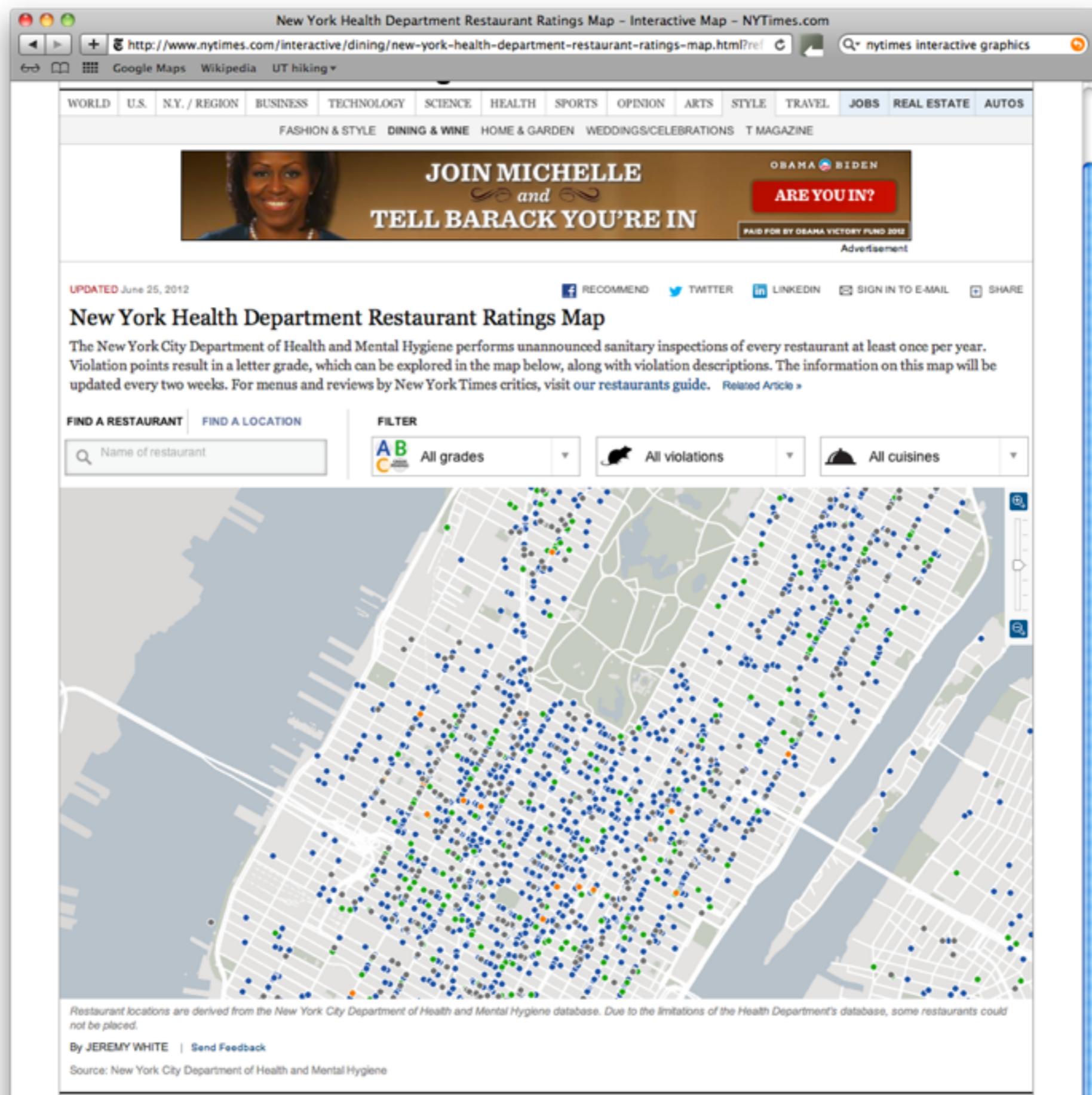


ITEM FILTERING



ITEM FILTERING

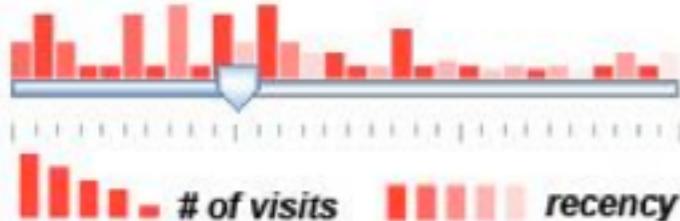




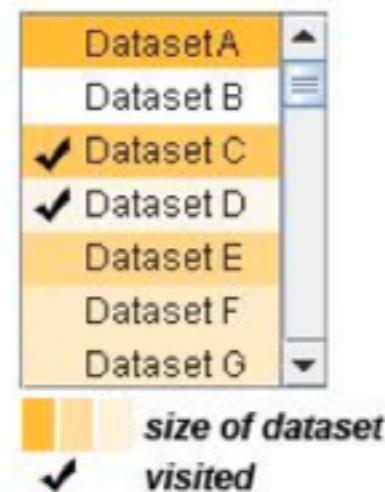
scented widgets

information scent: user's (imperfect) perception of data

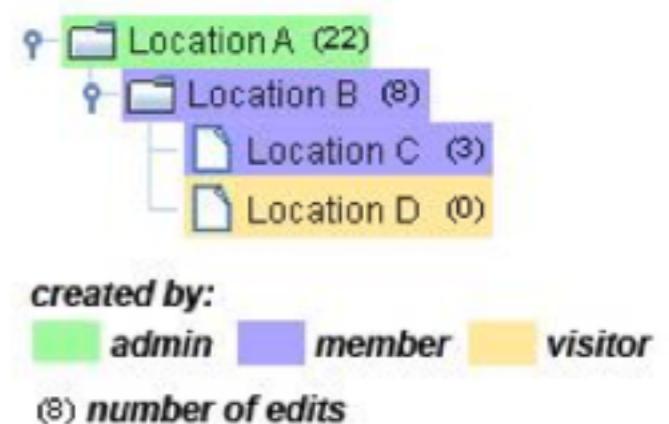
GOAL: lower the cost of information forging through better cues



★ rating nm ordered rank



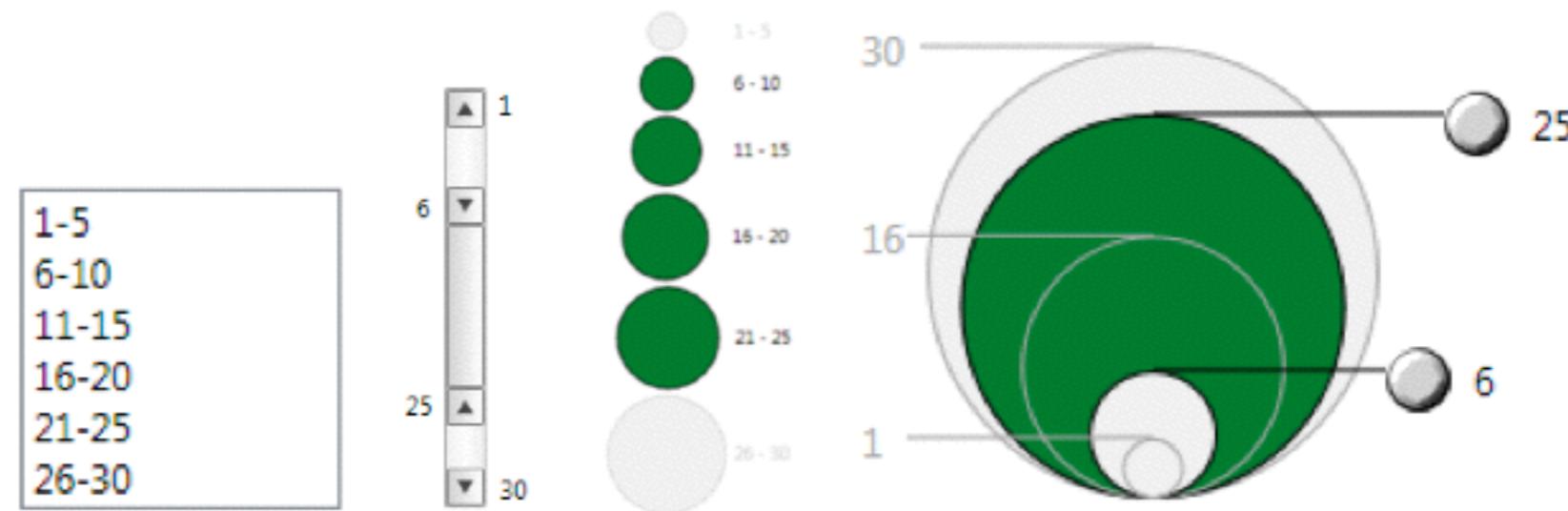
size of dataset
visited



interactive legends

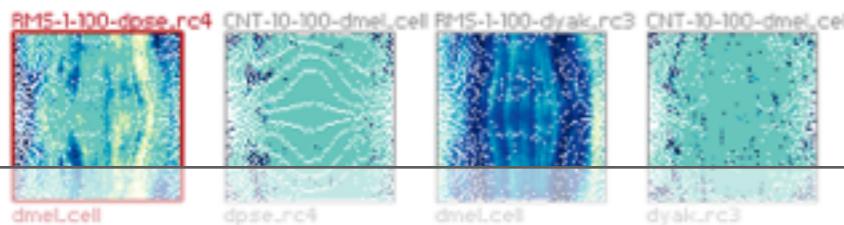
controls combining the visual representation of static legends with interaction mechanisms of widgets

define and control visual display together

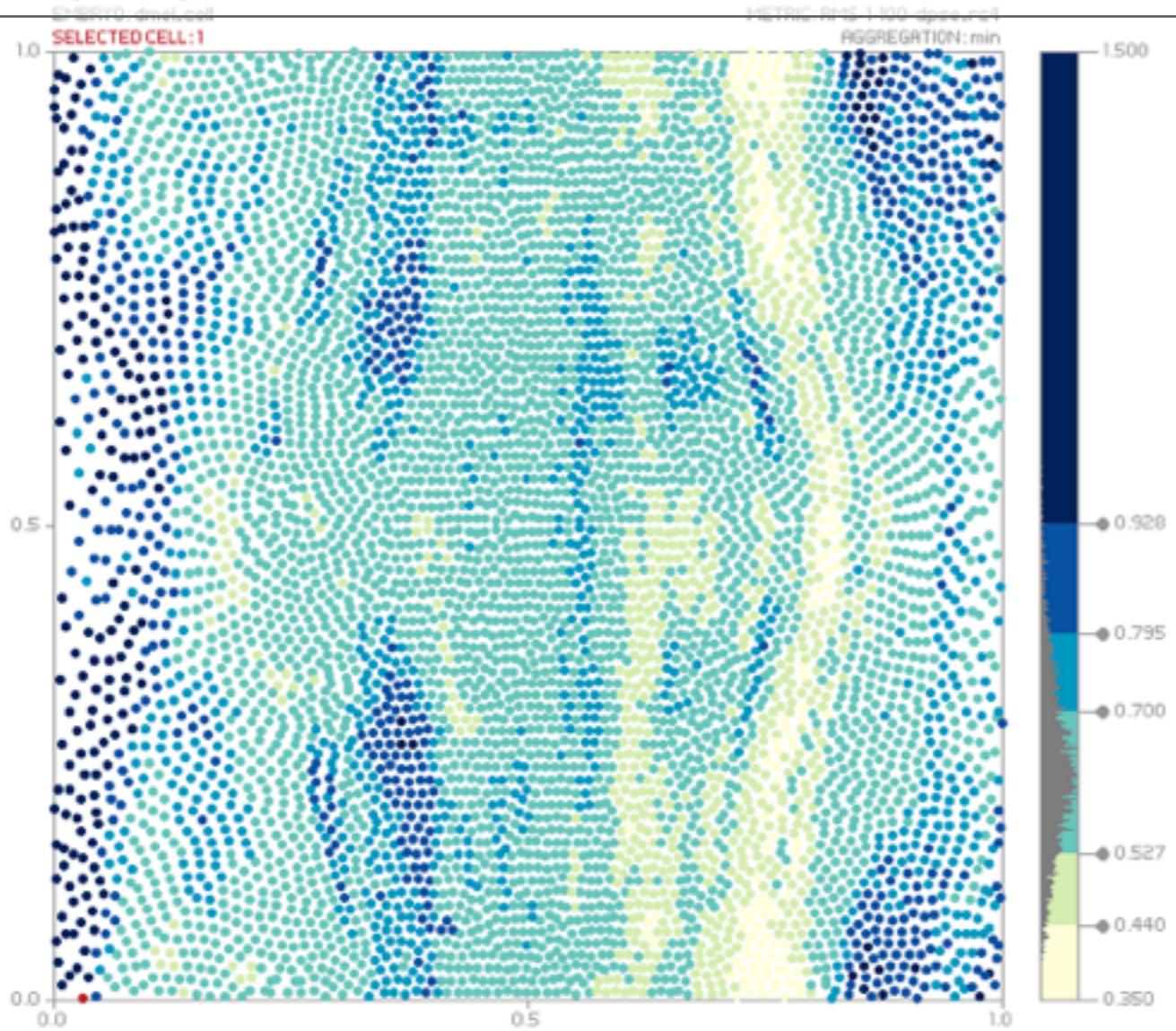


MulteeSum

Summaries

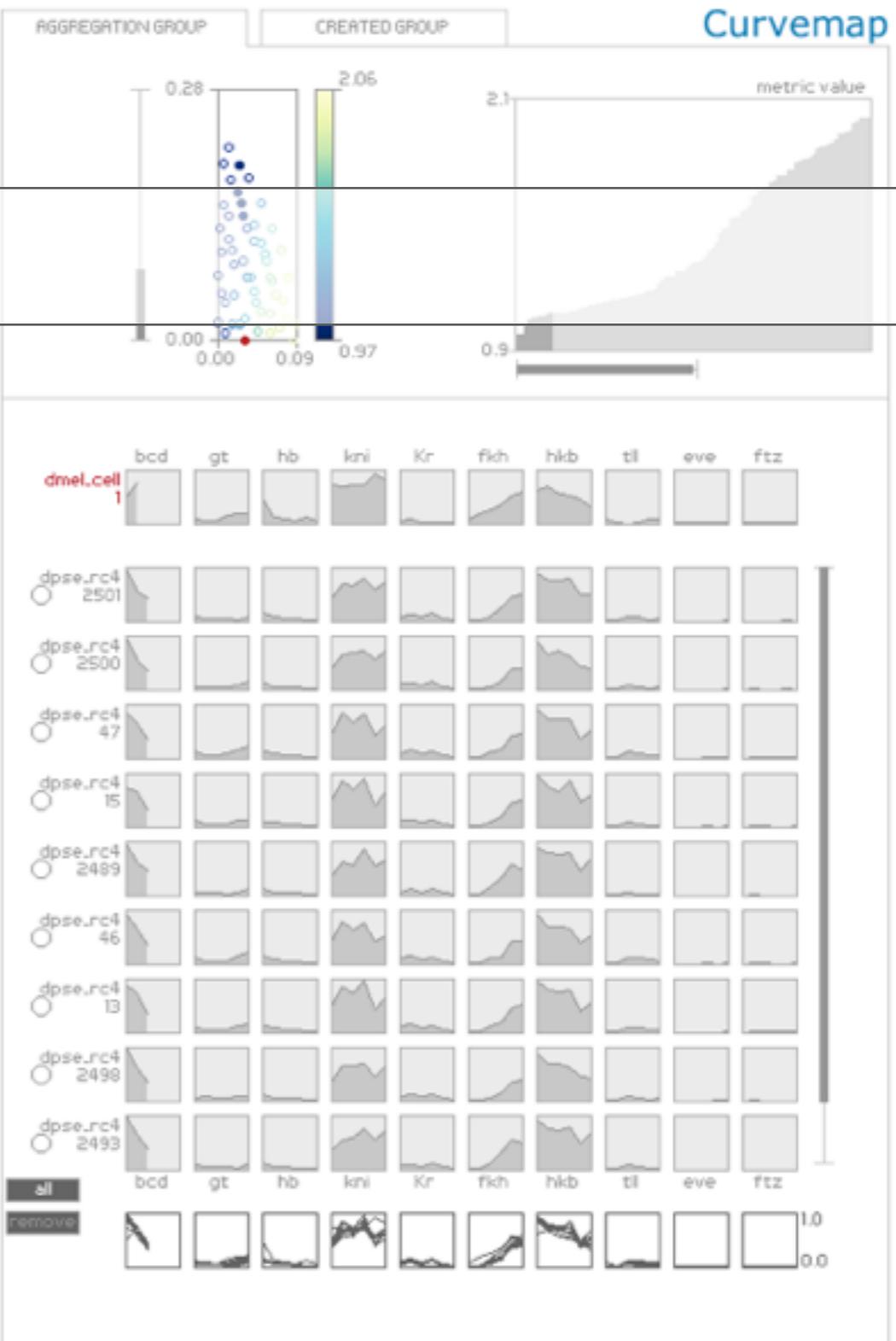


Embryo Map

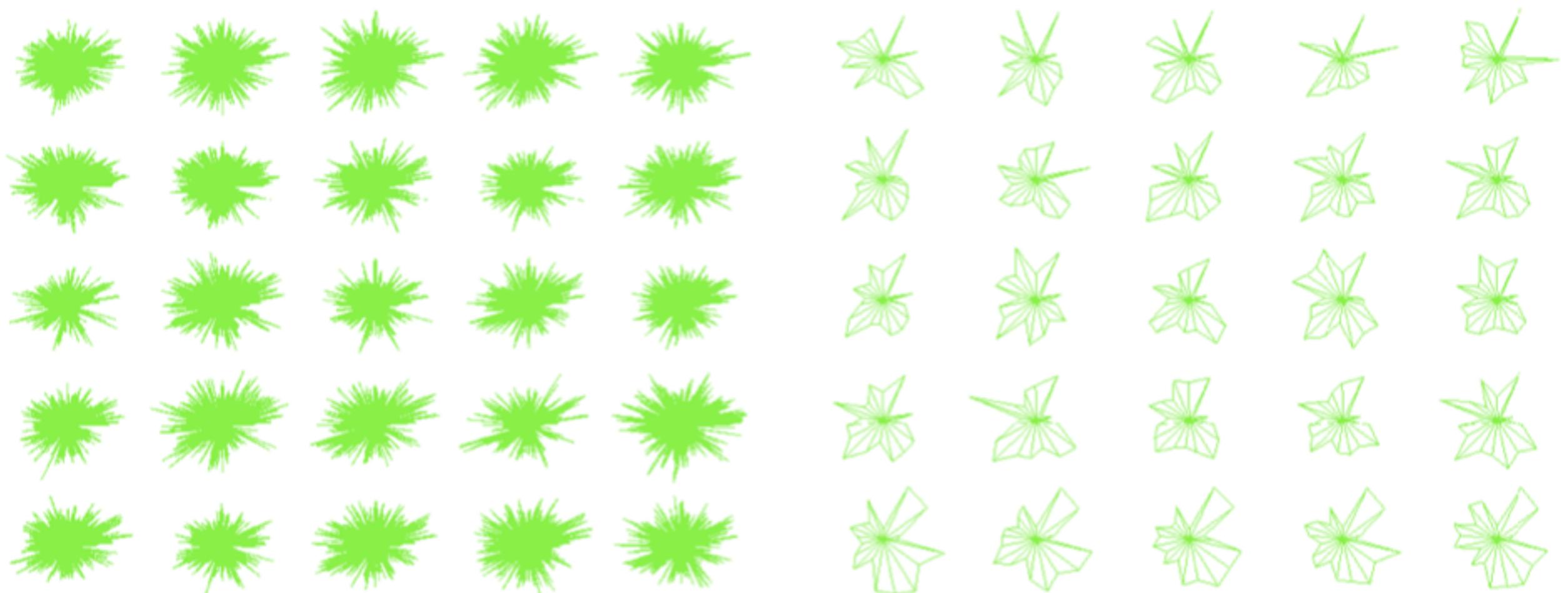


Colormaps▲

Demo



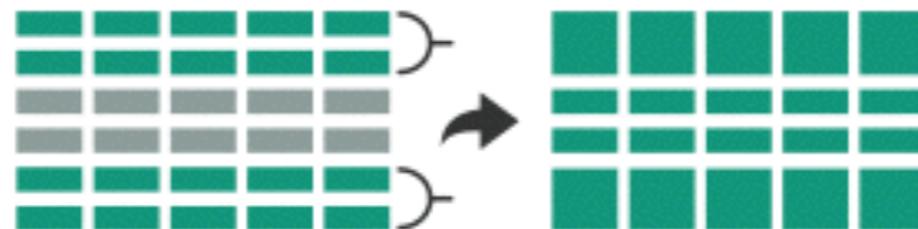
ATTRIBUTE FILTERING



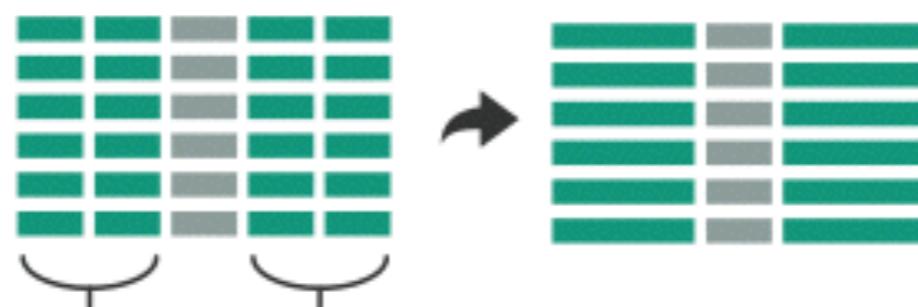
aggregate

a group of elements is represented by a new derived element that stands in for the entire group

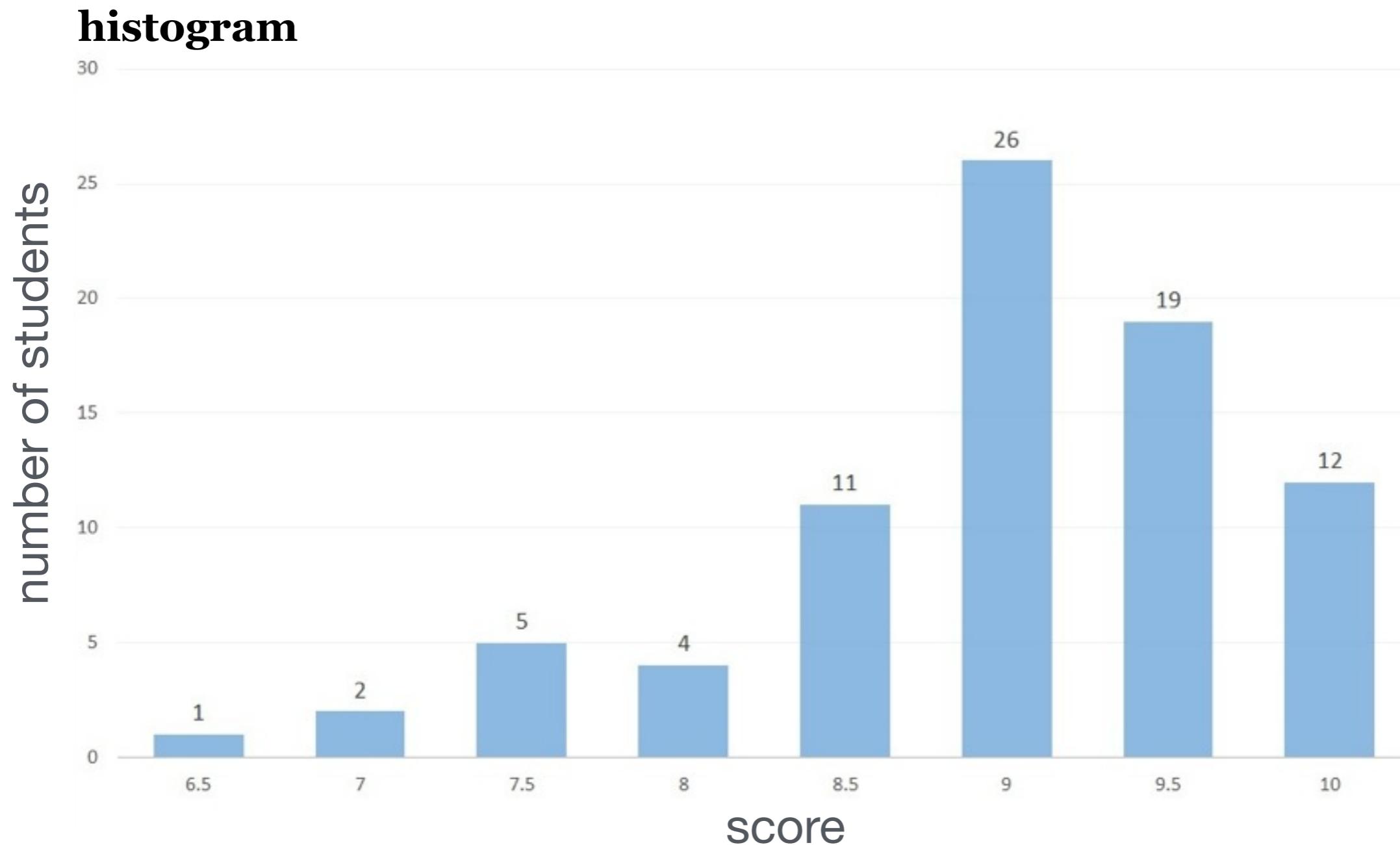
→ Items



→ Attributes

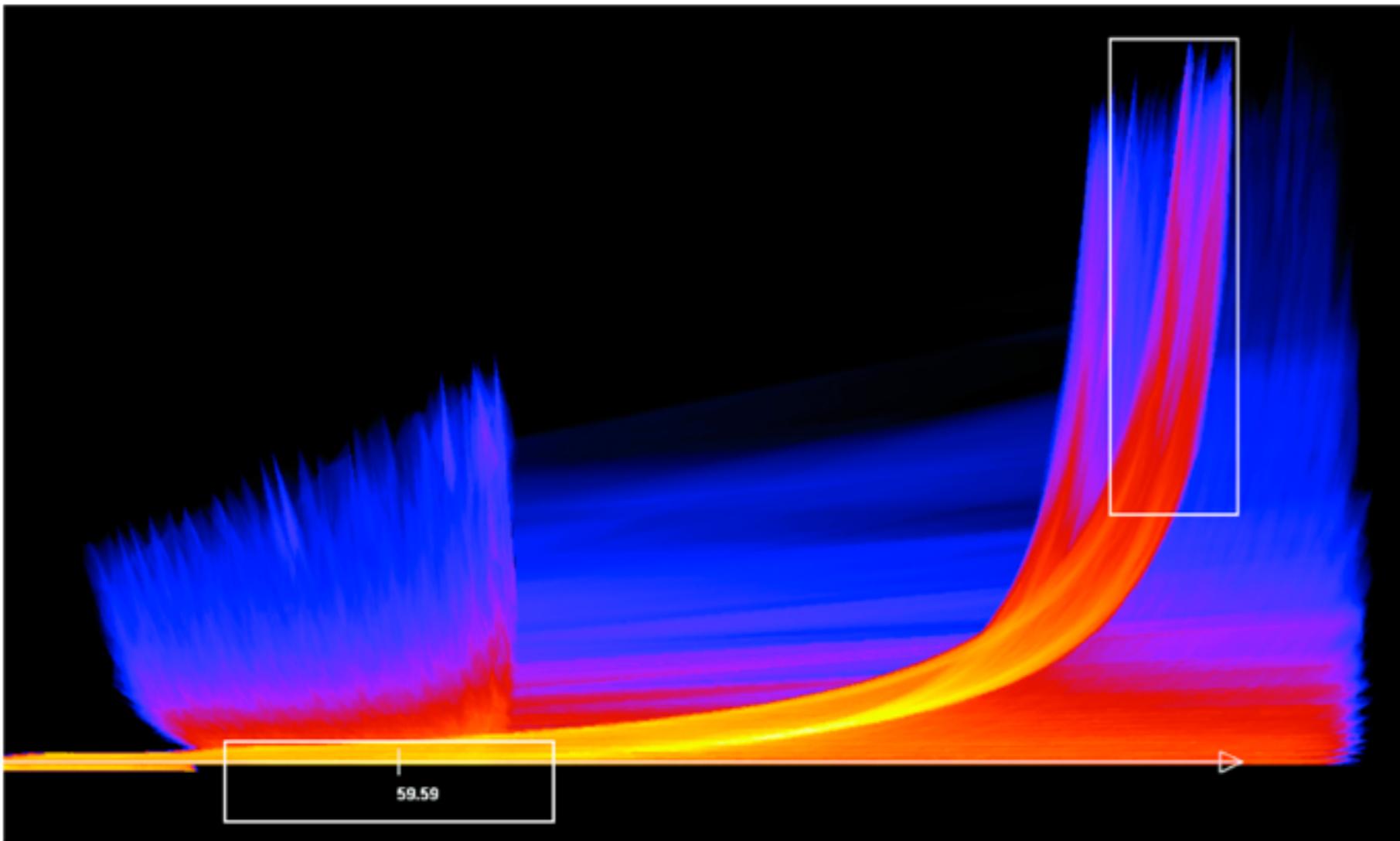


item aggregation



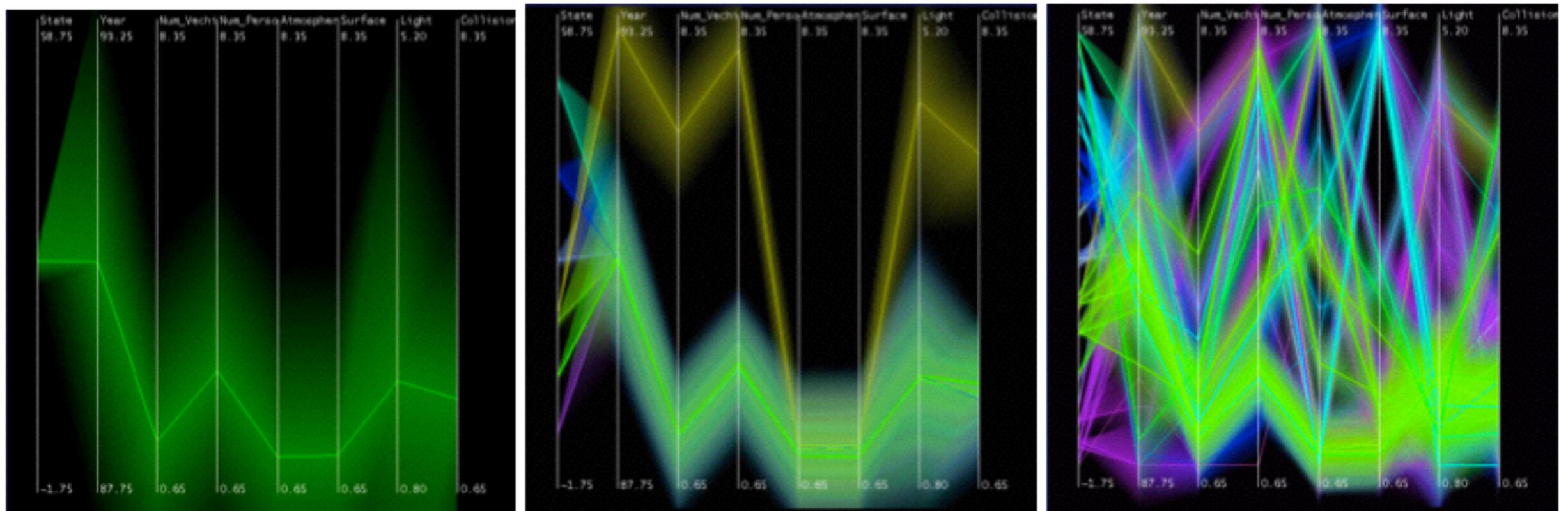
item aggregation

continuous scatterplot

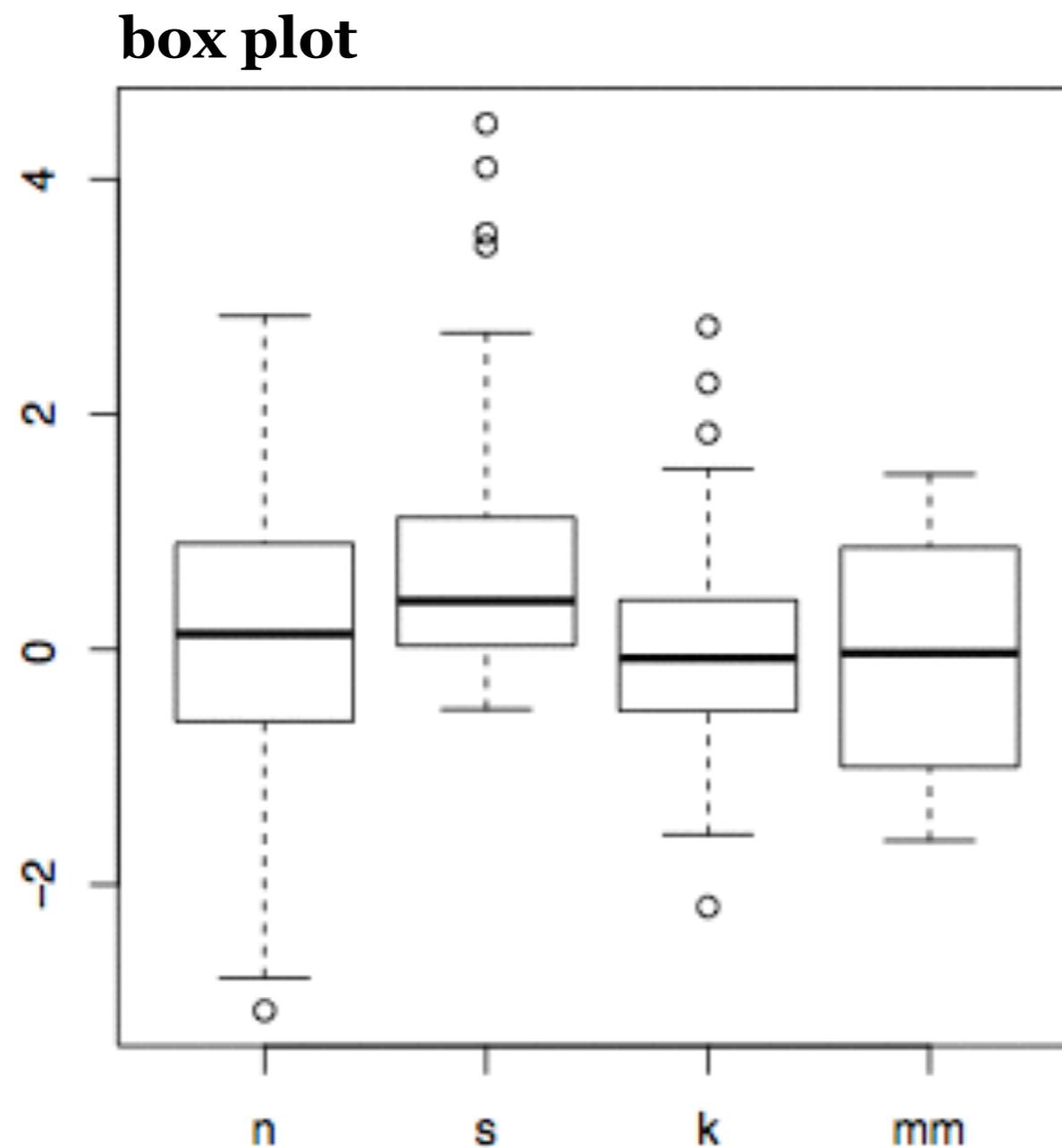


item aggregation

hierarchical parallel coordinates



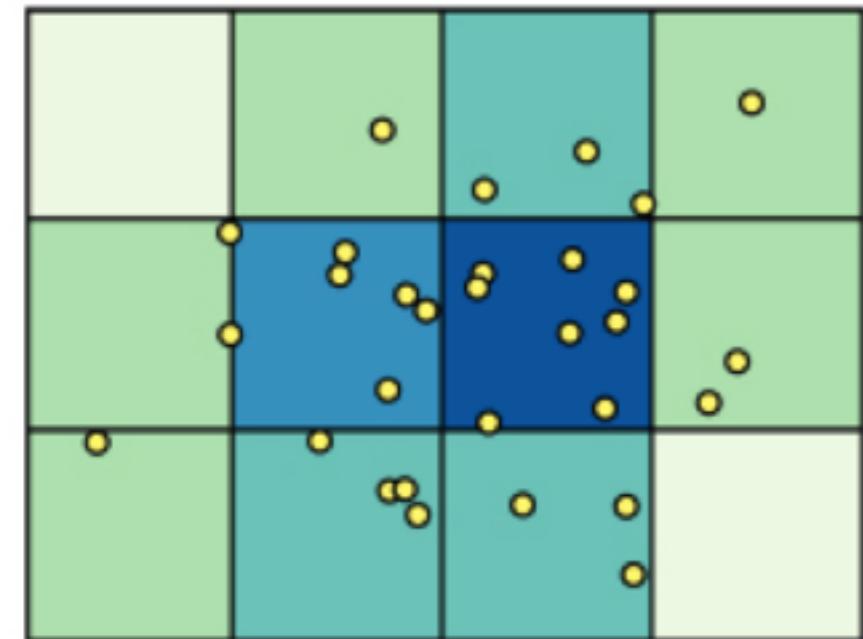
item aggregation



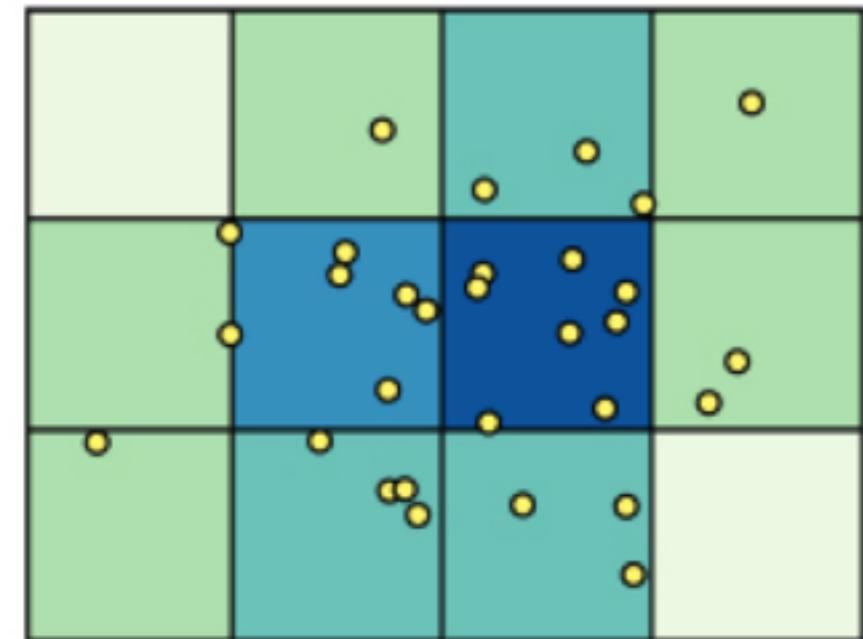
spatial aggregation



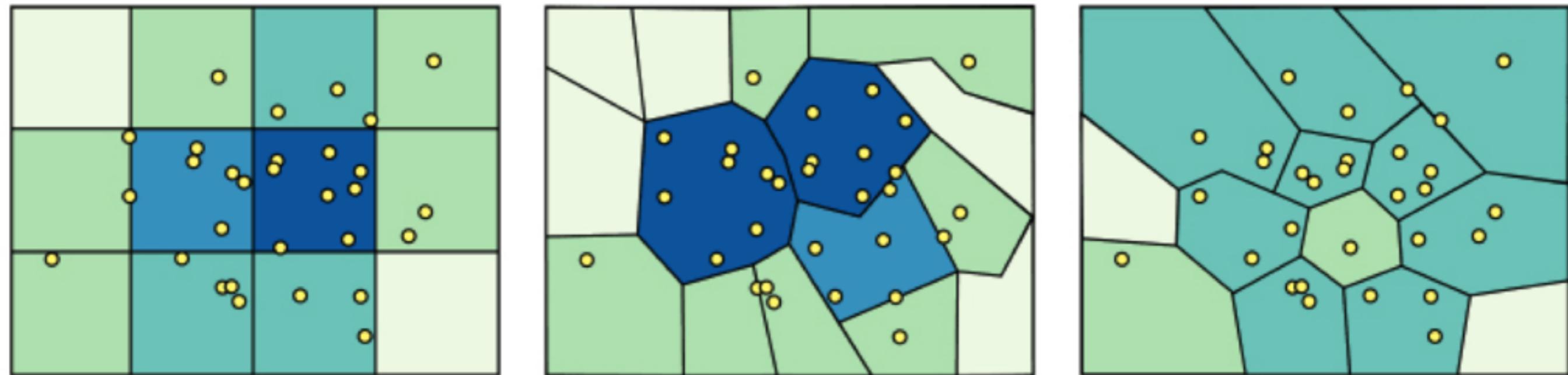
spatial aggregation



spatial aggregation



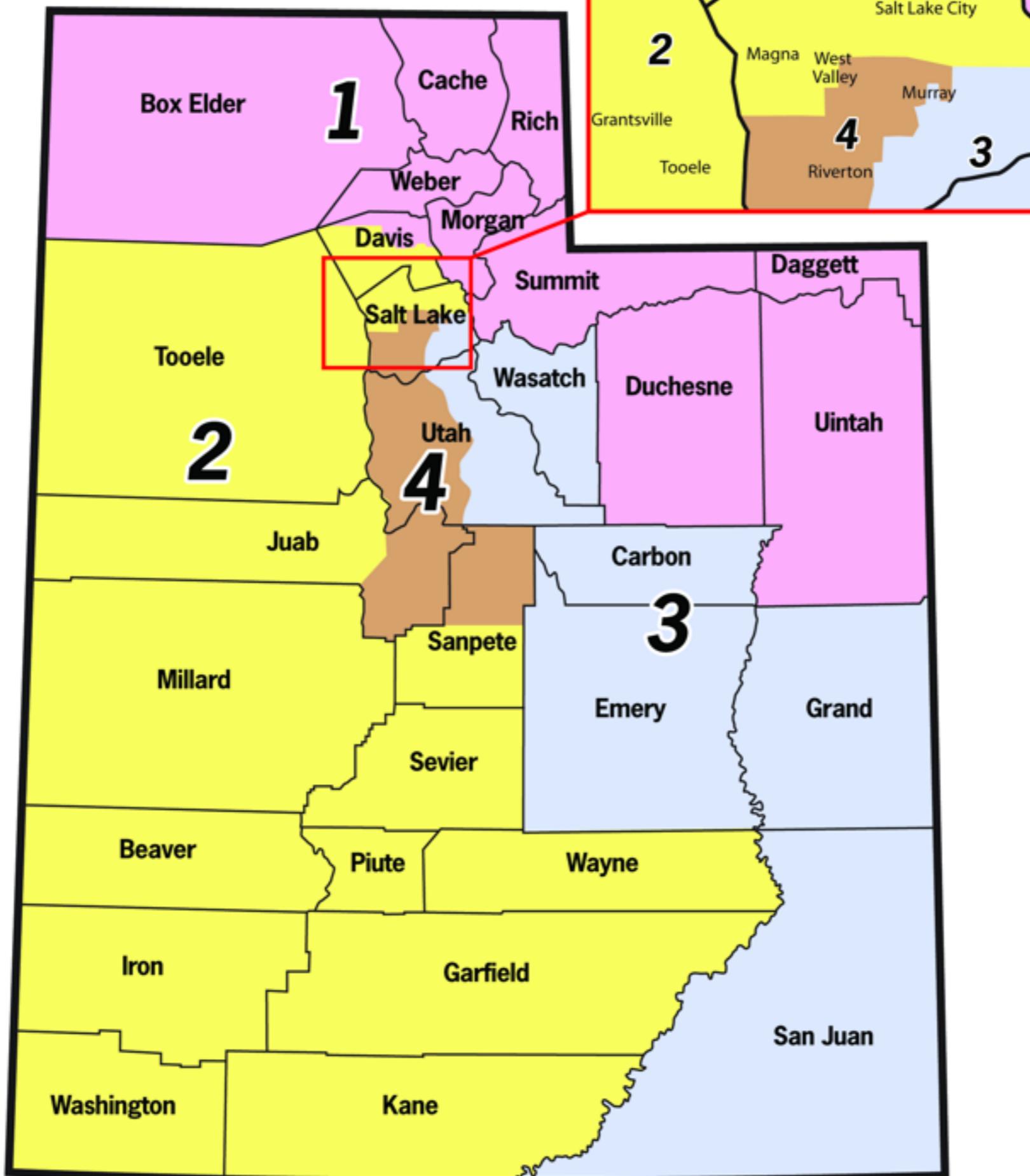
spatial aggregation



modifiable areal unit problem

in cartography, changing the boundaries of the regions used to analyze data can yield dramatically different results

Congressional Districts



attribute aggregation

- 1) group attributes and compute a similarity score across the set
- 2) dimensionality reduction, to preserve meaningful structure

attribute aggregation

- 1) group attributes and compute a similarity score across the set**
- 2) dimensionality reduction, to preserve meaningful structure**

MulteeSum

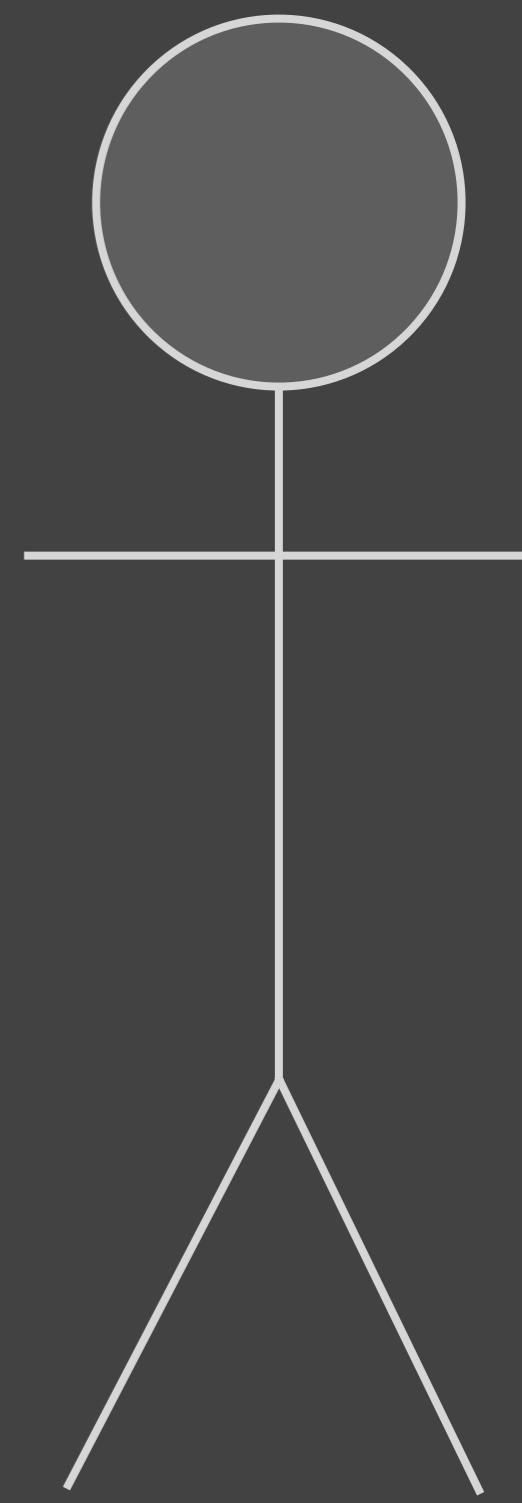
A Tool for Comparative Spatial and Temporal Gene Expression Data

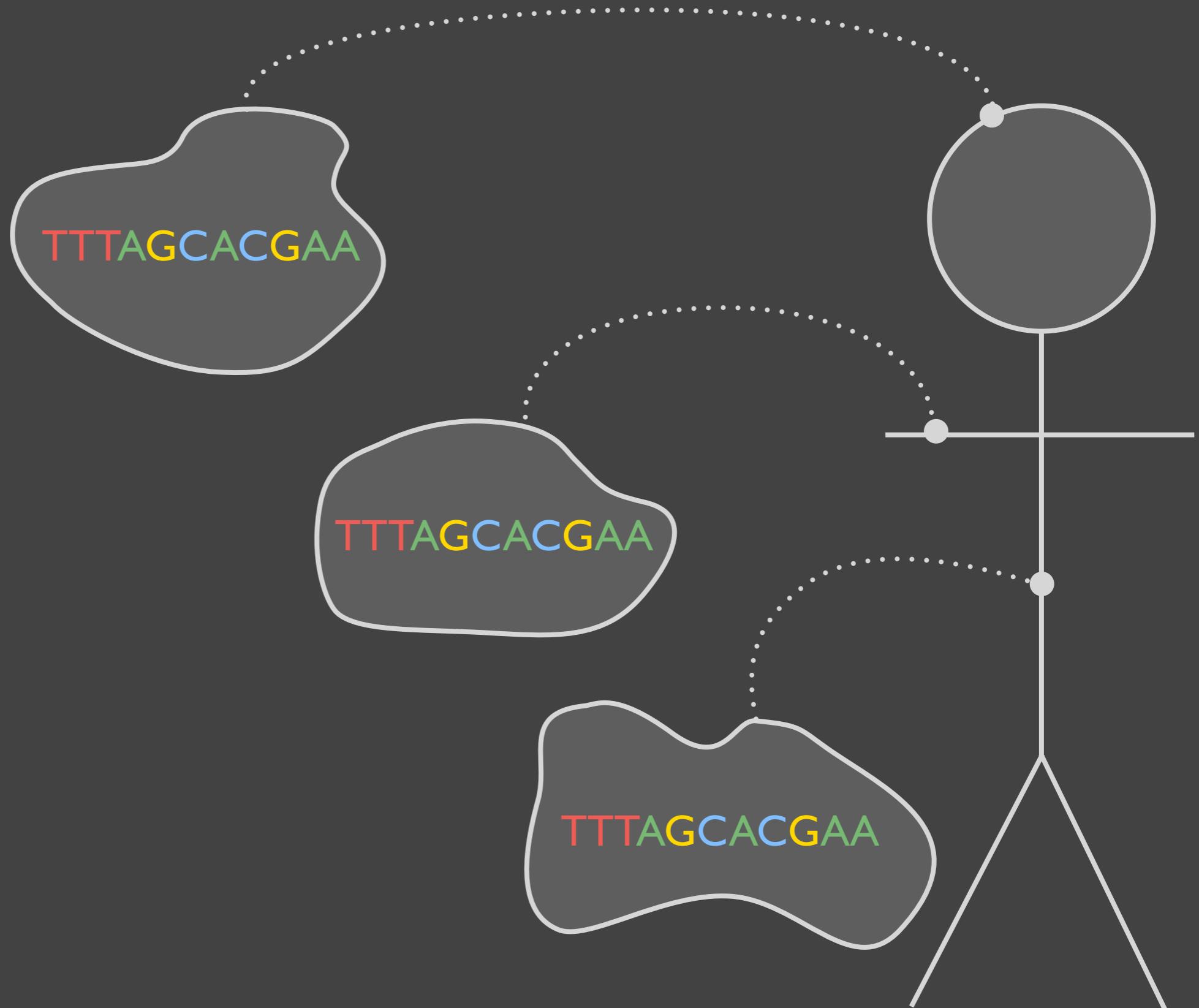
Miriah Meyer,¹ Tamara Munzner,² Angela DePace,³ Hanspeter Pfister¹

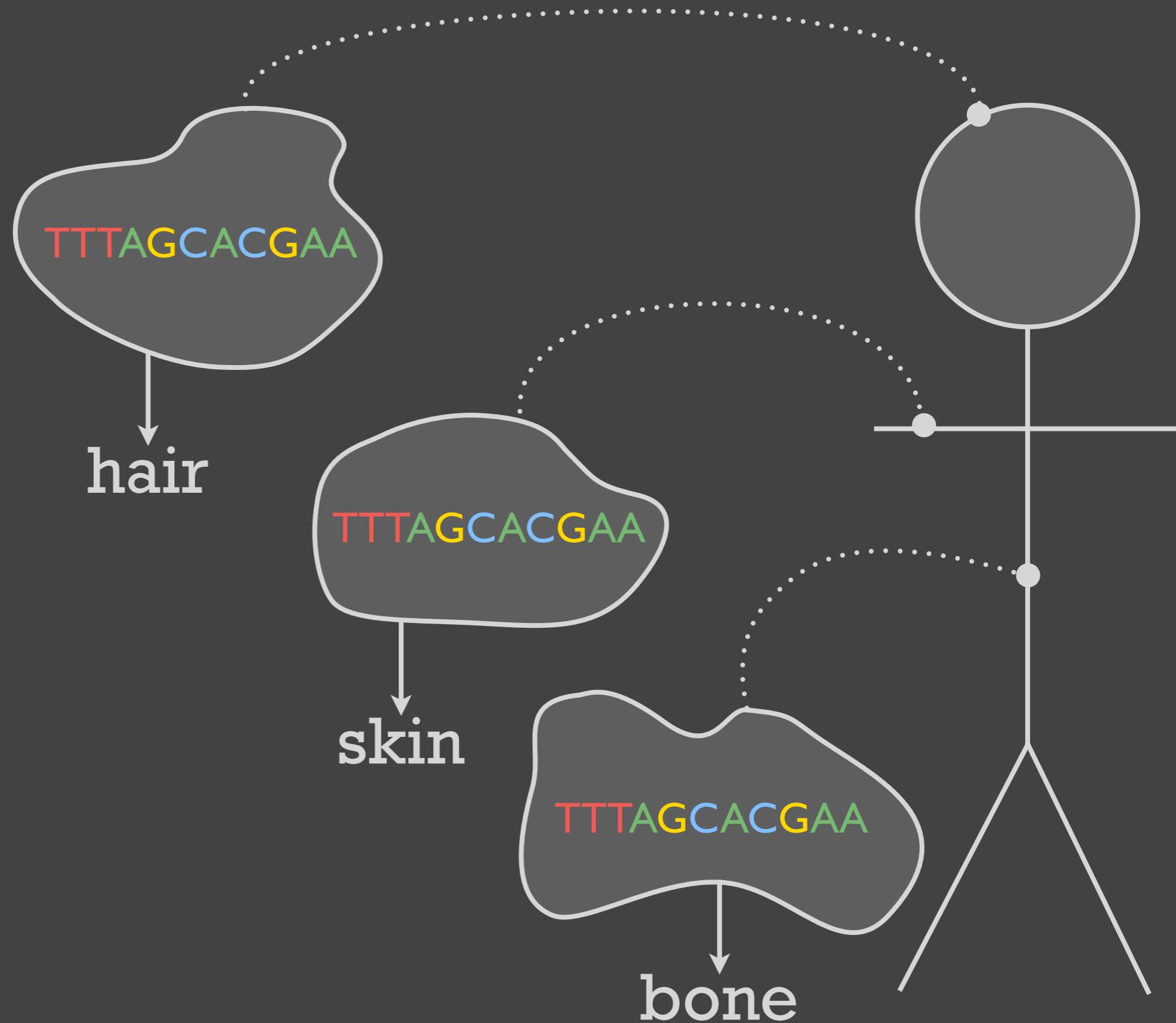
¹ Harvard University

² University of British Columbia

³ Harvard Medical School

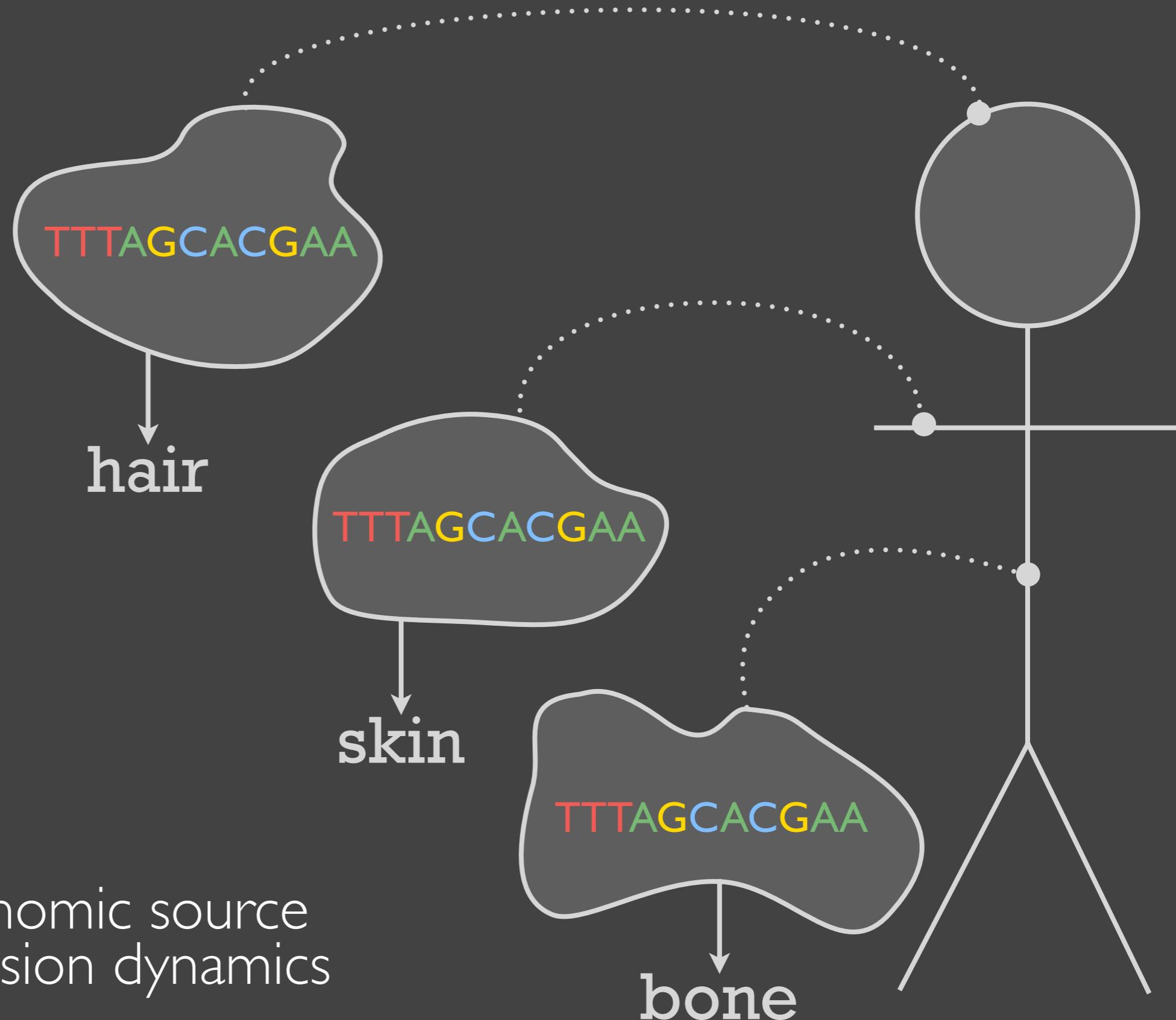






grand challenge in biology:

understand genomic source
of gene expression dynamics



collaborators: DePace lab at Harvard Medical School

model species: fruit fly



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scientific goal: link changes in the regulatory part of the genome to species variation



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requires: characterize differences in gene expression patterns between species



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process

two year collaboration

two early prototype systems

feedback from six biologists

informal interviews, emails

one day a week in biology lab

tool deployed

currently used several times a week

data & tool & tasks

summaries & groups

encodings & interaction

conclusions

gene expression is

gene expression is ...

*... the measured level of how
much a gene is on or off.*

gene expression is ...

*... the measured level of how
much a gene is on or off.*

... a single quantitative value.

0.2

gene expression is ...

... the measured level of how much a gene is on or off.

... a single quantitative value.



collaborators measure it ...

... for multiple time points.

gene expression is ...

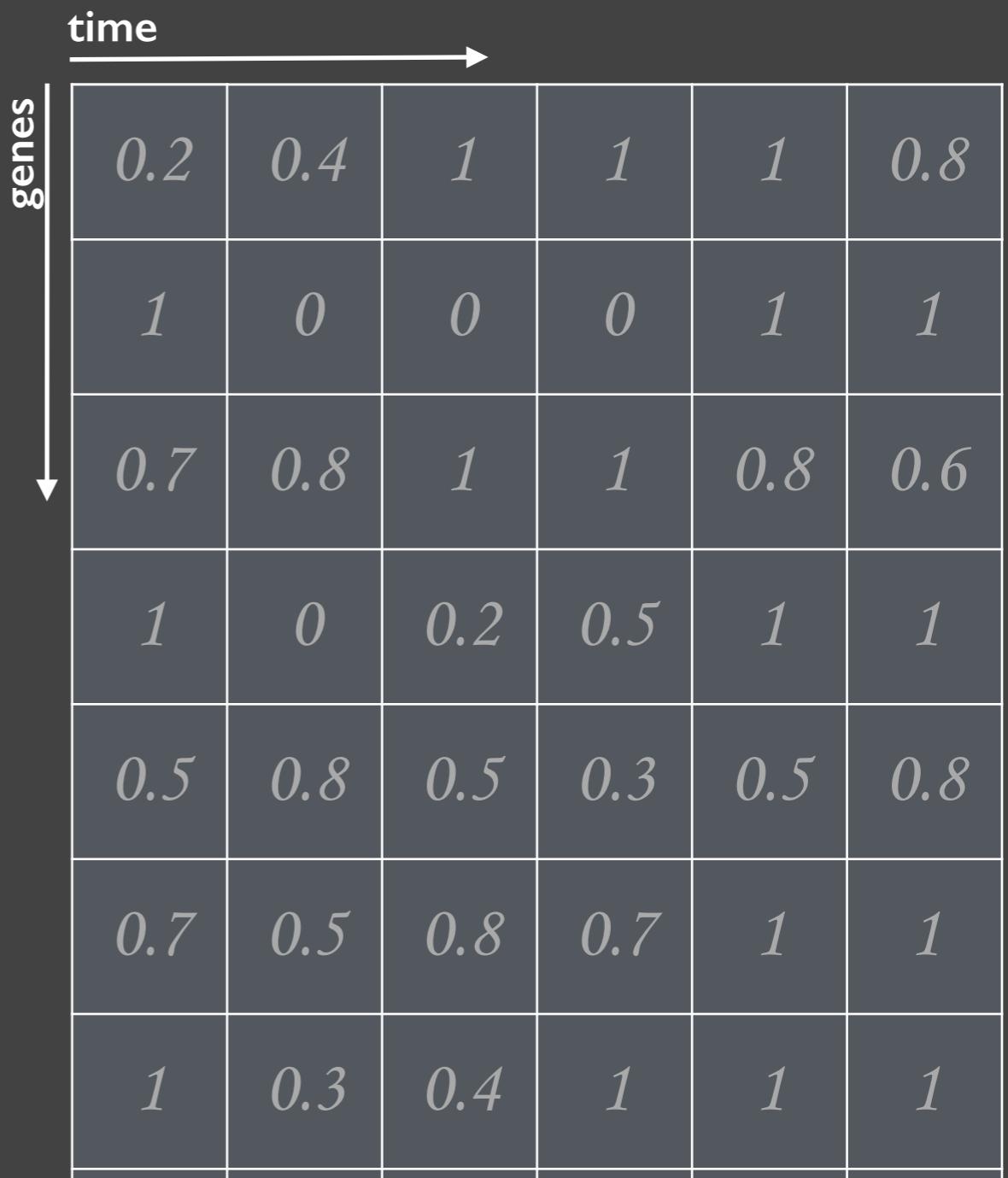
... the measured level of how much a gene is on or off.

... a single quantitative value.

collaborators measure it ...

... for multiple time points.

... for multiple genes.





gene expression is ...

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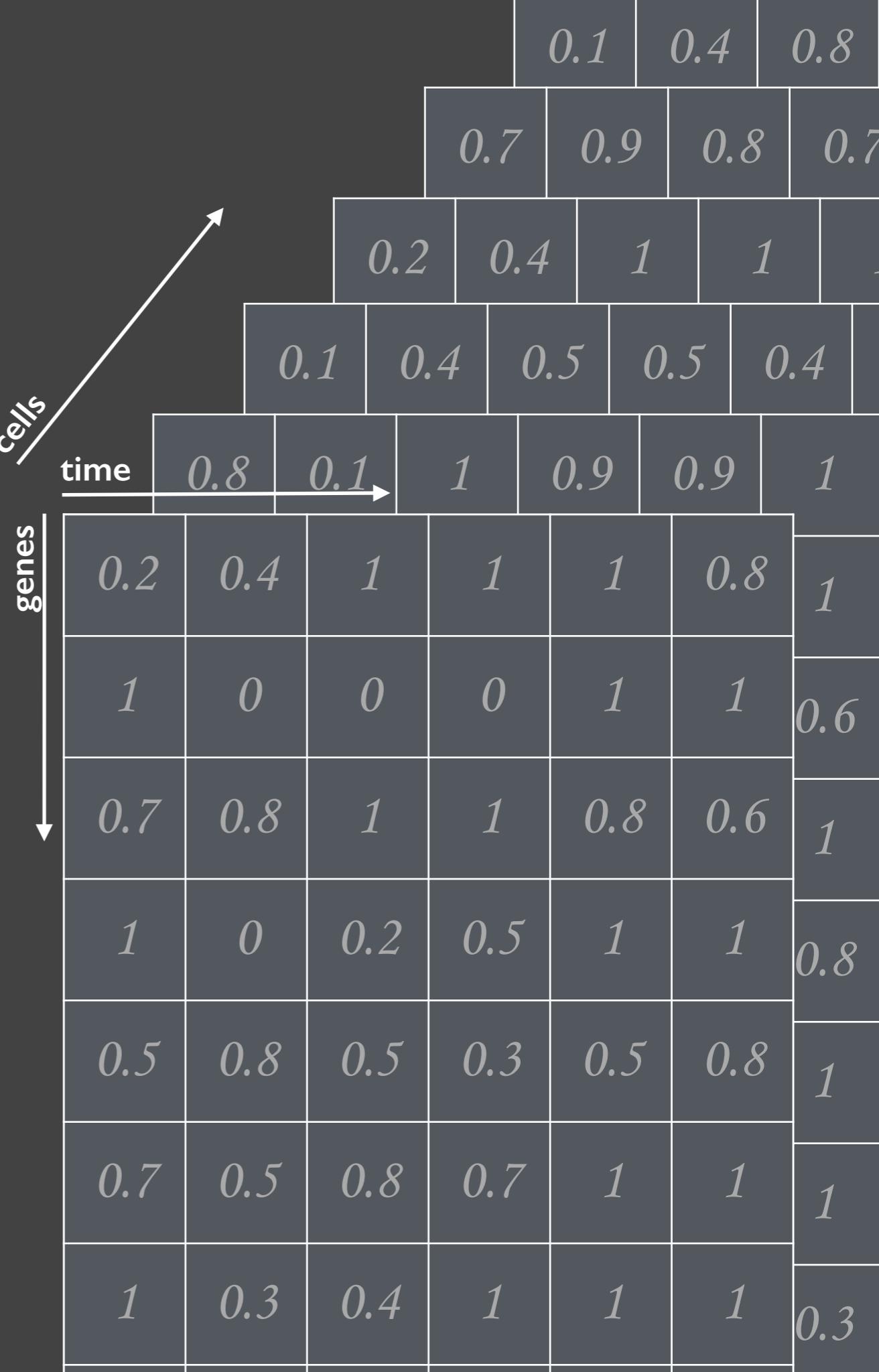
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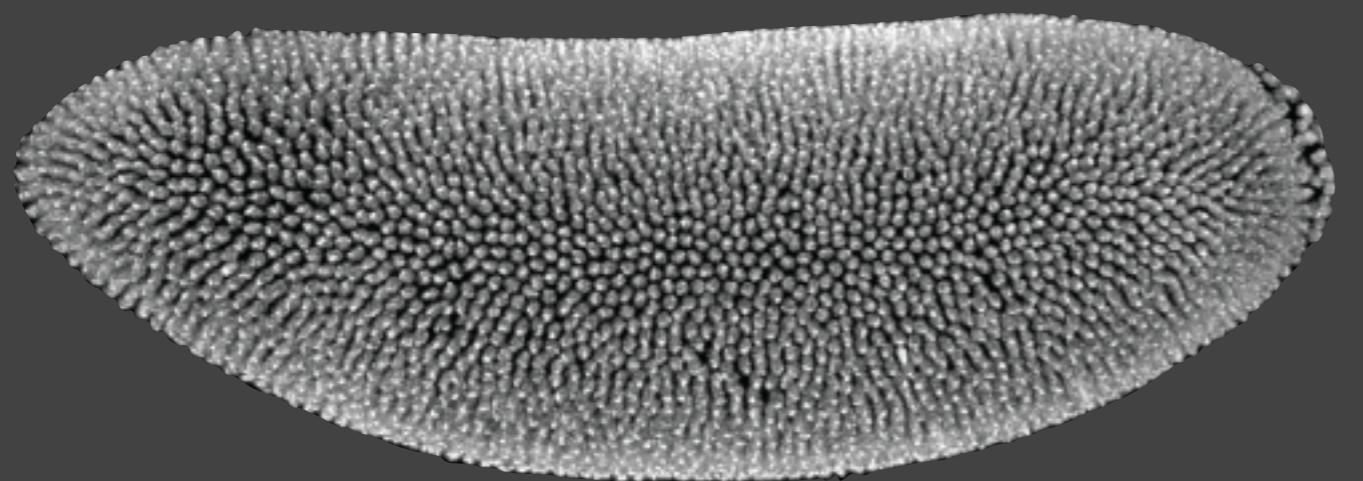
collaborators measure it ...

... for multiple time points.

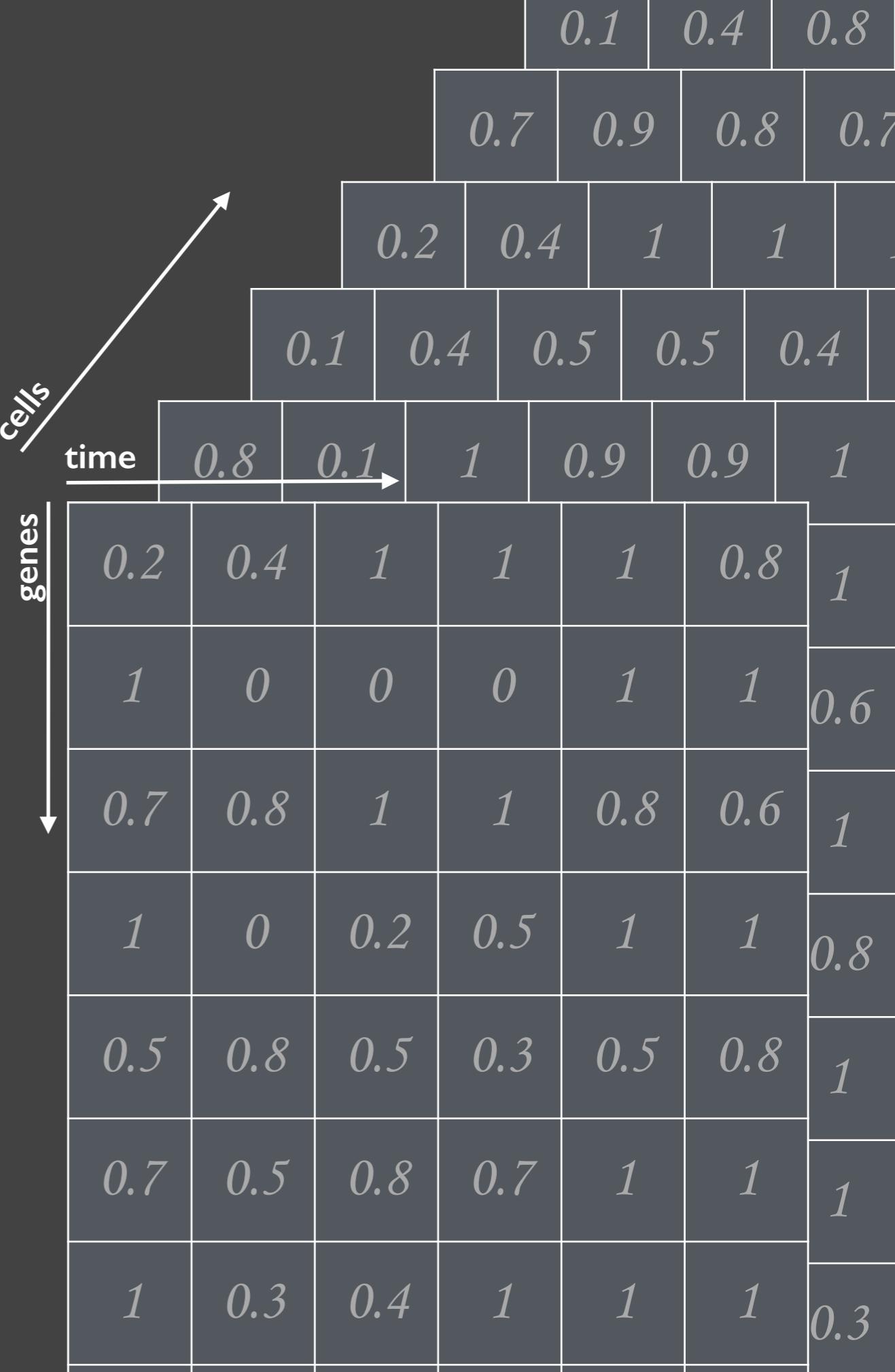
... for multiple genes.

... in many cells.

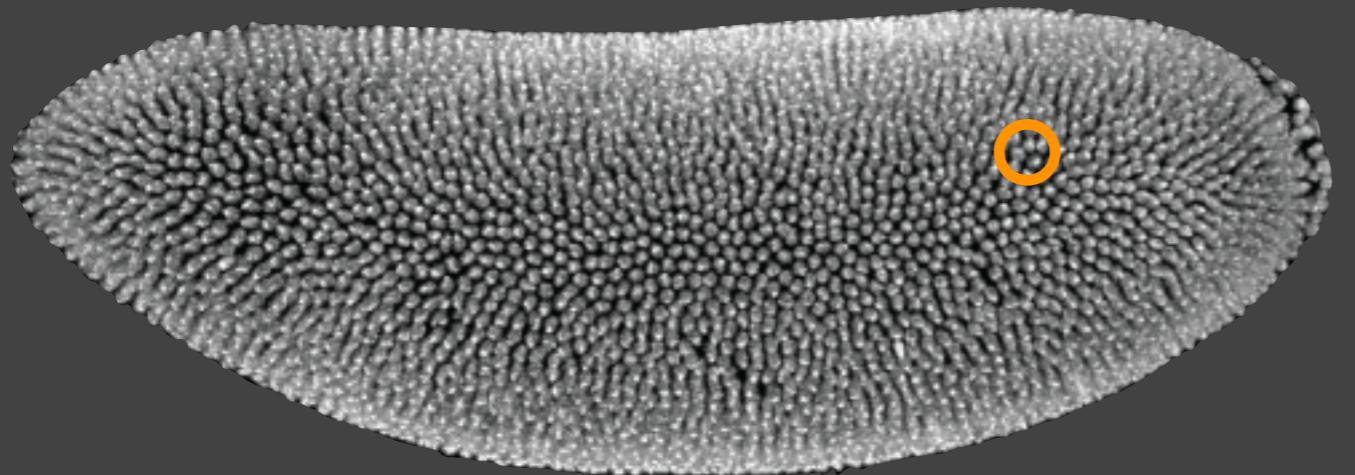




**measurements taken for every
cell in a single fruit fly embryo.**



**measurements taken for every
cell in a single fruit fly embryo.**



**correlate gene expression with
spatial location.**



virtual embryo

virtual embryo

several thousand cells
 $[5,000 \pm 1,000]$

virtual embryo

several thousand cells
 $[5,000 \pm 1,000]$

each cell has:

virtual embryo

several thousand cells
 $[5,000 \pm 1,000]$

each cell has:
expression profile
[6 time points \times 50 genes]

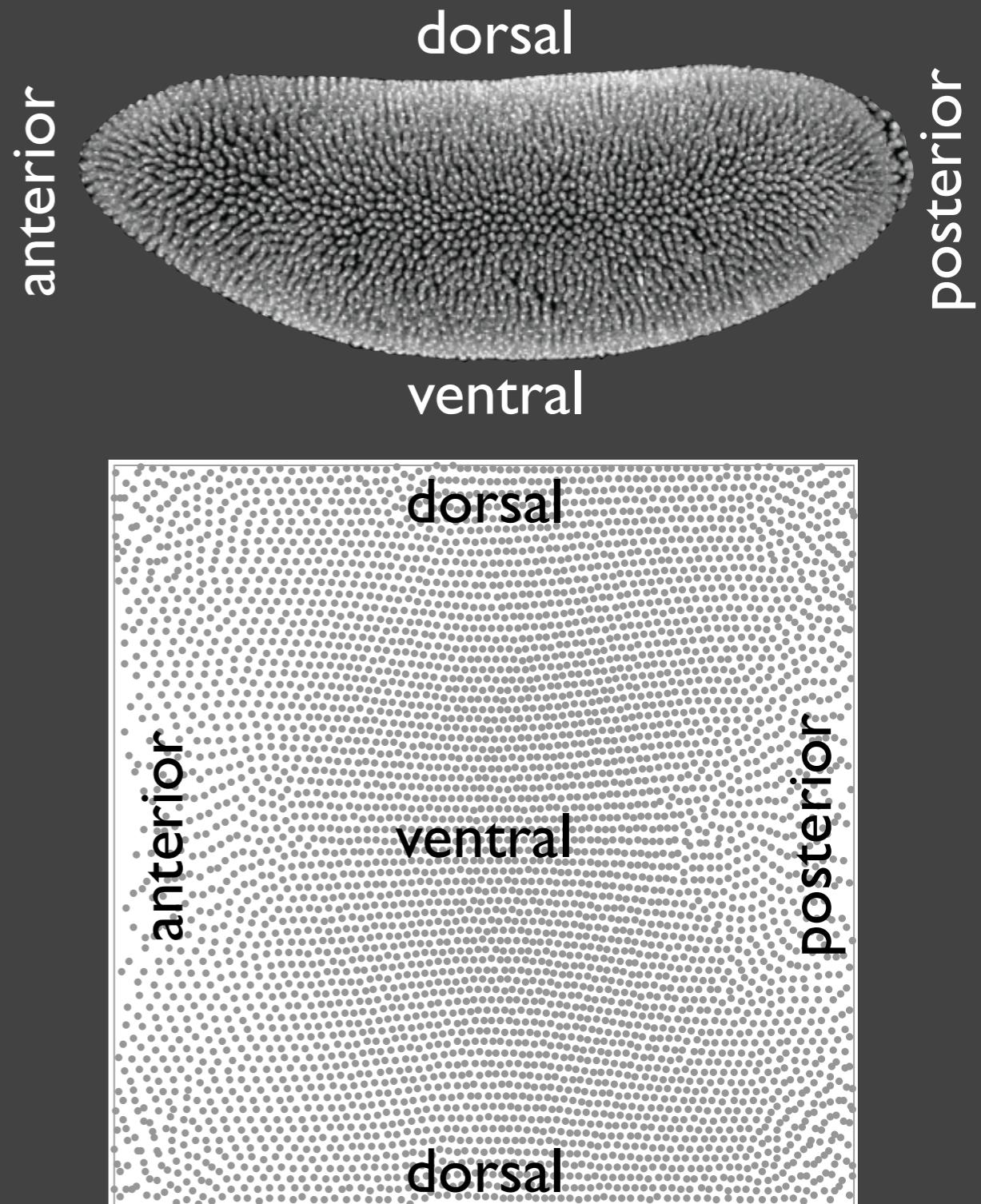
genes	time					
	0.2	0.4	1	1	1	0.8
1	0	0	0	1	1	1
0.7	0.8	1	1	0.8	0.6	
1	0	0.2	0.5	1	1	1
0.5	0.8	0.5	0.3	0.5	0.8	
0.7	0.5	0.8	0.7	1	1	1
1	0.3	0.4	1	1	1	1
0.5	0	0	0.7	0.5	0.3	

virtual embryo

several thousand cells
 $[5,000 \pm 1,000]$

each cell has:
expression profile
[6 time points \times 50 genes]

spatial position
[3D and 2D coordinates]



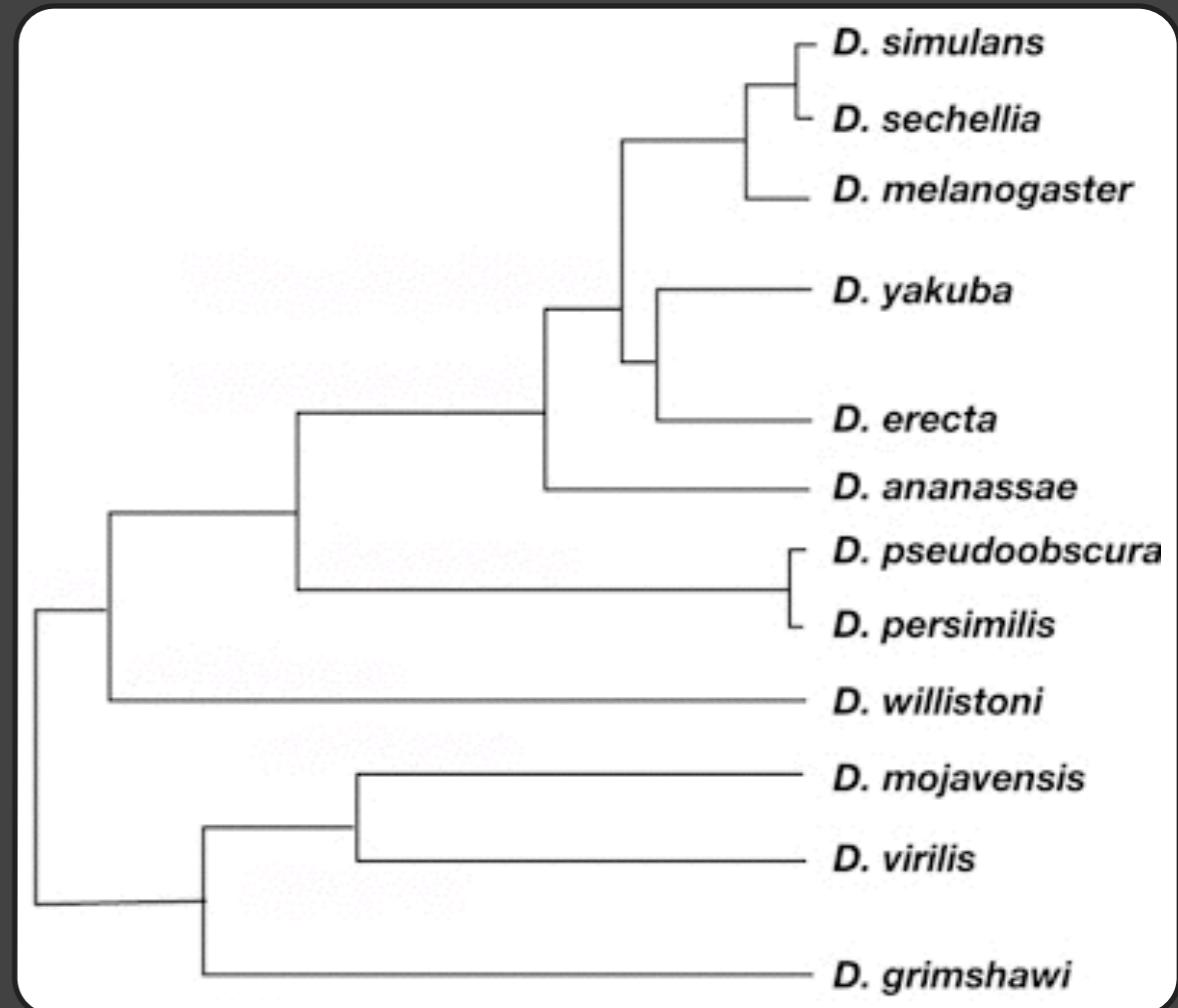
virtual embryo

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expression profile
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spatial position
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12 related species



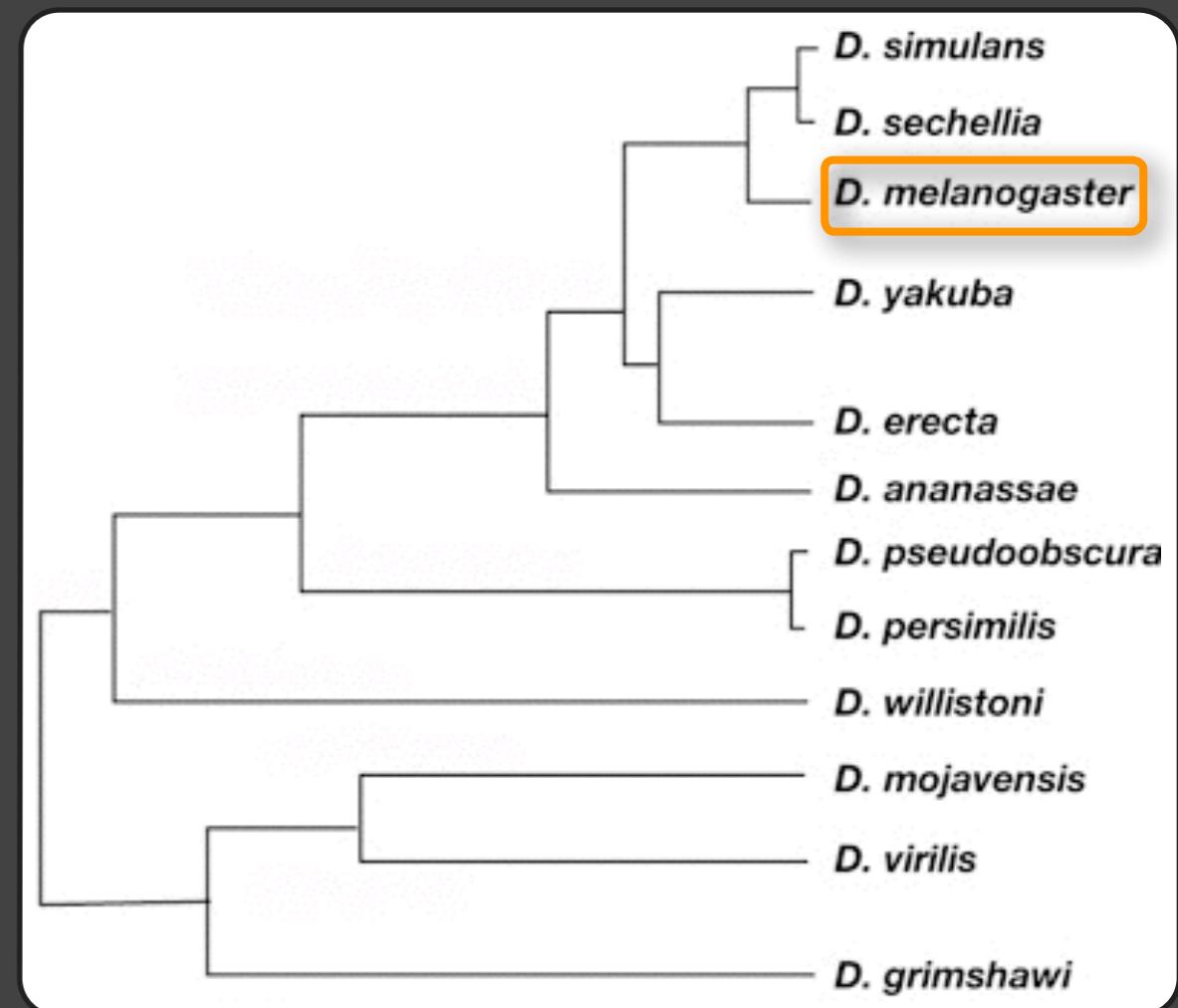
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12 related species
one complete



virtual embryo

several thousand cells
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expression profile

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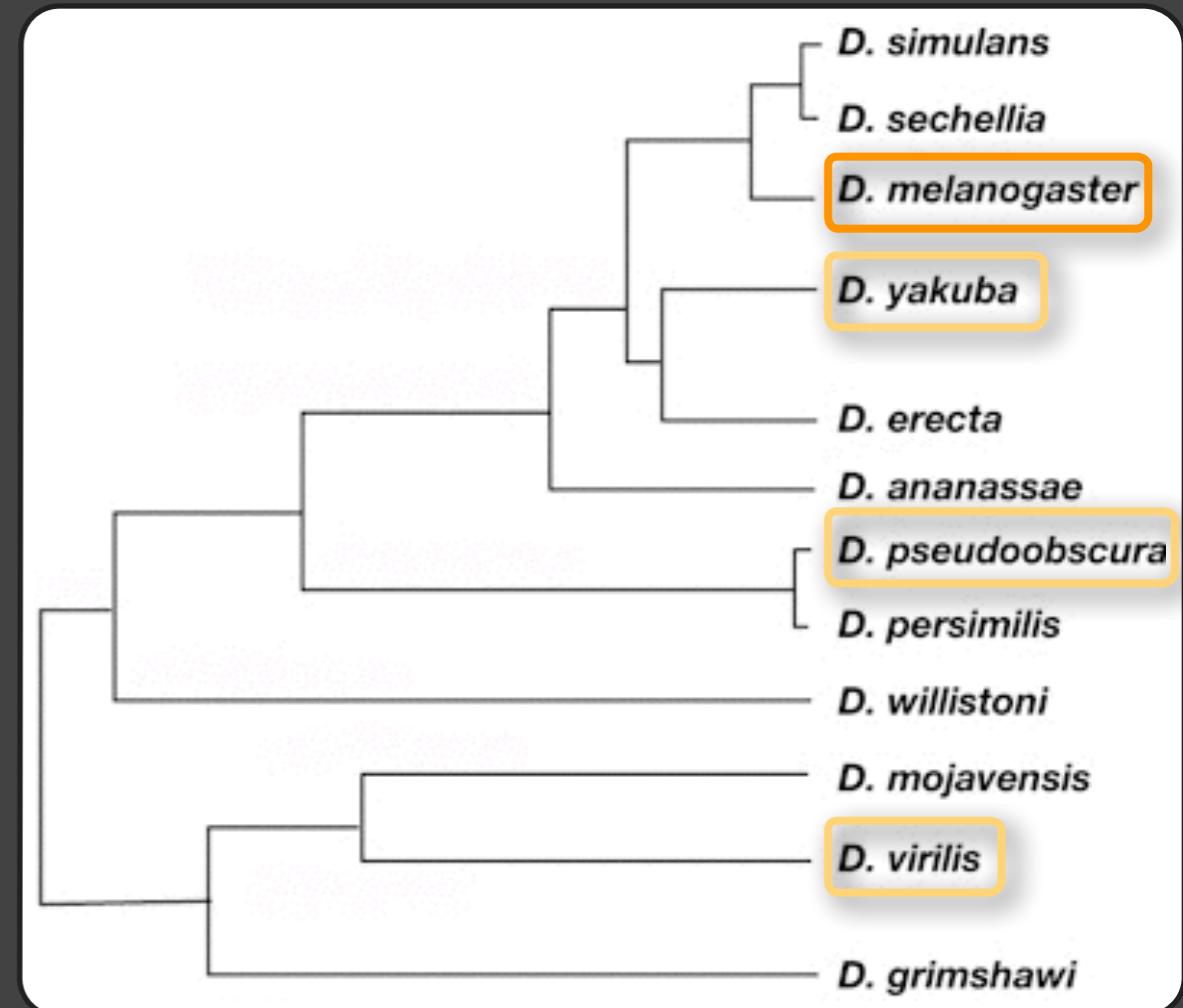
spatial position

[3D and 2D coordinates]

12 related species

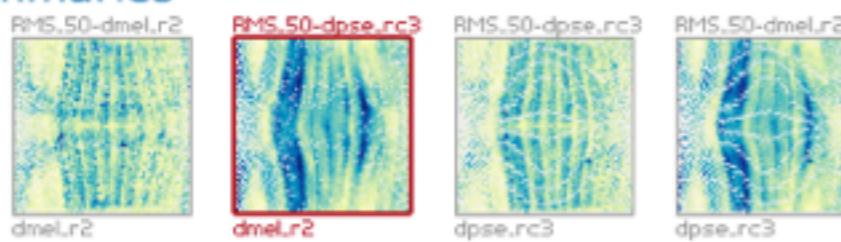
one complete

three preliminary

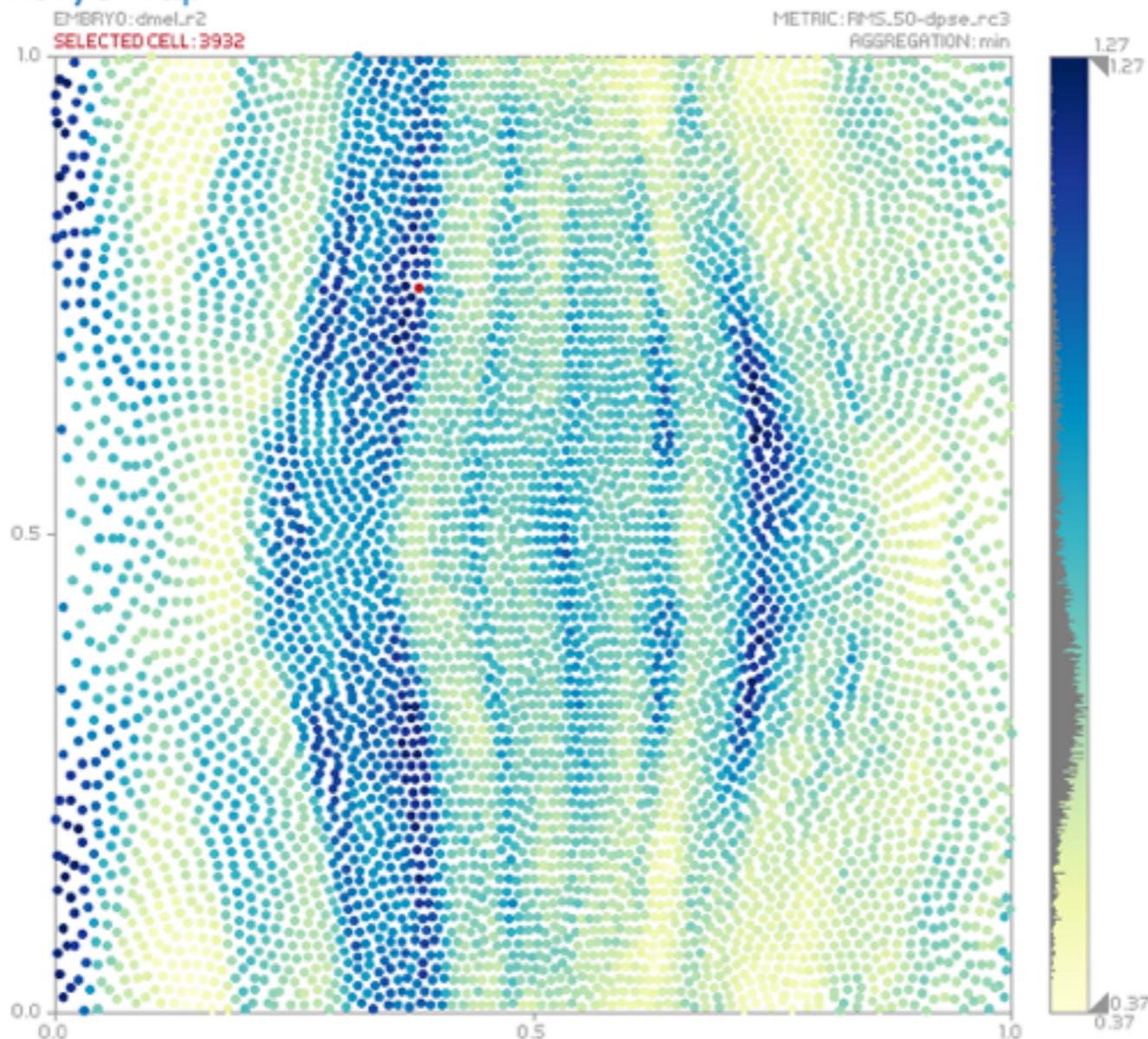


MulteeSum

Summaries

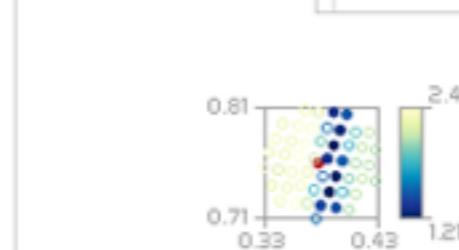


Embryo Map

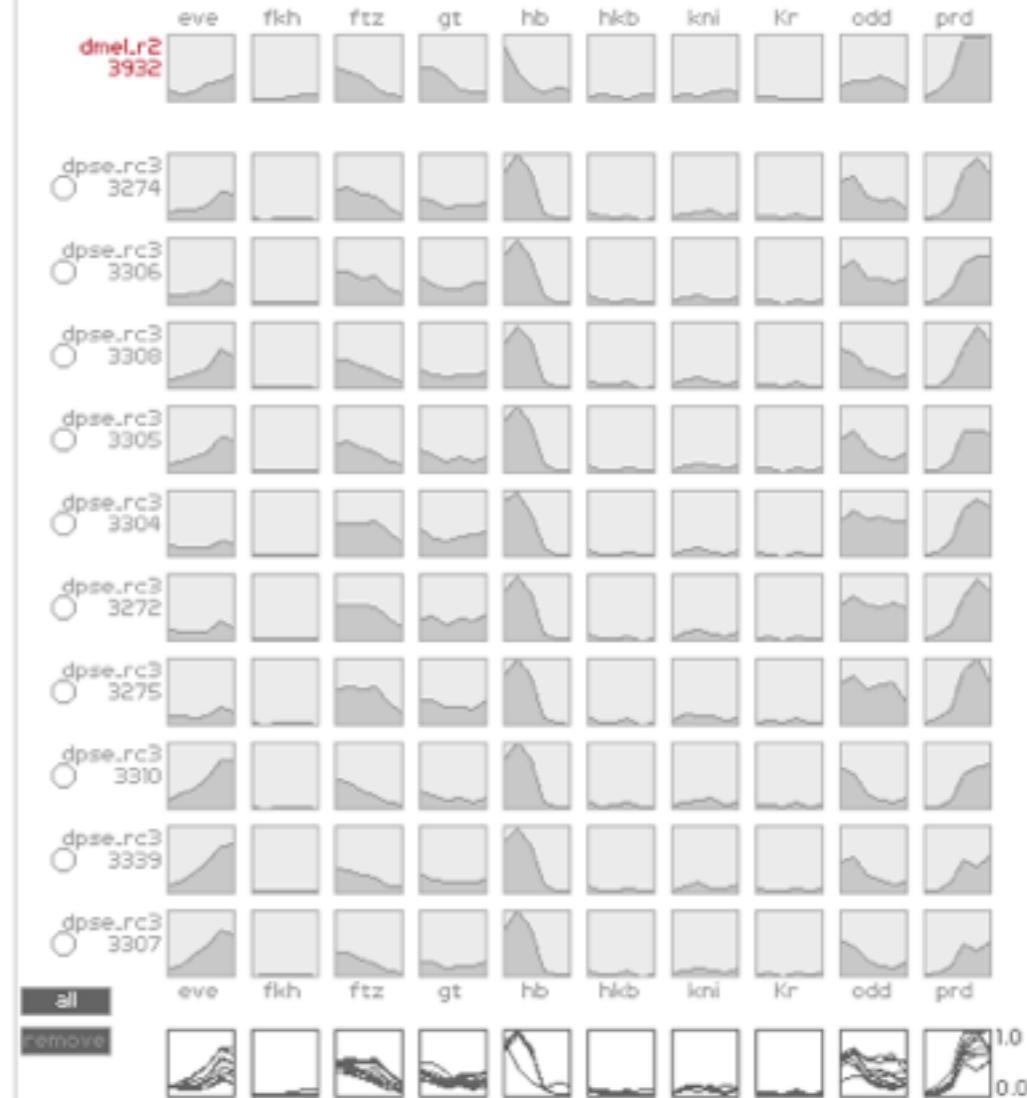
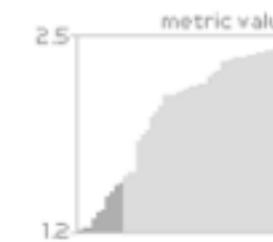


AGGREGATION GROUP

CREATED GROUP

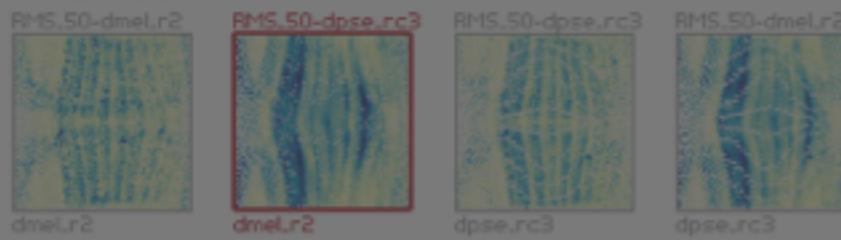


Curvemap

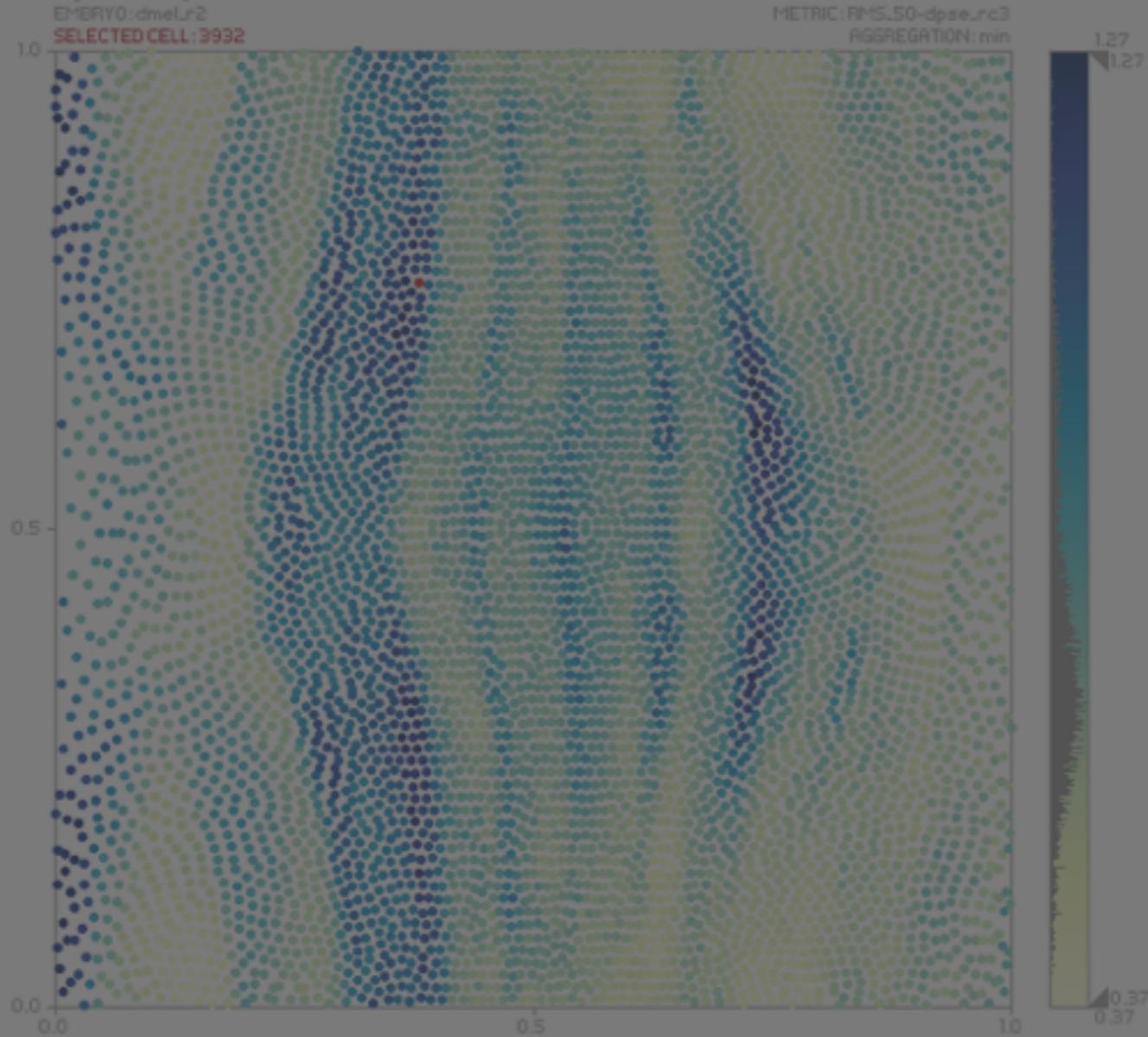


MulteeSum

Summaries

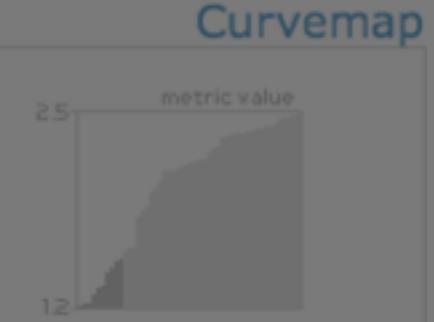
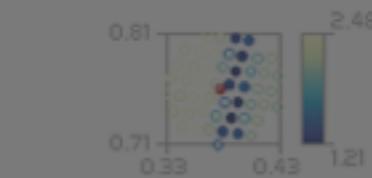


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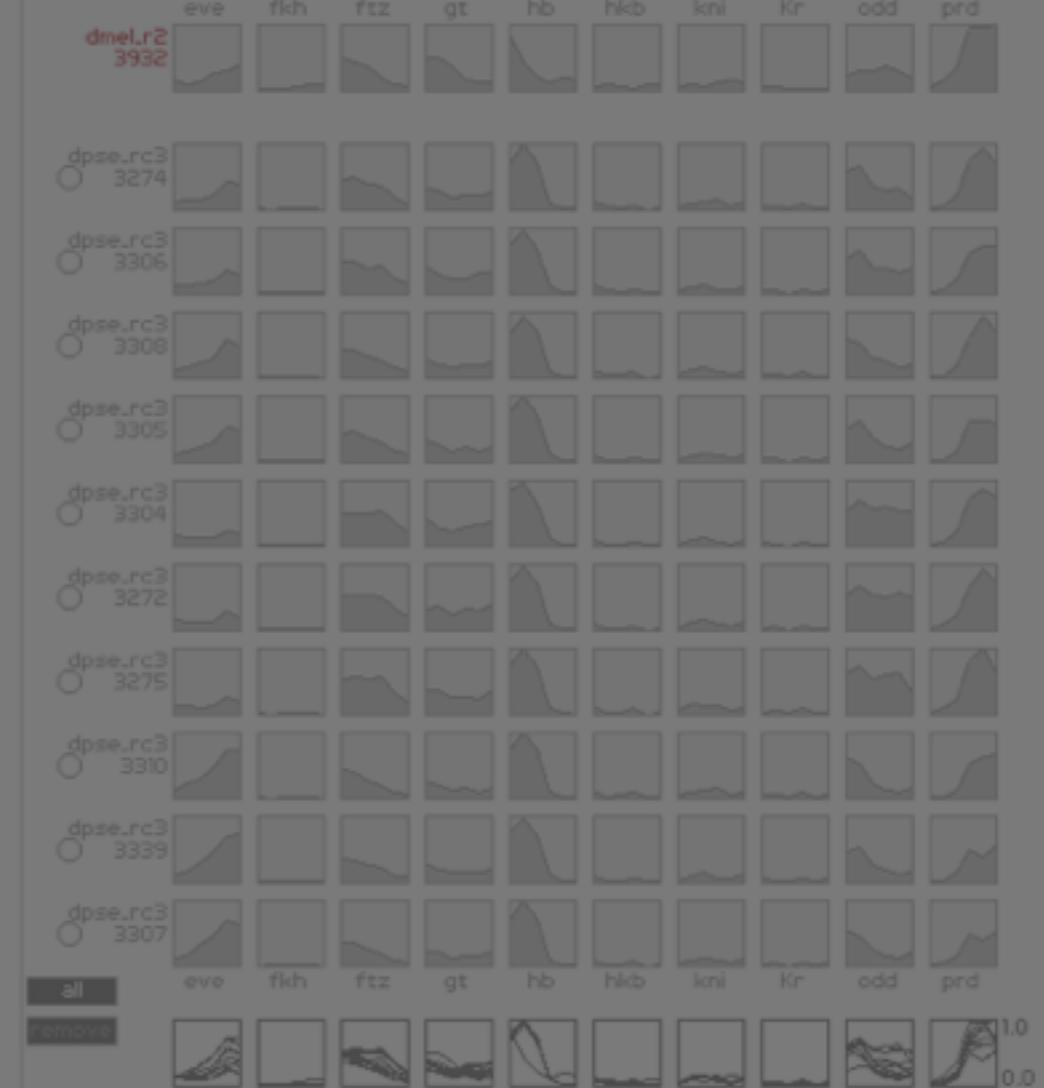


AGGREGATION GROUP

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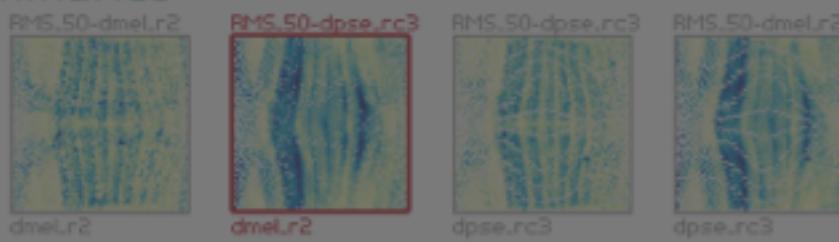


Curvemap



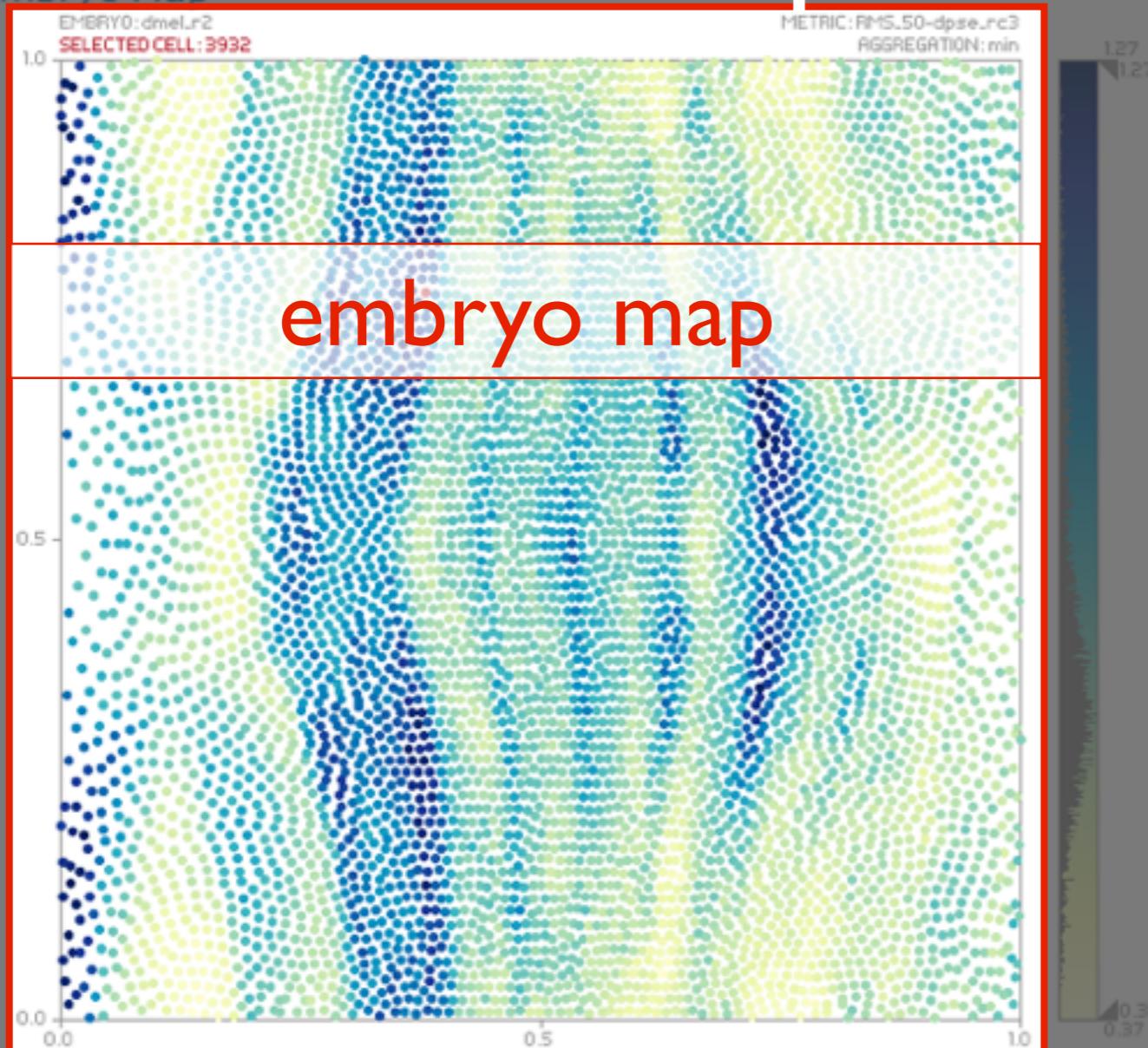
MulteeSum

Summaries



spatial

Embryo Map

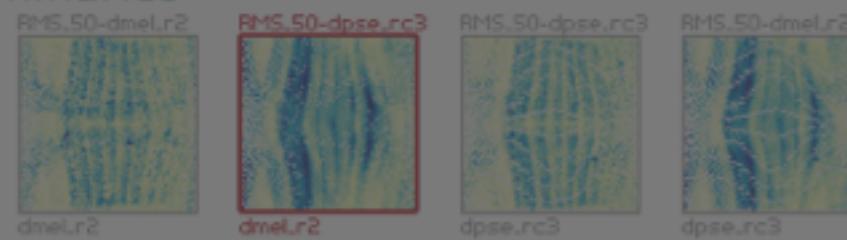


embryo map

Curvemap

MulteeSum

Summaries



Embryo Map

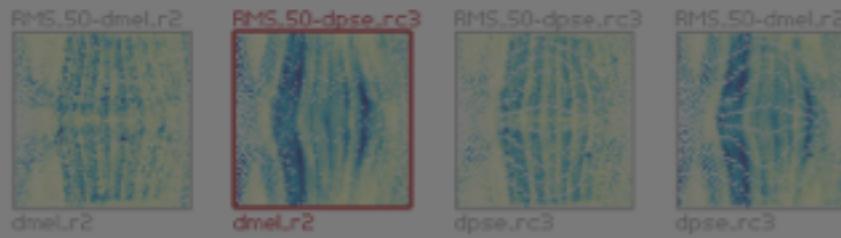
Curvemap

expression

curvemap [Meyer 10]

MulteeSum

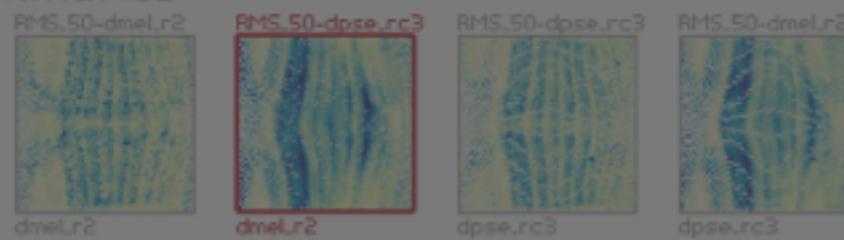
Summaries



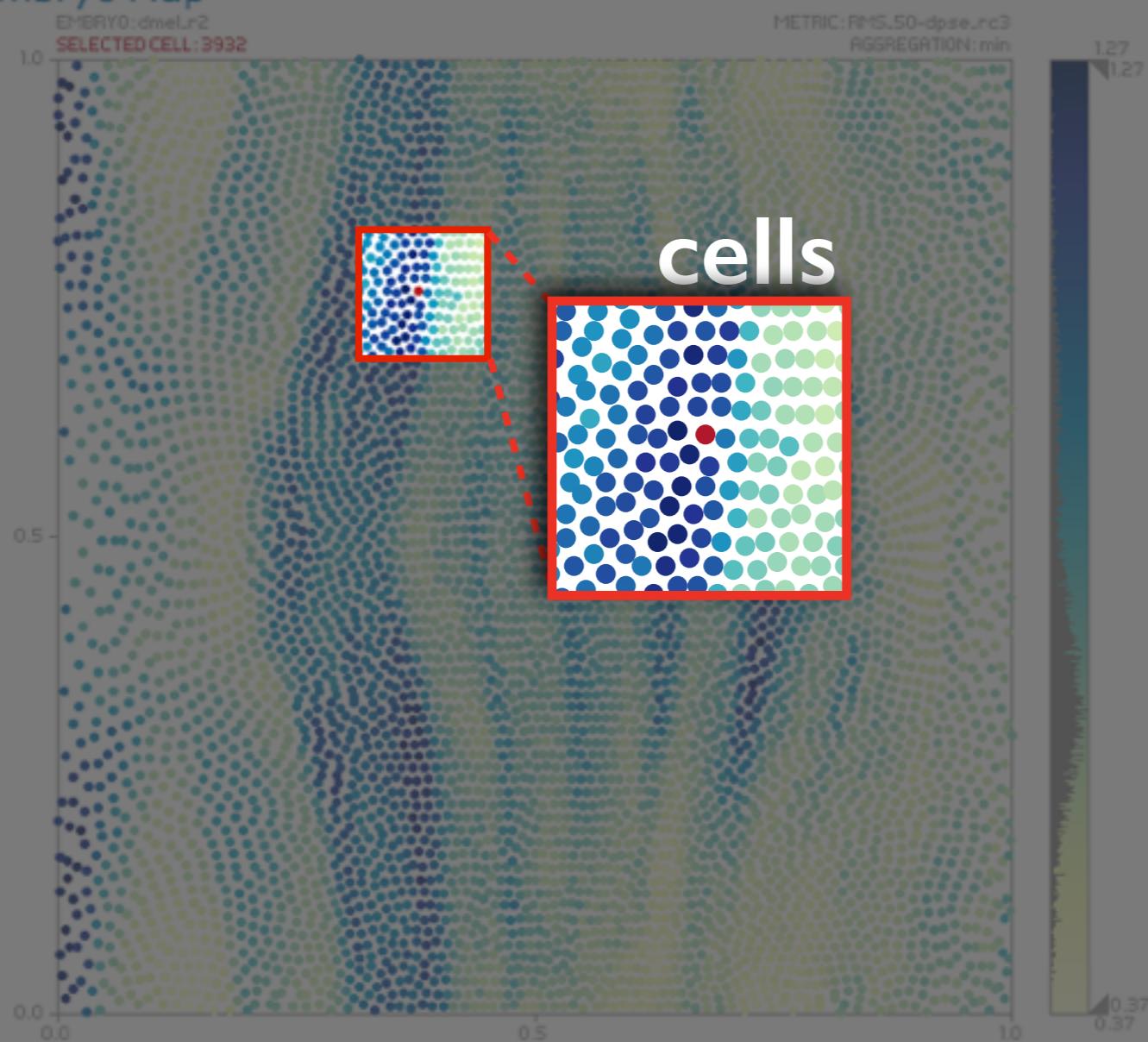
Embryo Map

MulteeSum

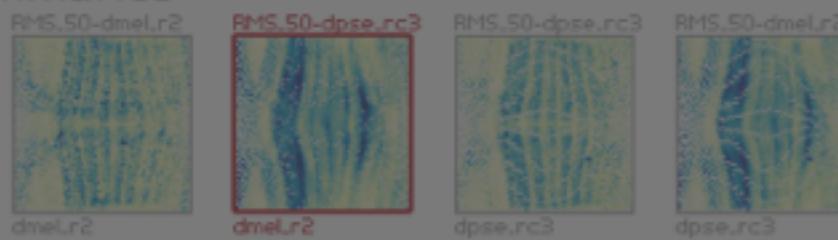
Summaries



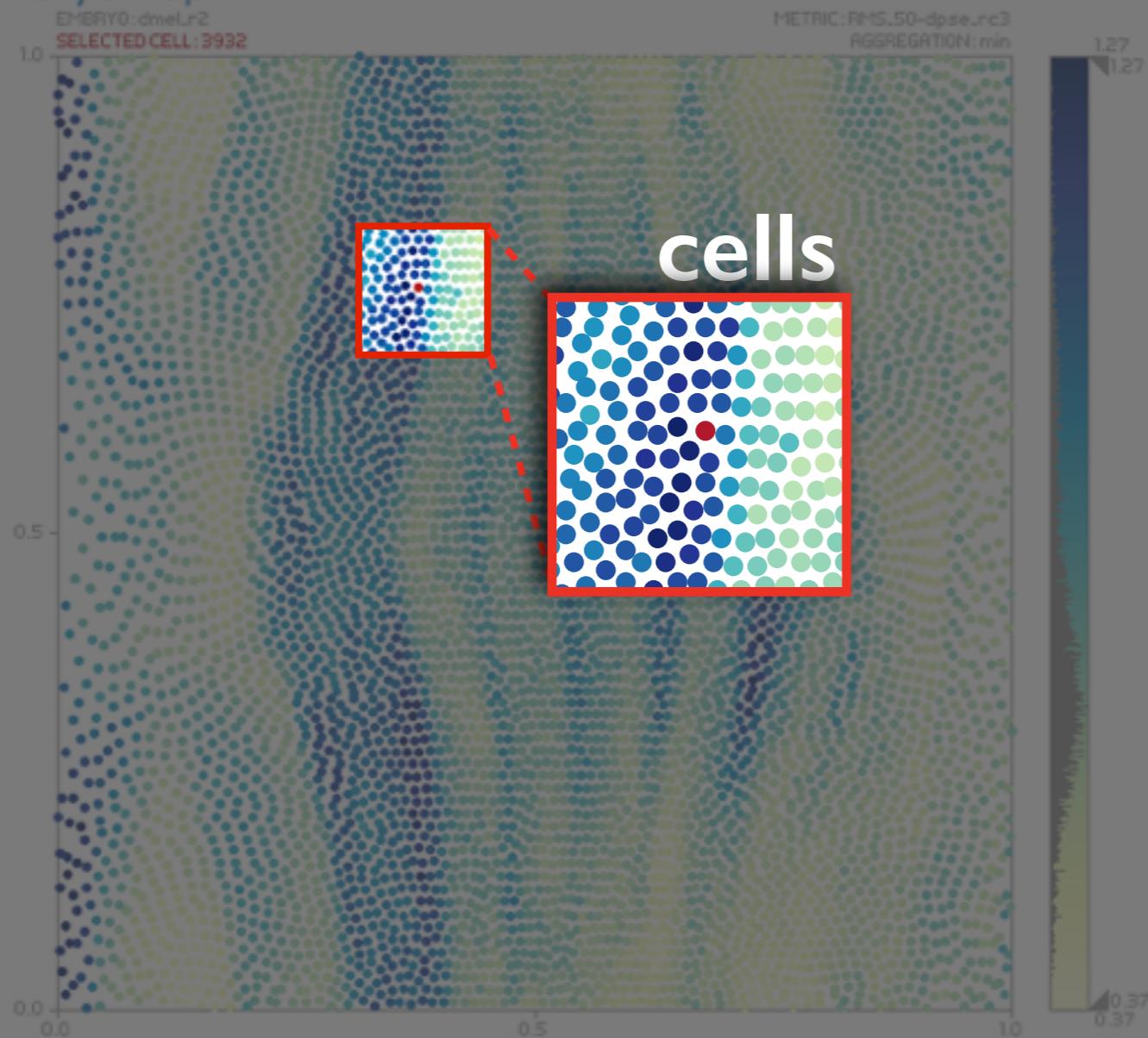
Embryo Map



Summaries



Embryo Map



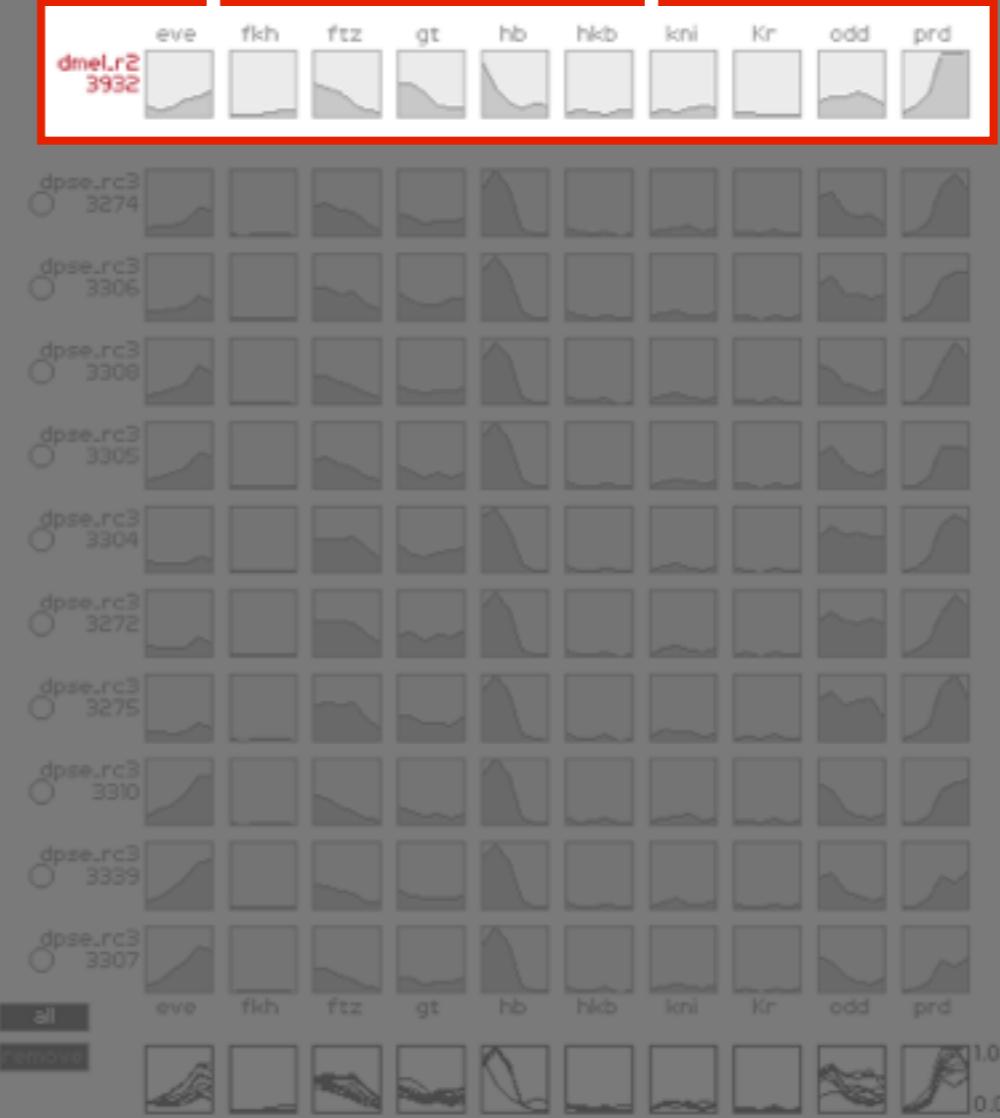
AGGREGATION GROUP

CREATED GROUP

Curvemap

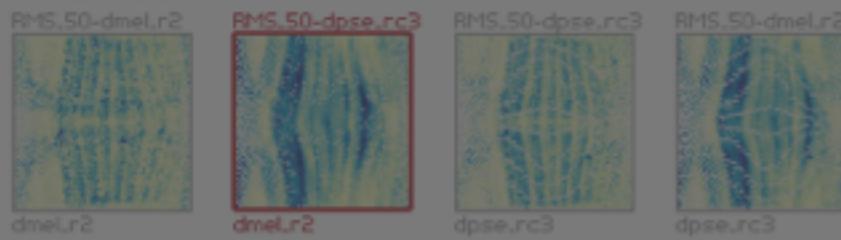


expression profile

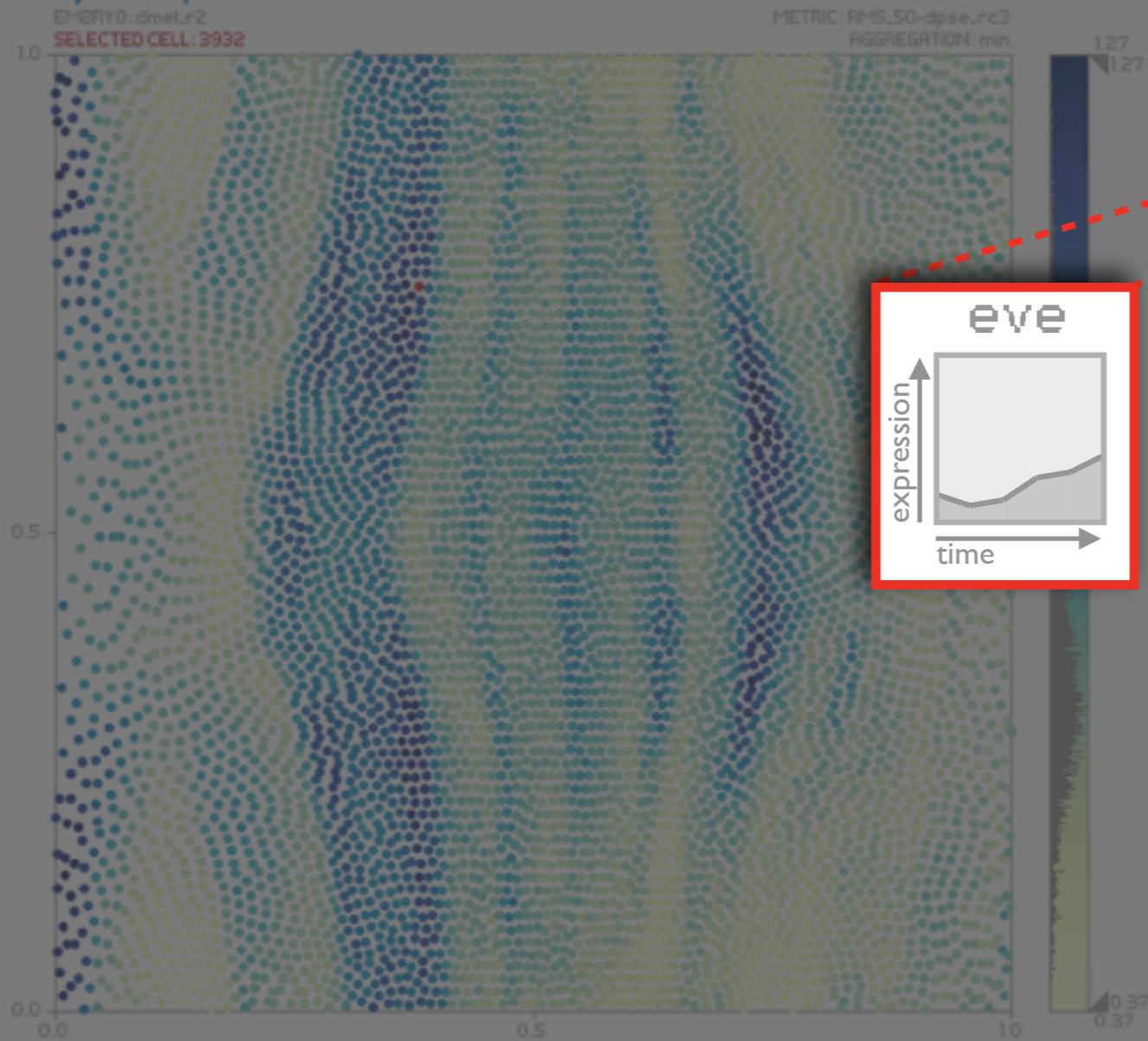


MulteeSum

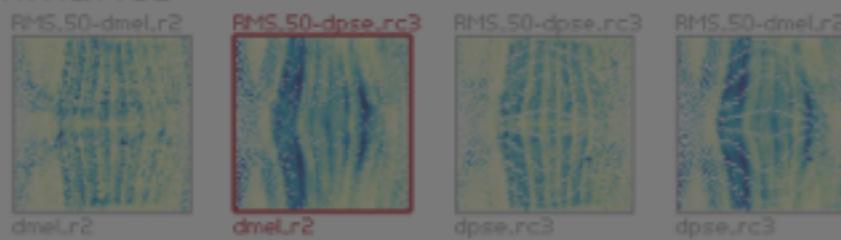
Summaries



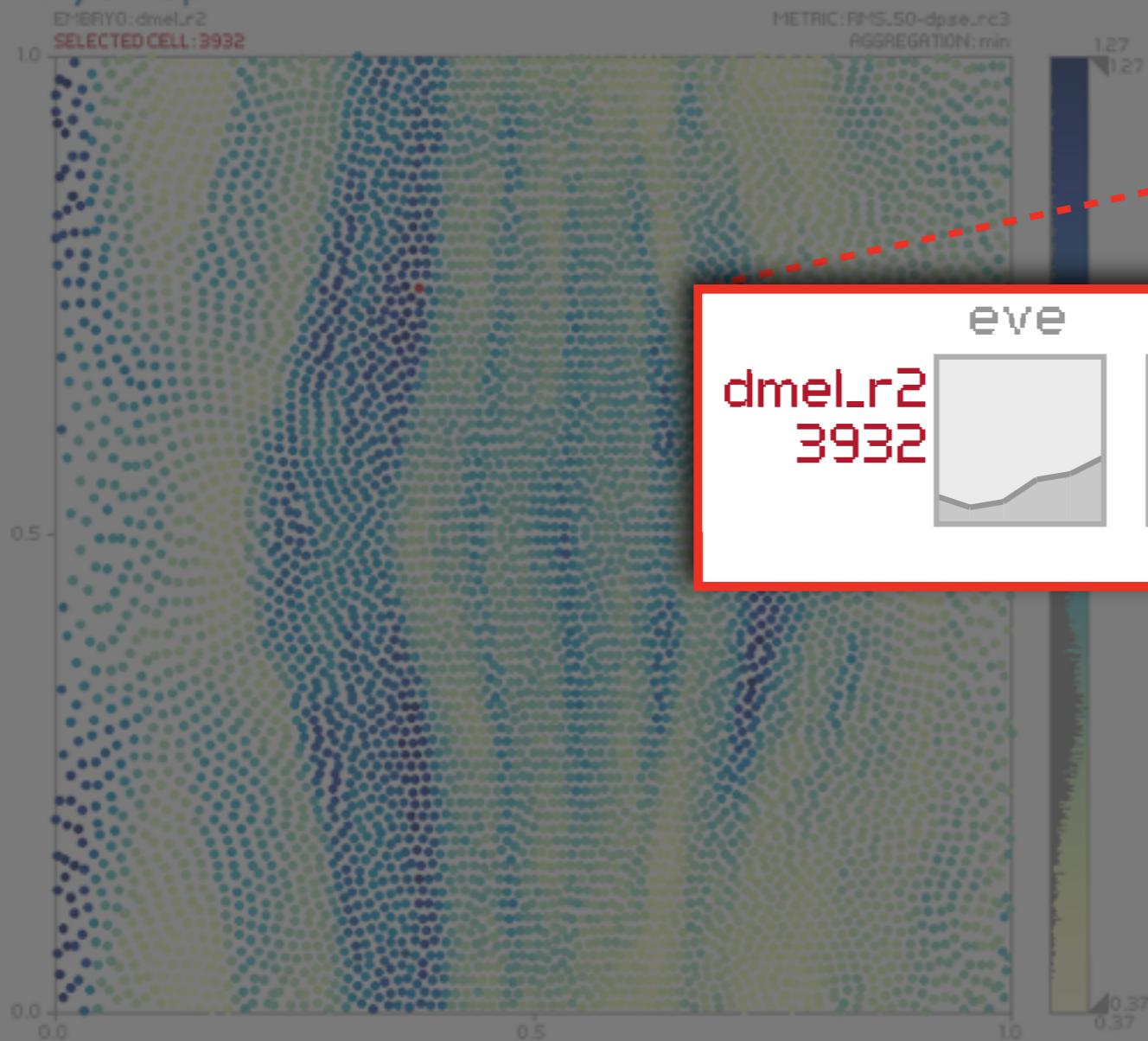
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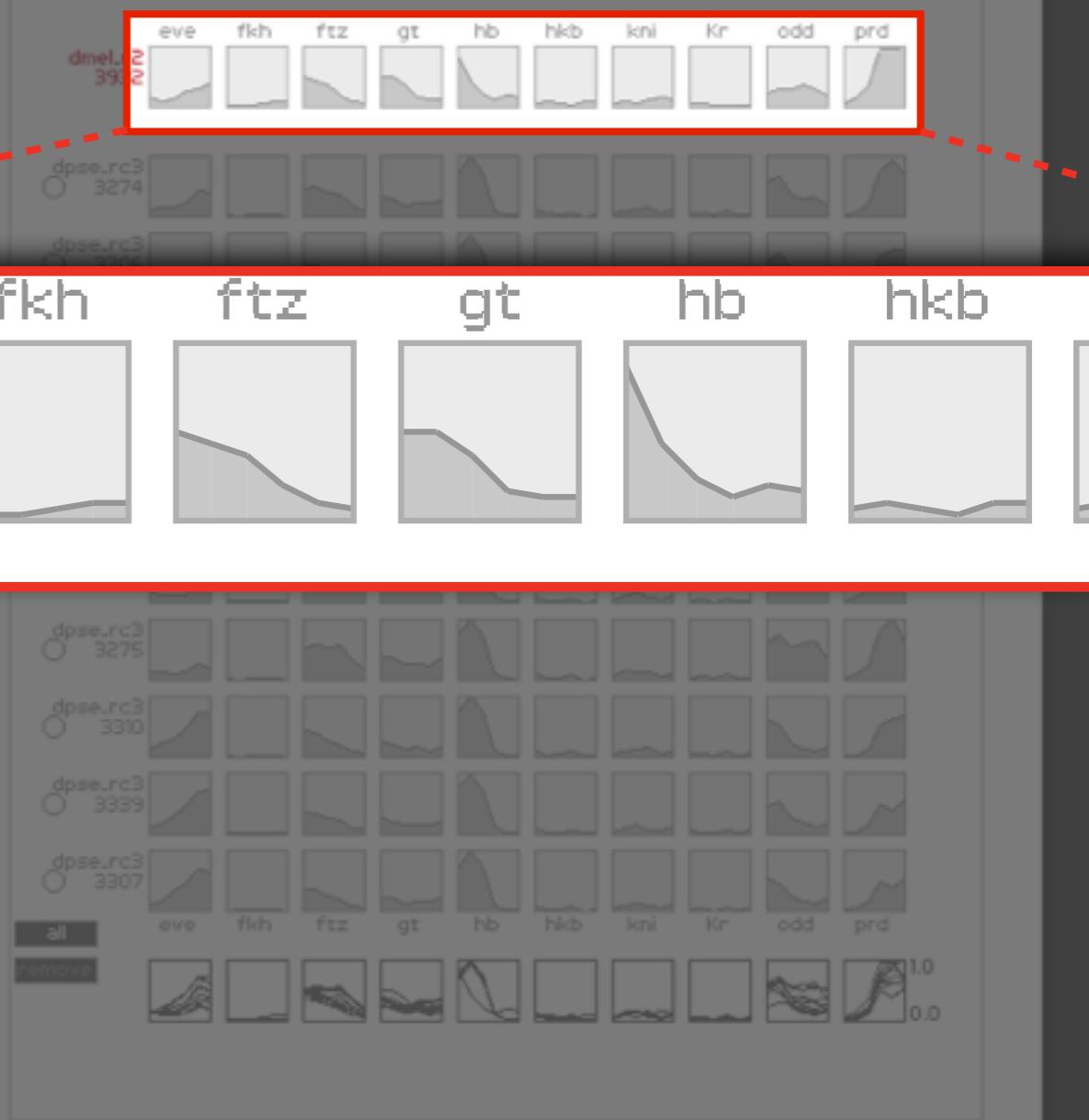
Summaries



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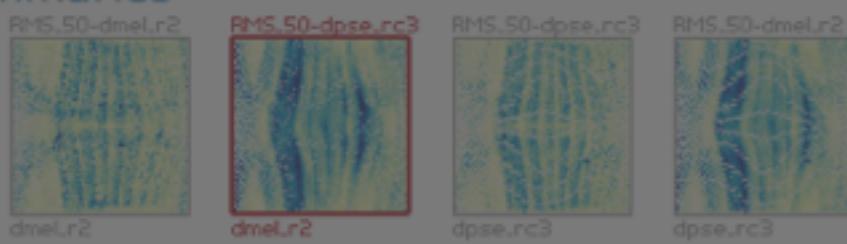


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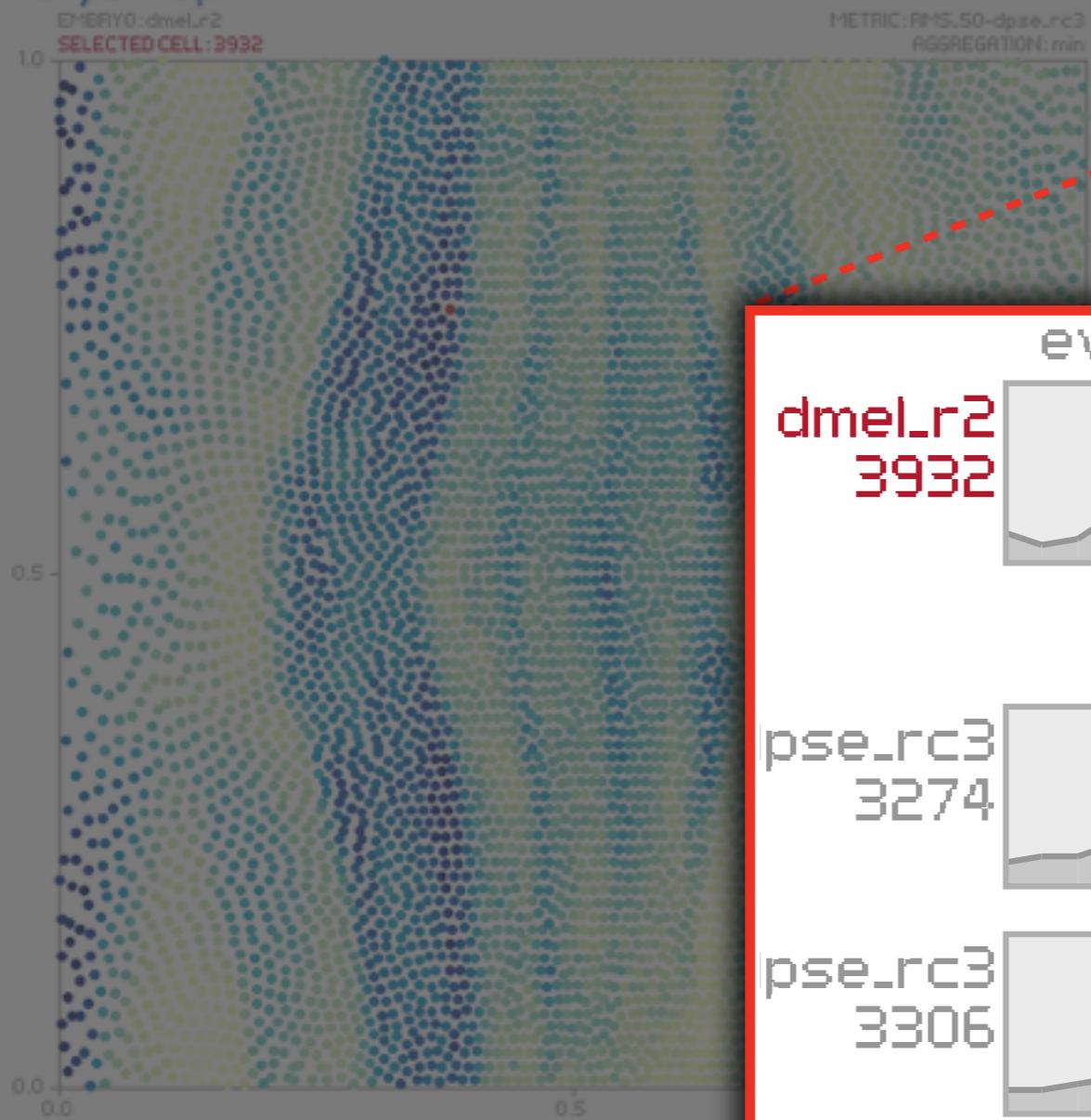


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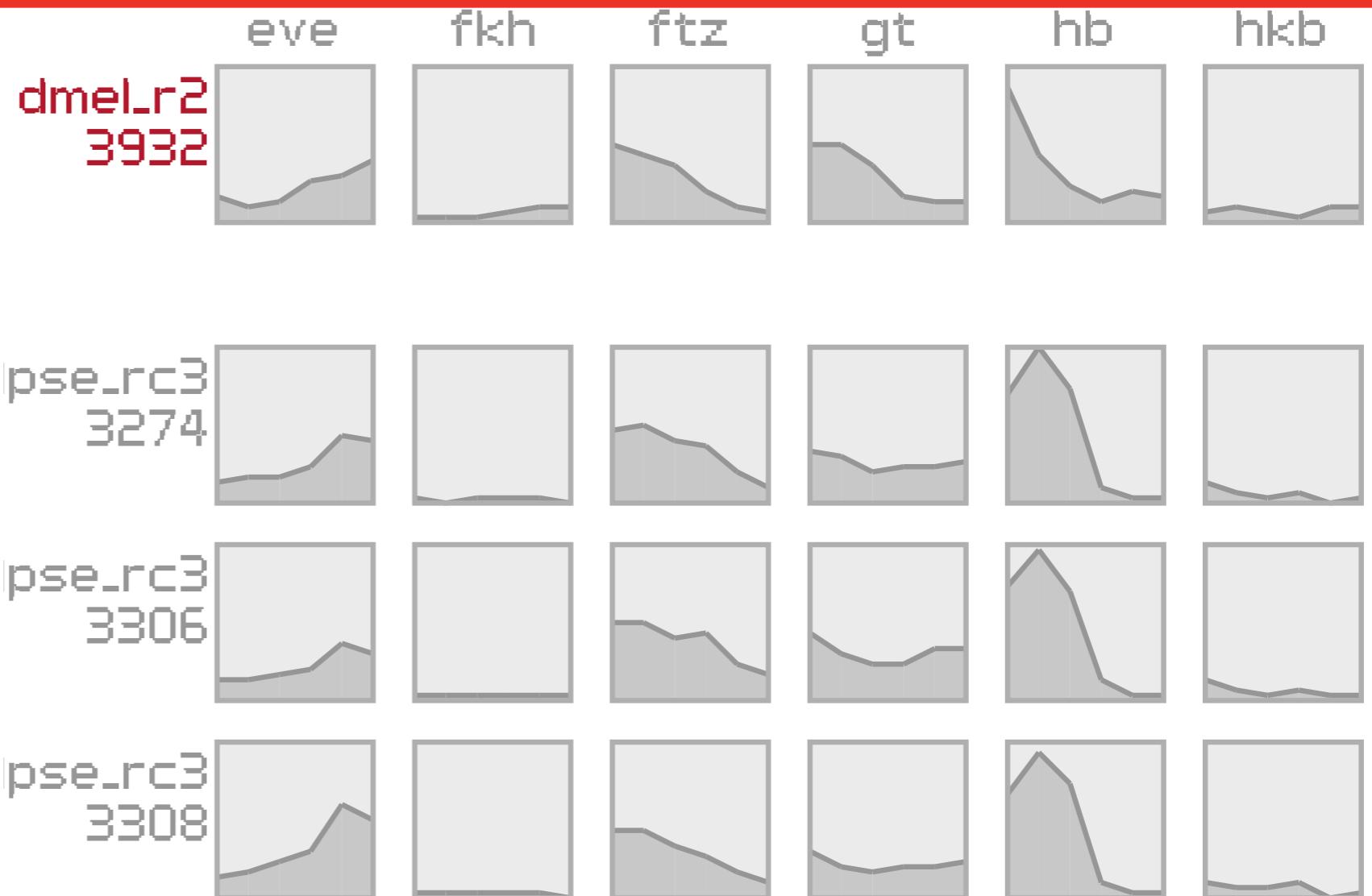
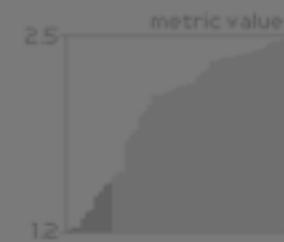
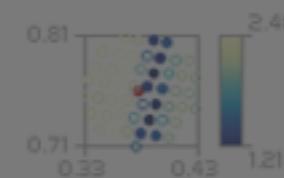
Embryo Map

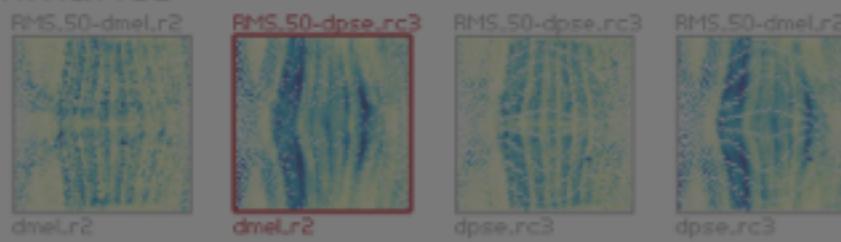
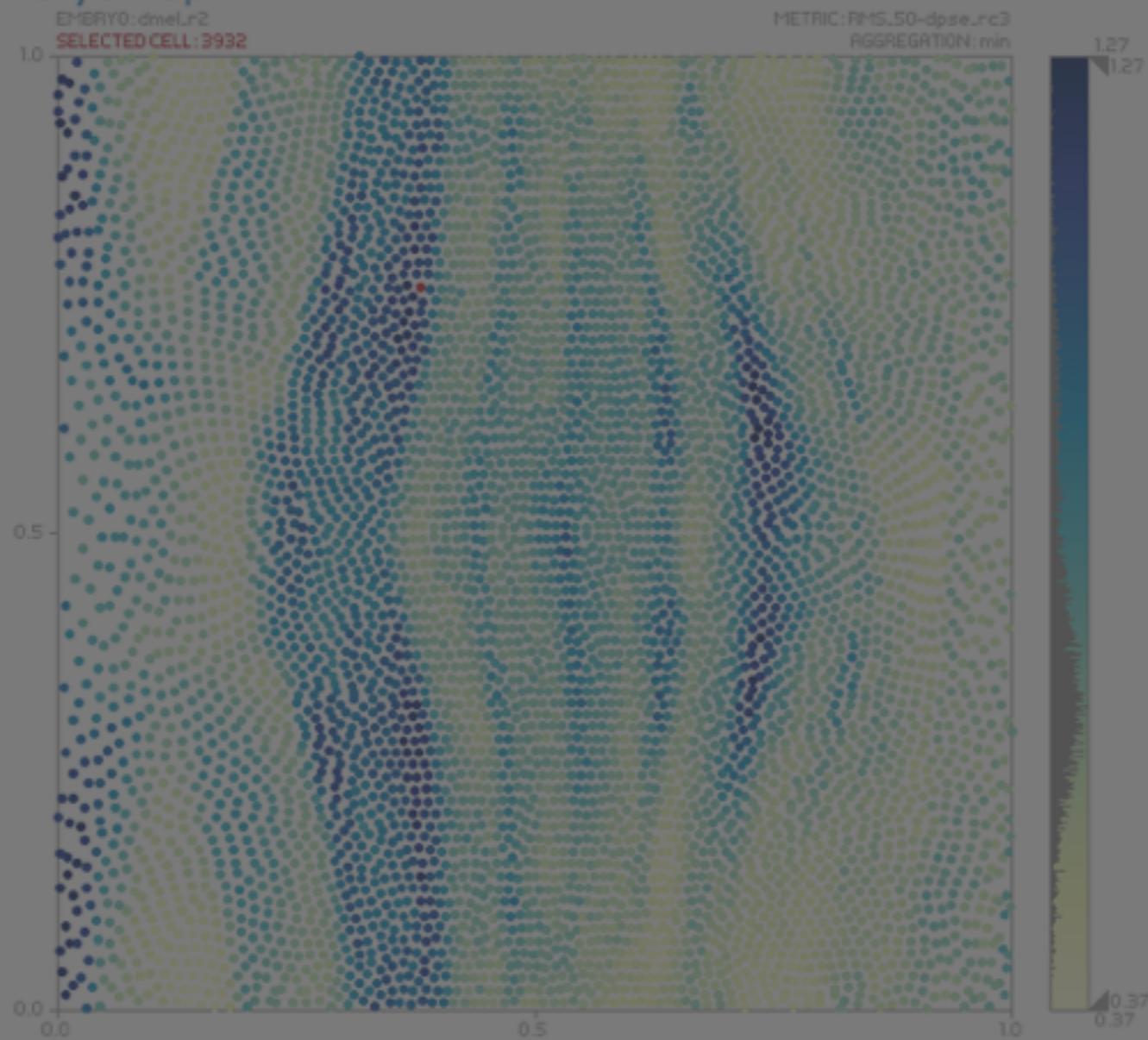
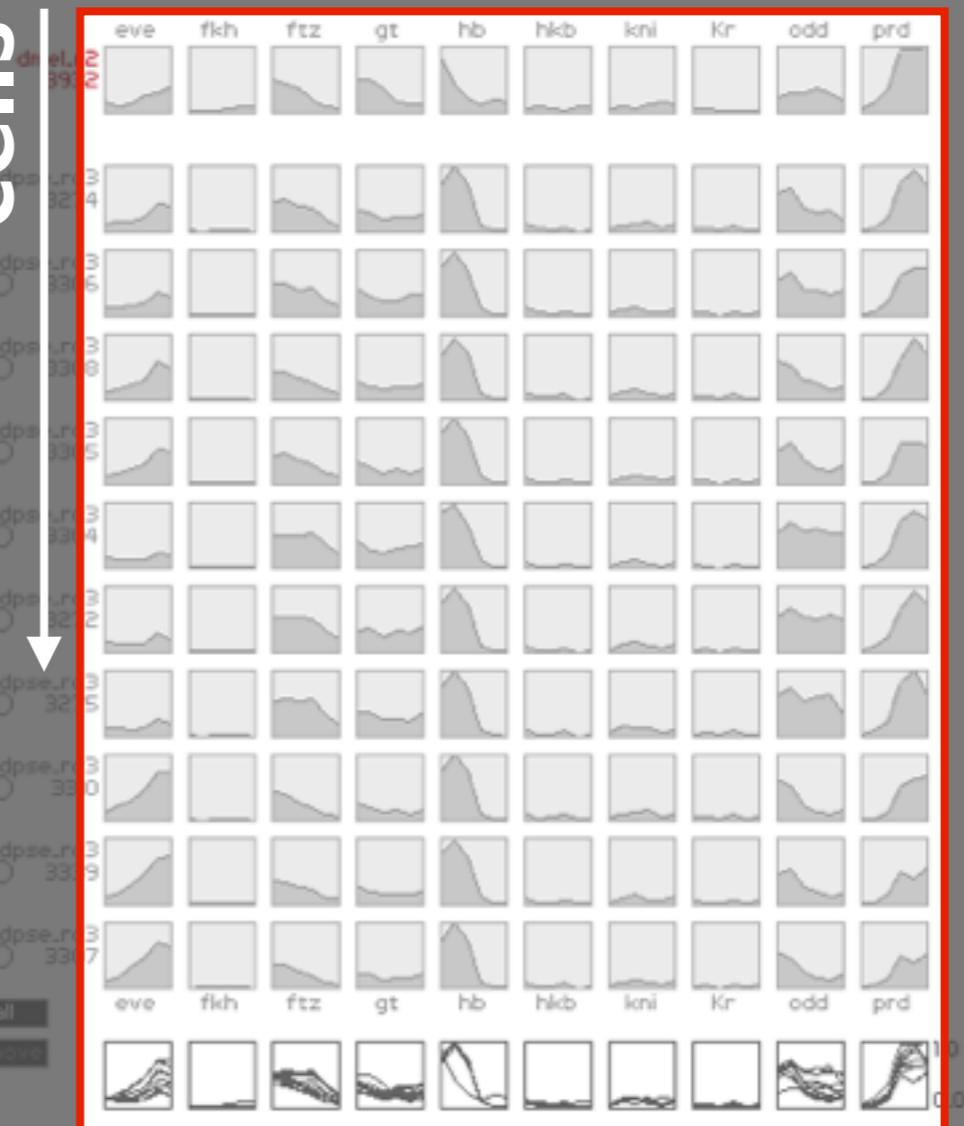


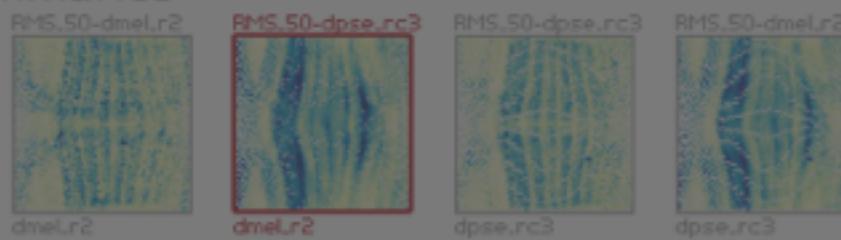
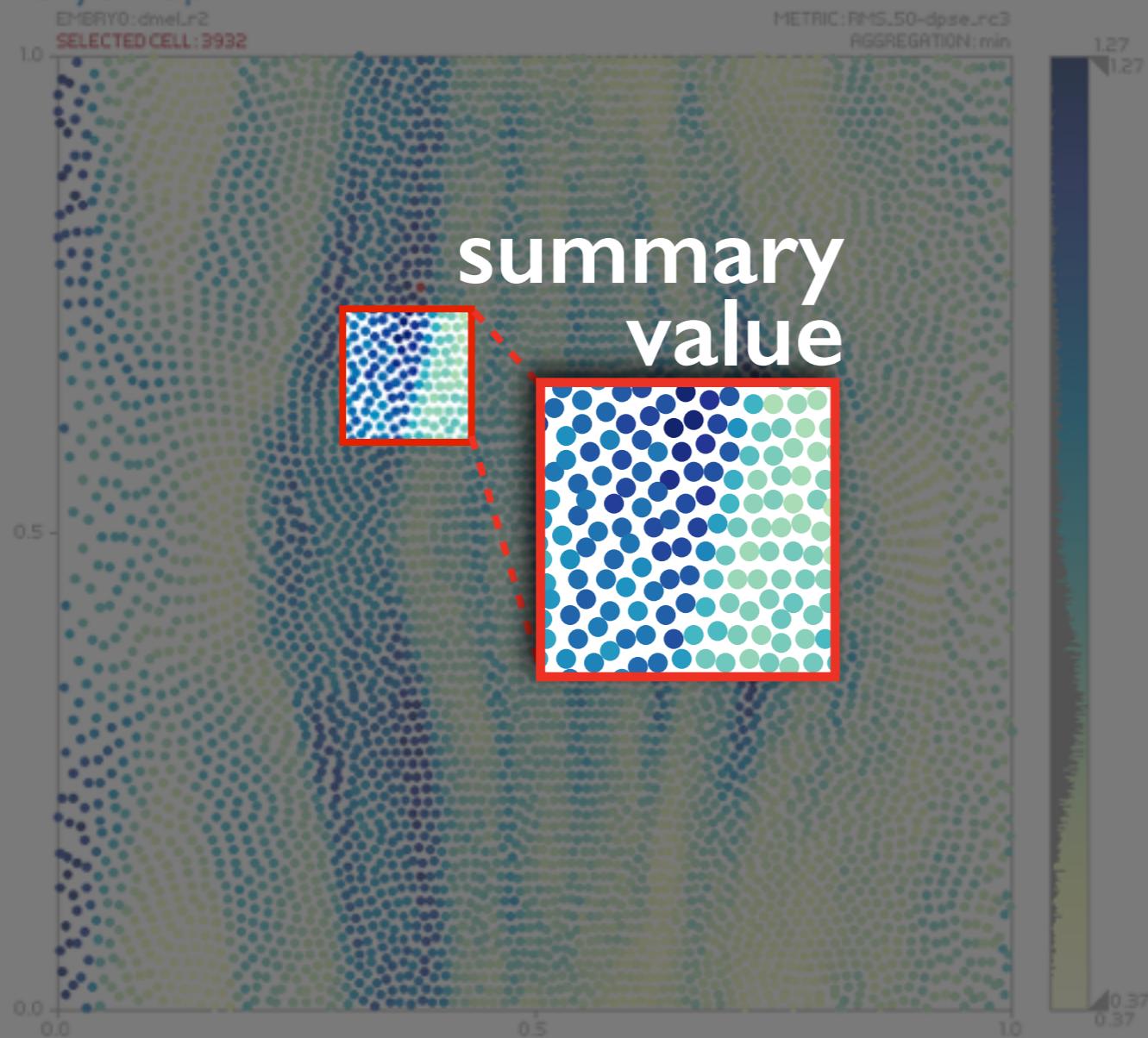
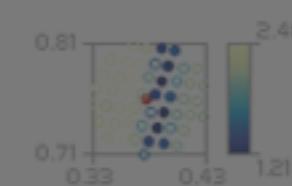
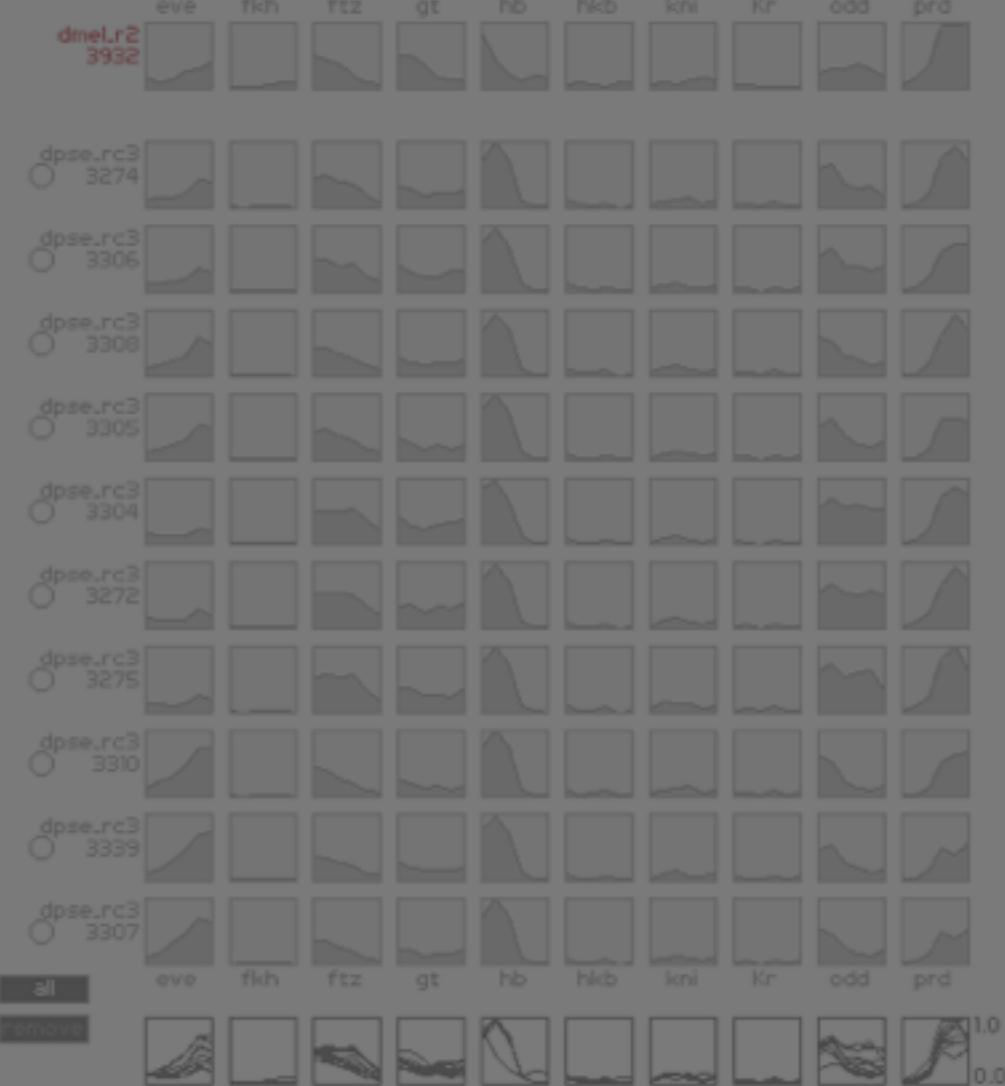
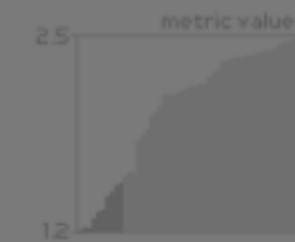
AGGREGATION GROUP

CREATED GROUP

Curvemap

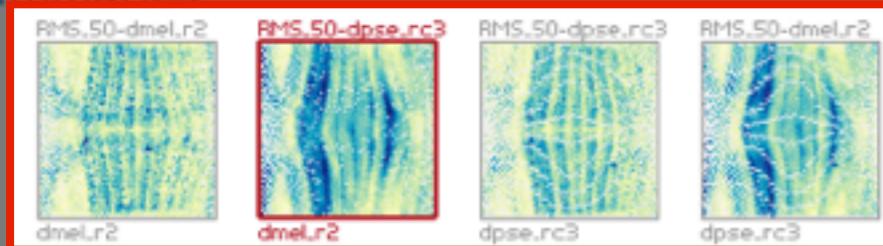


Summaries**Embryo Map****genes****cells**

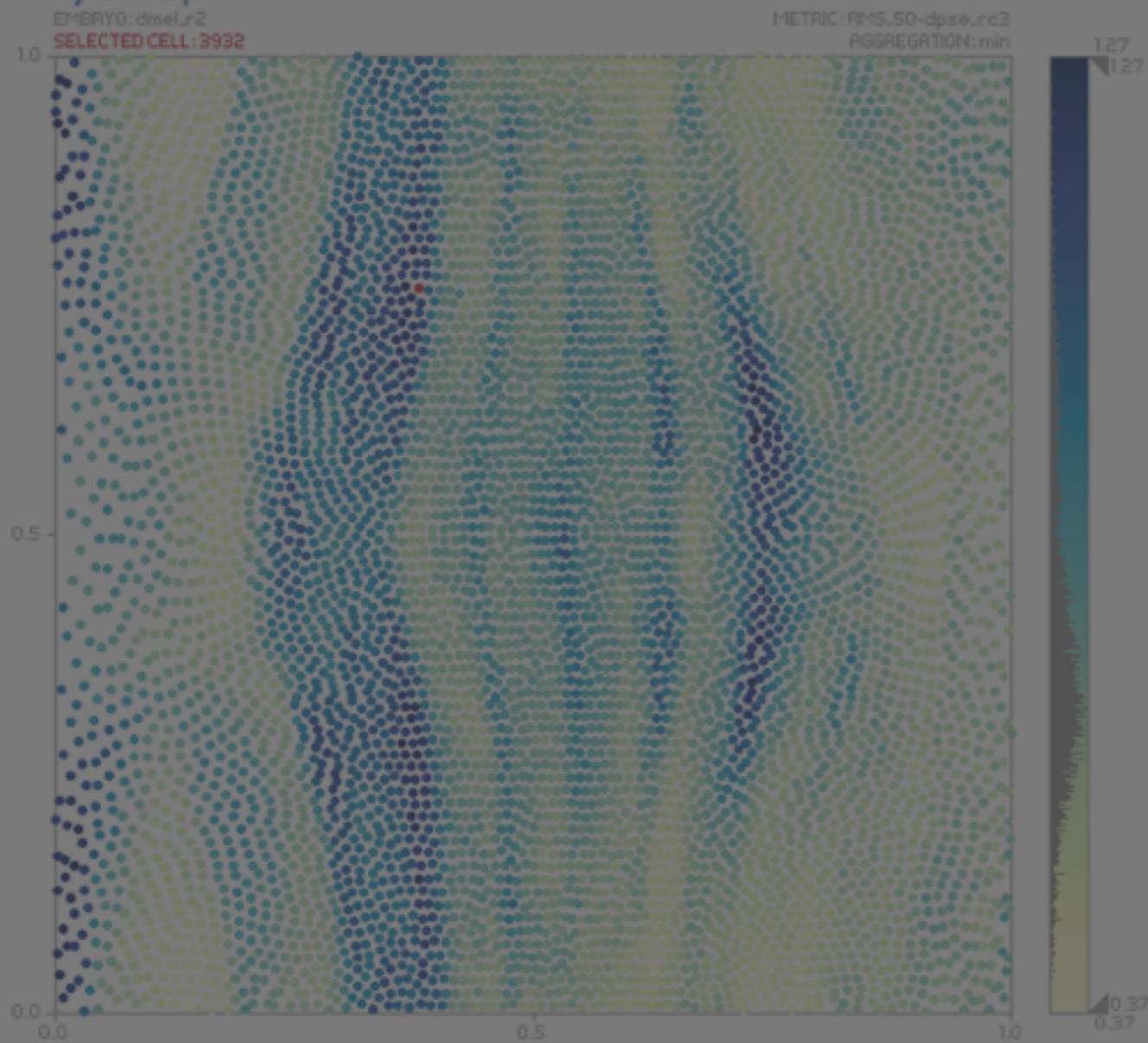
Summaries**Embryo Map****AGGREGATION GROUP****CREATED GROUP****Curvemap**

summaries

summaries



Embryo Map



AGGREGATION GROUP

CREATED GROUP

Curvemap



Characterize differences in gene expression patterns between species.

Characterize differences in gene expression patterns between species.

differences related to:

spatial position

gene expression profiles

complex combination

challenging to characterize manually

support mechanisms:

summaries, groups

data & tool & tasks

summaries & groups

encodings & interaction

conclusions

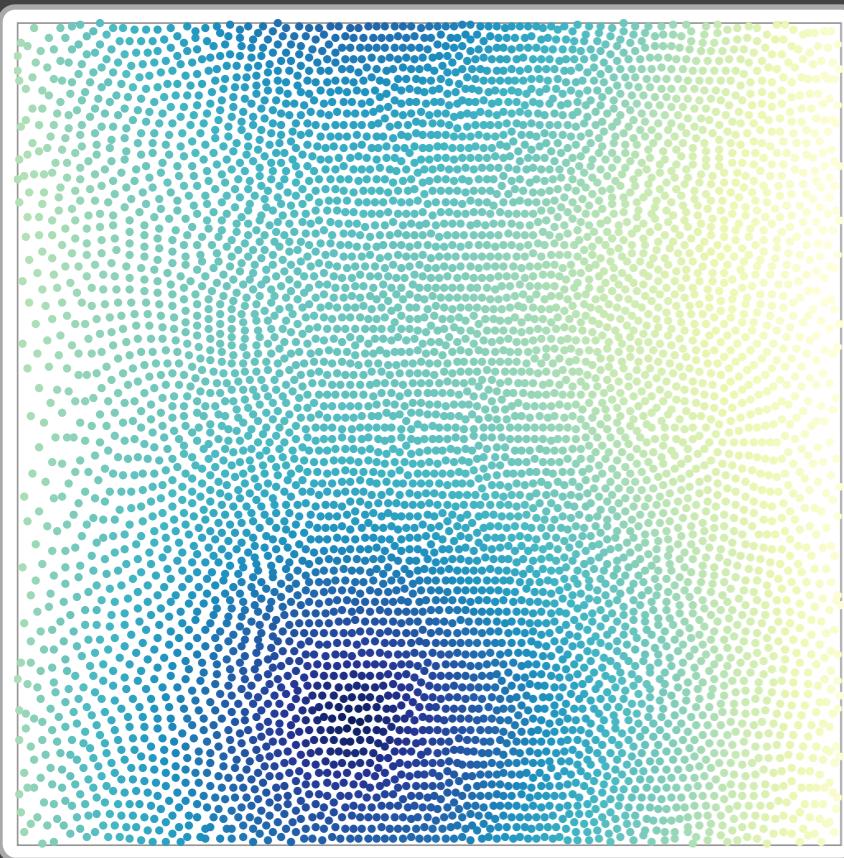
summary

a value for each cell that summarizes the underlying data

summary

a value for each cell that summarizes the underlying data

spatial

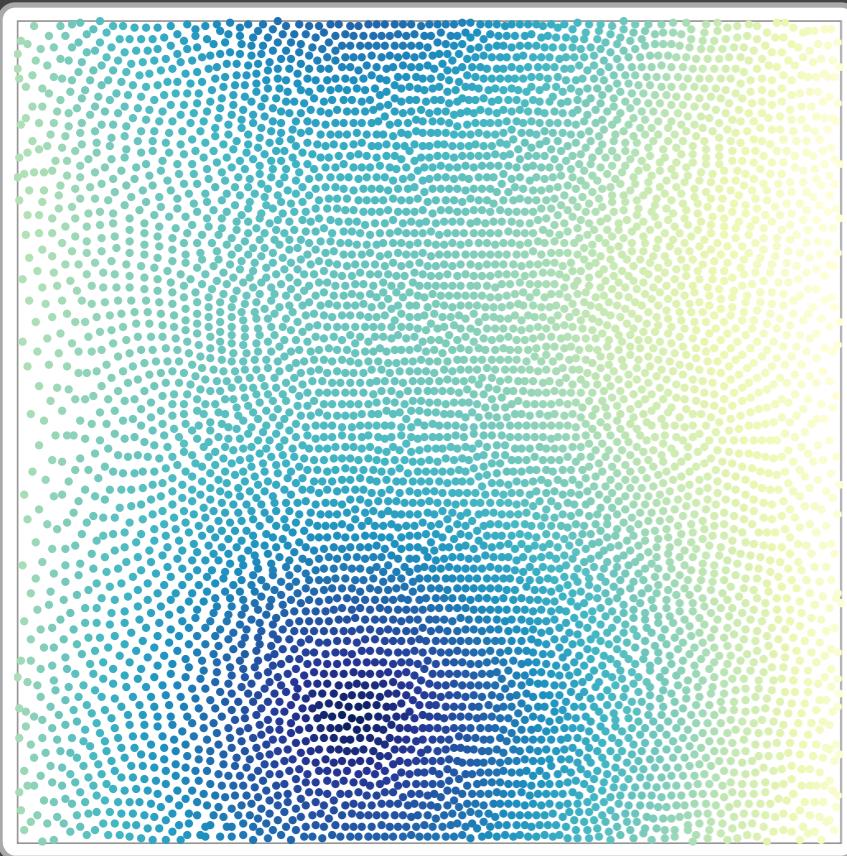


distance to cell i

summary

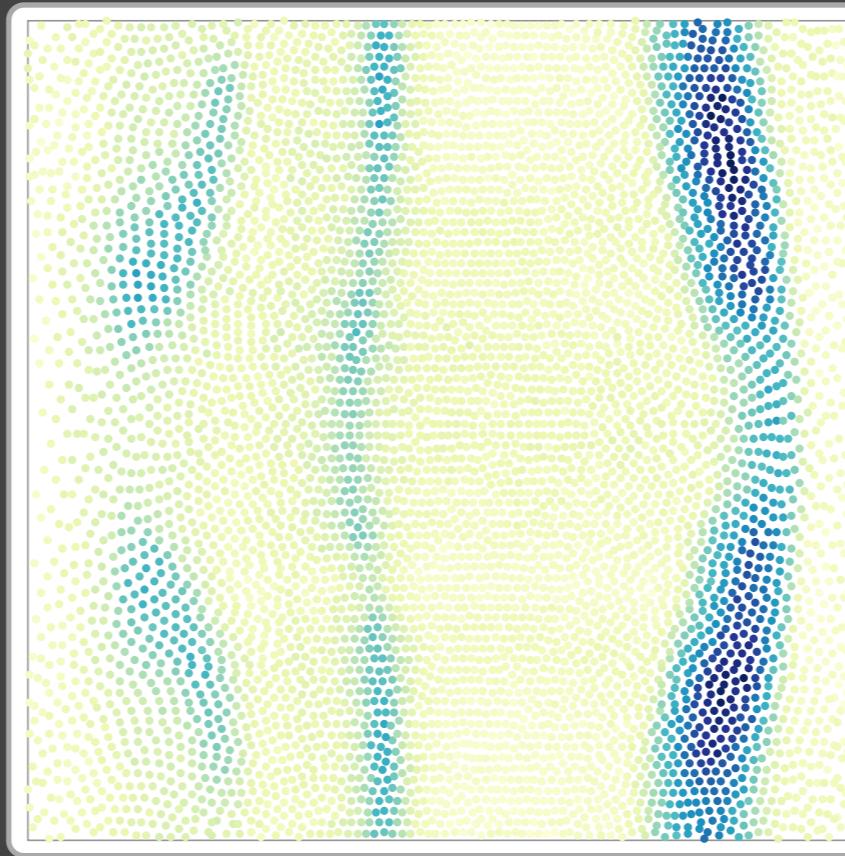
a value for each cell that summarizes the underlying data

spatial



distance to cell i

expression

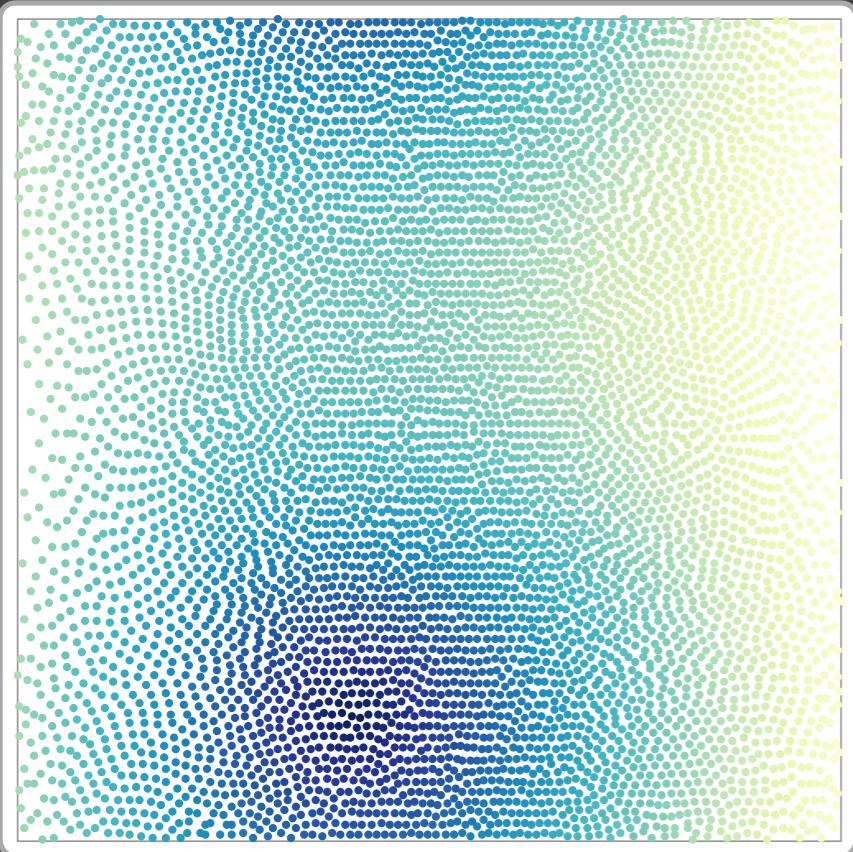


gene j at timepoint k

summary

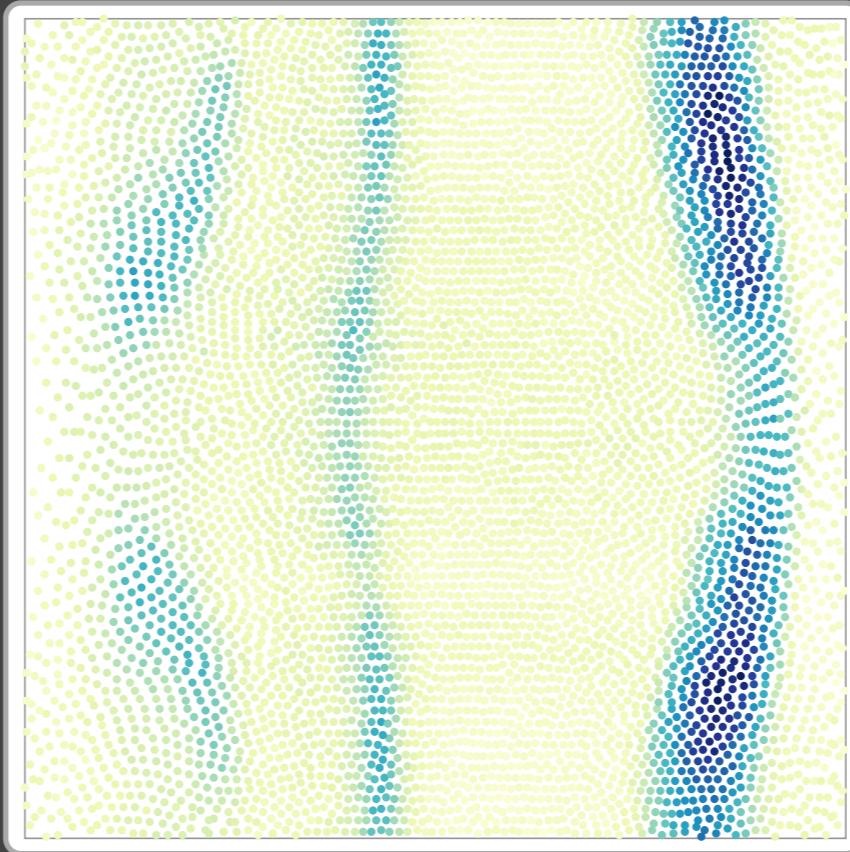
a value for each cell that summarizes the underlying data

spatial



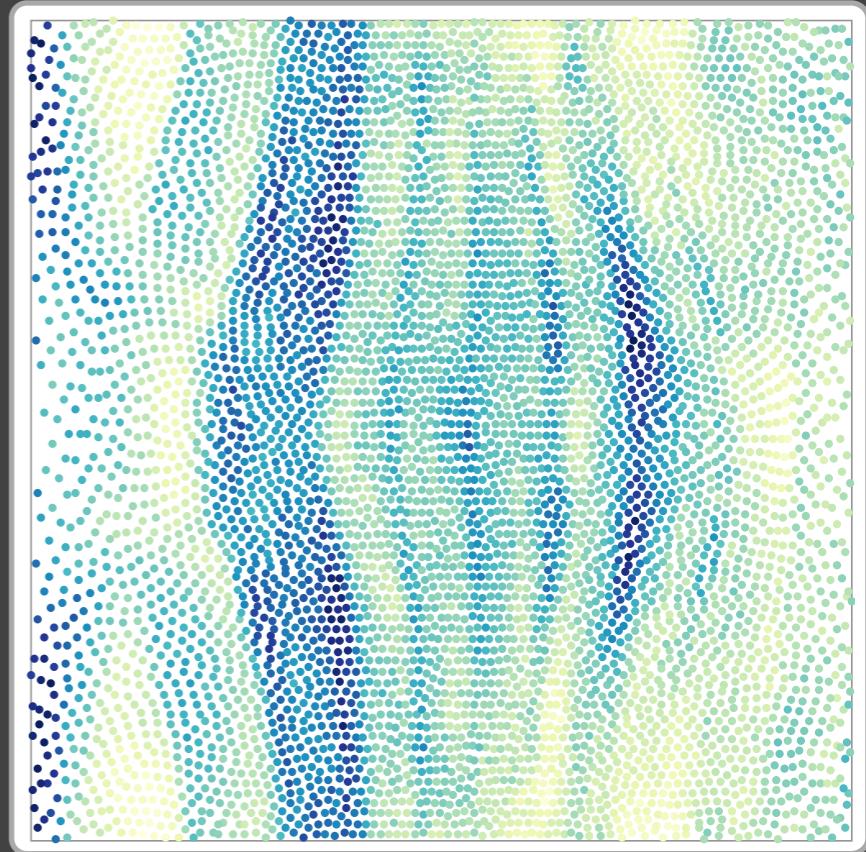
distance to cell i

expression



gene j at timepoint k

combination

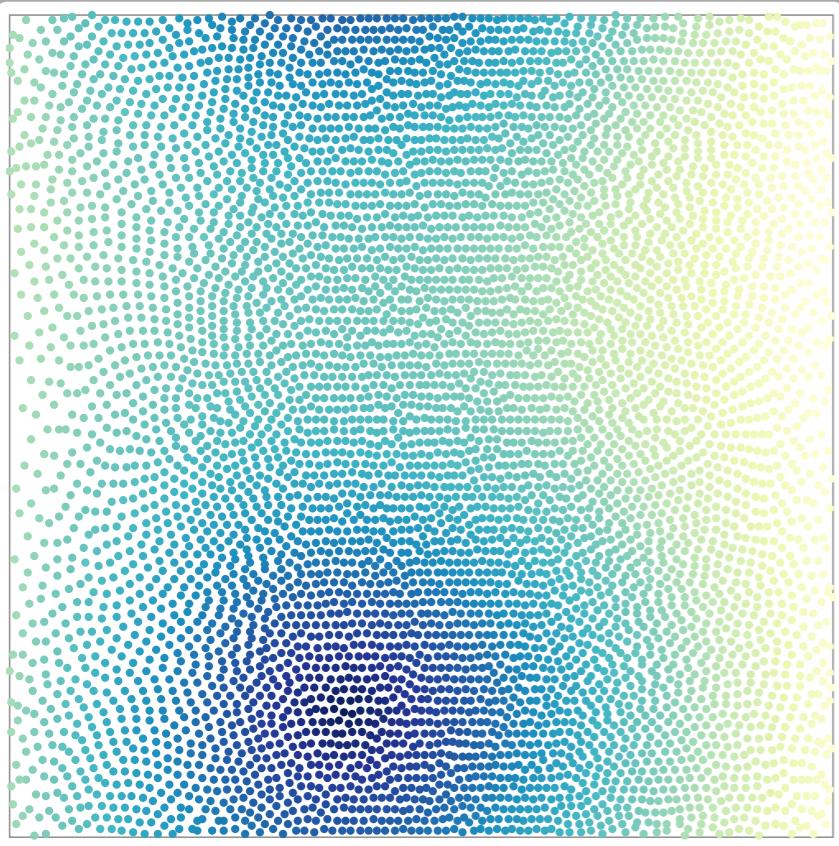


compare embryos
 A and B

summary

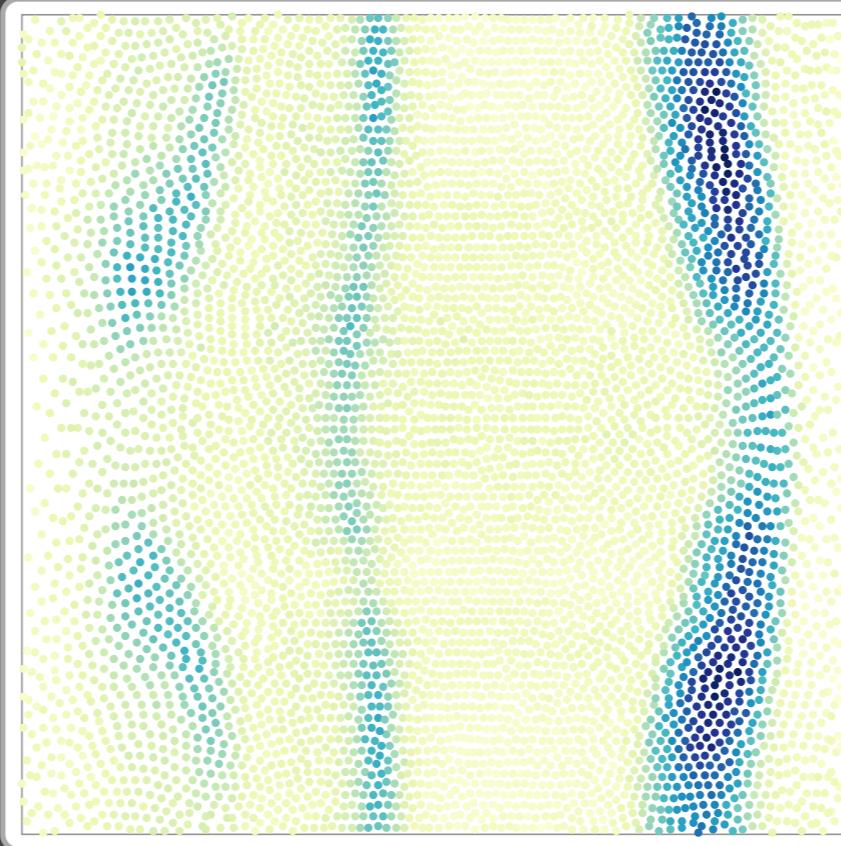
a value for each cell that summarizes the underlying data

spatial



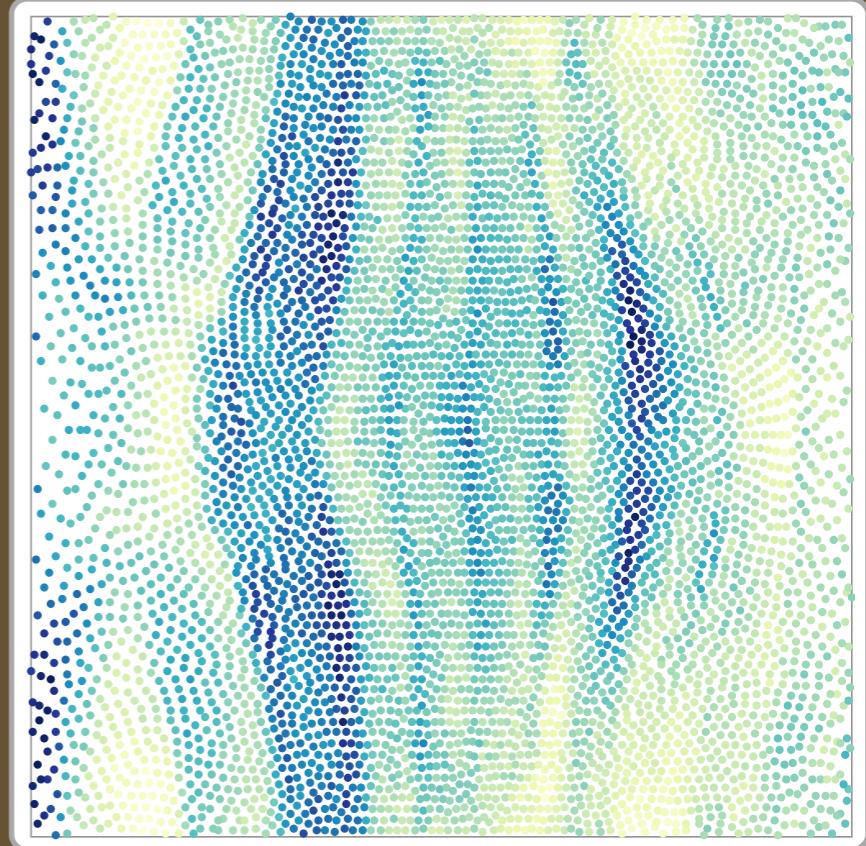
distance to cell i

expression



gene j at timepoint k

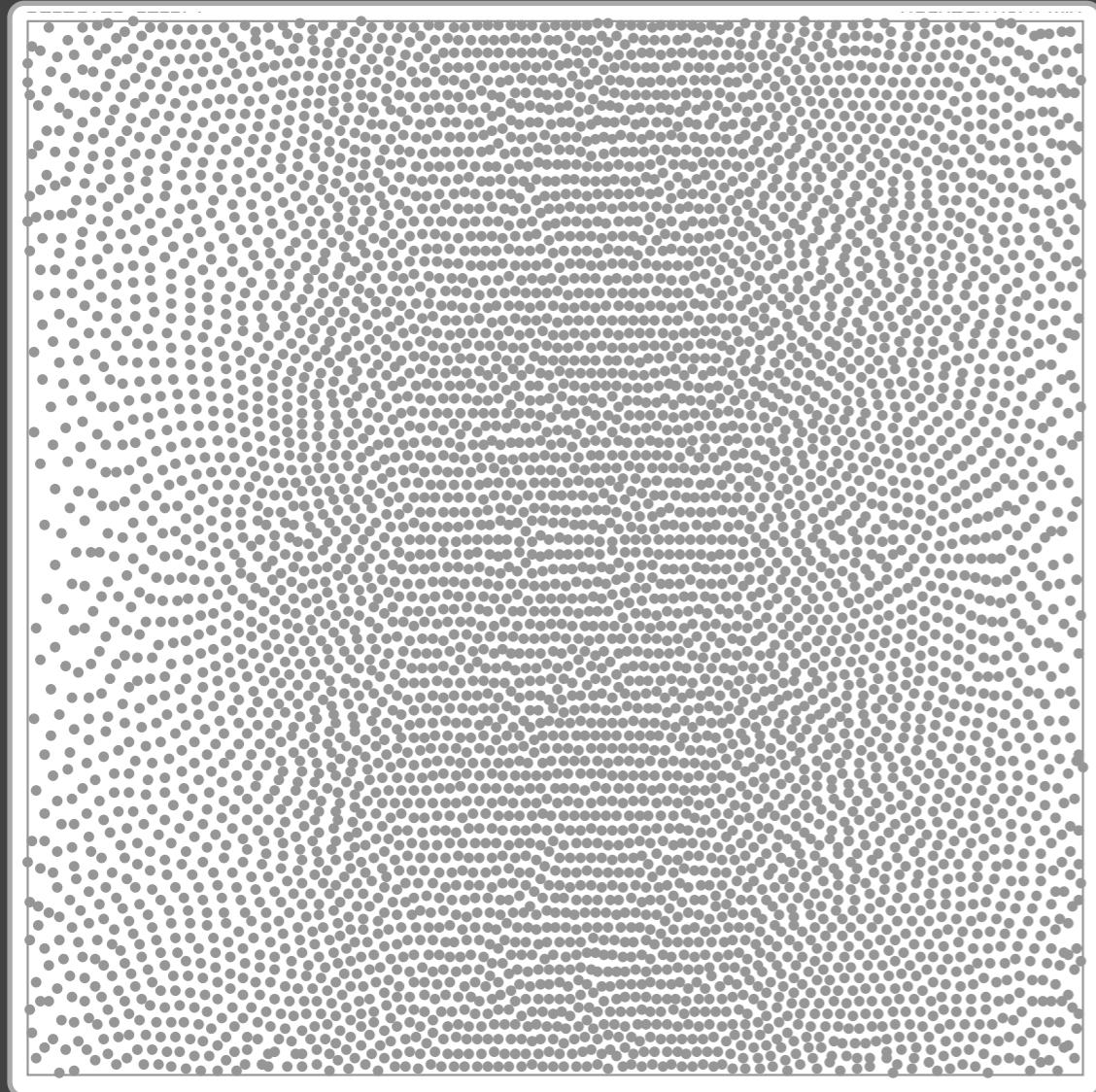
combination



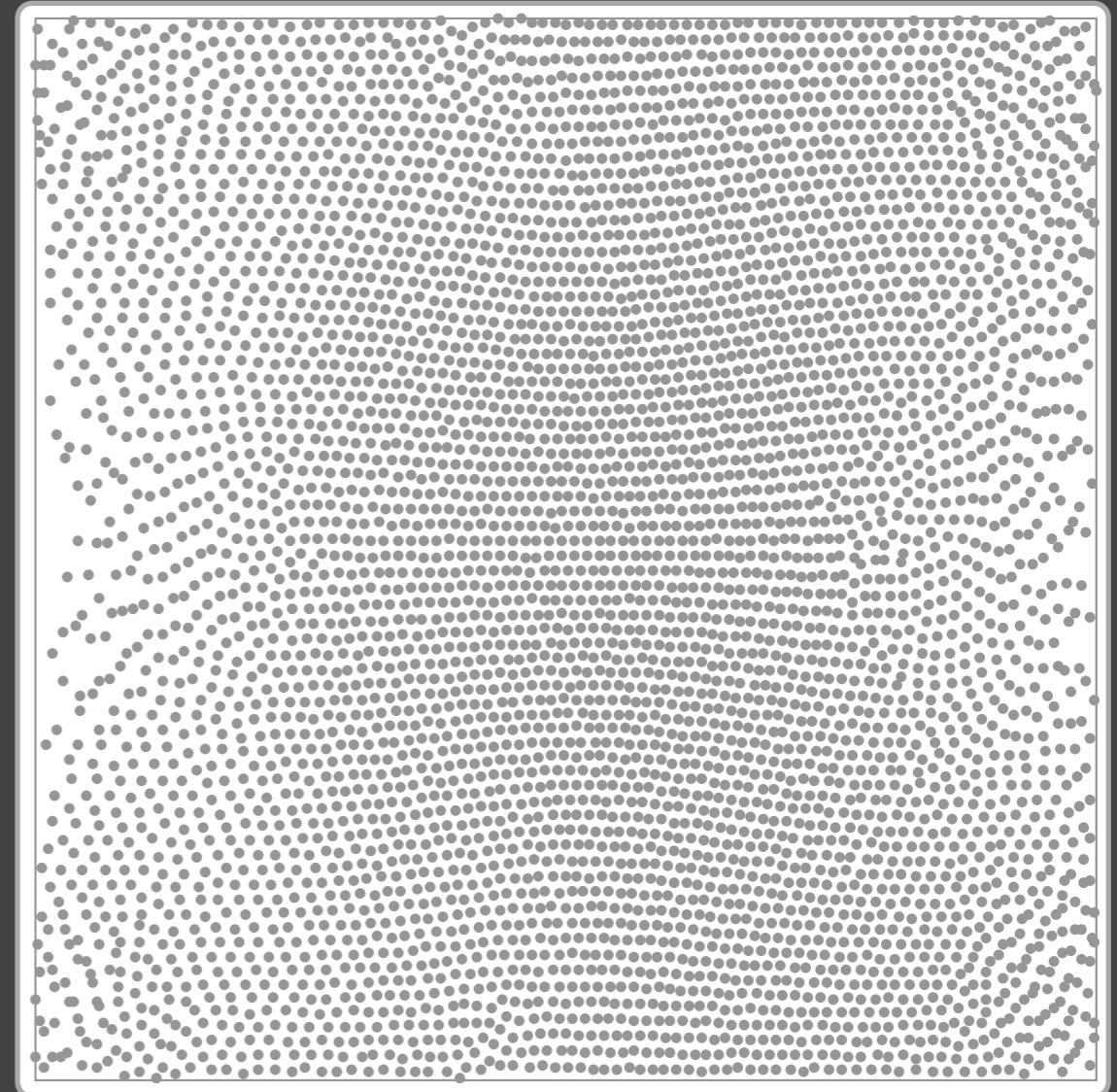
compare embryos
 A and B

creating a comparison summary

creating a comparison summary

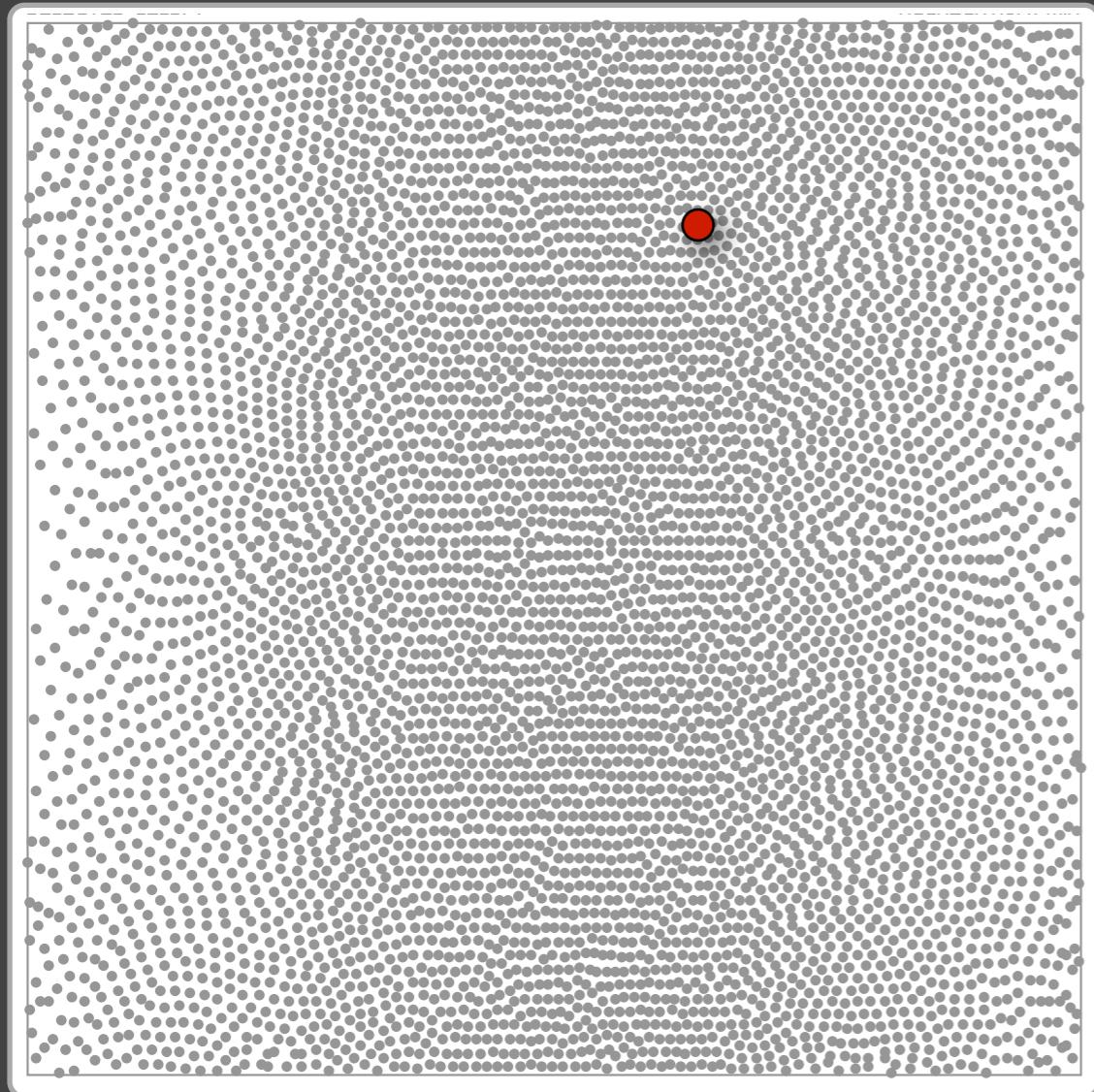


embryo A

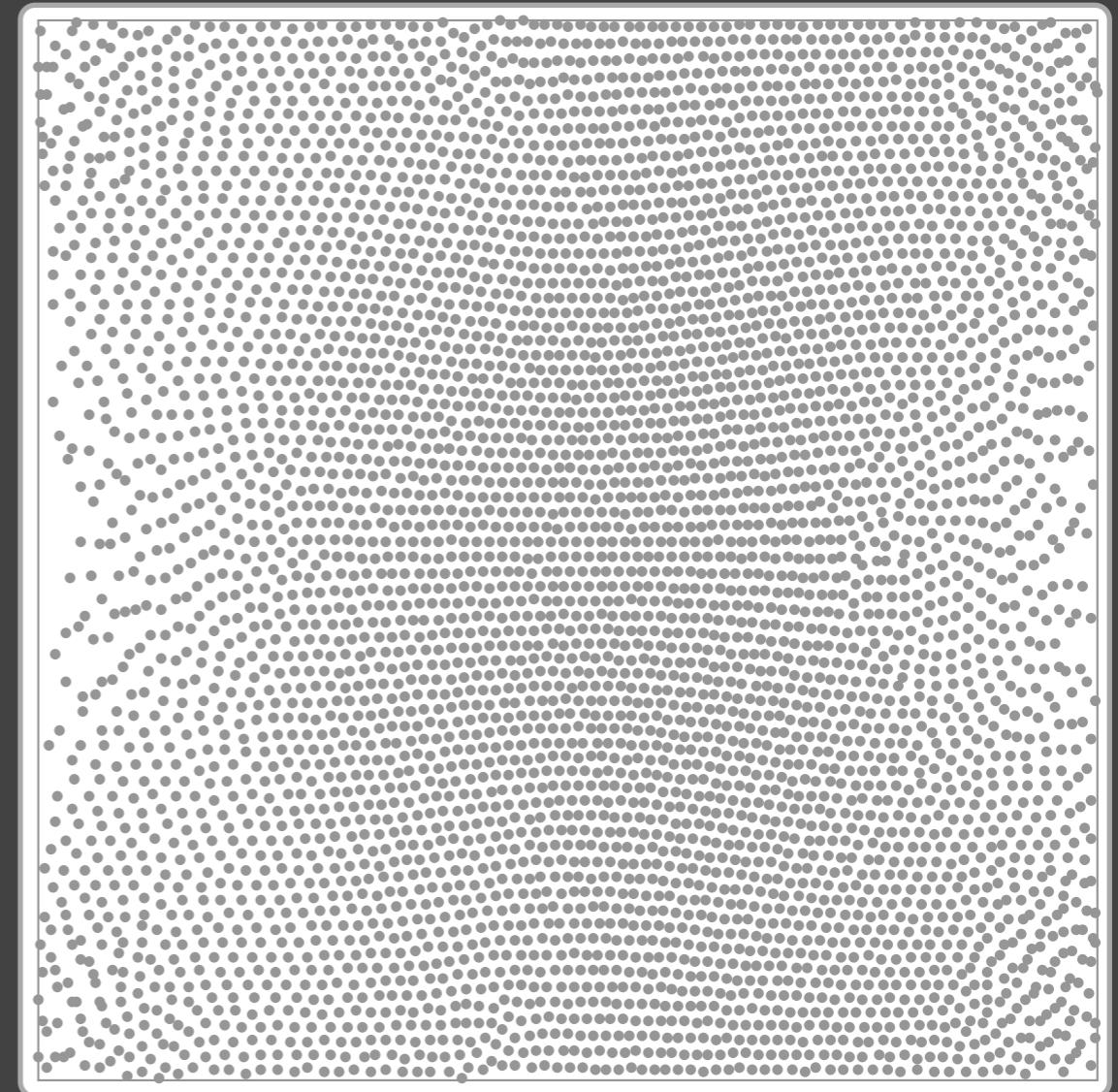


embryo B

creating a comparison summary

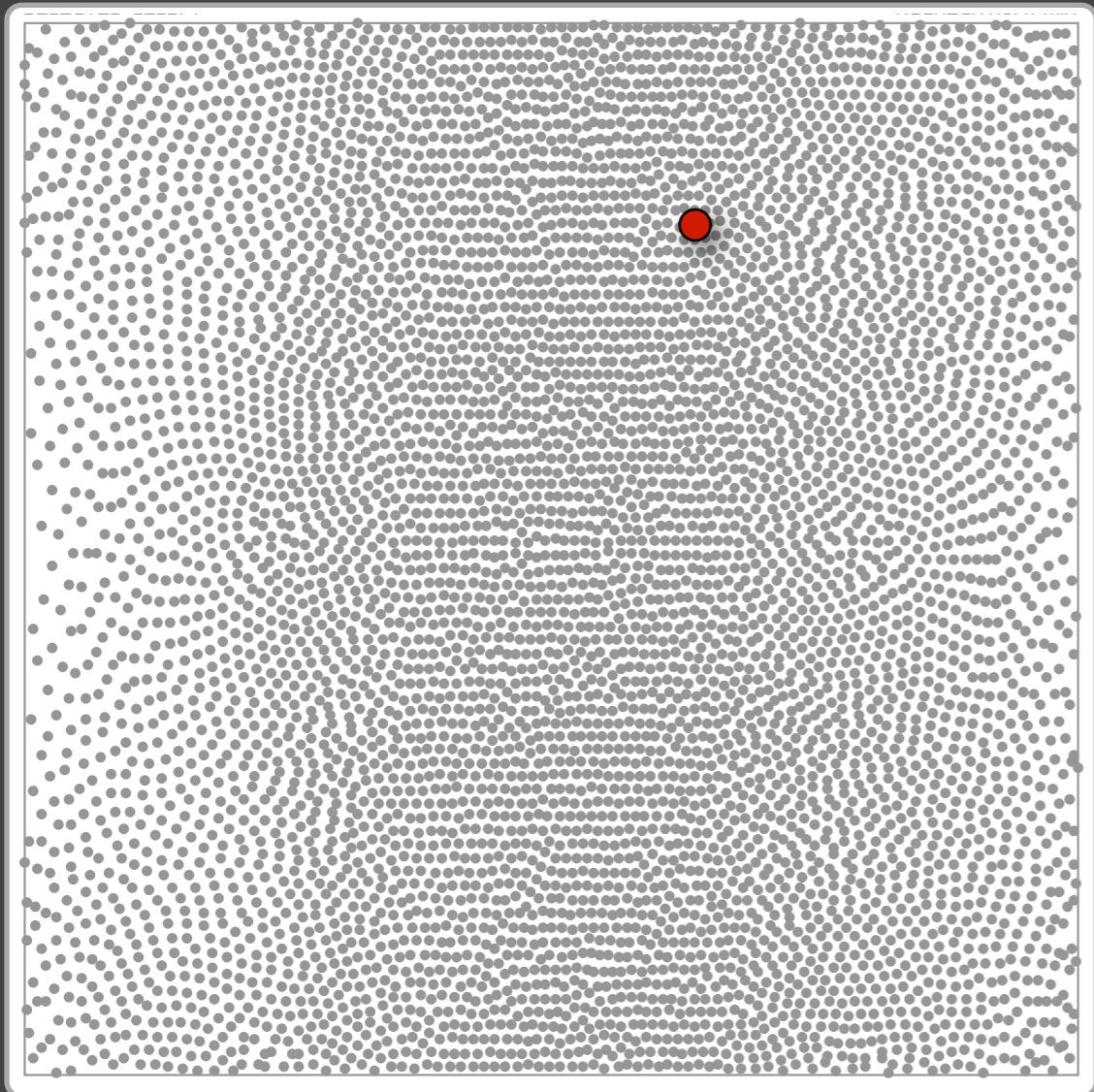


embryo A

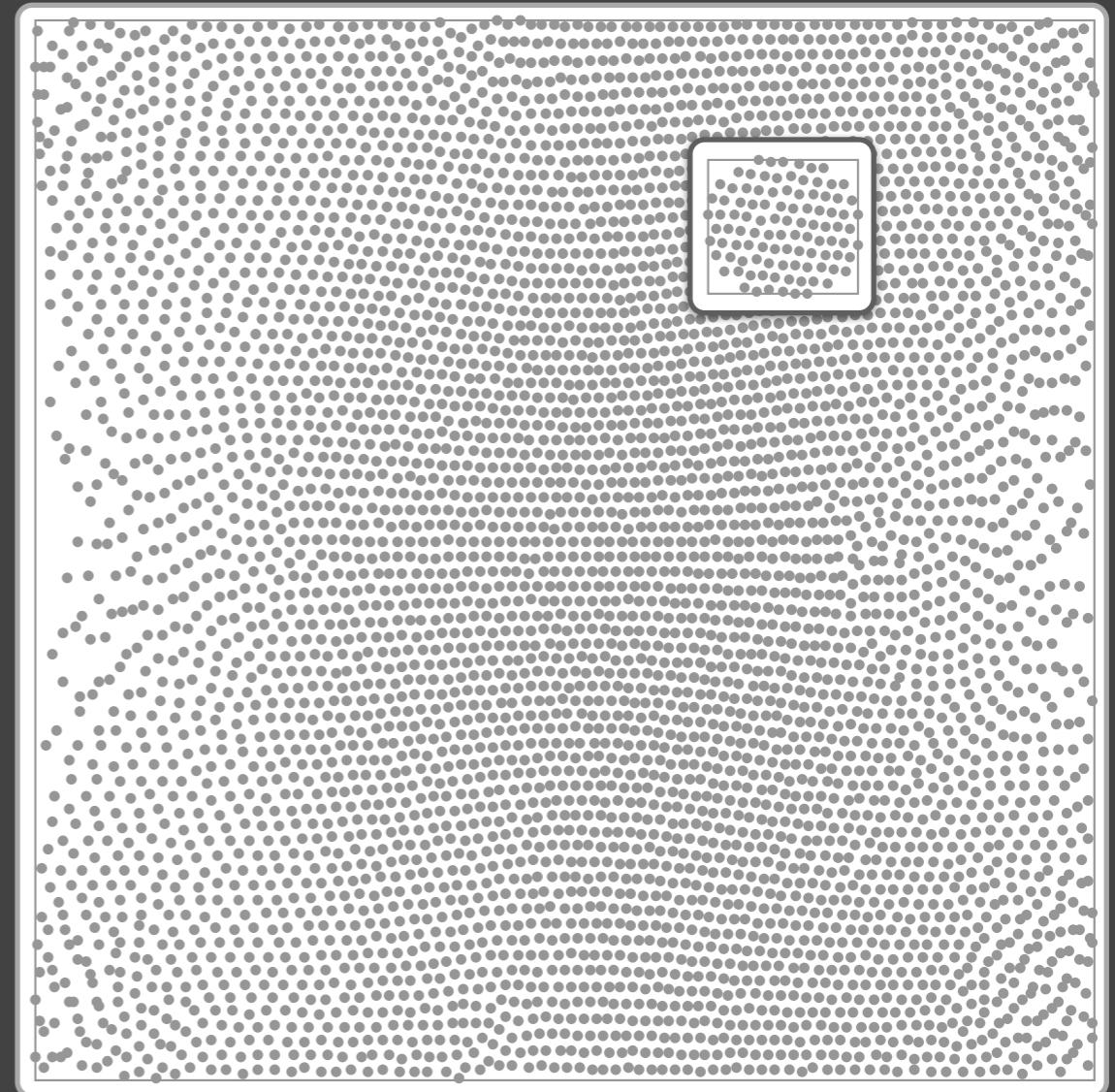


embryo B

creating a comparison summary



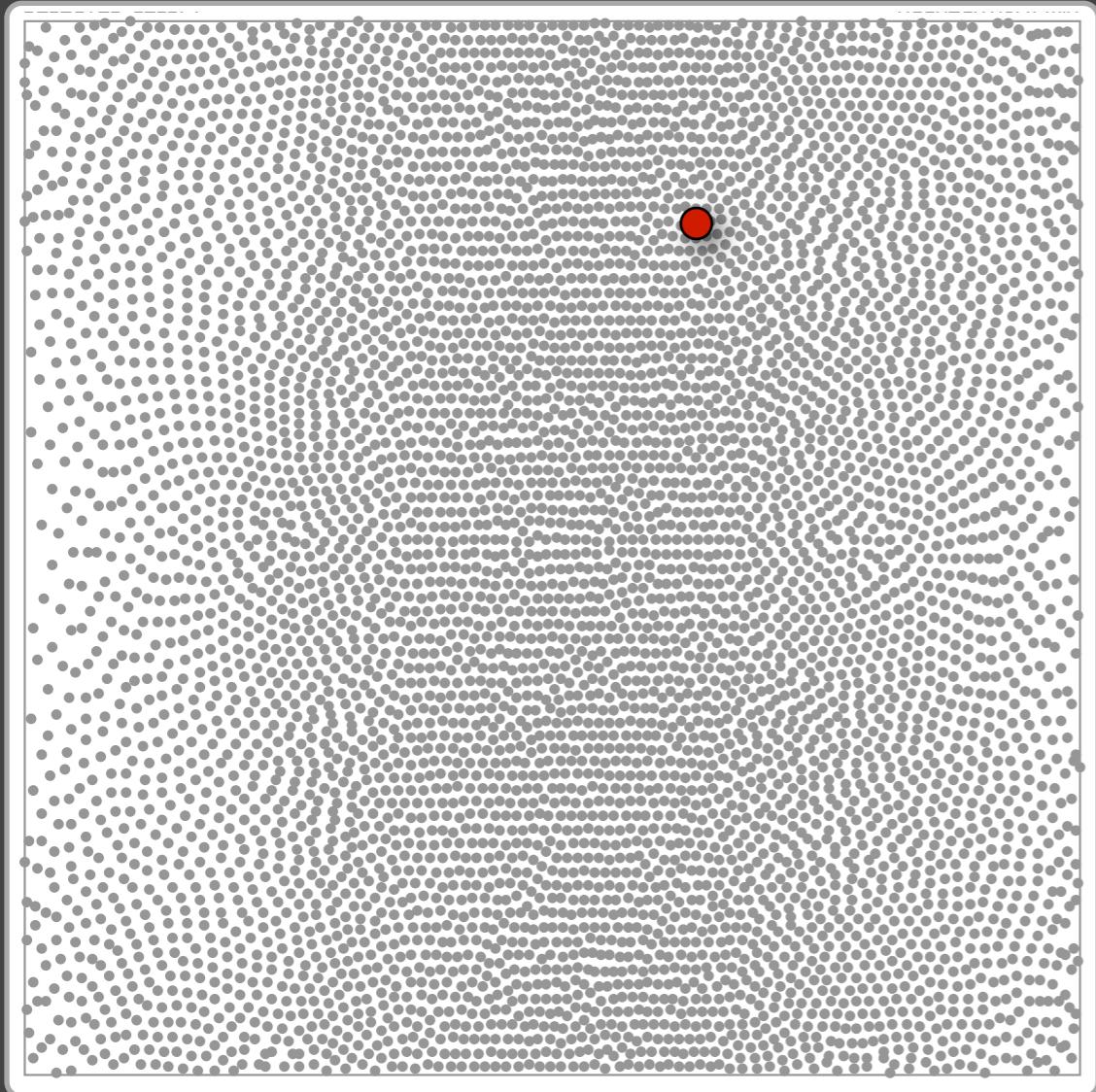
embryo A



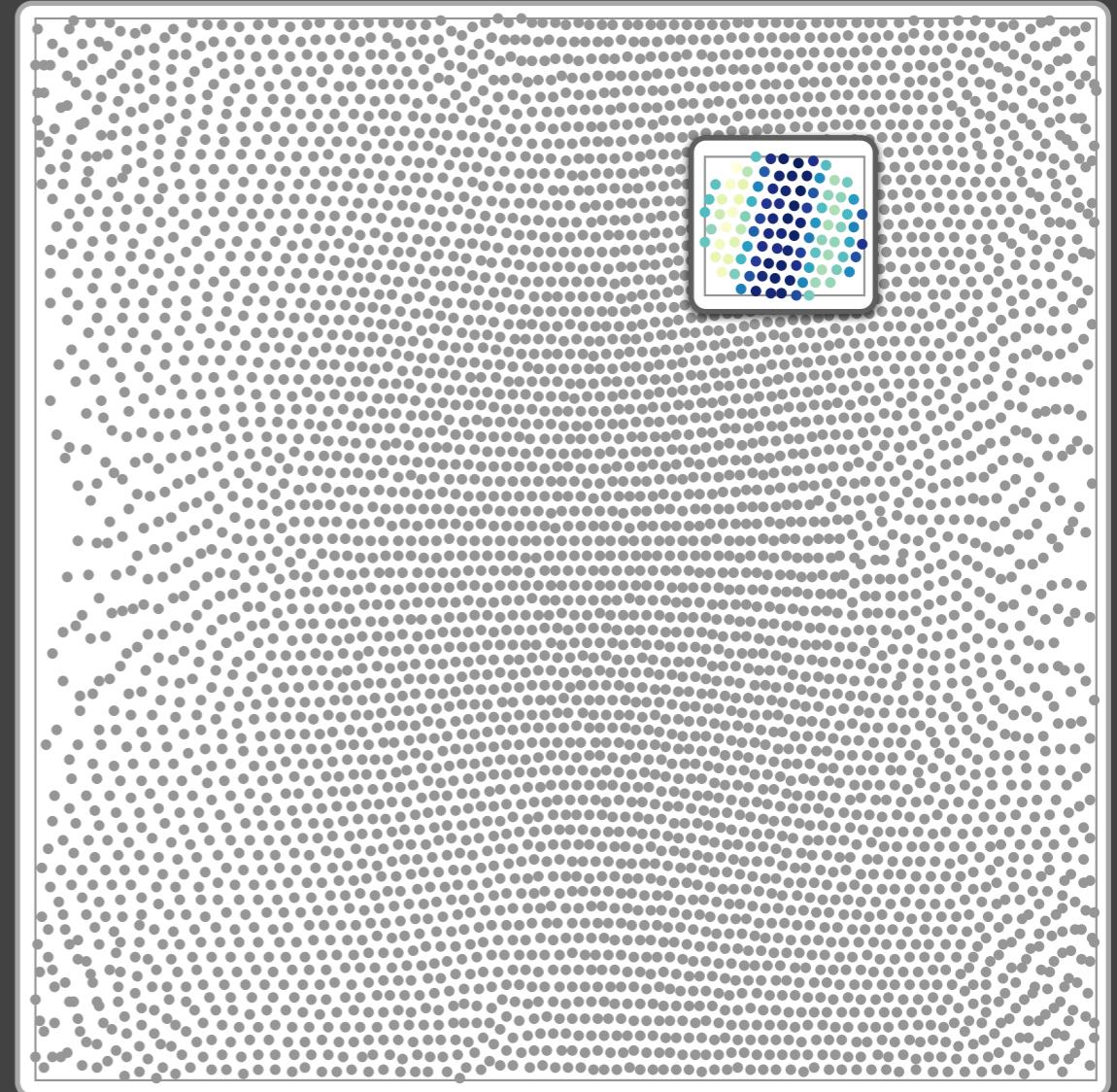
embryo B

creating a comparison summary

root-mean-square distance



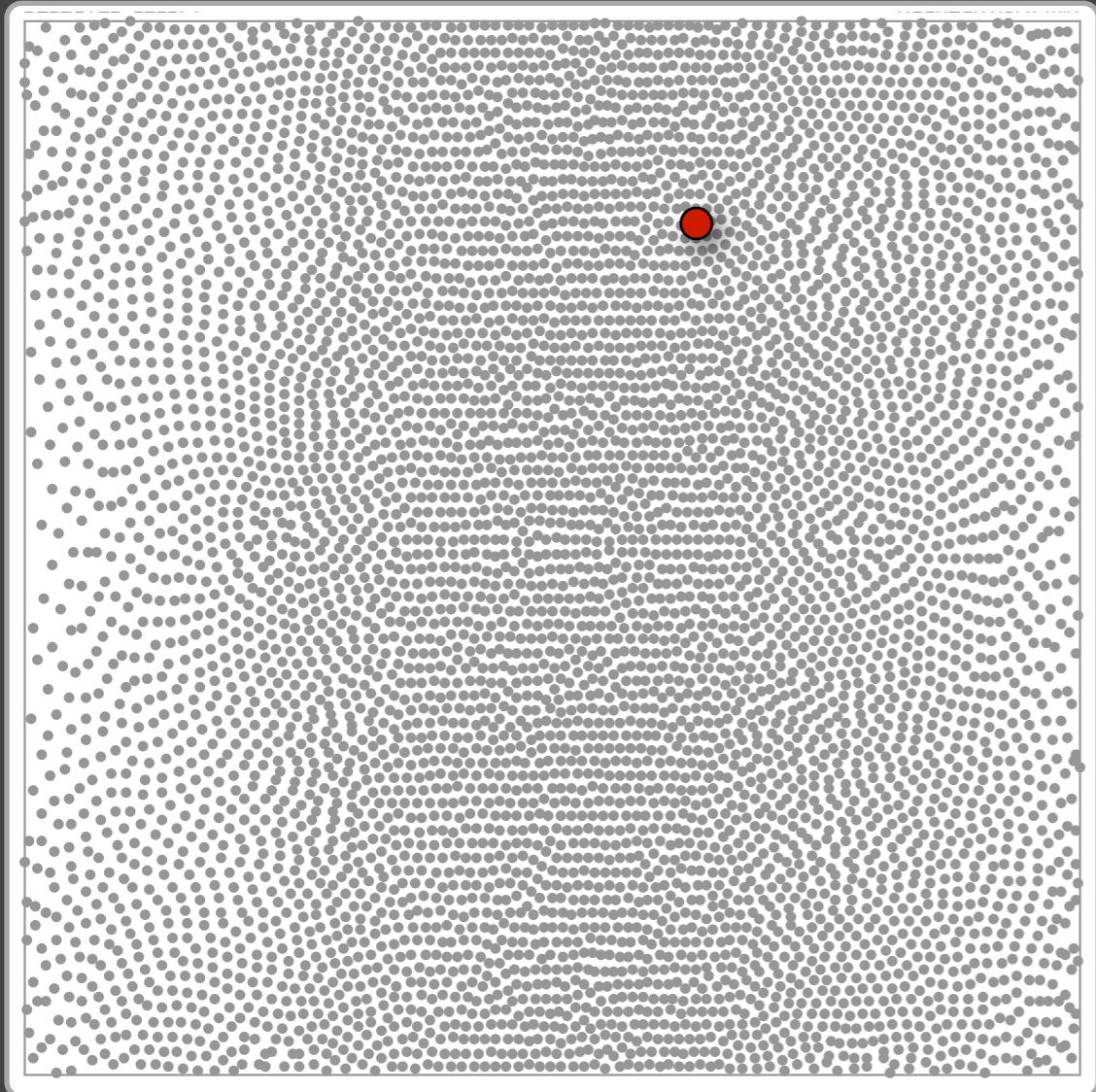
embryo A



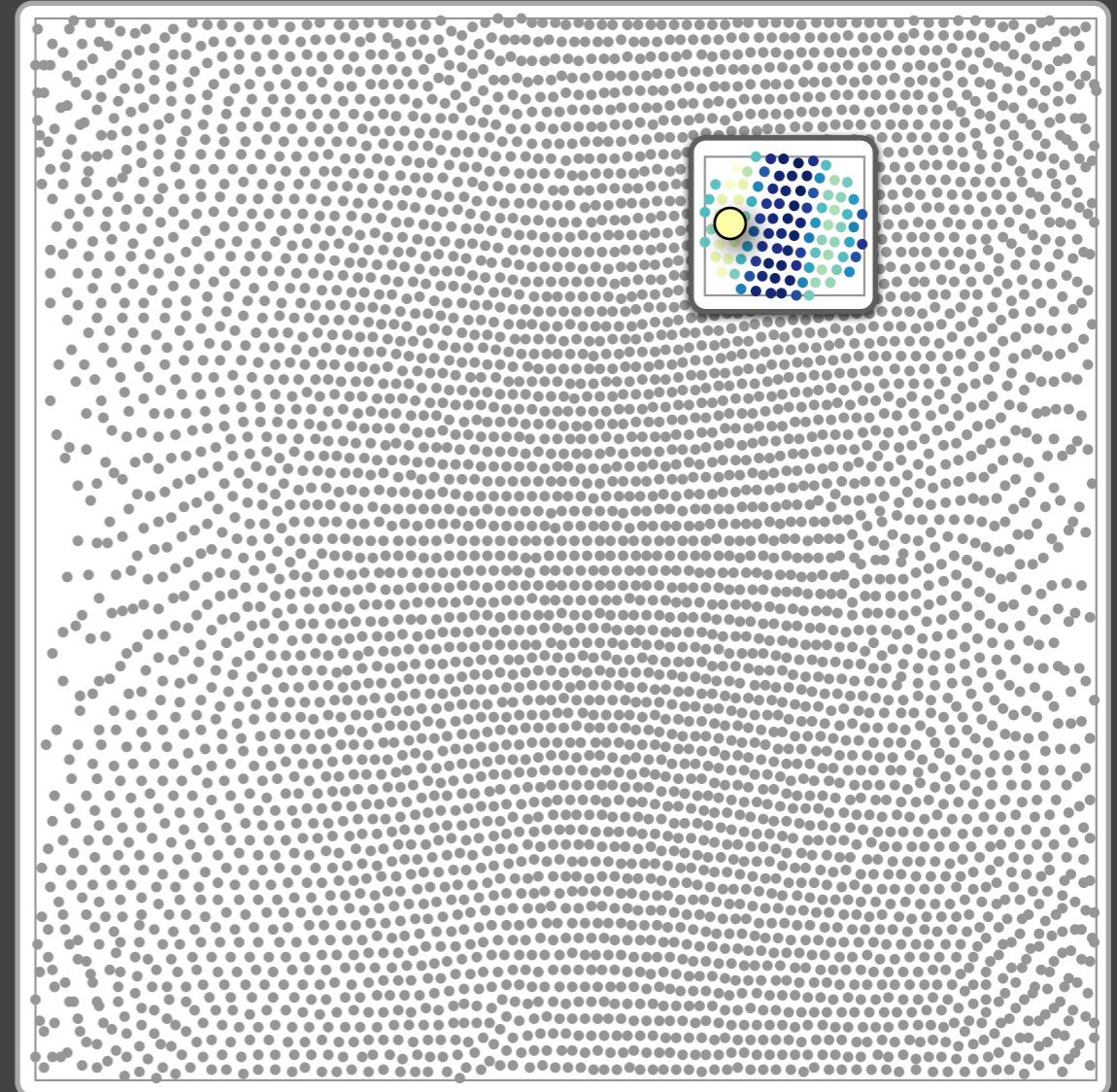
embryo B

creating a comparison summary

root-mean-square distance



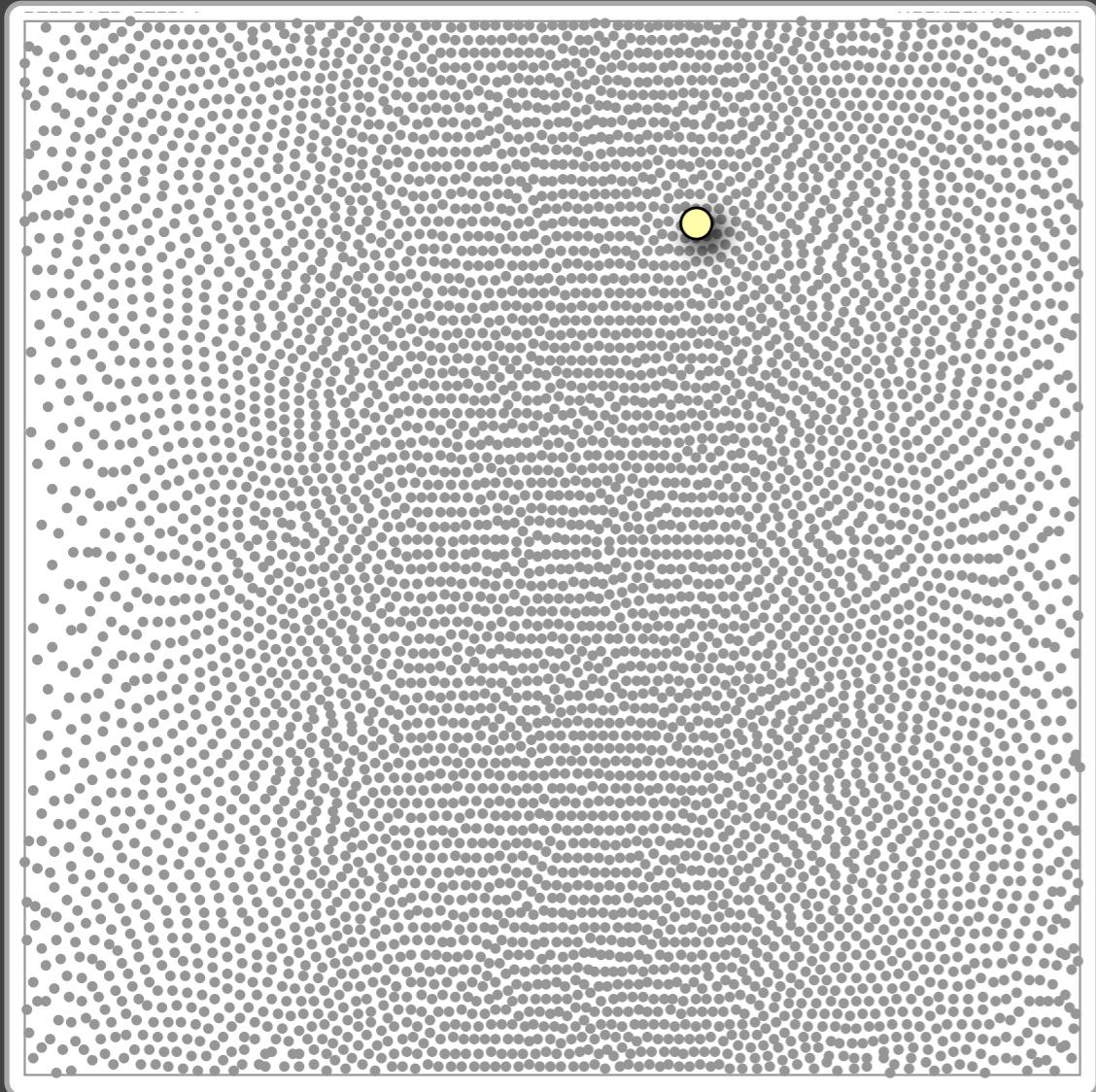
embryo A



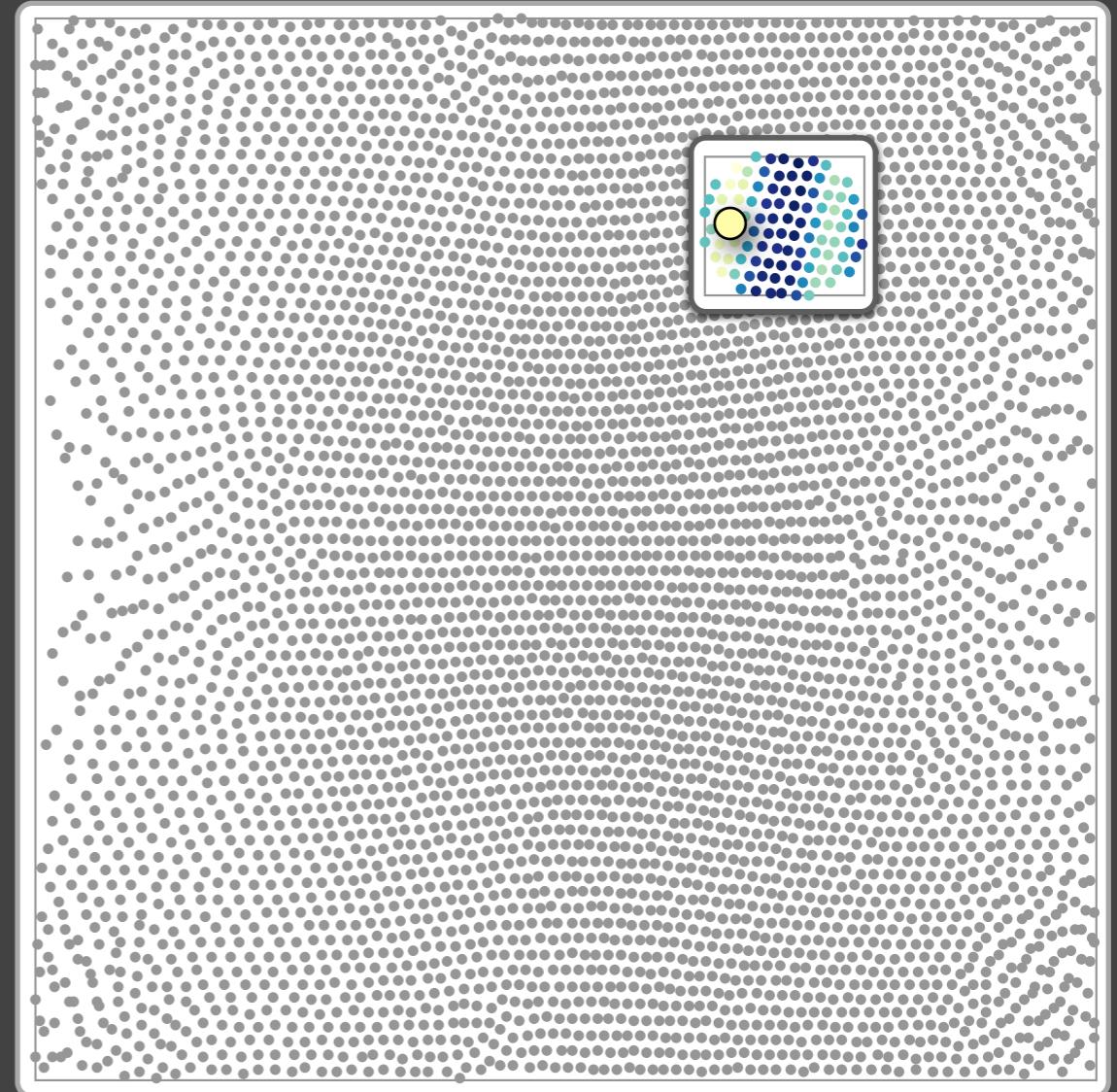
embryo B

creating a comparison summary

root-mean-square distance



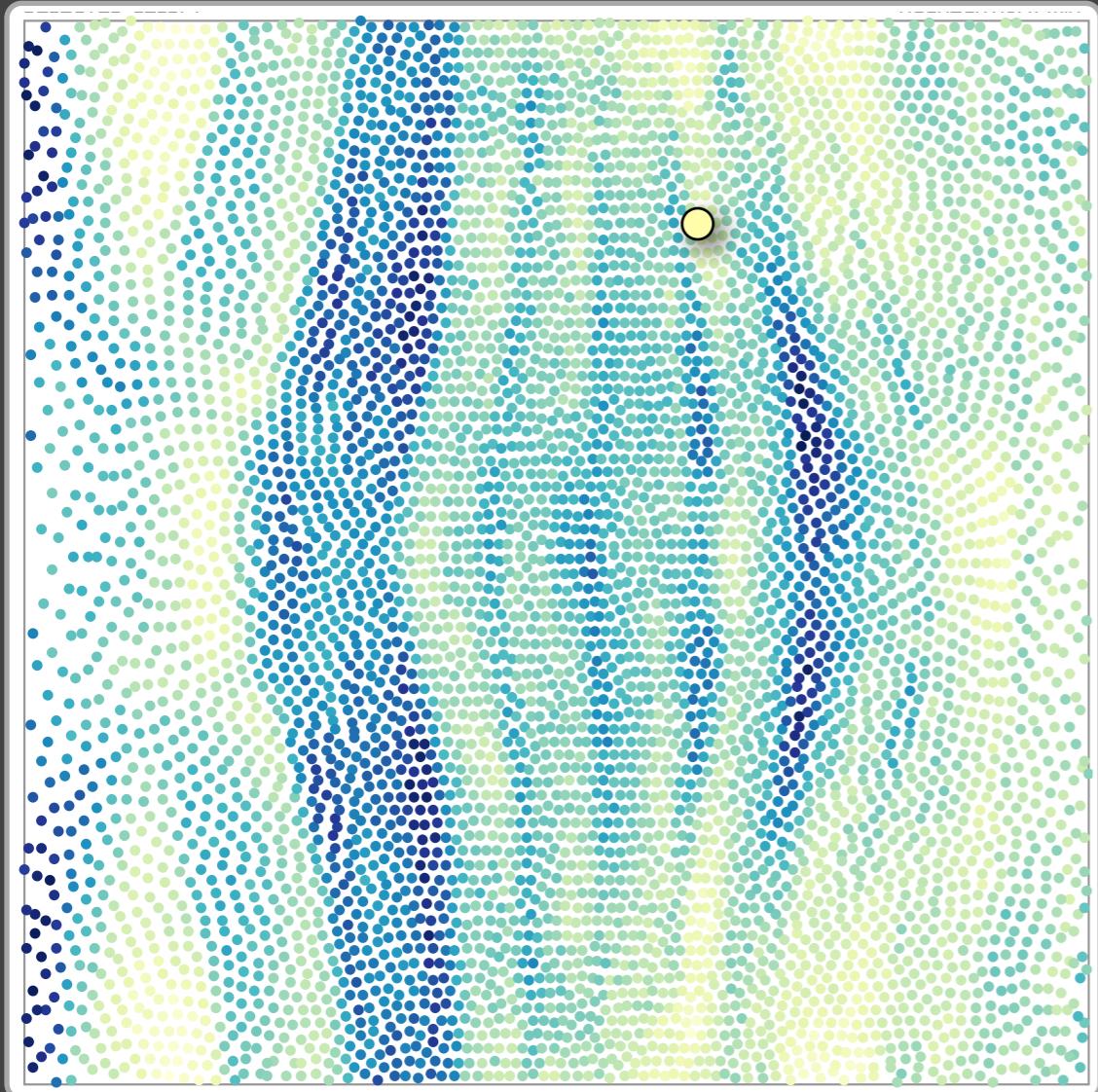
embryo A



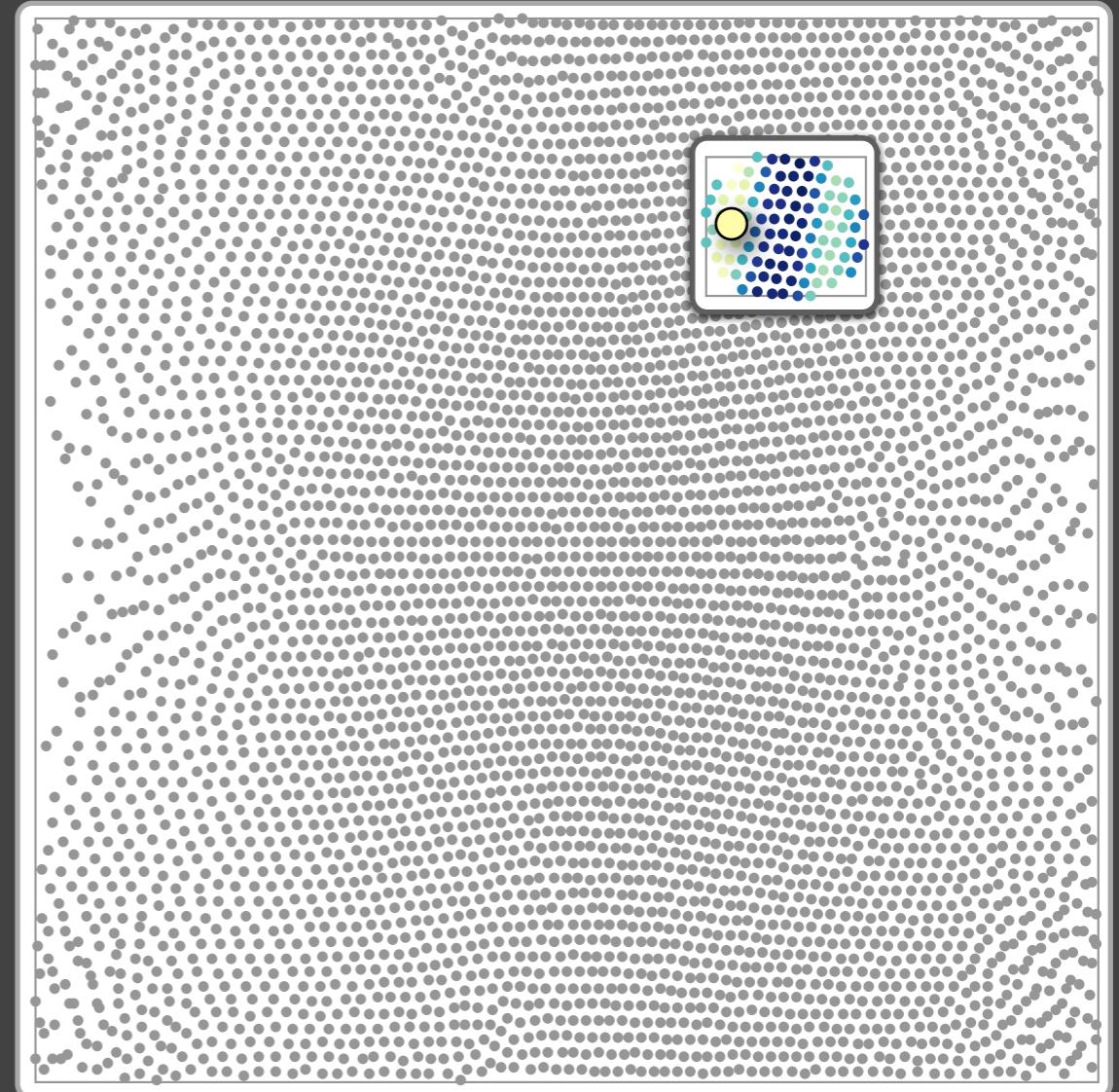
embryo B

creating a comparison summary

root-mean-square distance



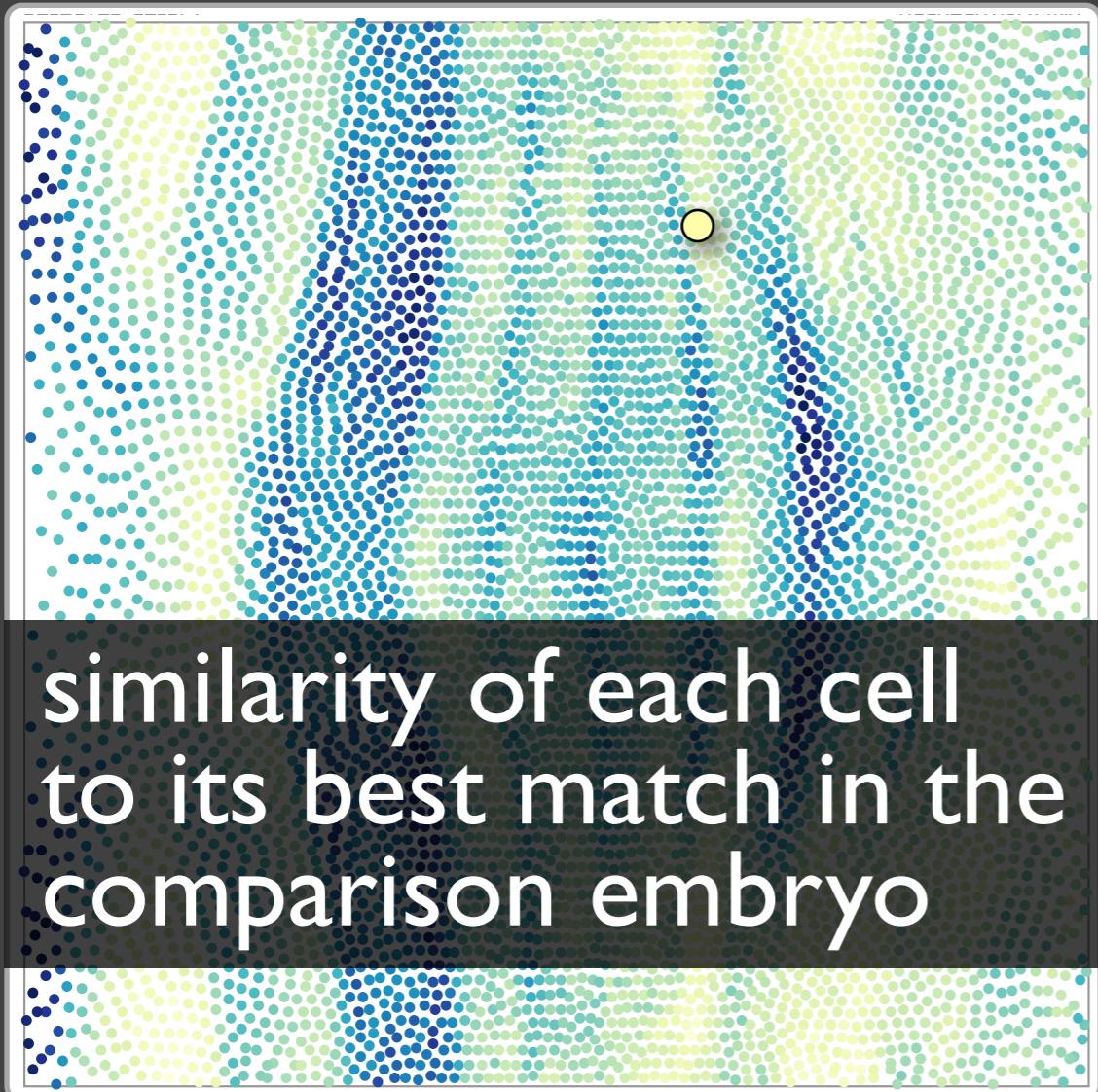
embryo A



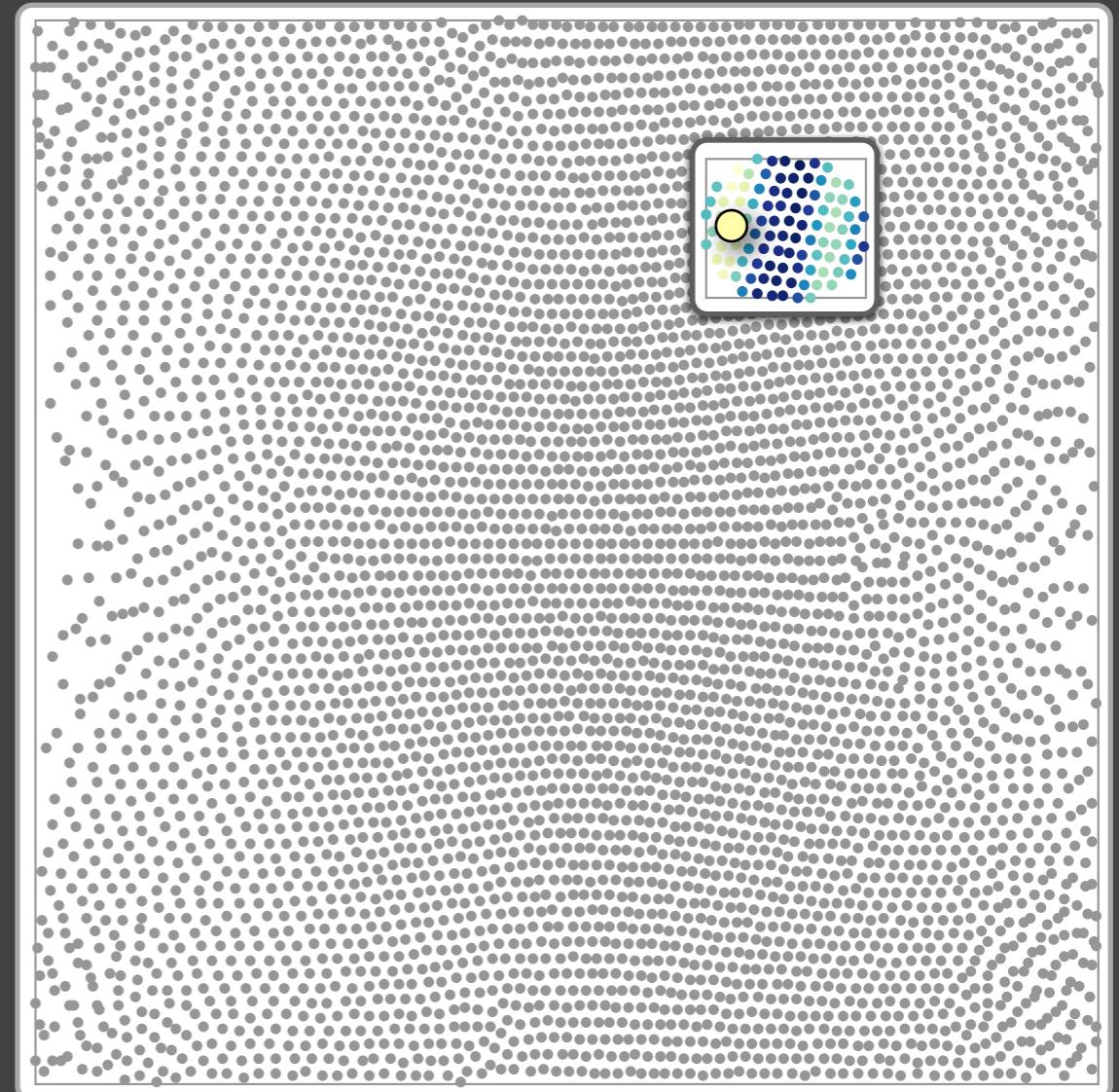
embryo B

creating a comparison summary

root-mean-square distance



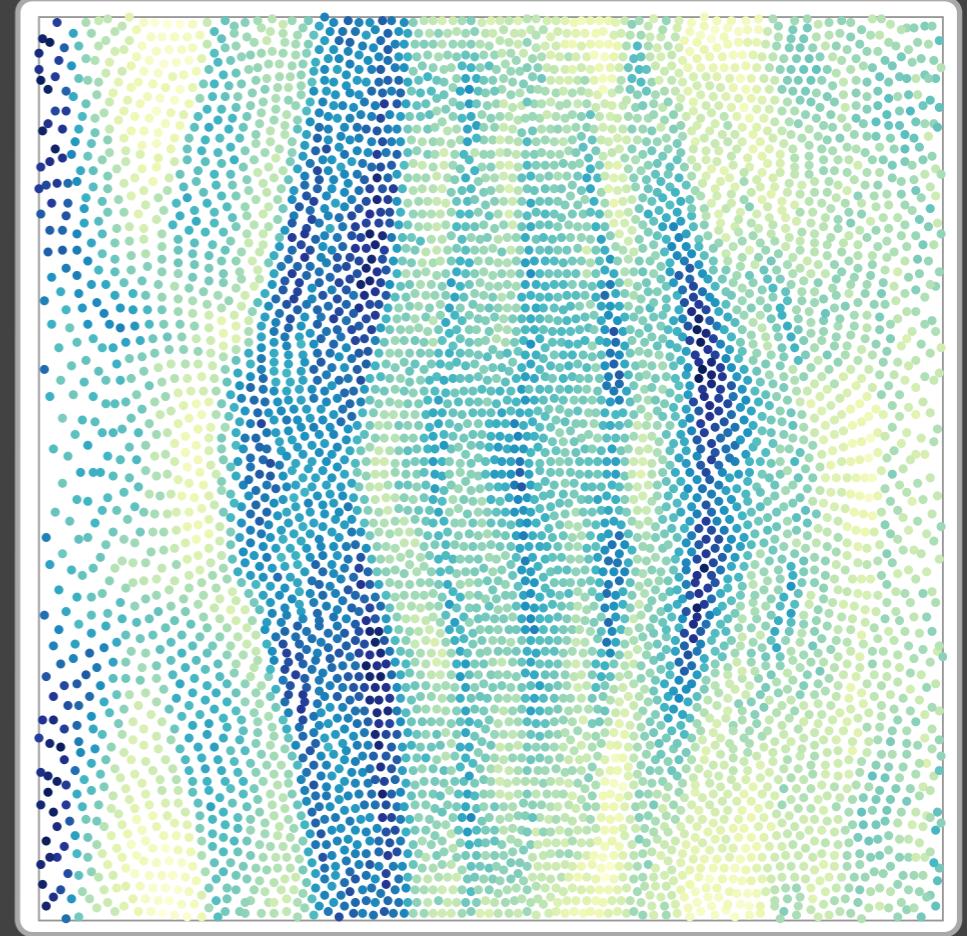
embryo A



embryo B

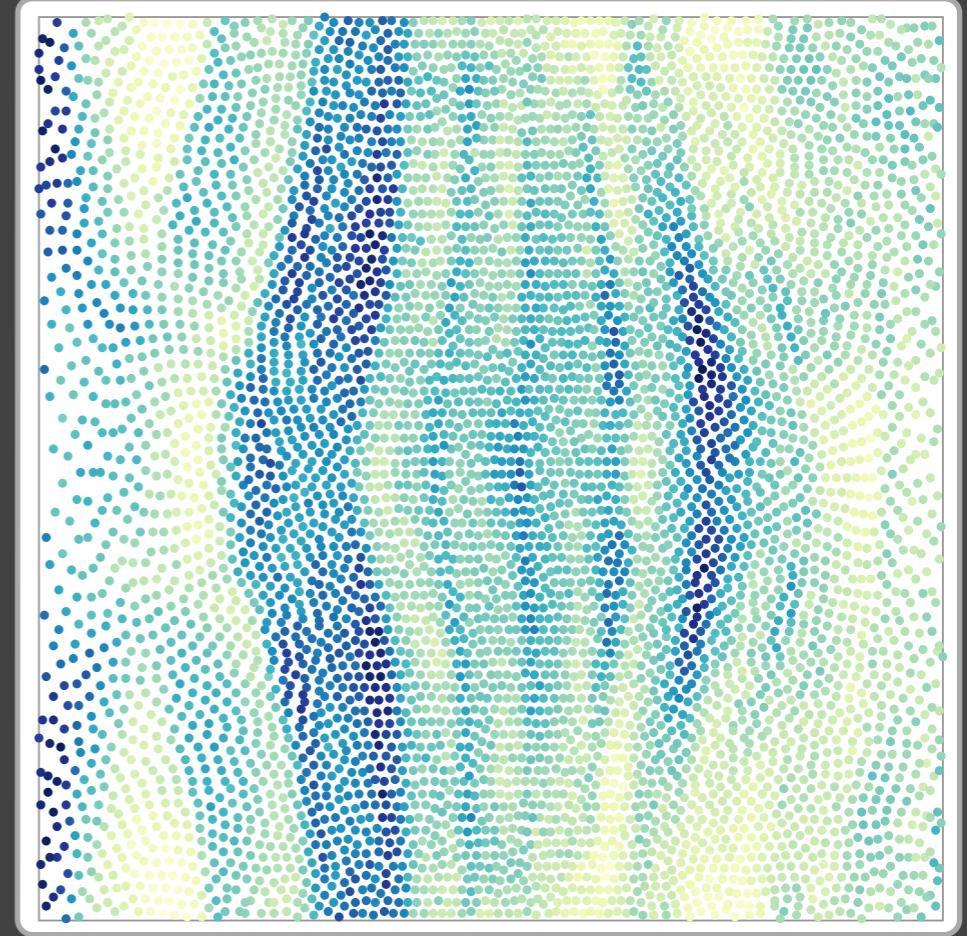
comparative summary components

aggregation group
metric
aggregation

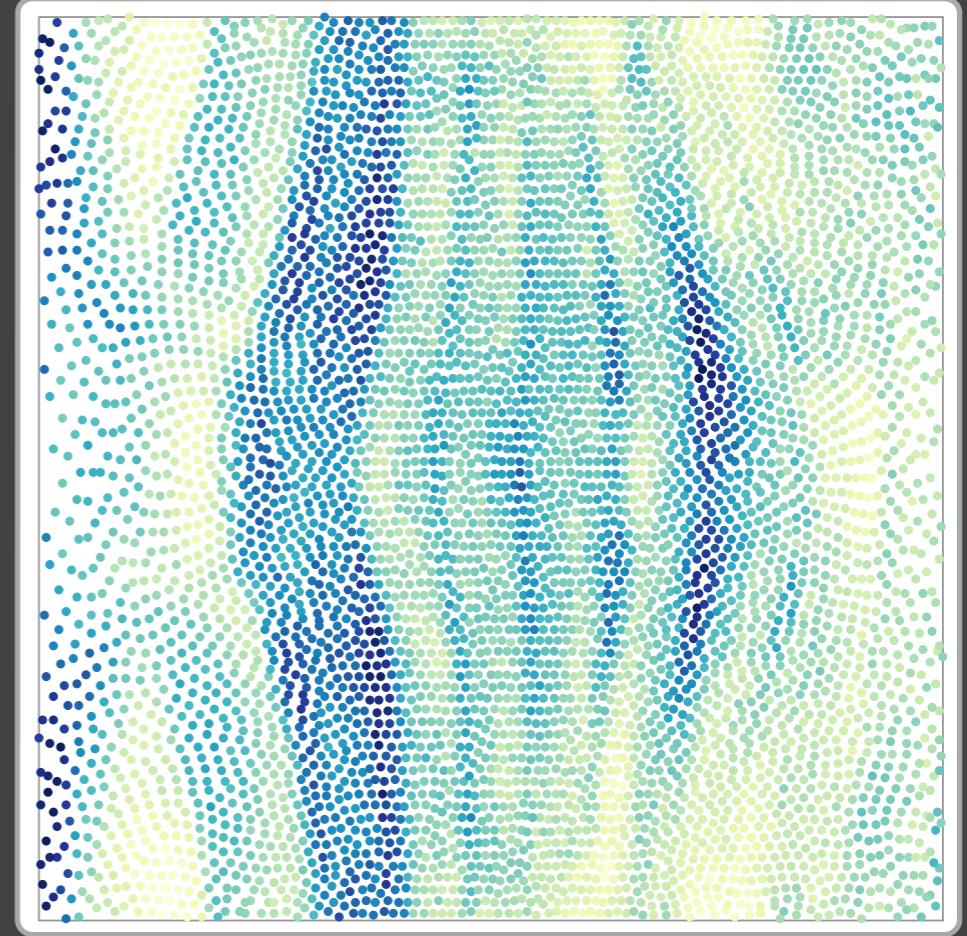


comparative summary components

aggregation group: *100 spatially closest cells*
metric
aggregation



comparative summary components

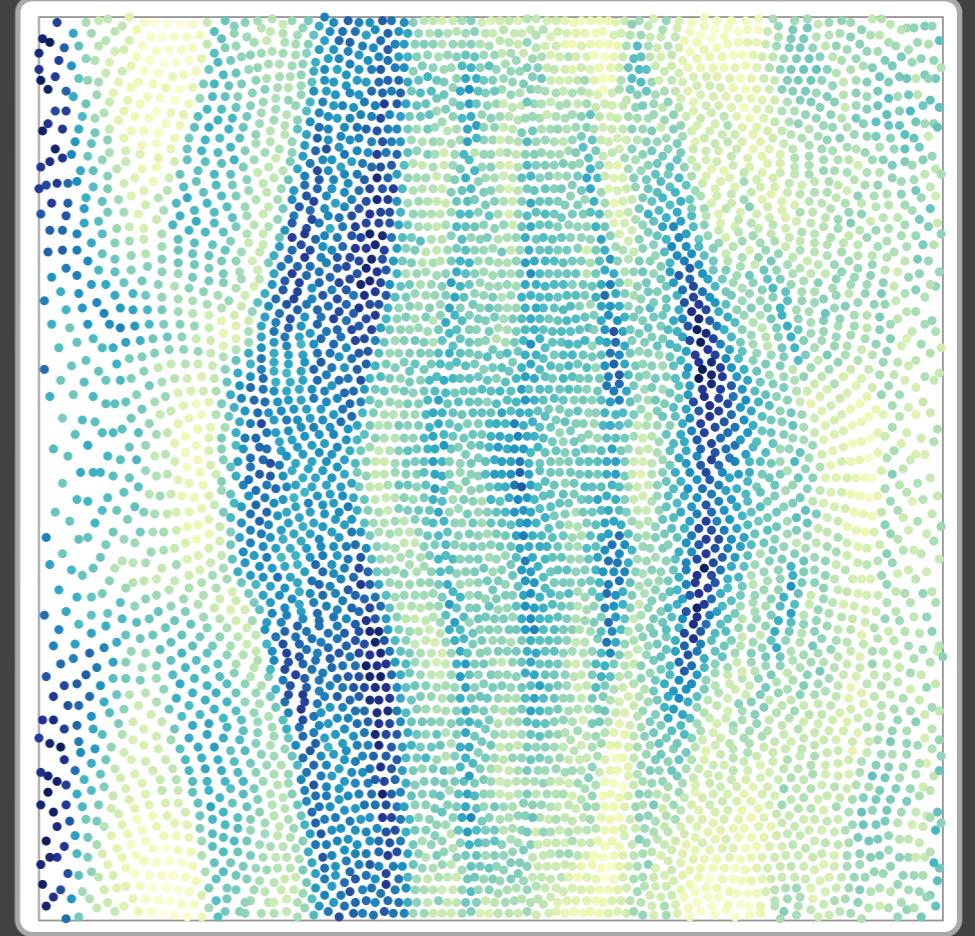


aggregation group: 100 *spatially closest cells*

metric: *root-mean-square distance*

aggregation

comparative summary components

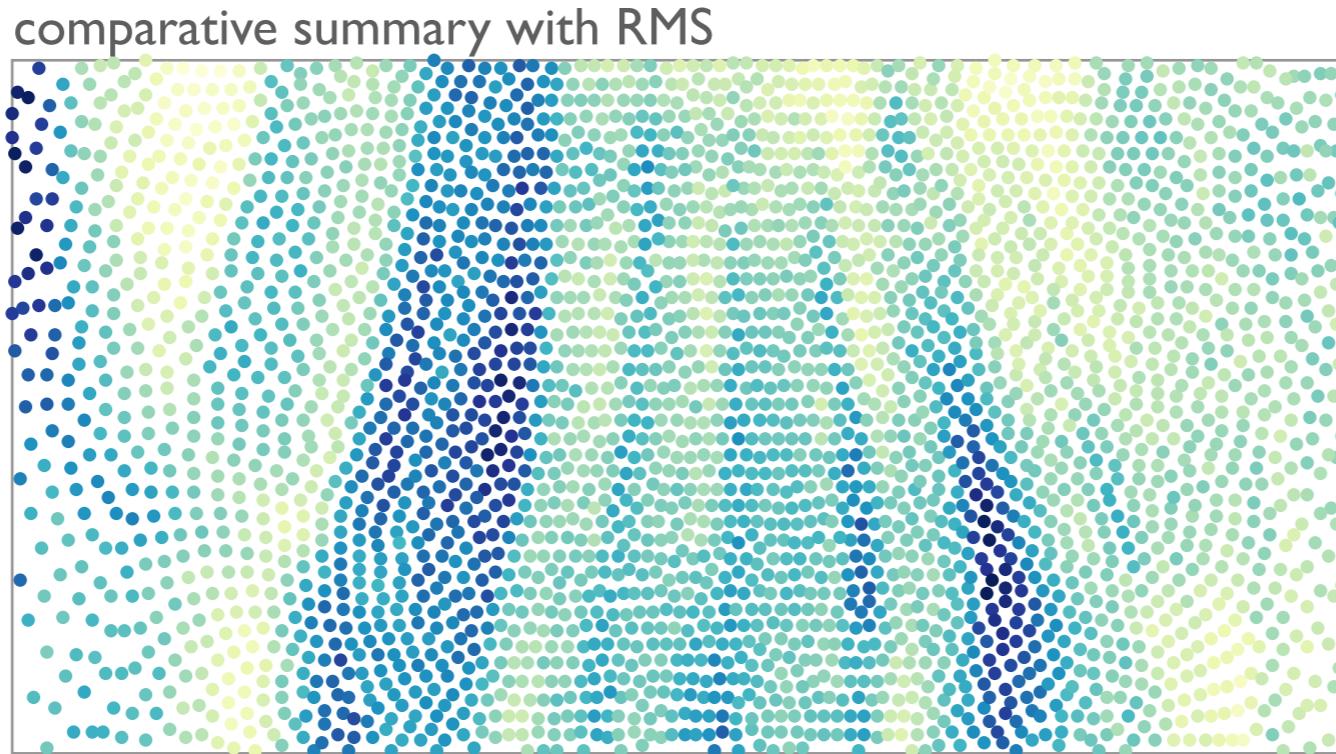


aggregation group: 100 *spatially closest cells*

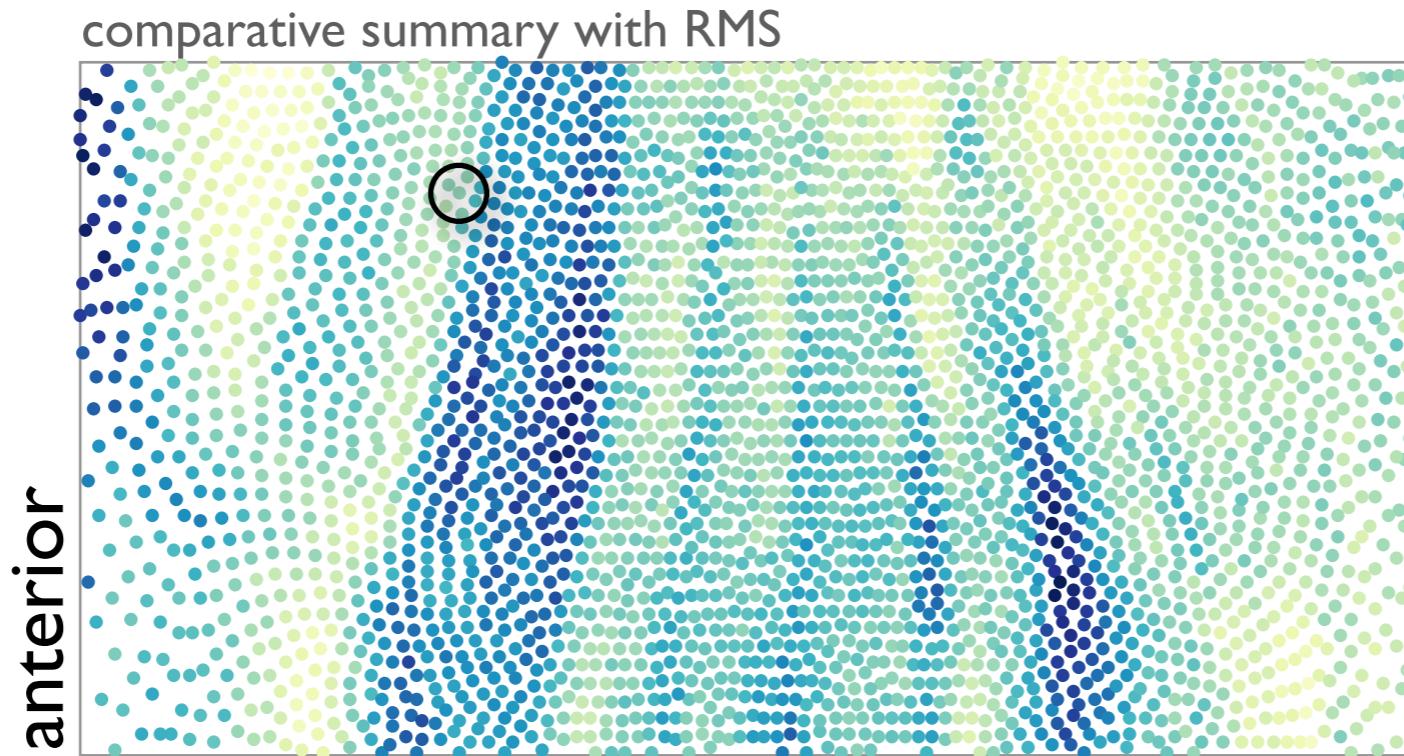
metric: *root-mean-square distance*

aggregation: *min operator*

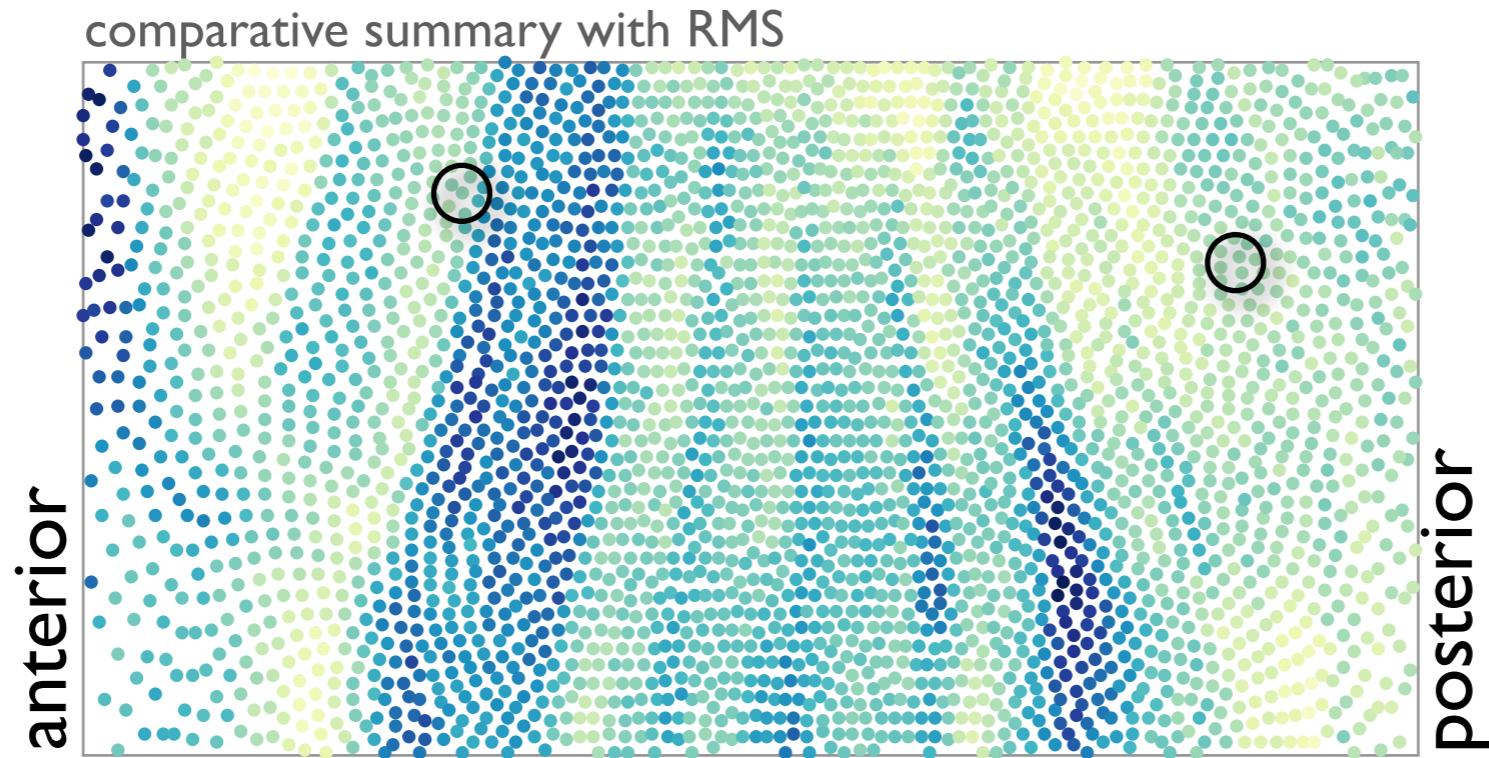
case study one: characterizing a summary



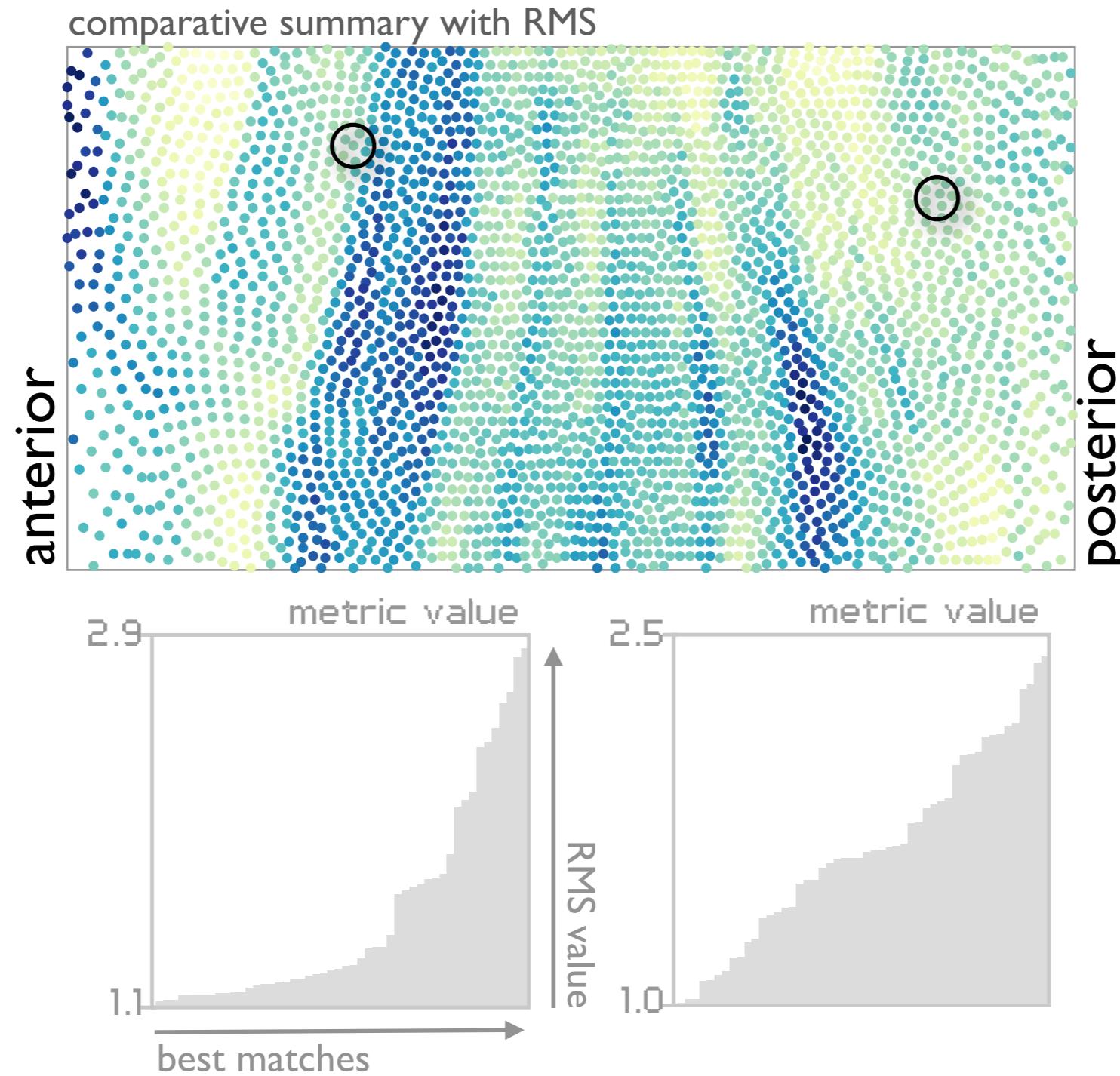
case study one: characterizing a summary



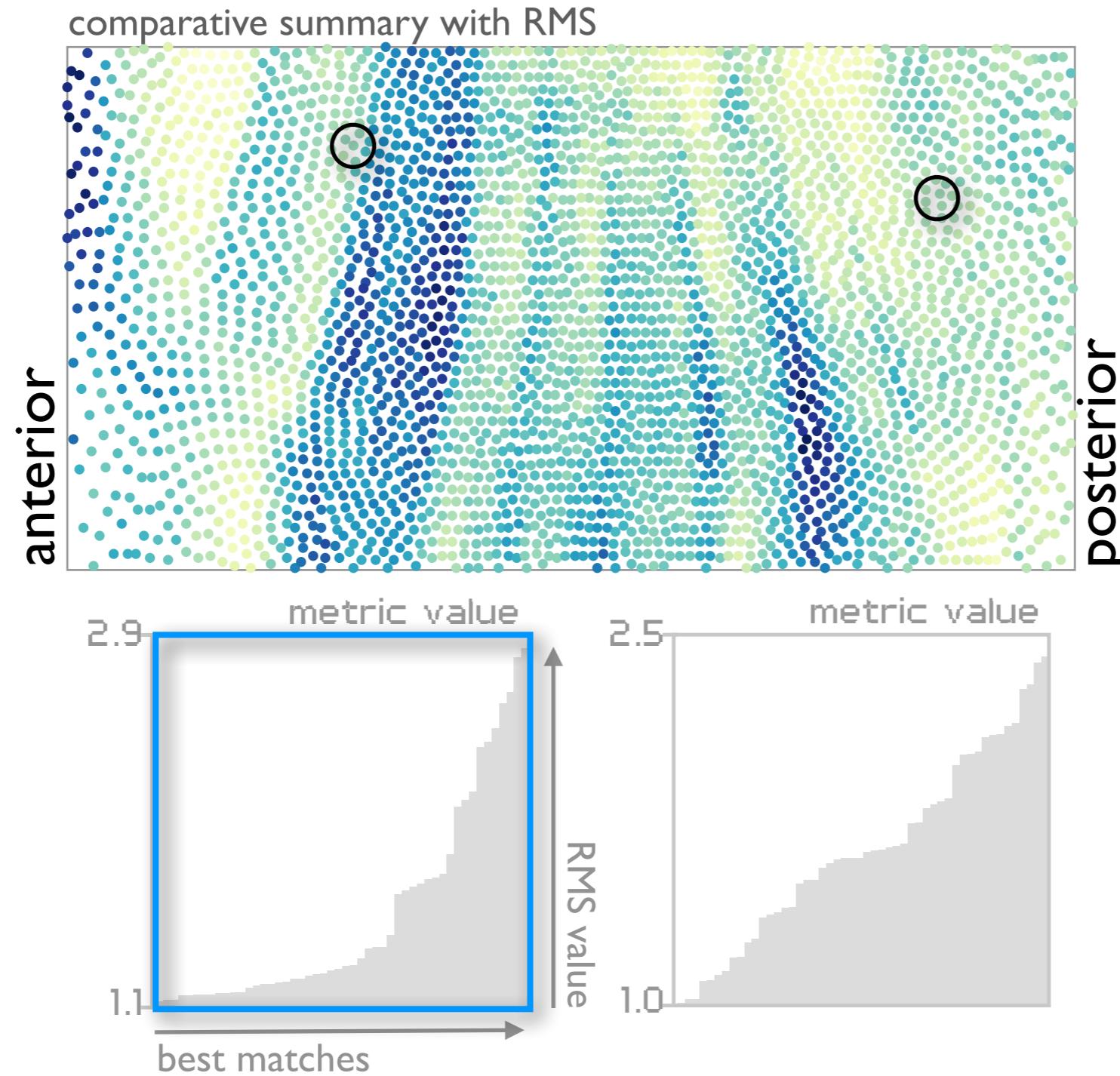
case study one: characterizing a summary



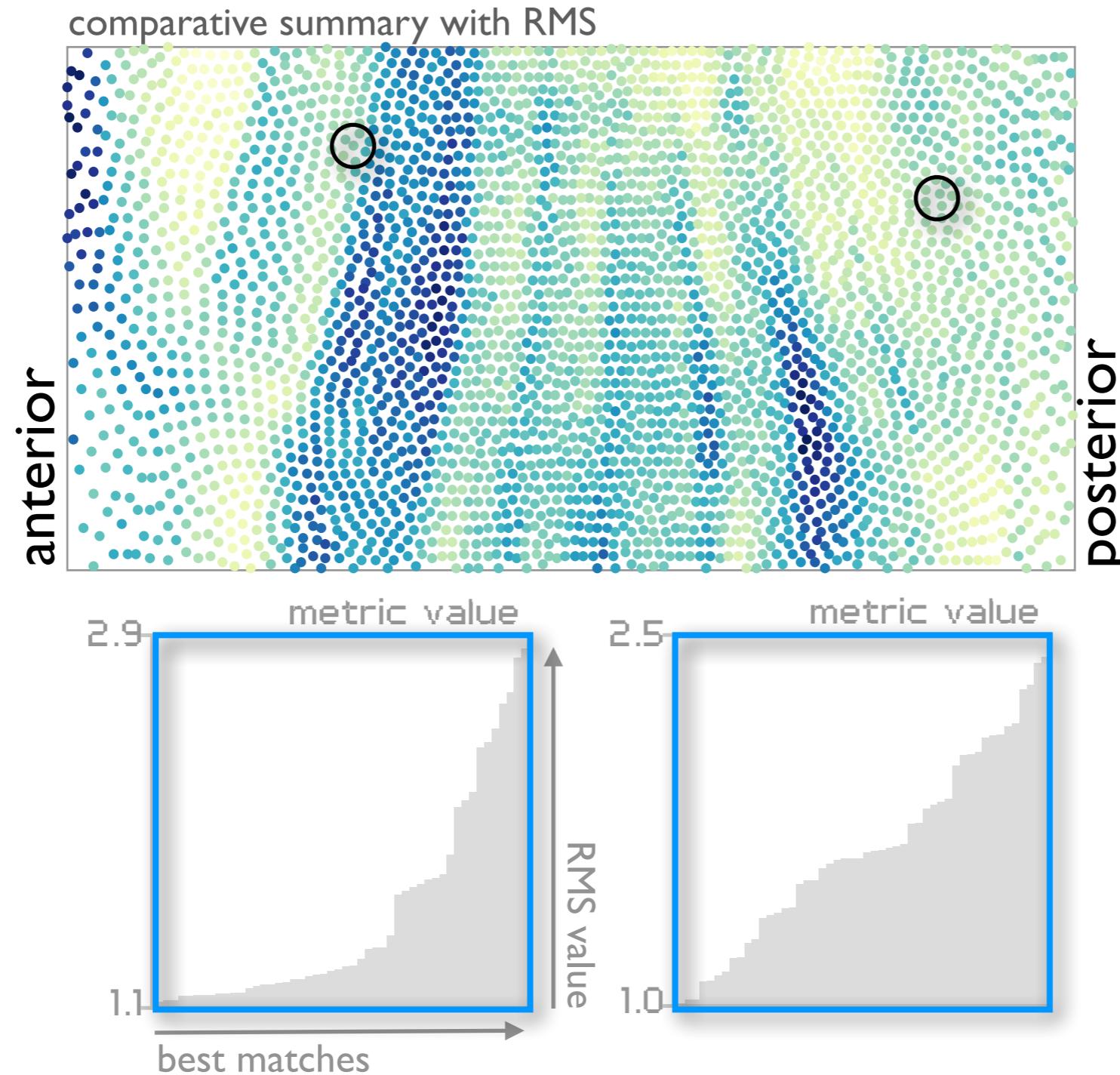
case study one: characterizing a summary



case study one: characterizing a summary



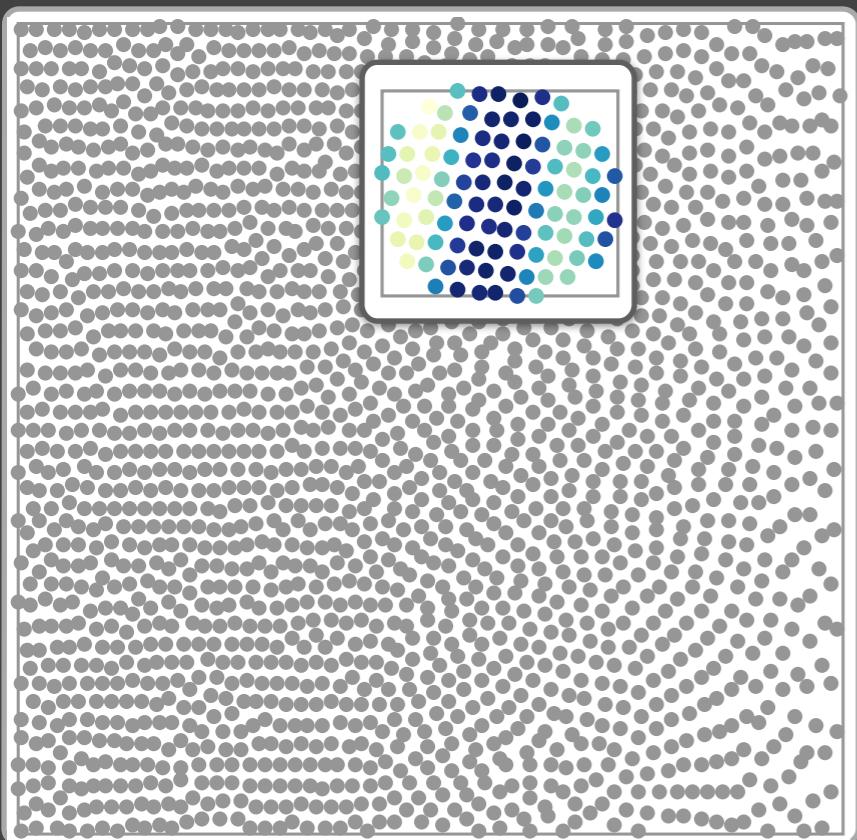
case study one: characterizing a summary



group
a set of cells

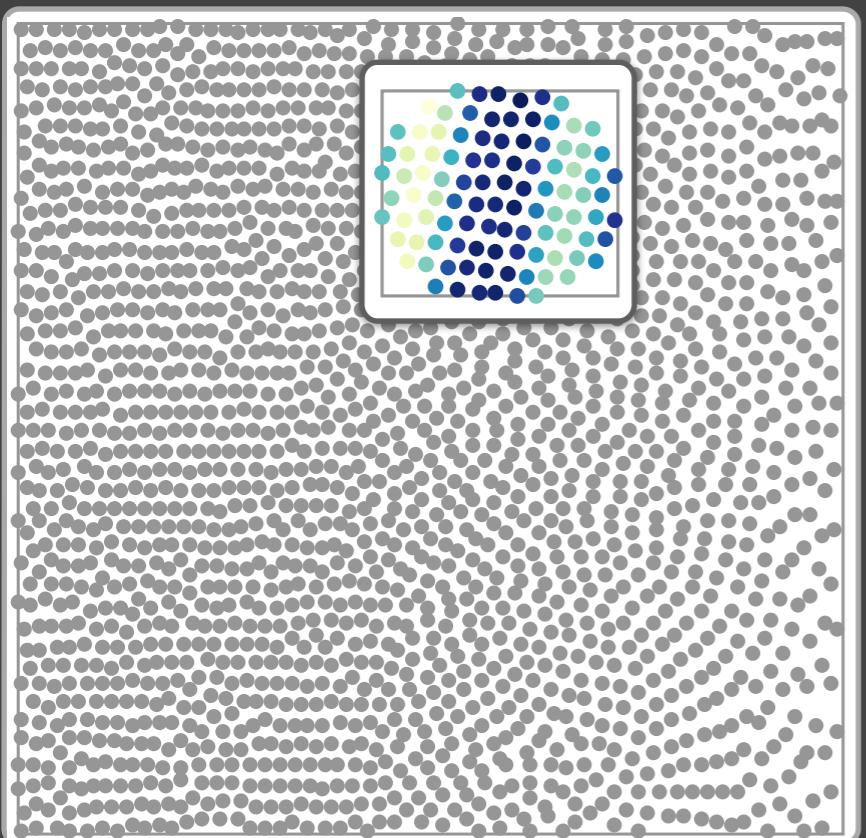
group
a set of cells

aggregation group

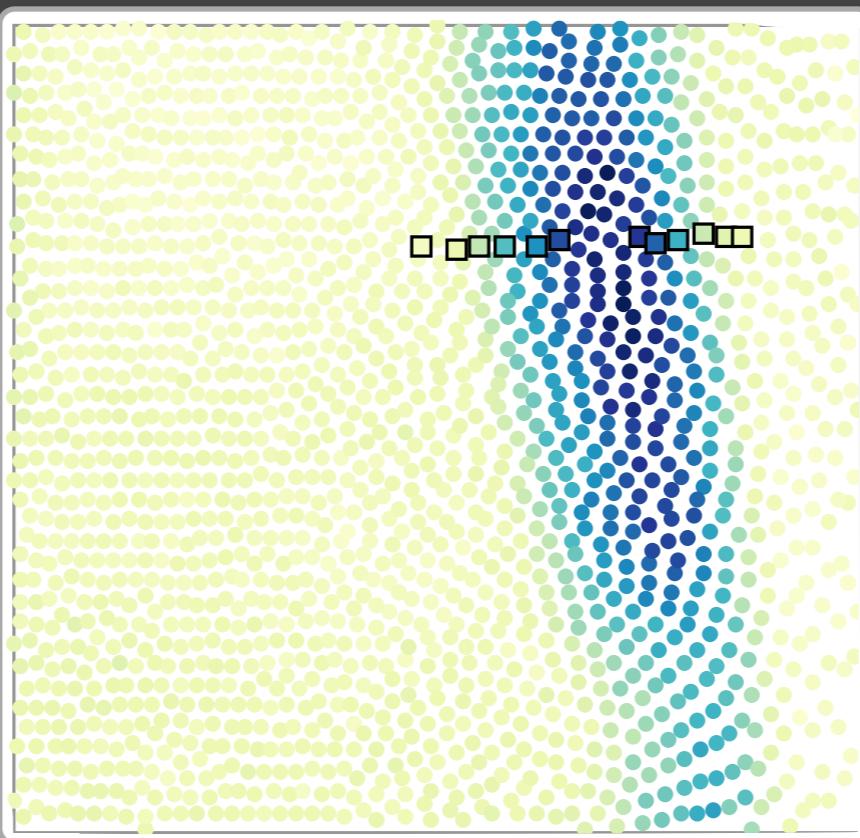


group
a set of cells

aggregation group

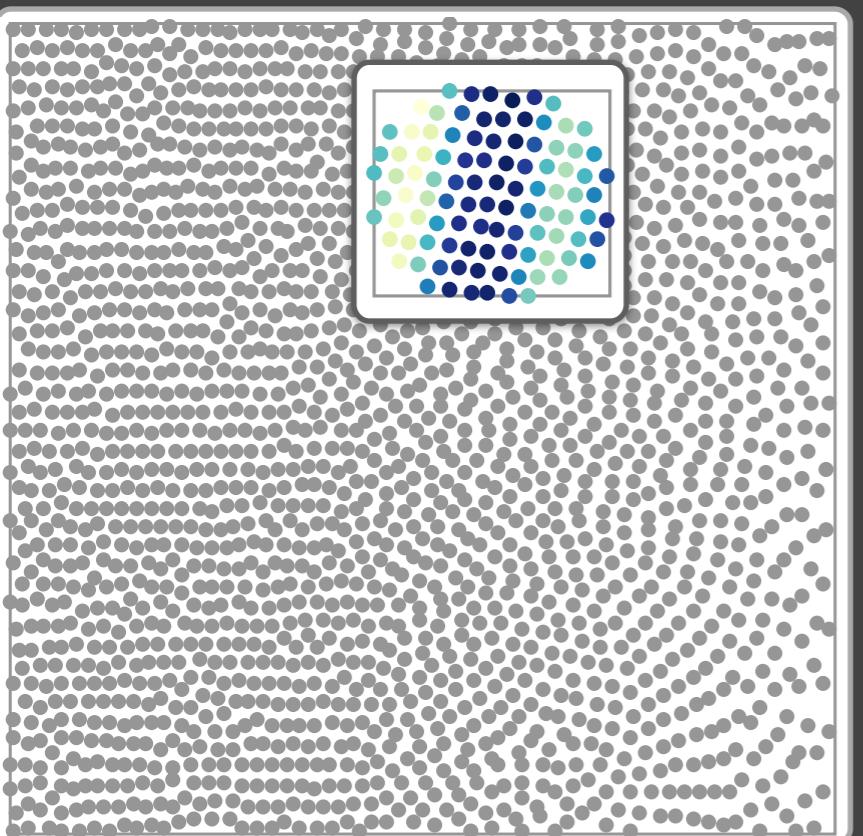


created group

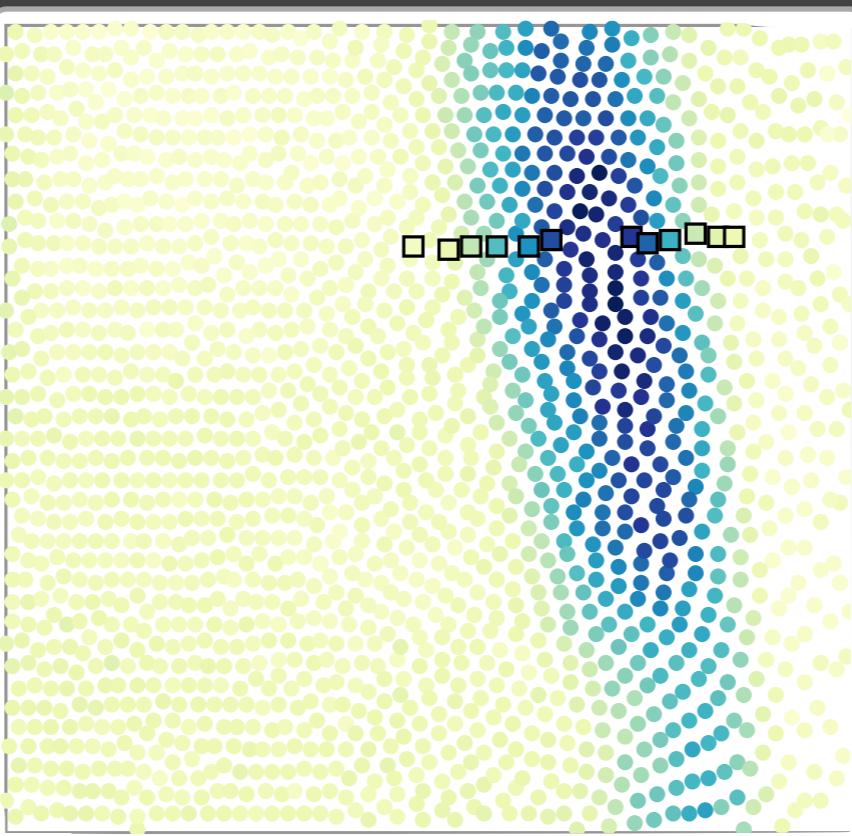


group
a set of cells

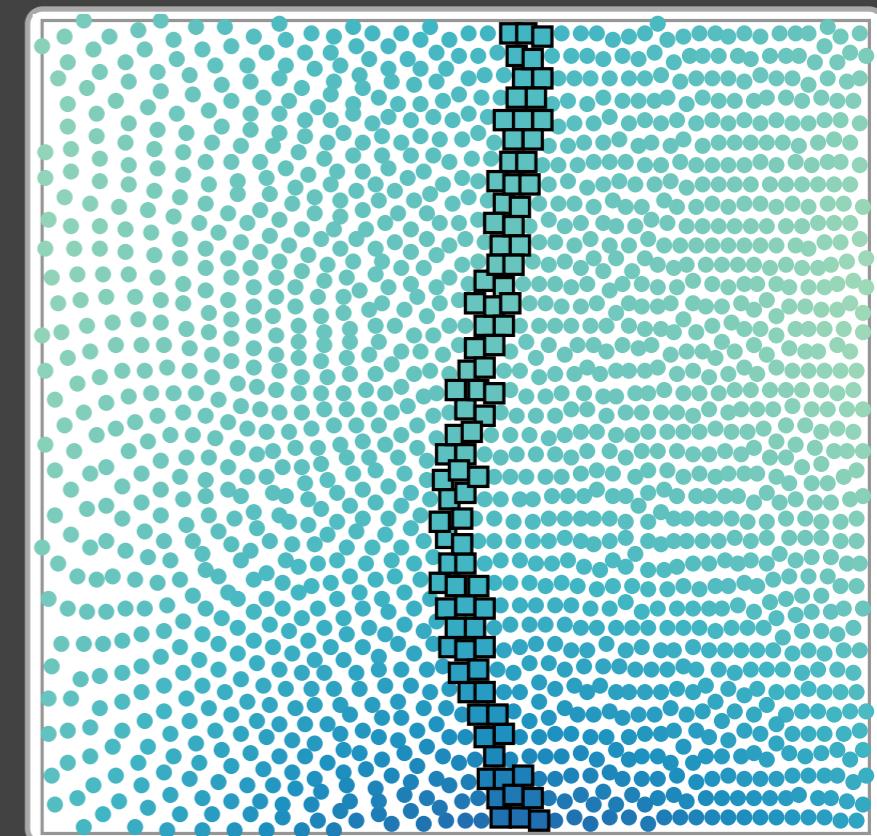
aggregation group



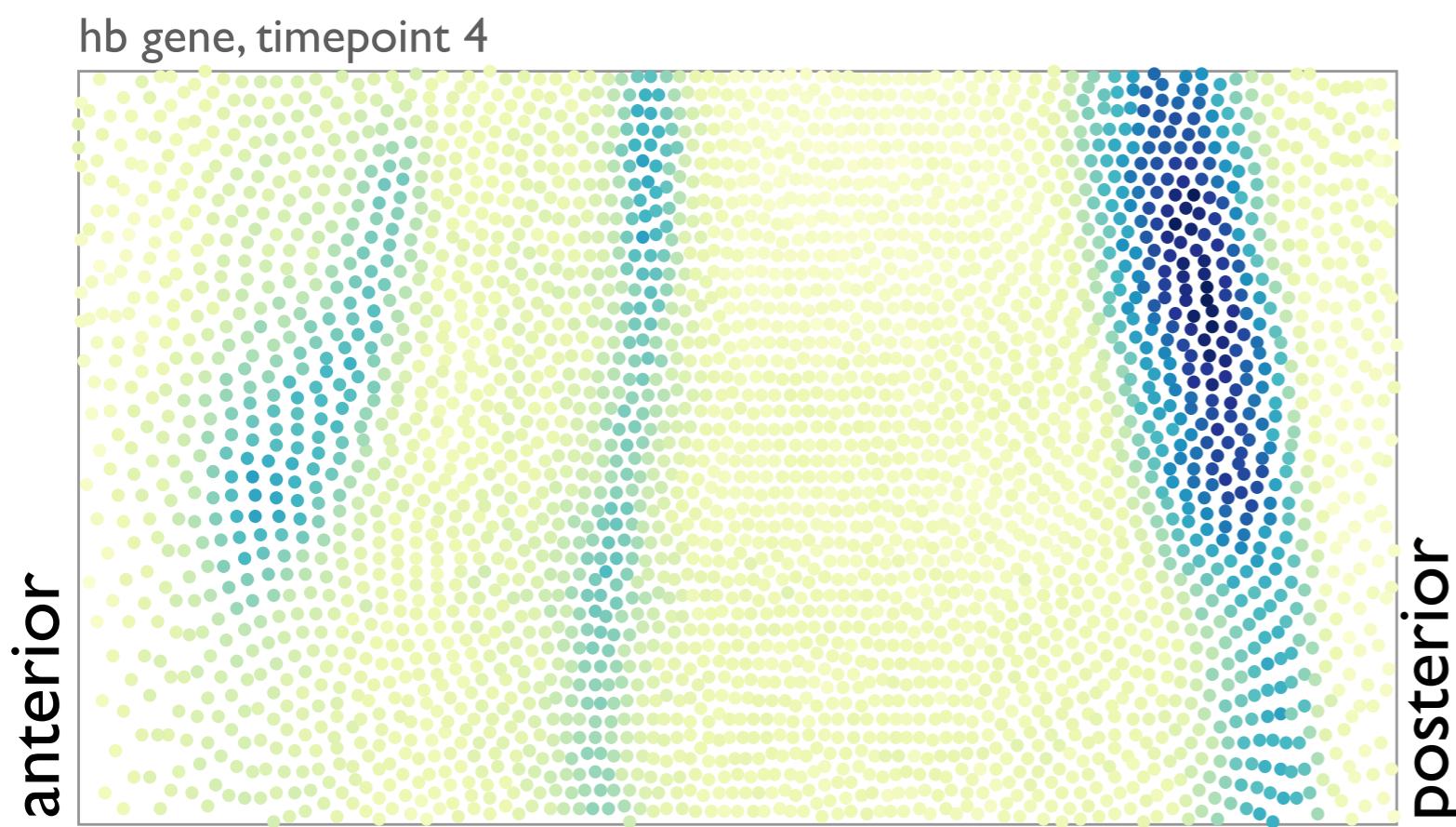
created group



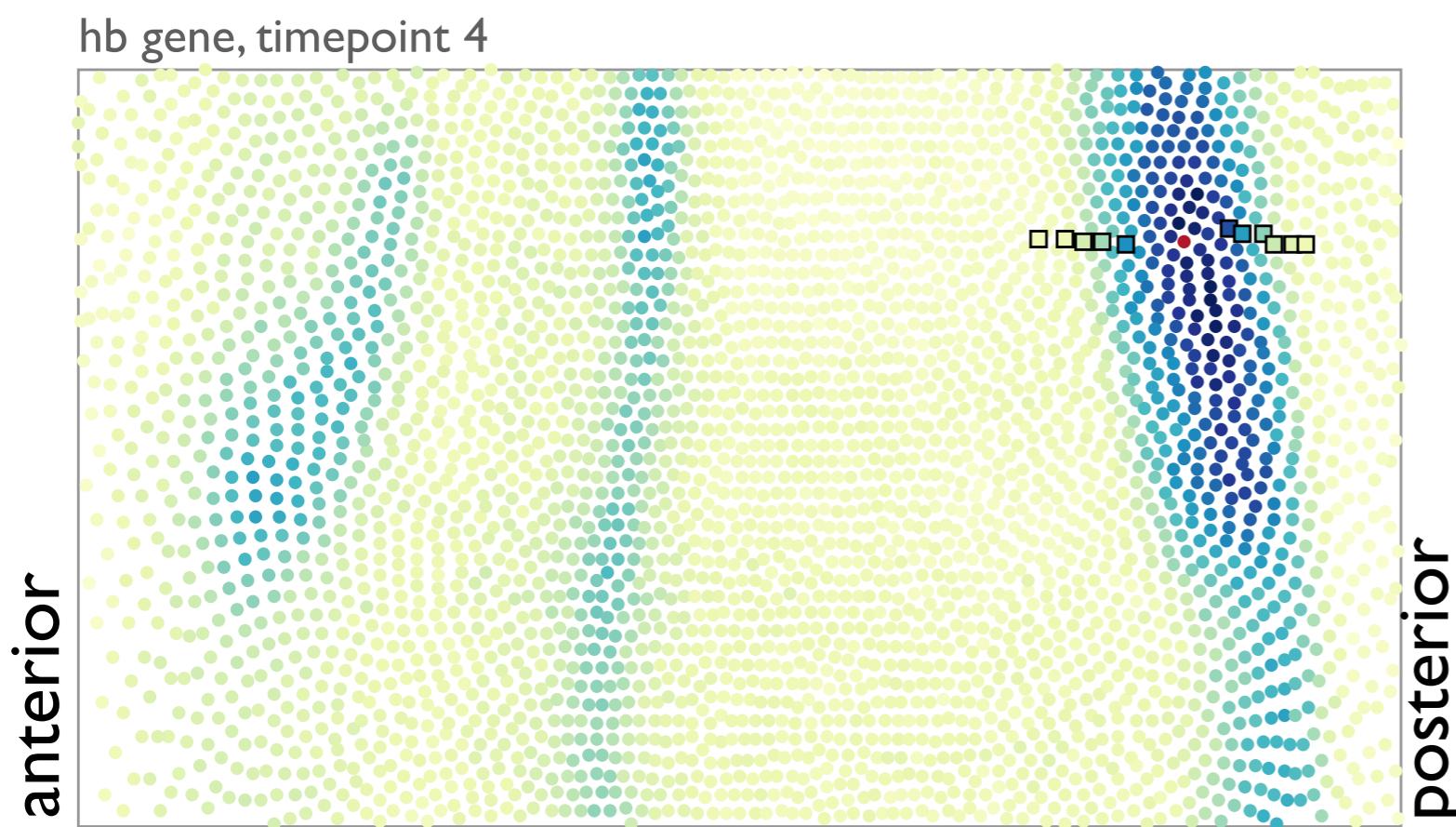
existing group



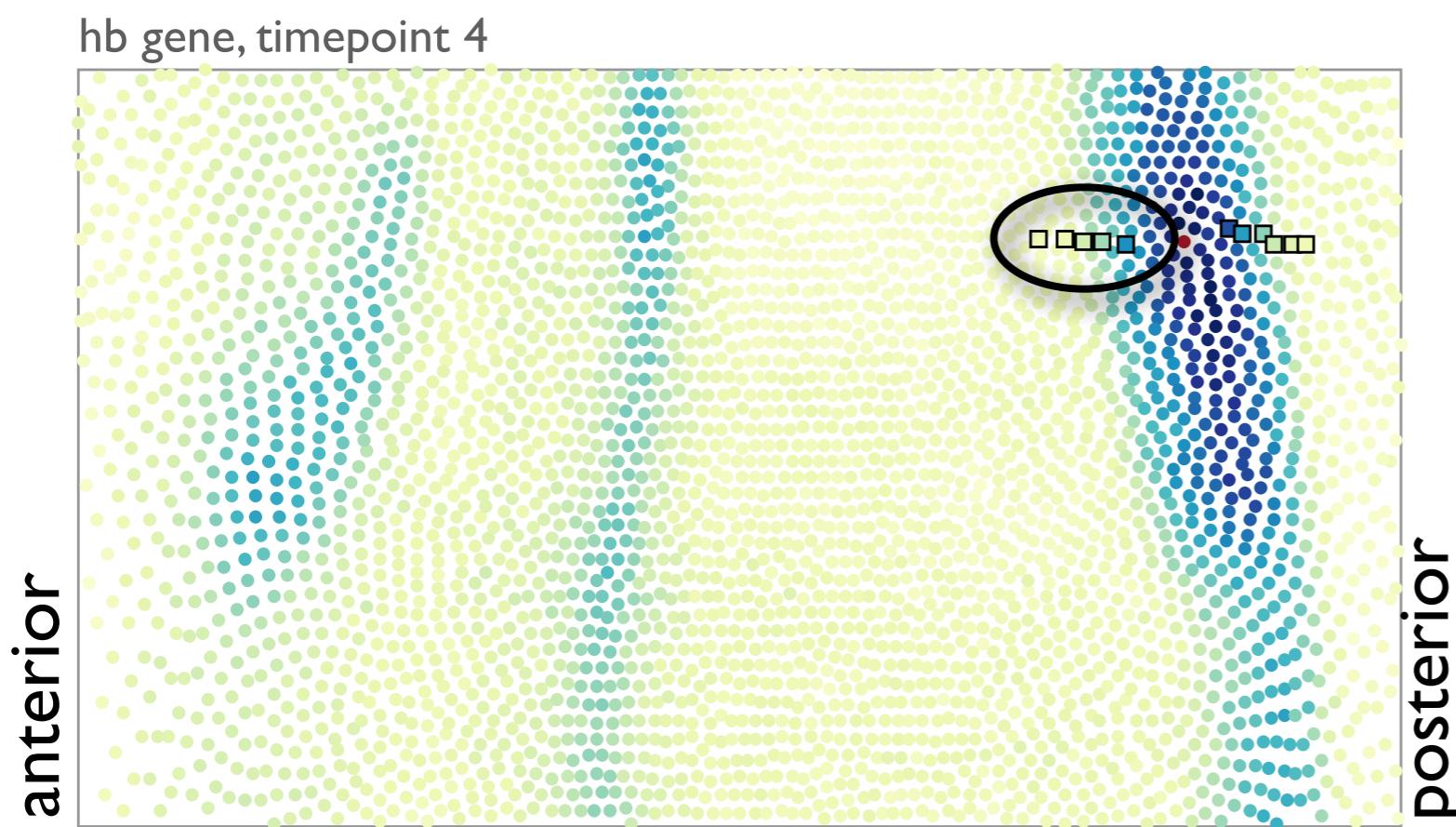
case study two: exploring groups



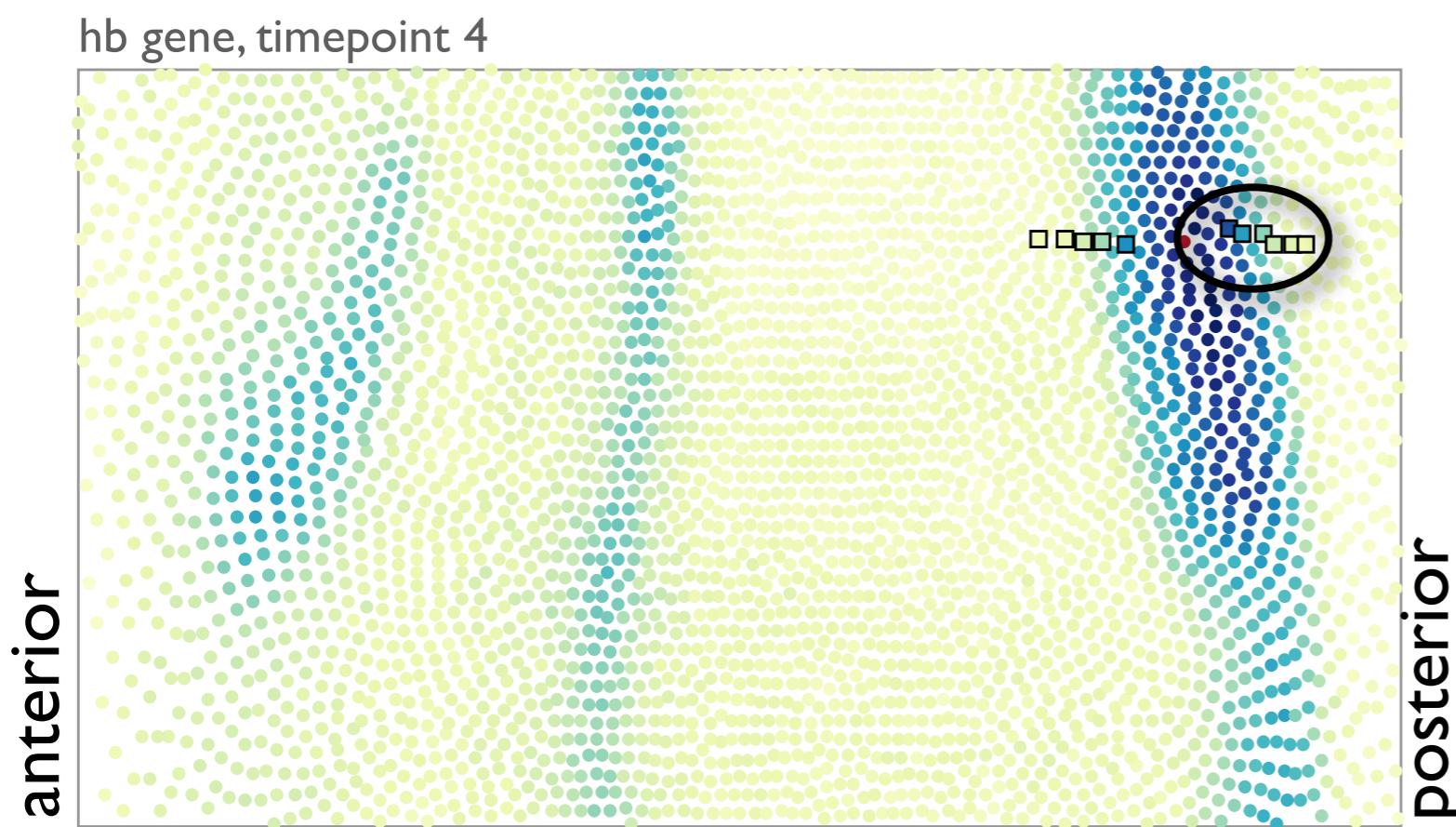
case study two: exploring groups



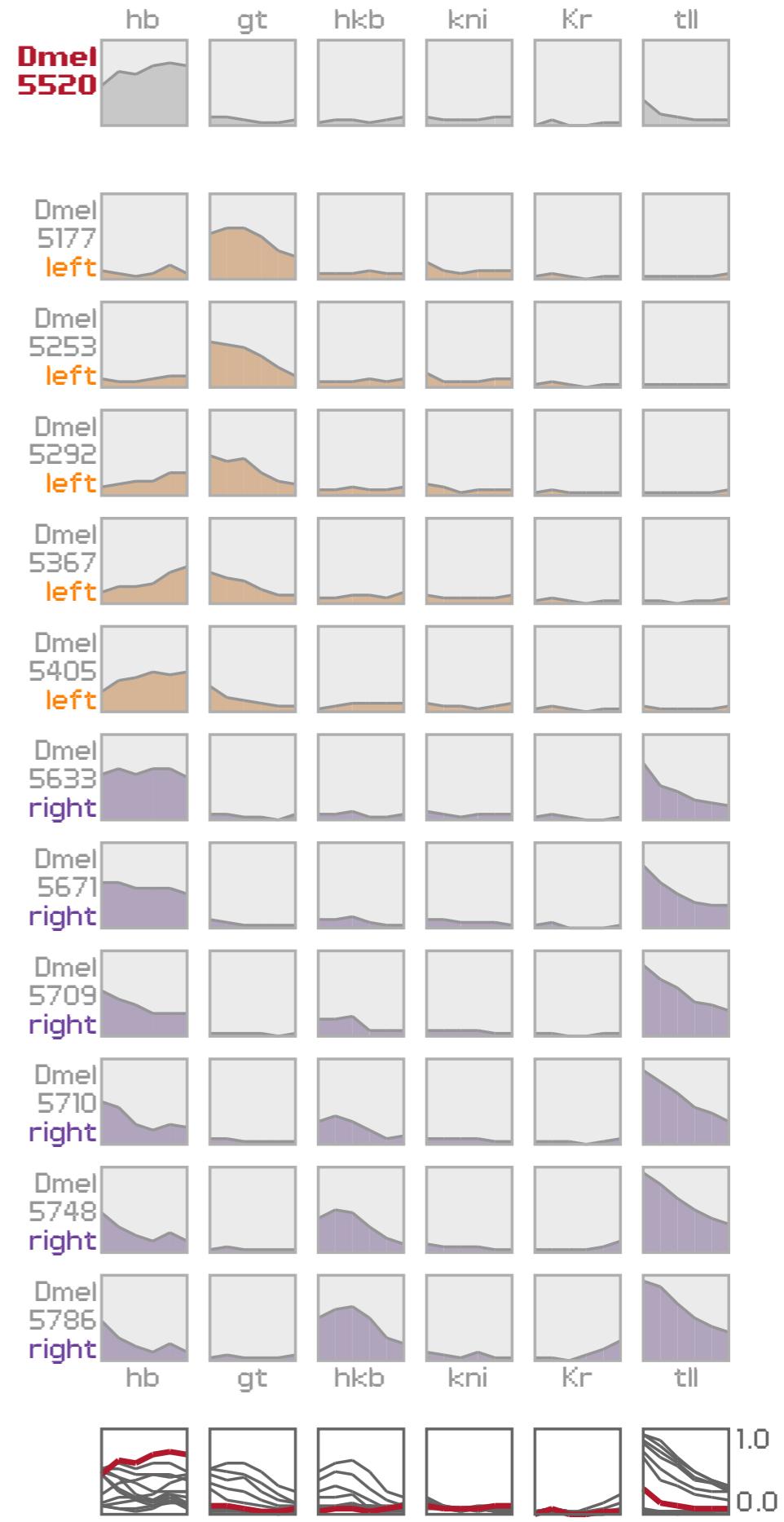
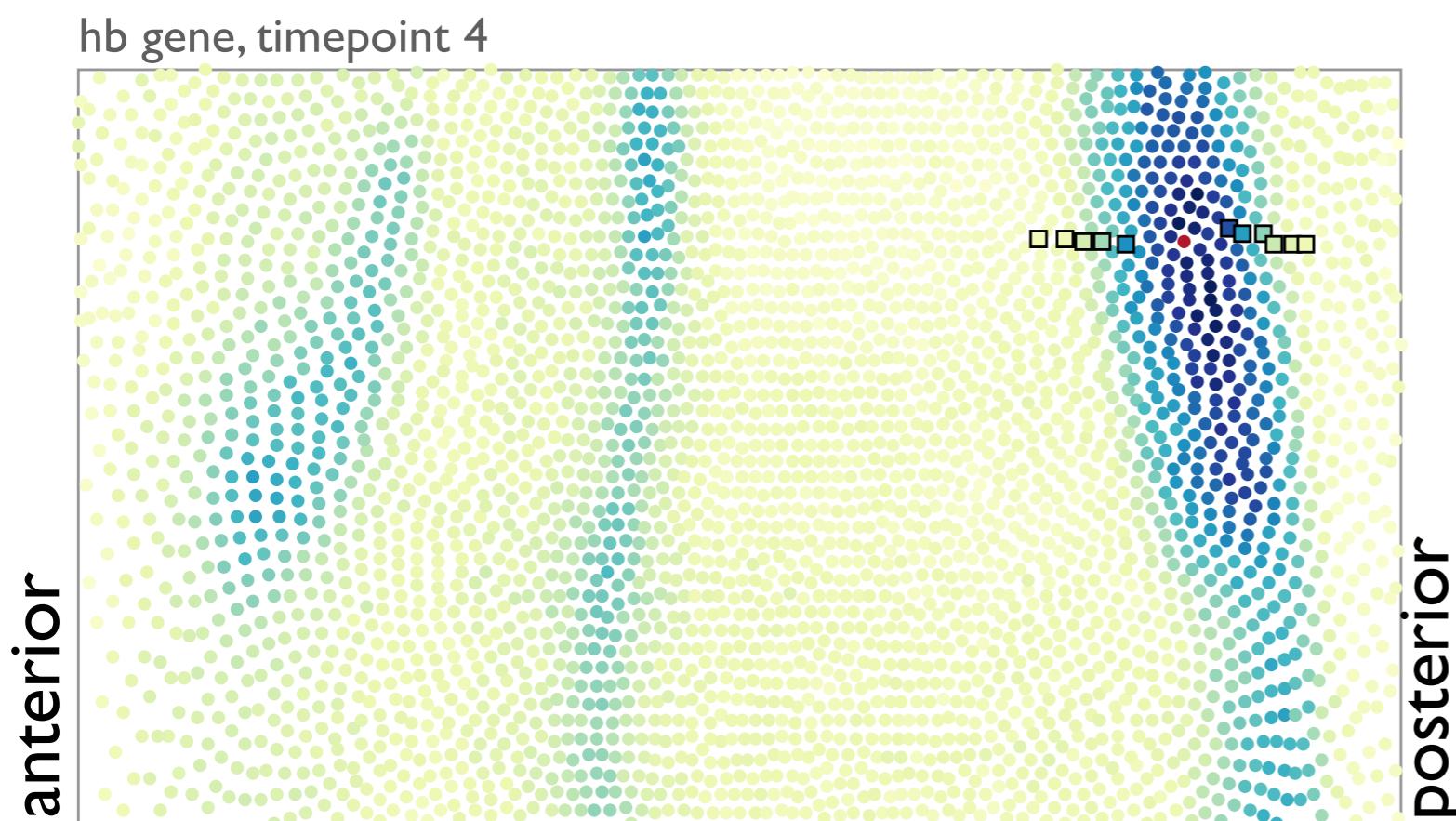
case study two: exploring groups



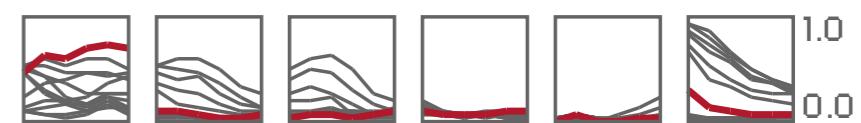
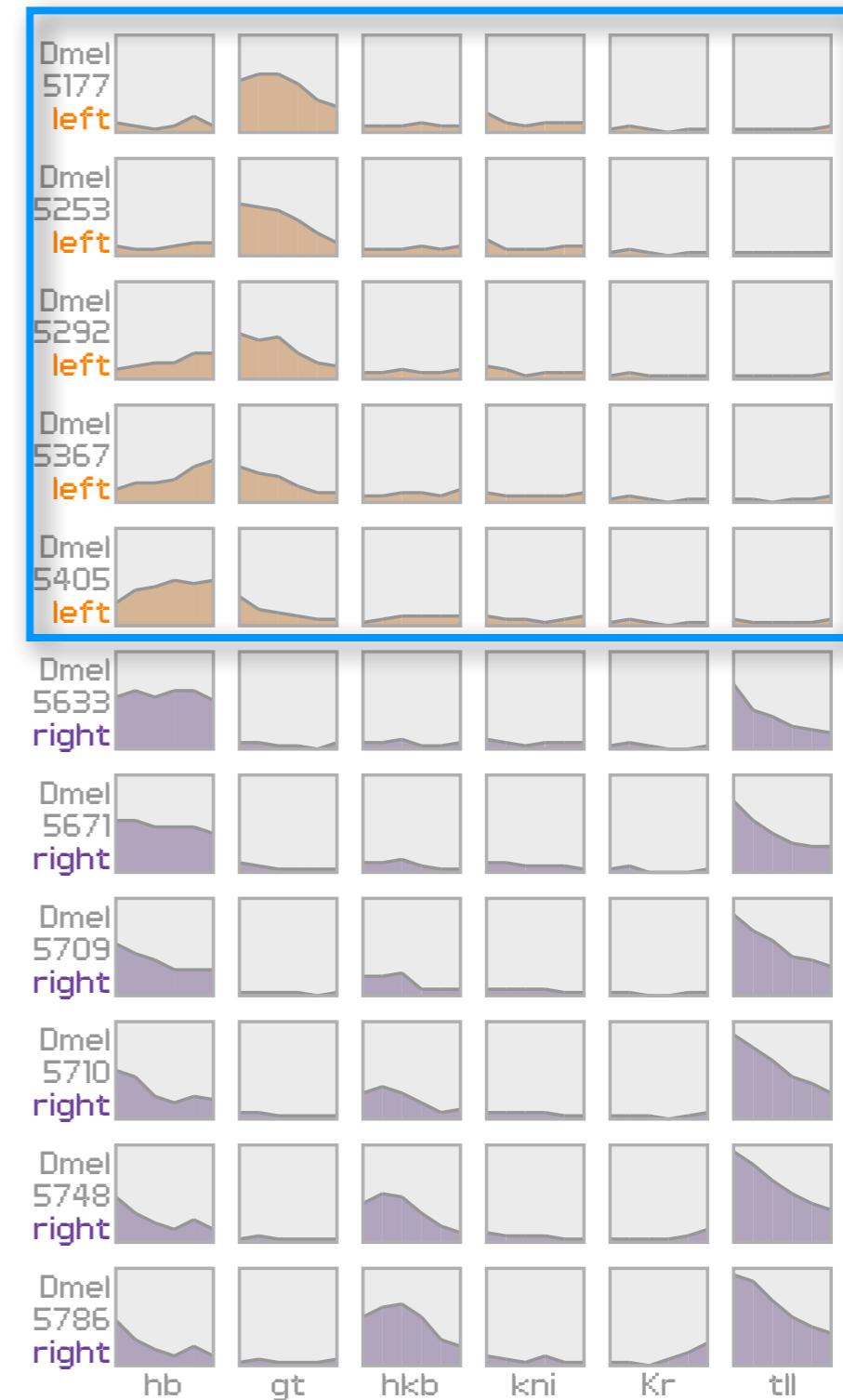
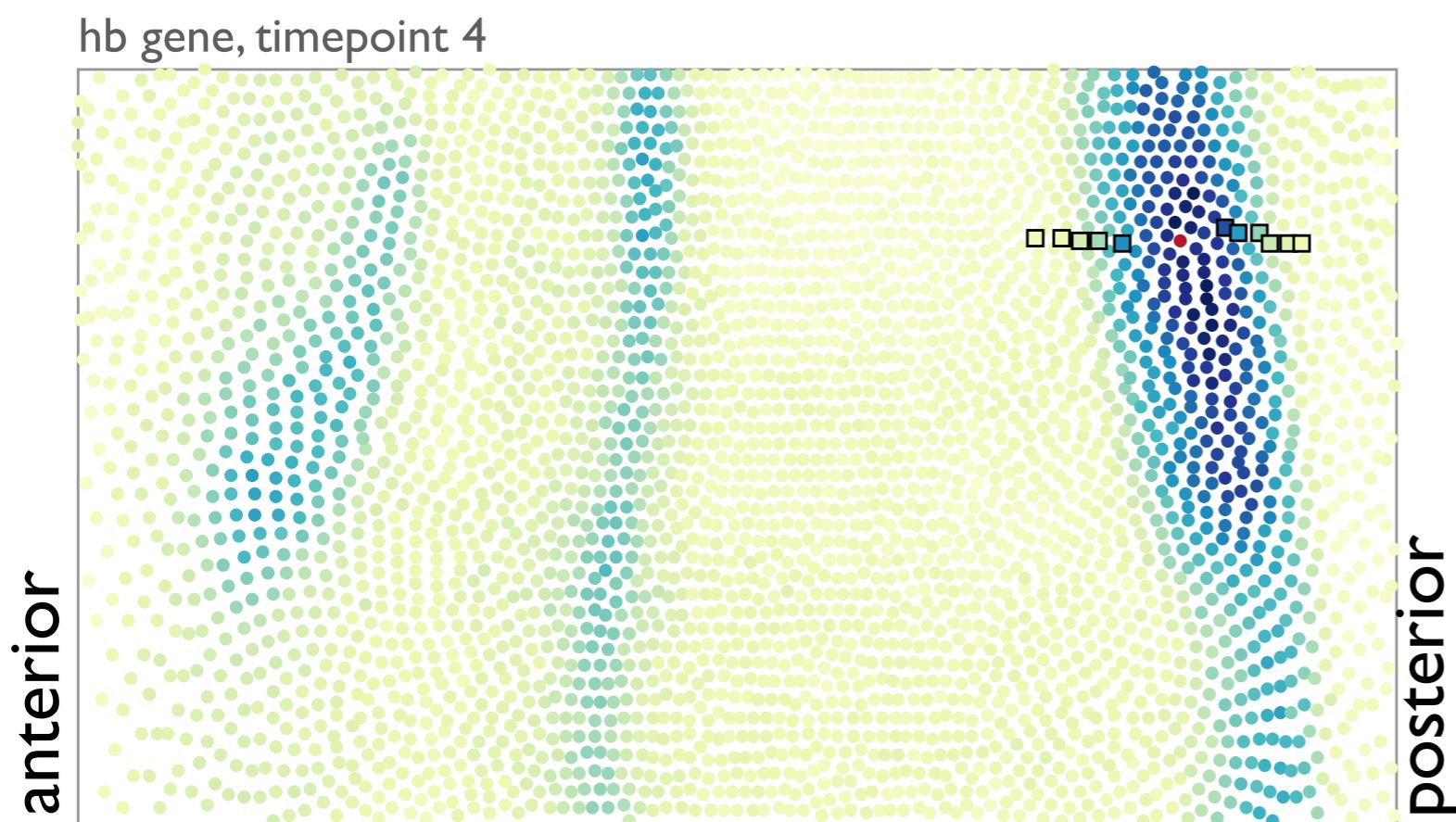
case study two: exploring groups



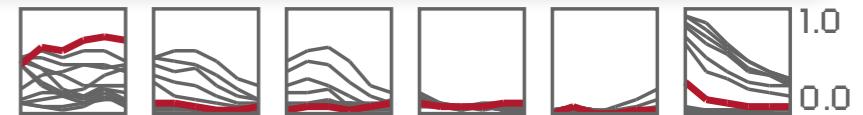
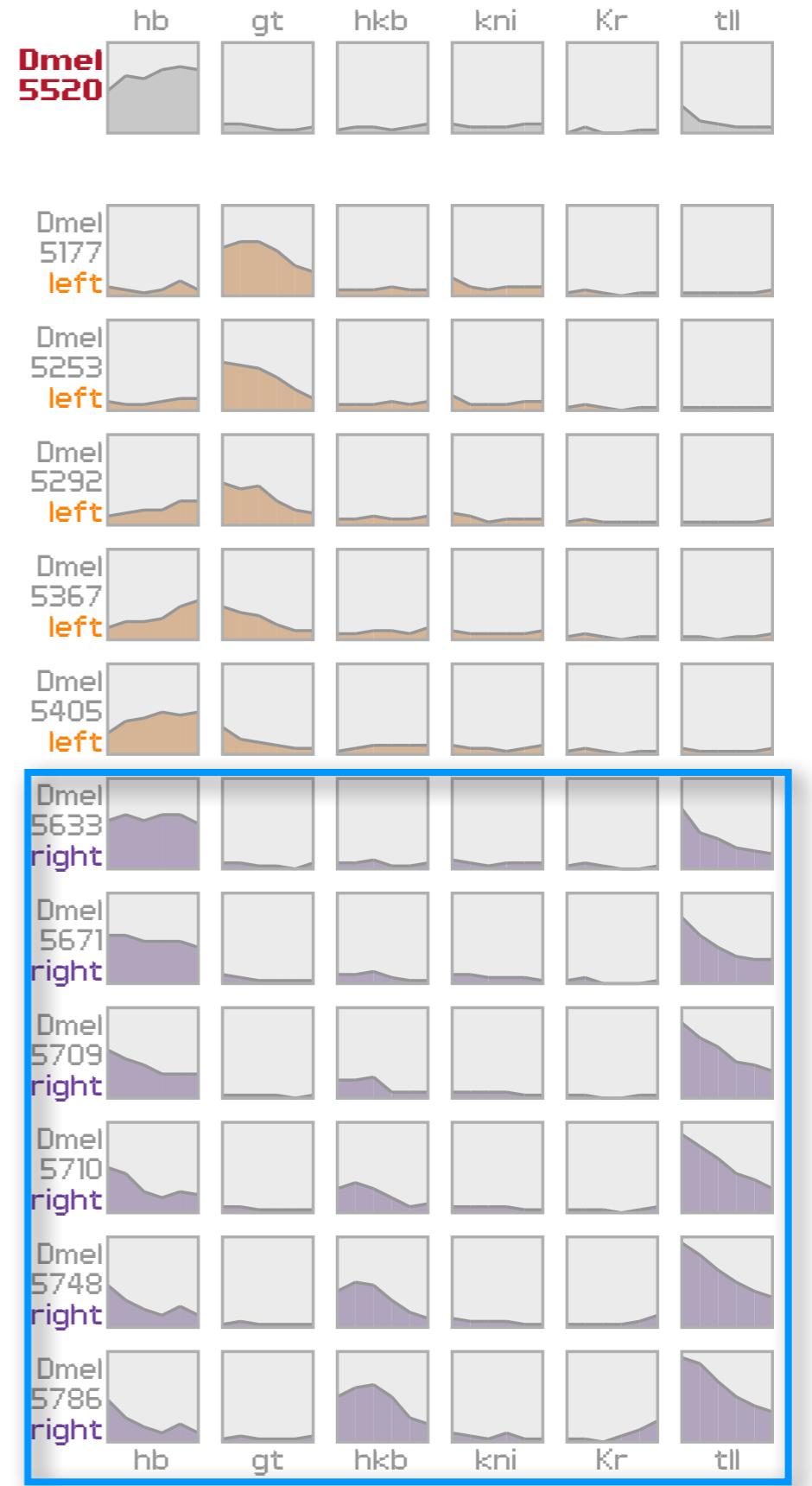
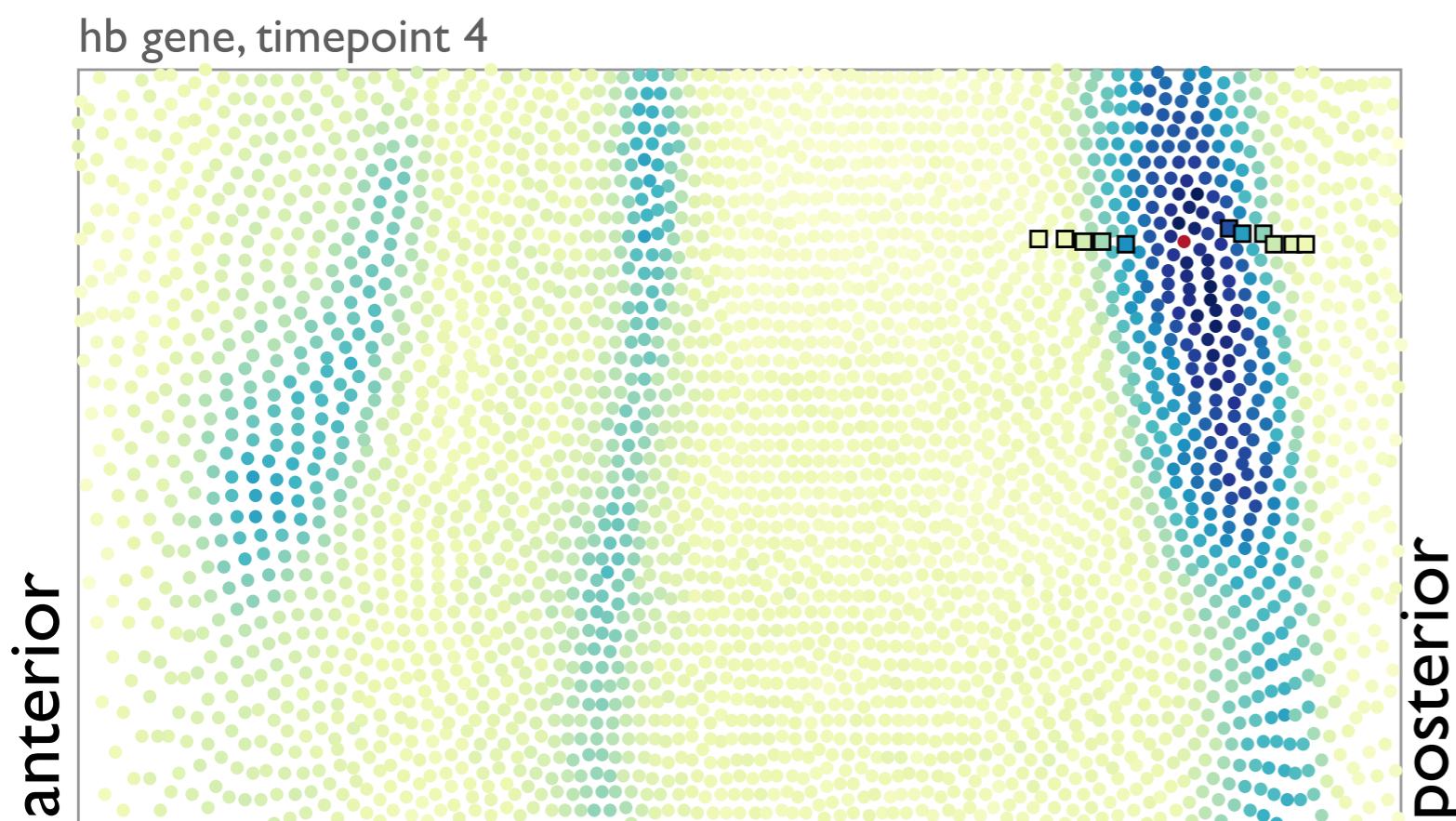
case study two: exploring groups



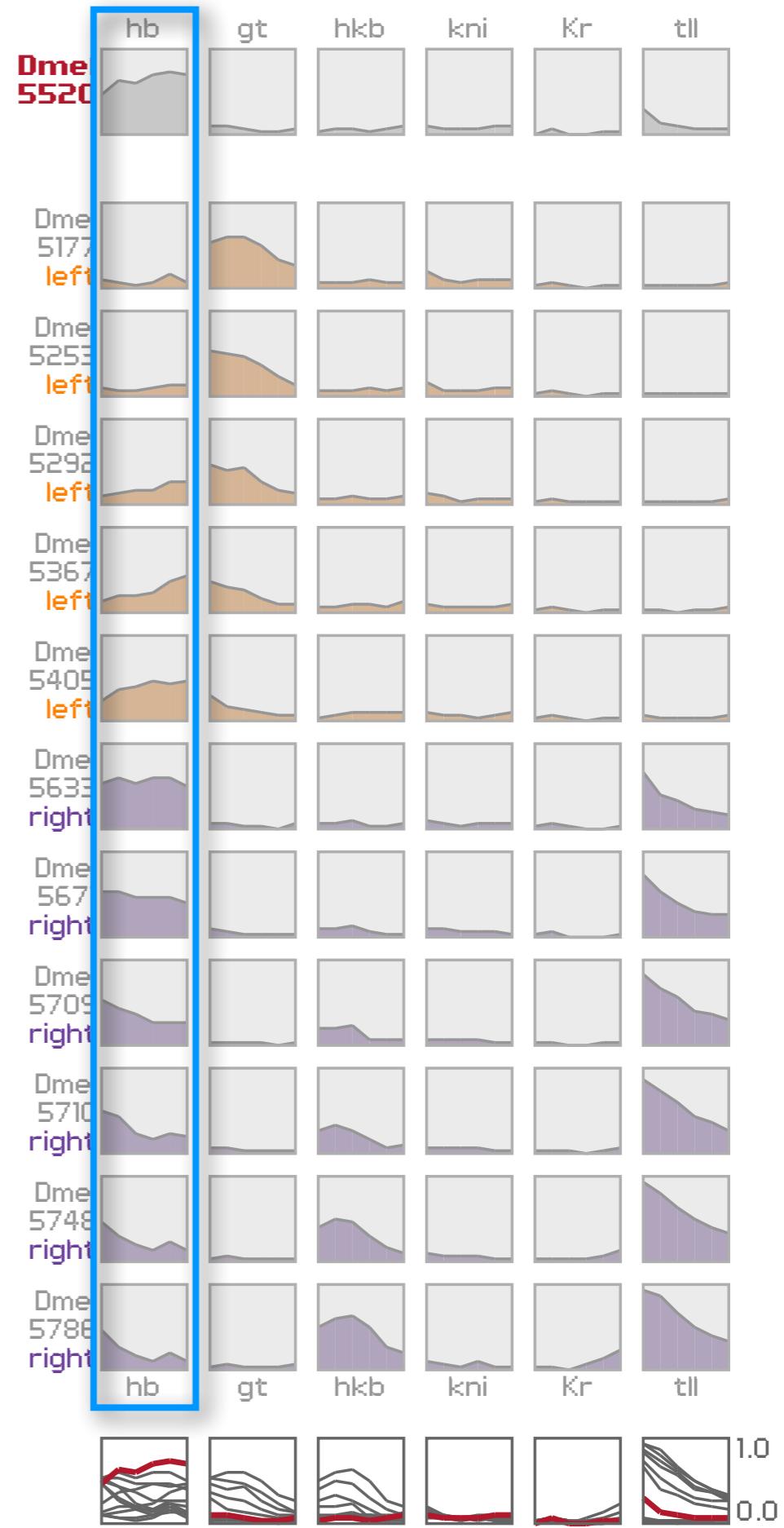
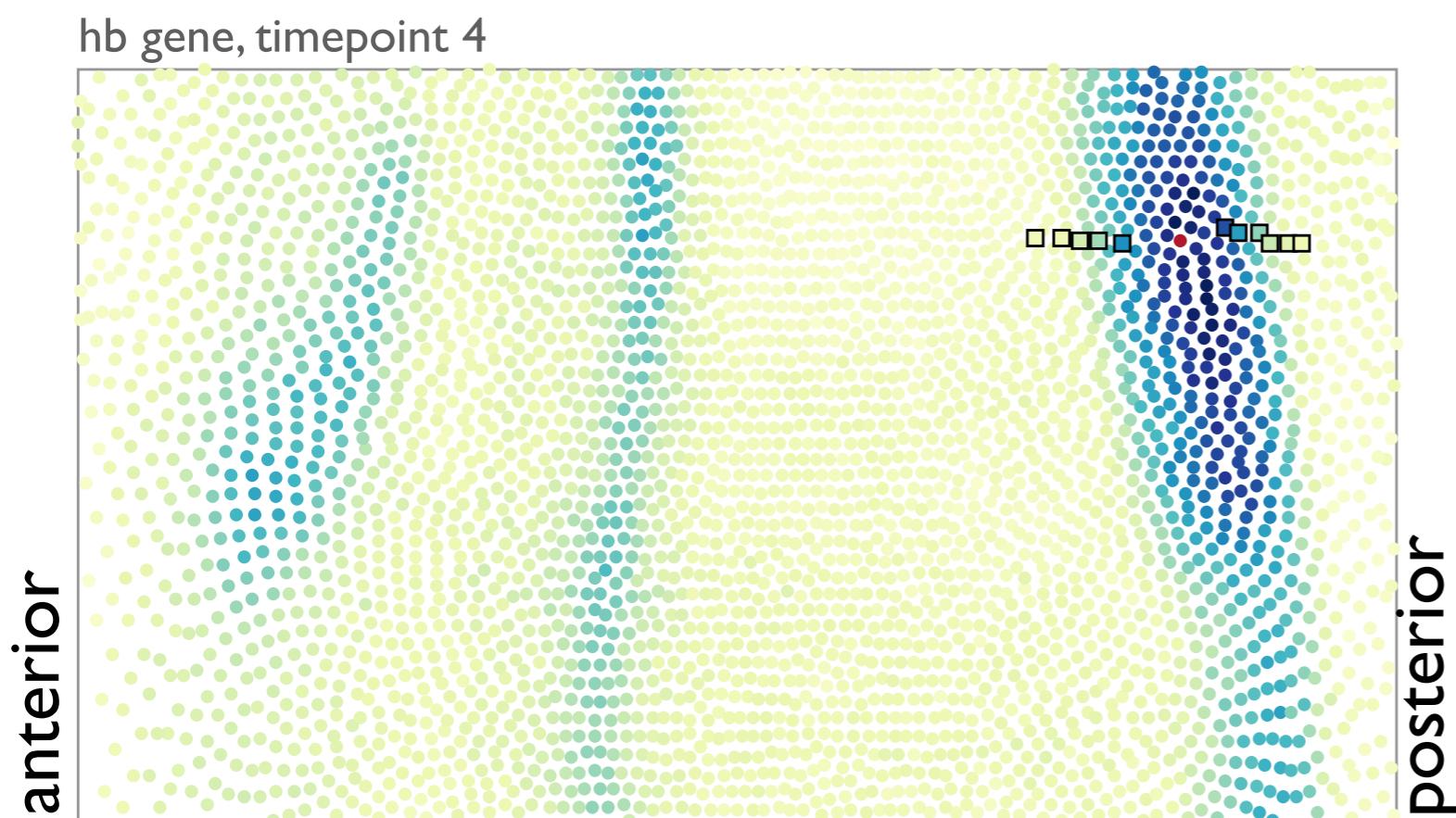
case study two: exploring groups



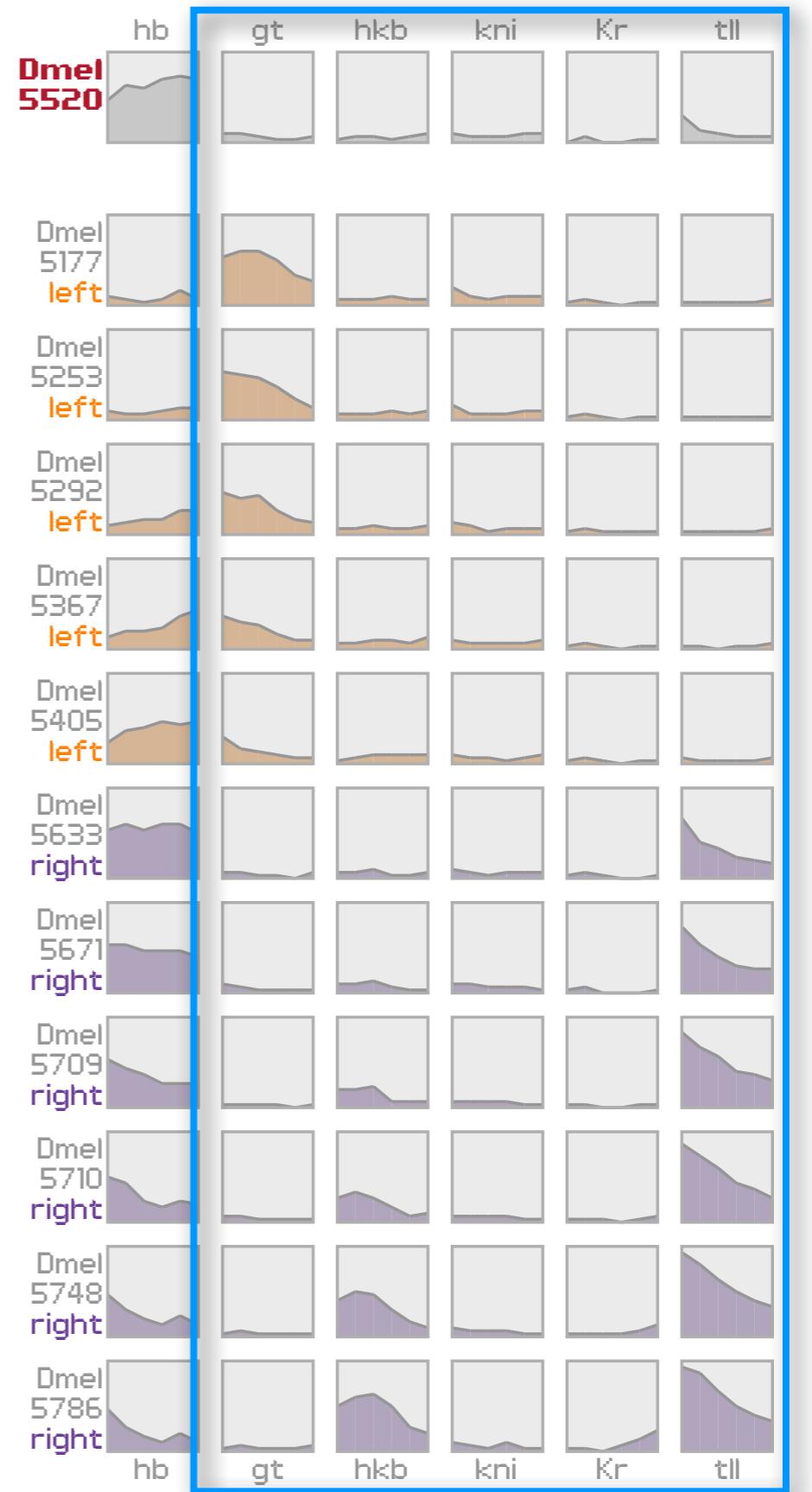
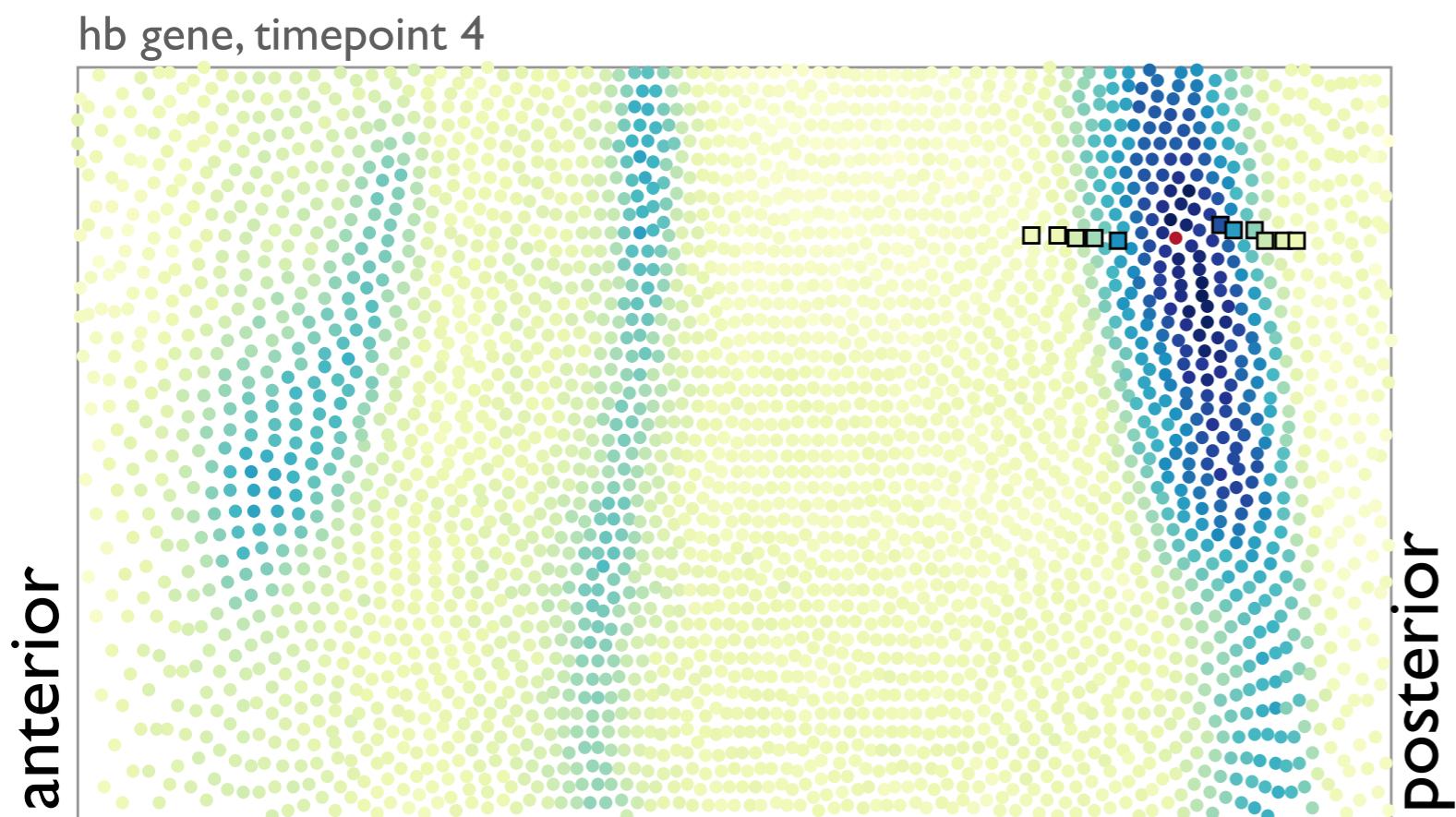
case study two: exploring groups



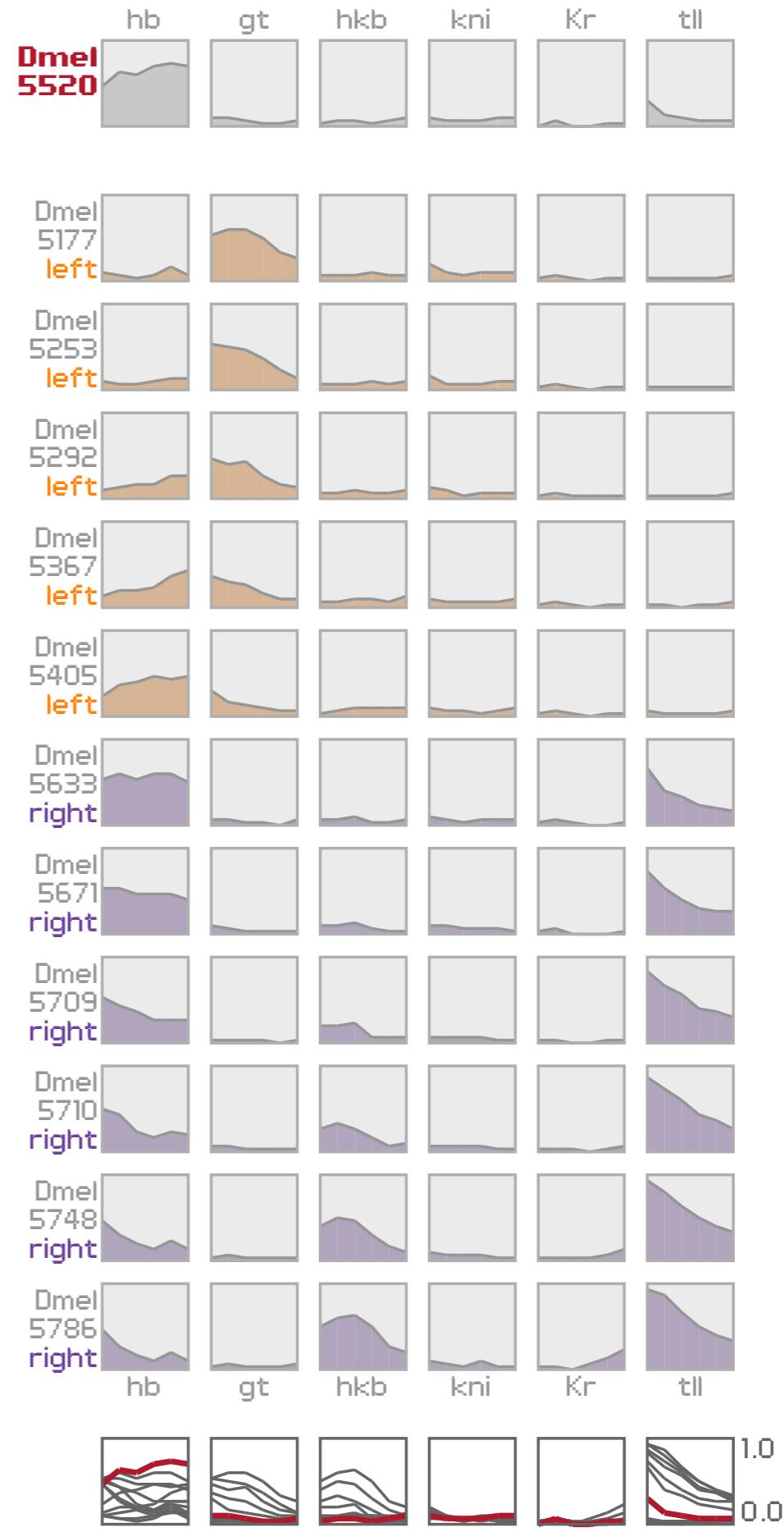
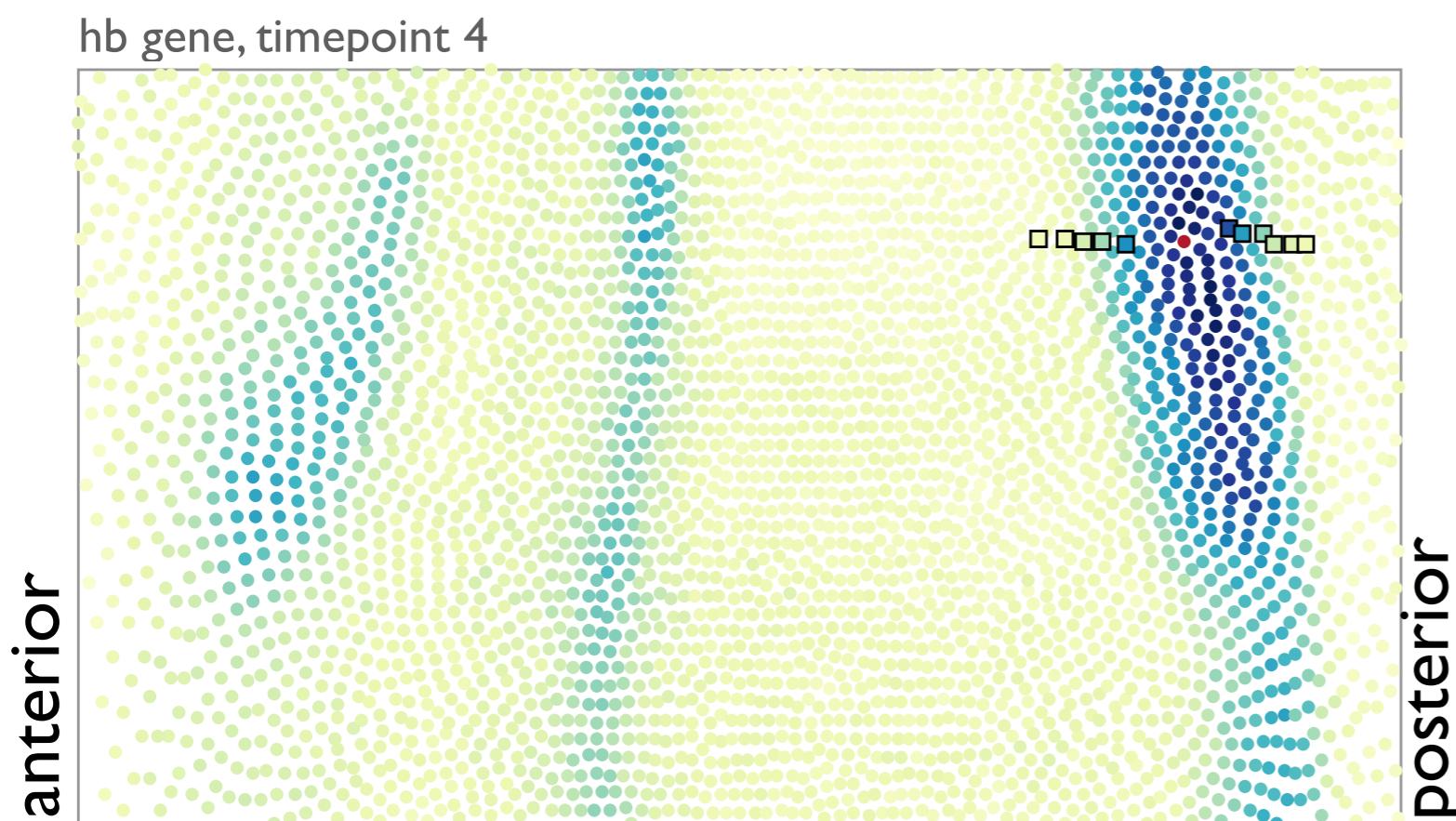
case study two: exploring groups



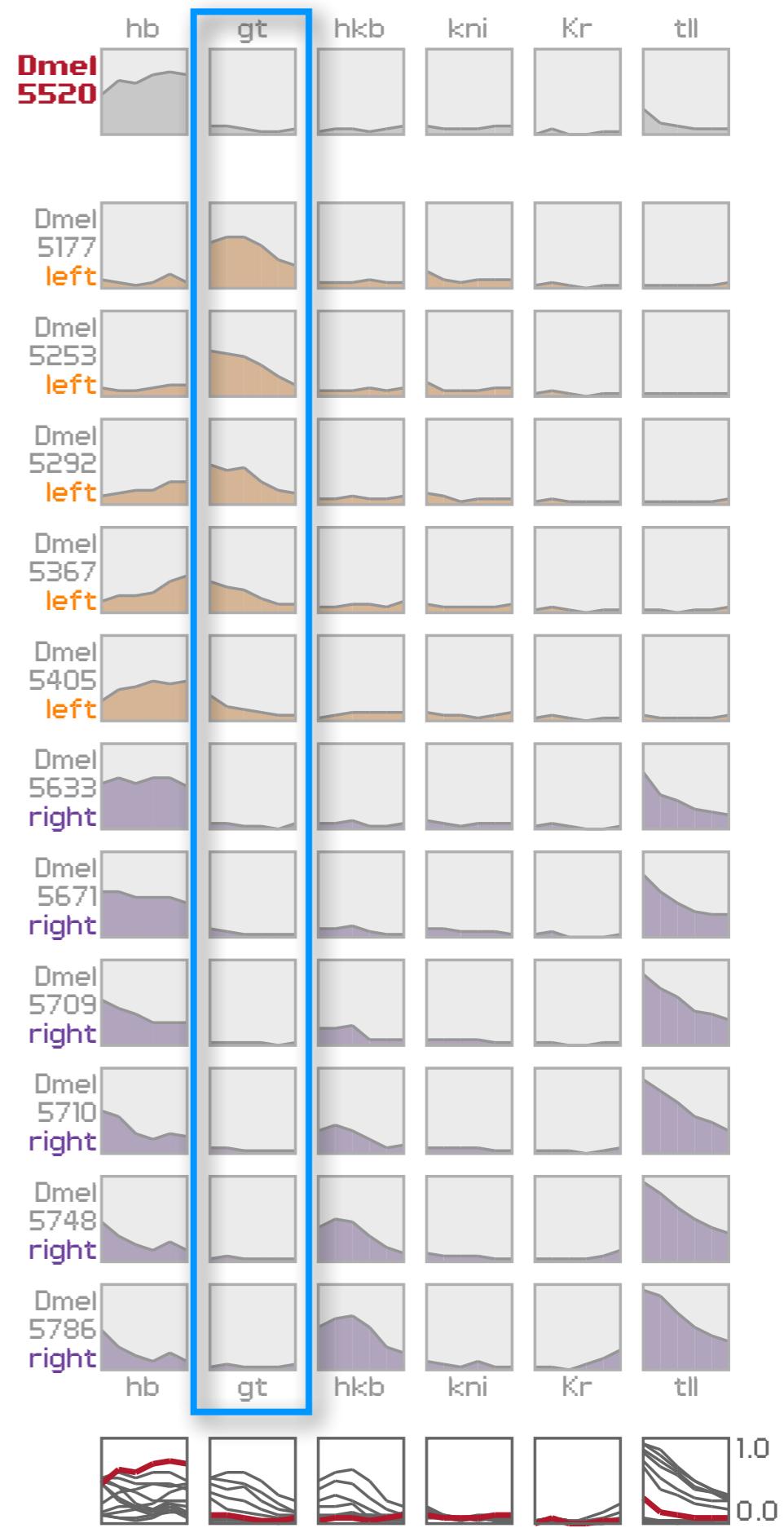
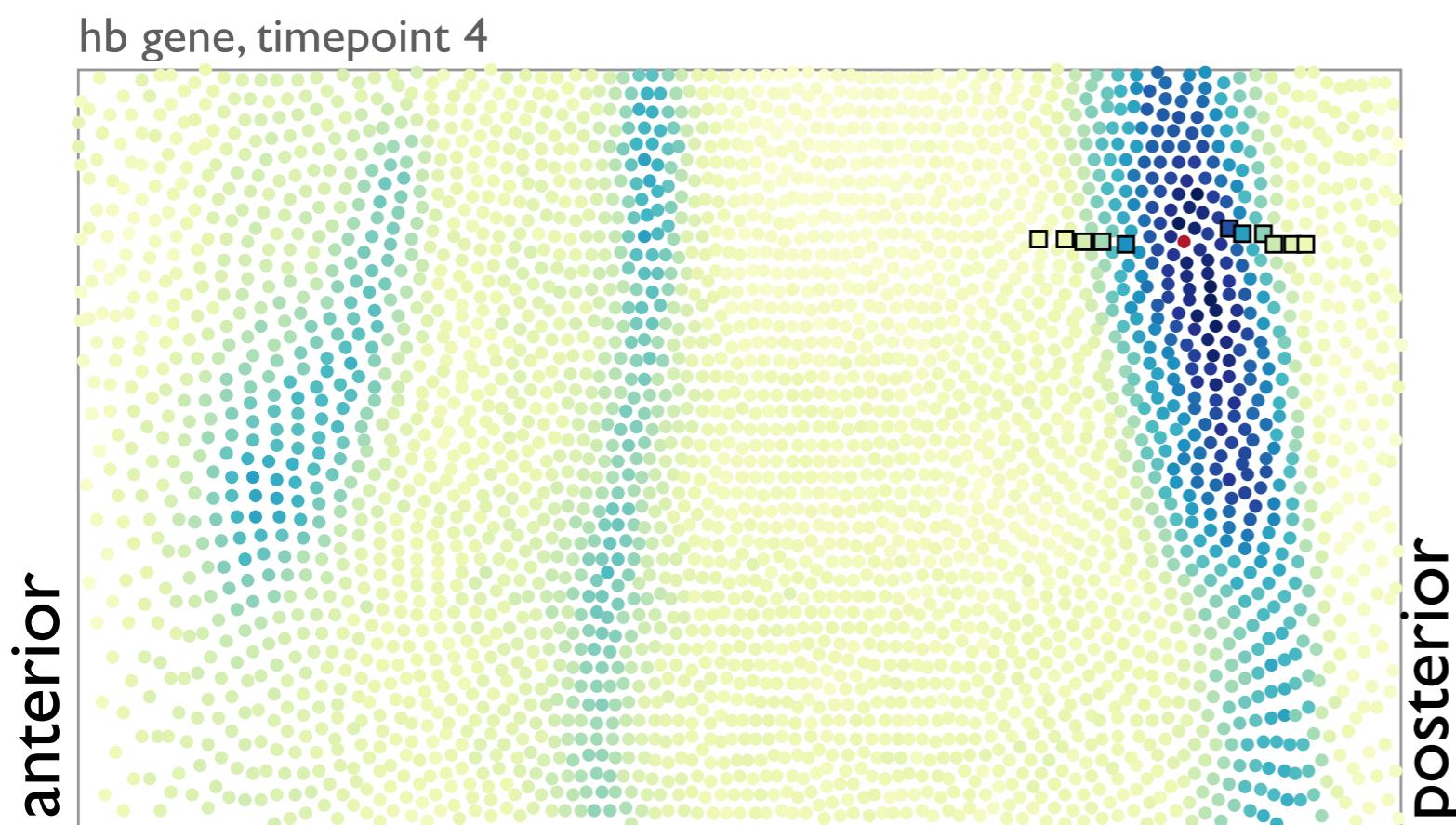
case study two: exploring groups



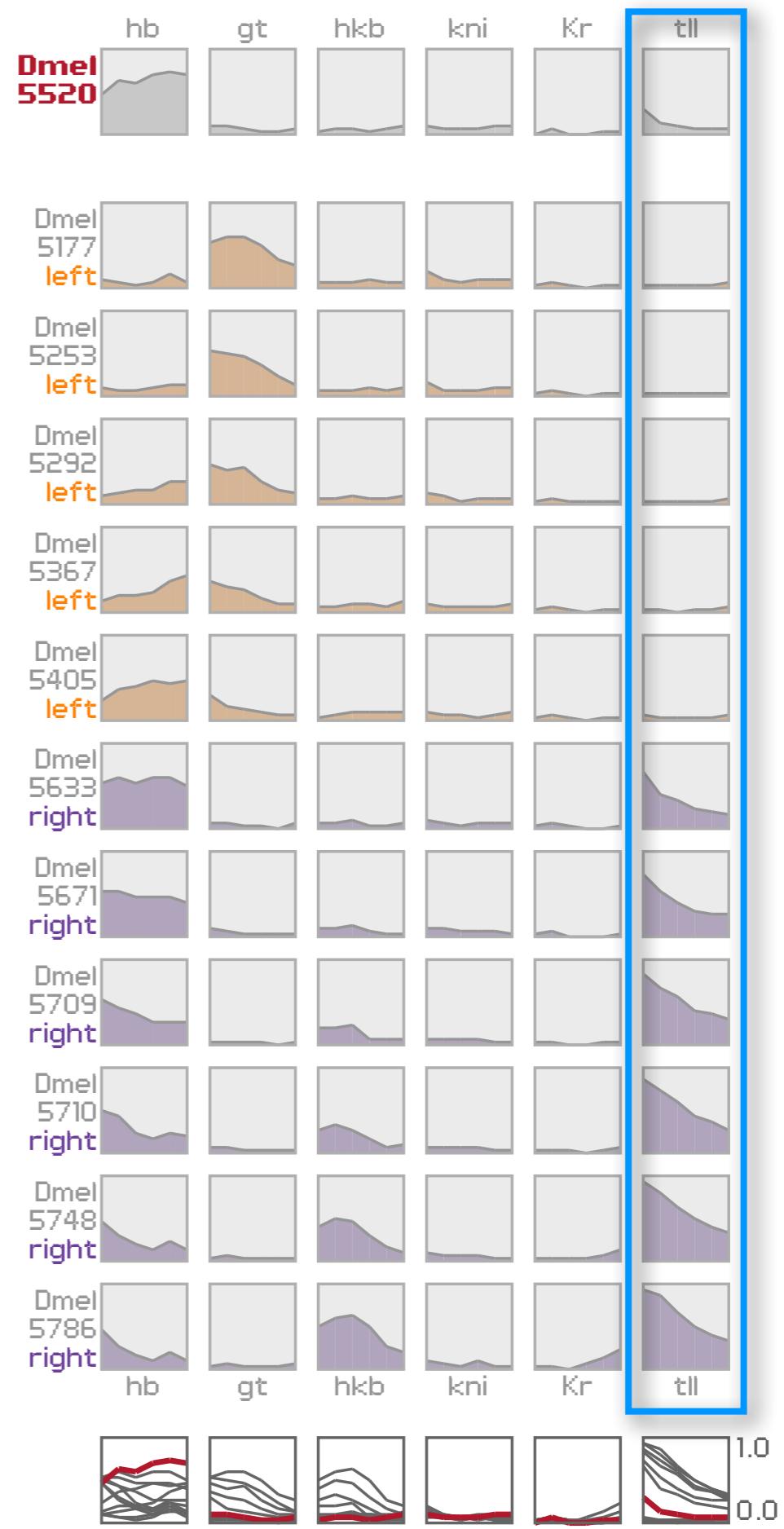
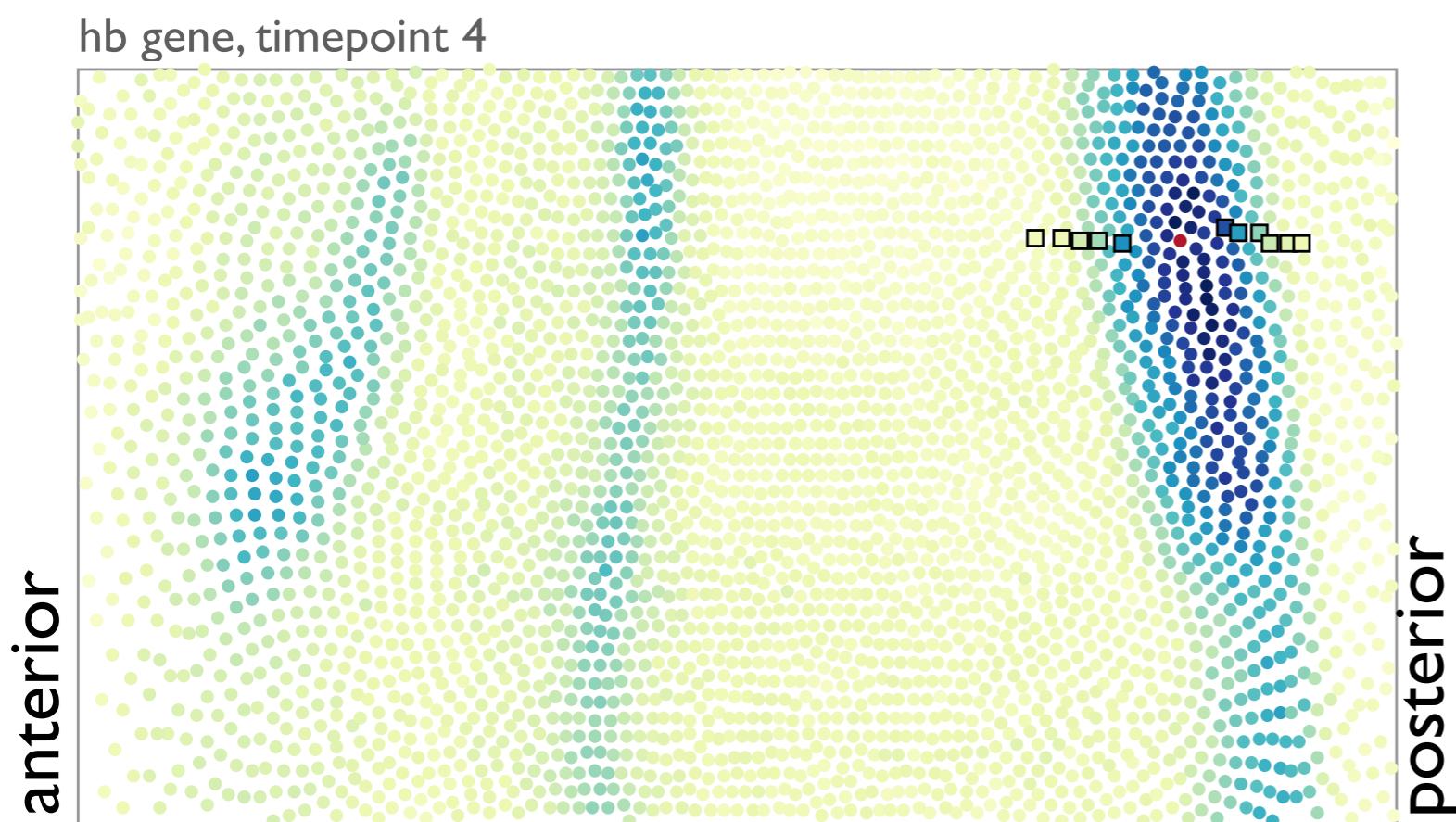
case study two: exploring groups



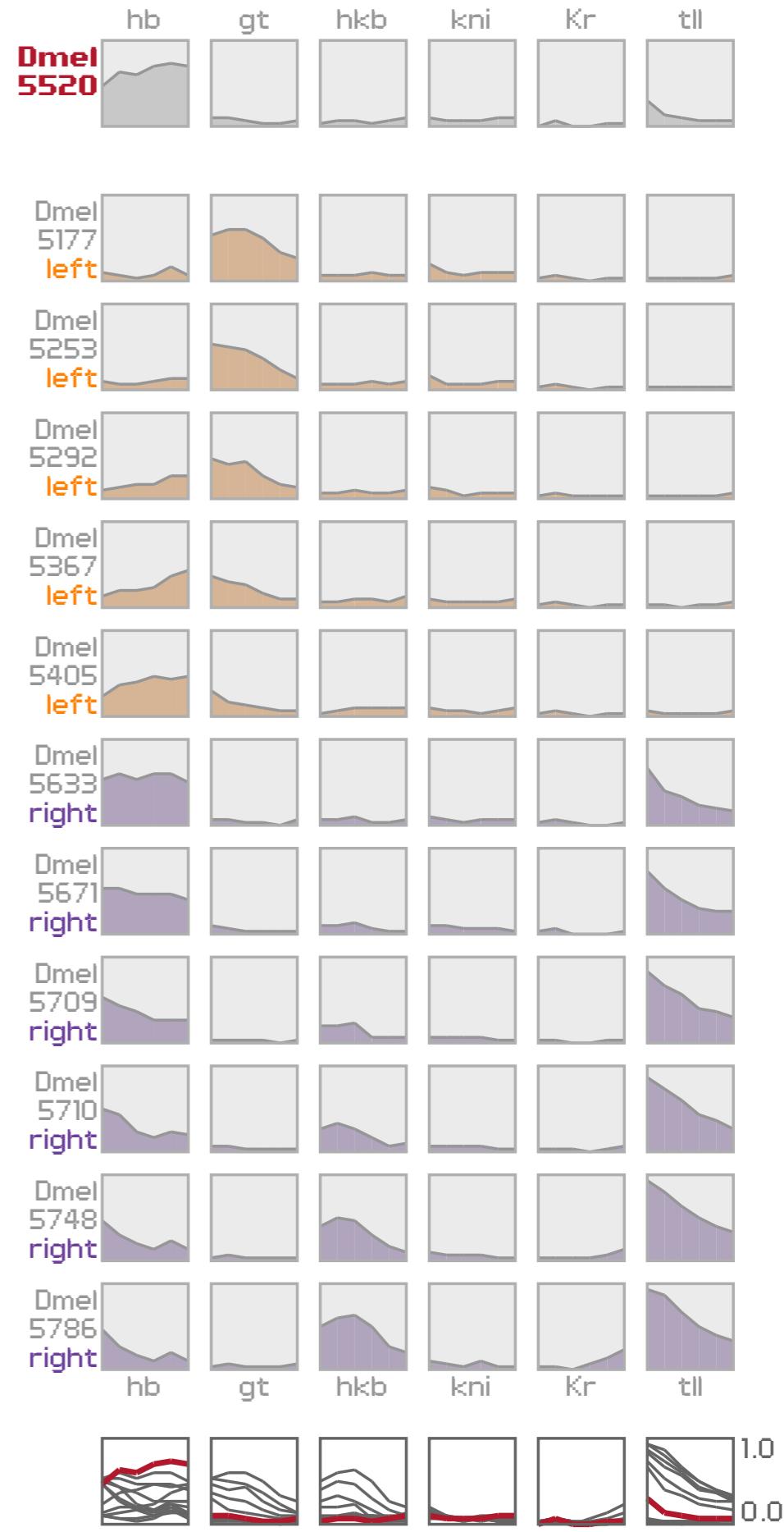
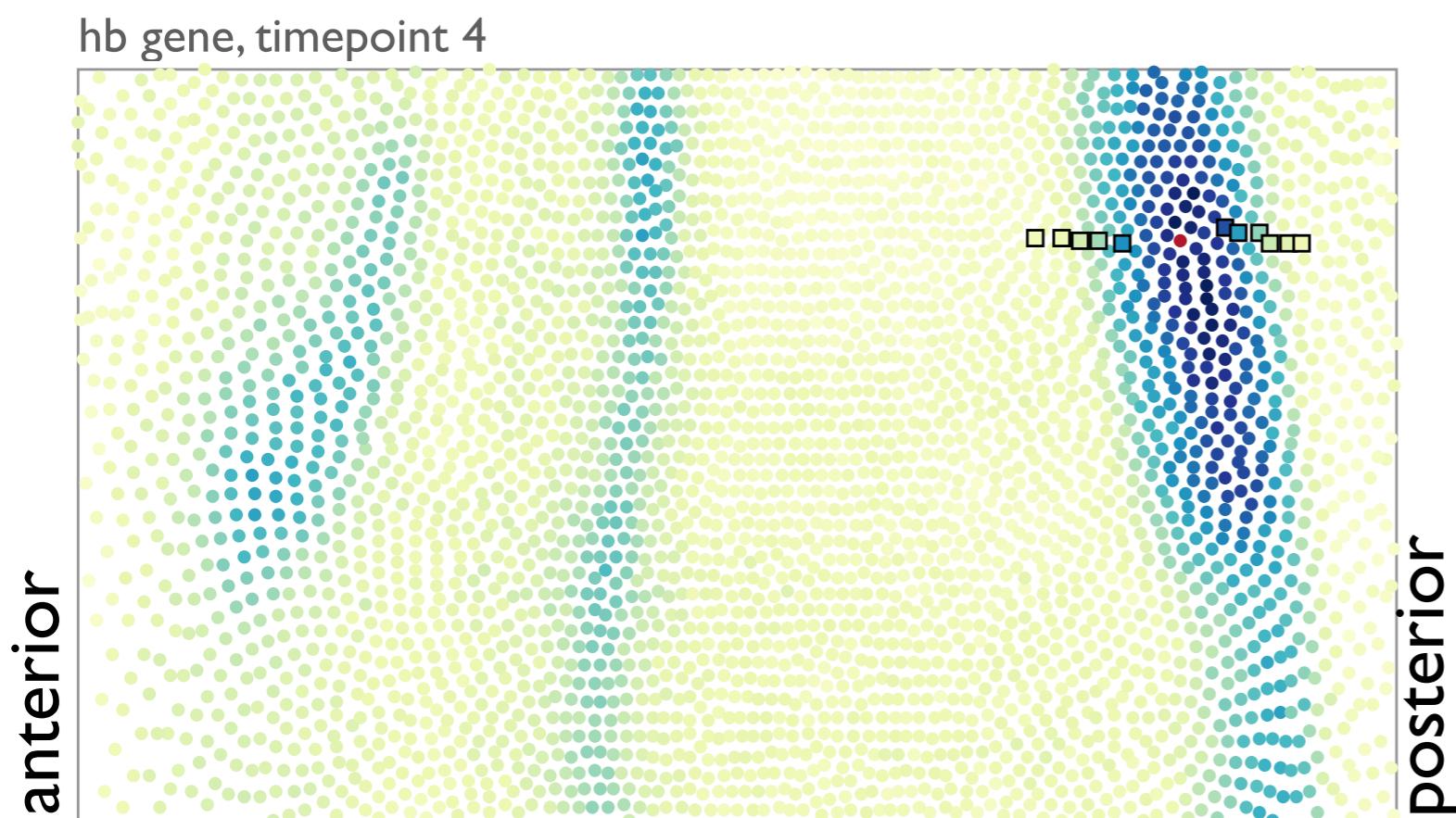
case study two: exploring groups



case study two: exploring groups



case study two: exploring groups



data & tool & tasks

summaries & groups

encodings & interaction

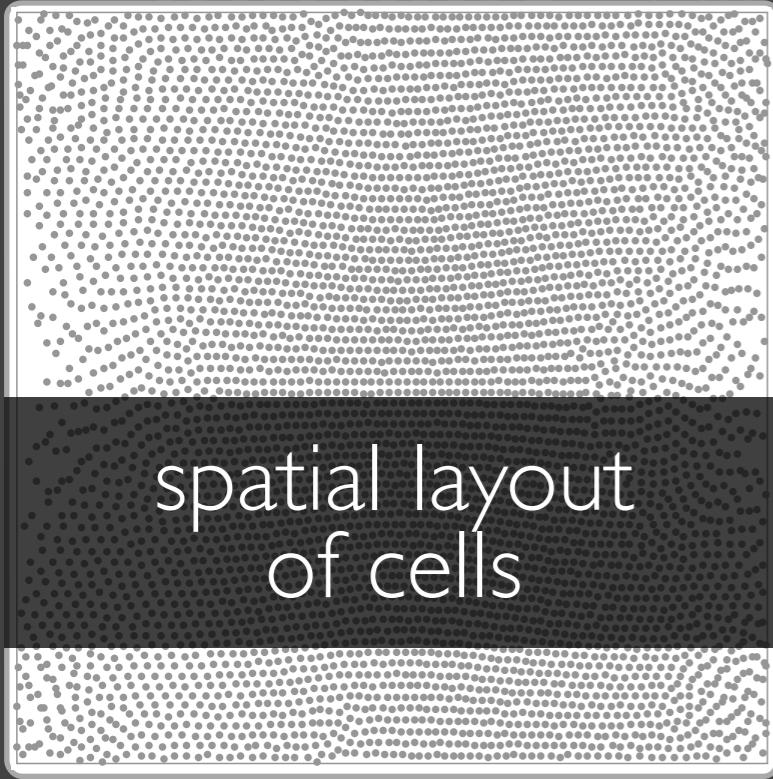
conclusions

initially: biologists relied on computational methods to explore data

goal: link underlying data with computational results

visualize: triad of data

initially, biologists relied on
anal methods to explore data

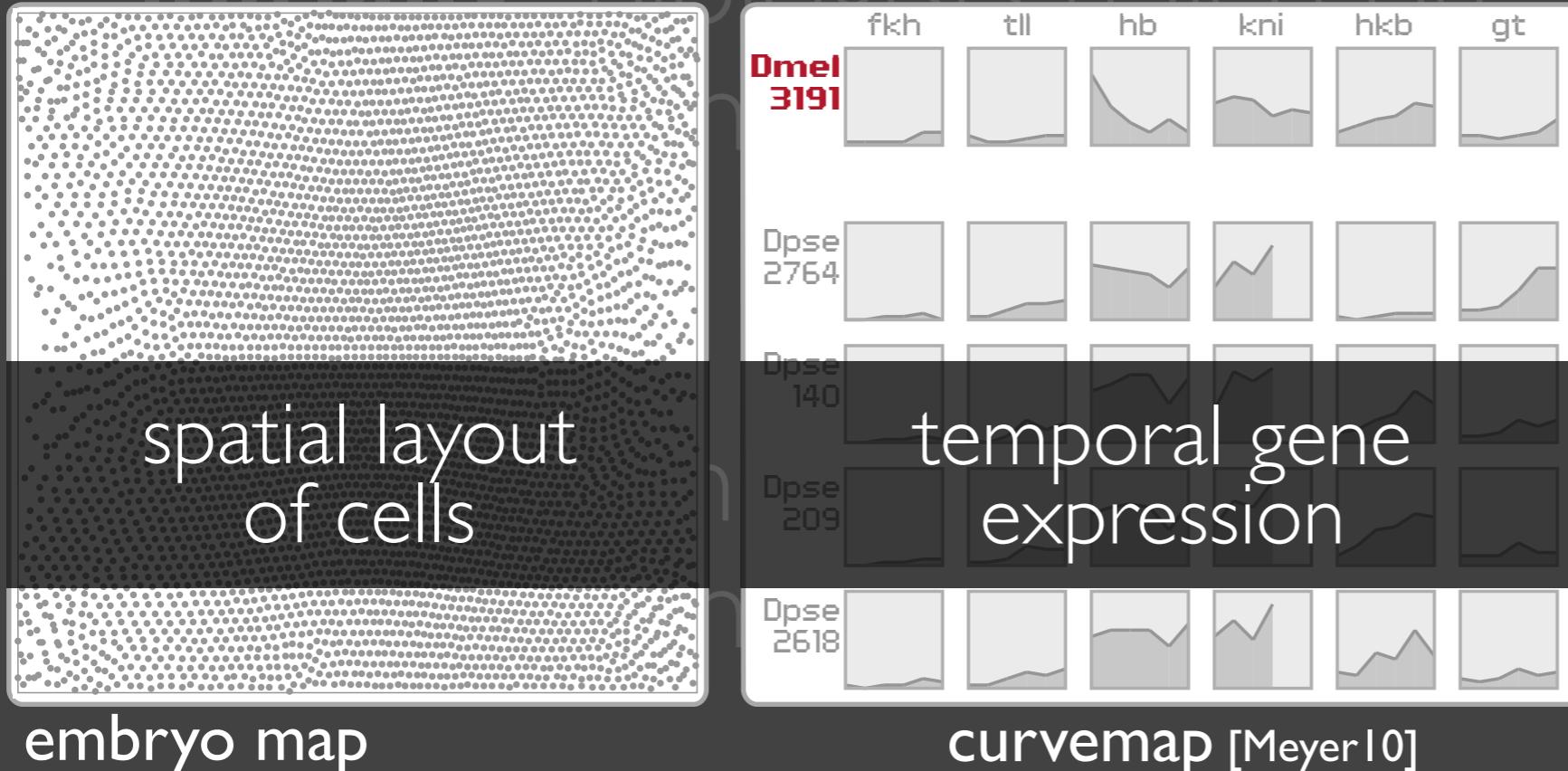


underlying data with
anal results

visualize: triad of data

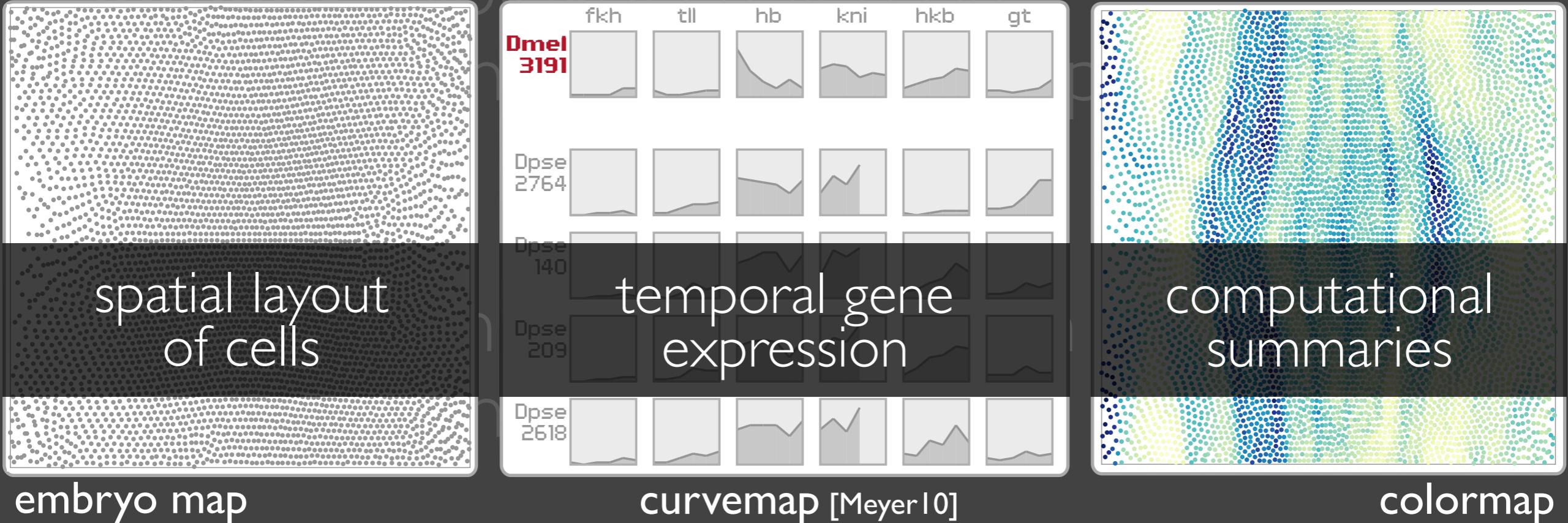
initially, biologists relied on

explore data



visualize: triad of data

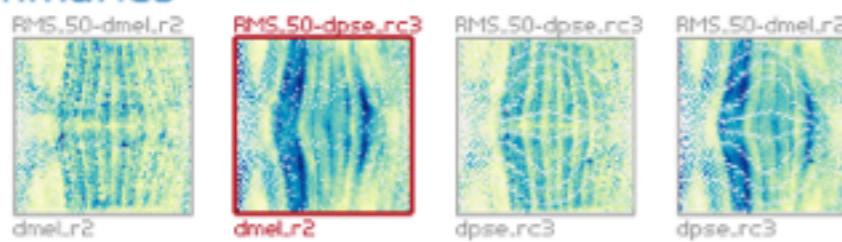
initially, biologists relied on



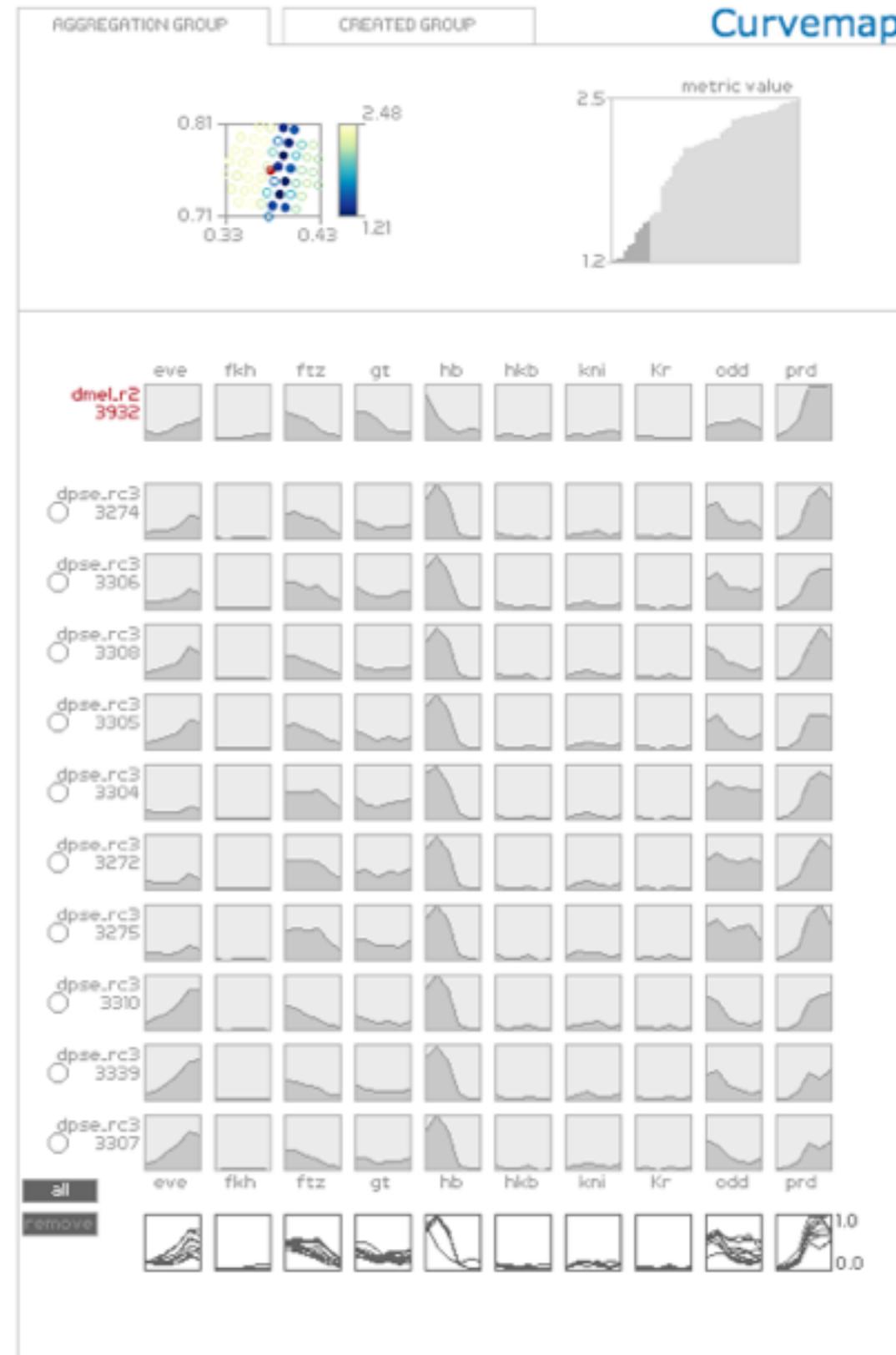
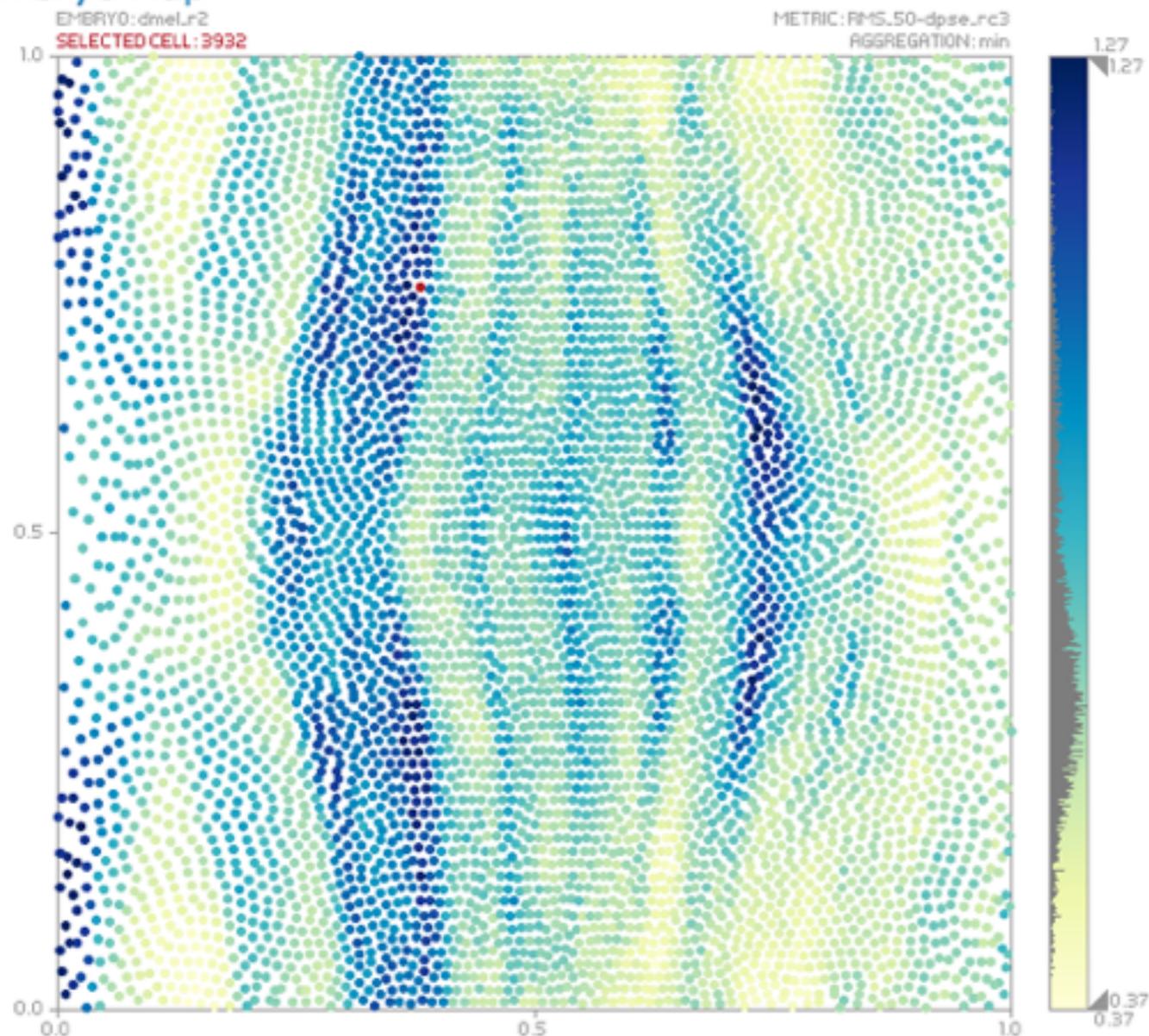
visualize: triad of data

MulteeSum

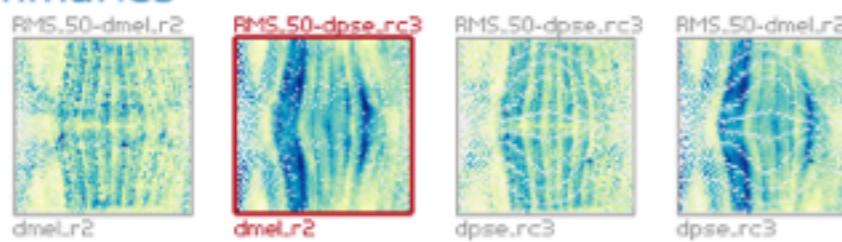
Summaries



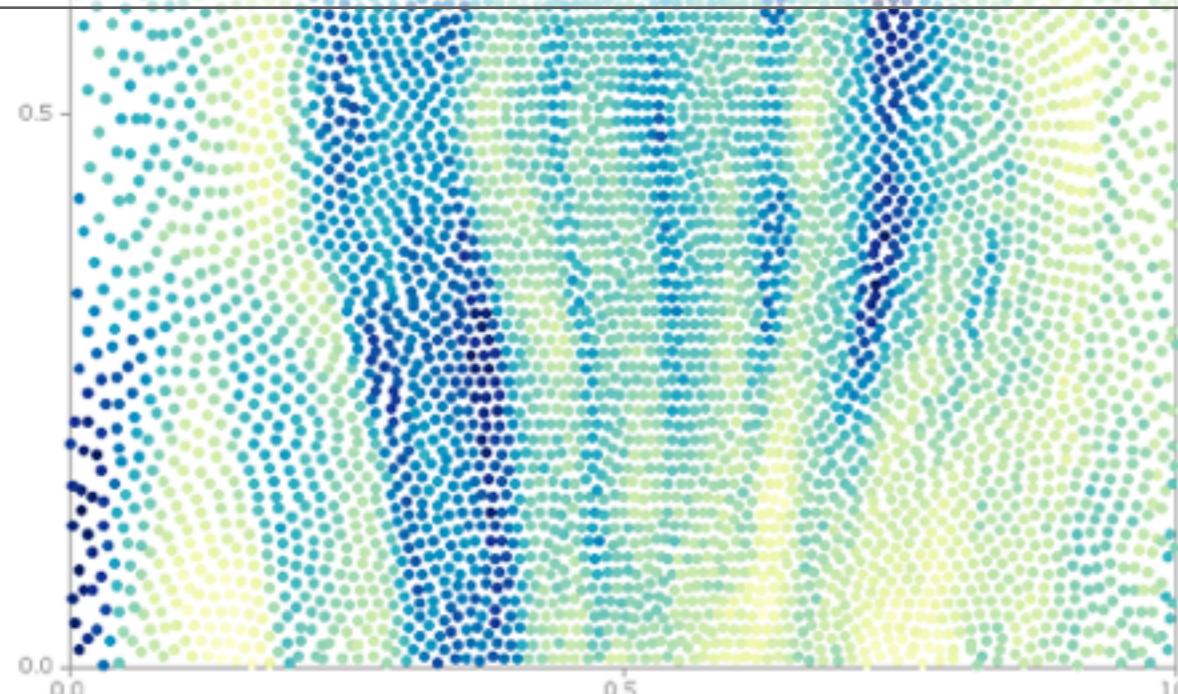
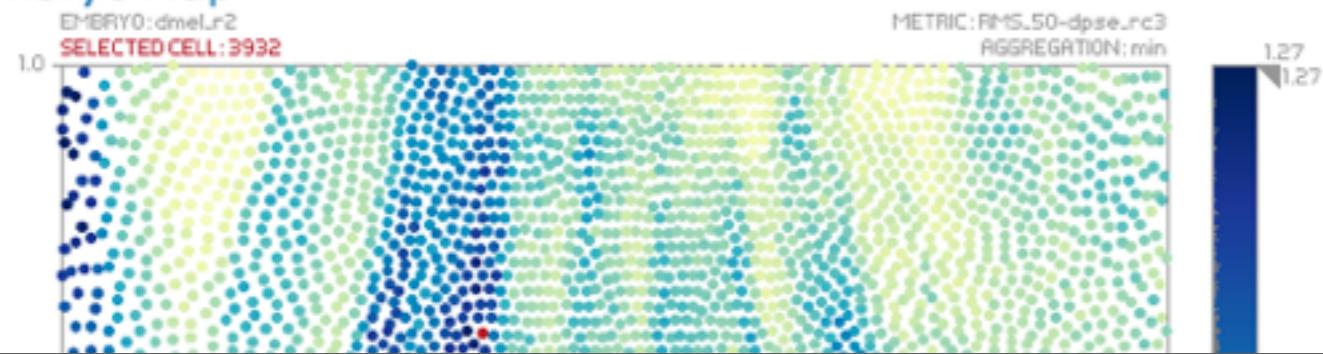
Embryo Map



Summaries

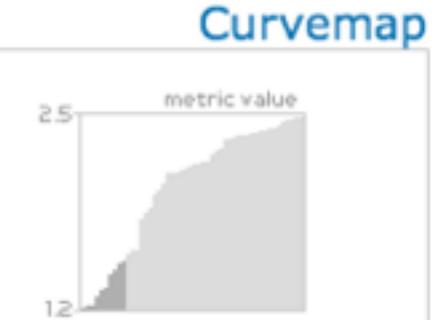
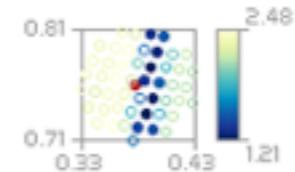


Embryo Map

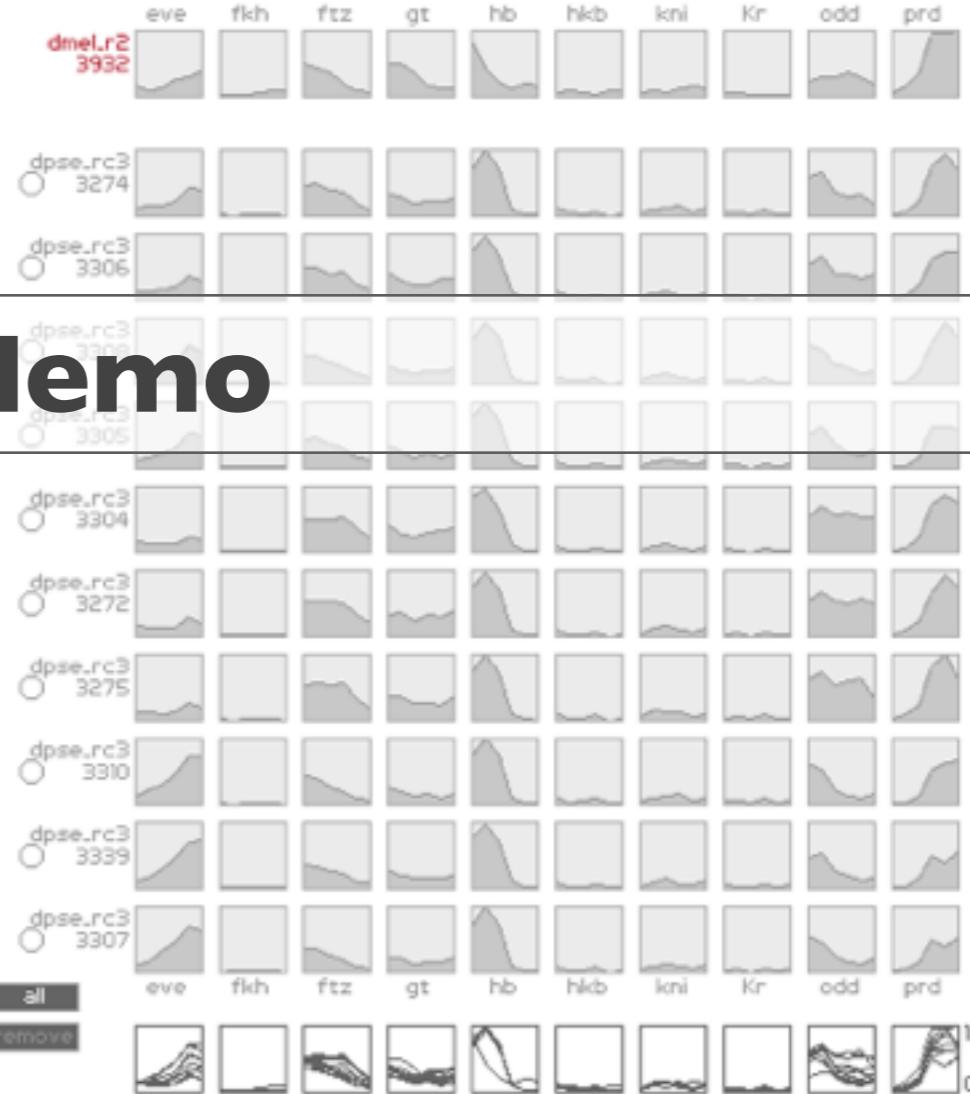


AGGREGATION GROUP

CREATED GROUP



Curvemap



interaction demo

data & tool & tasks

summaries & groups

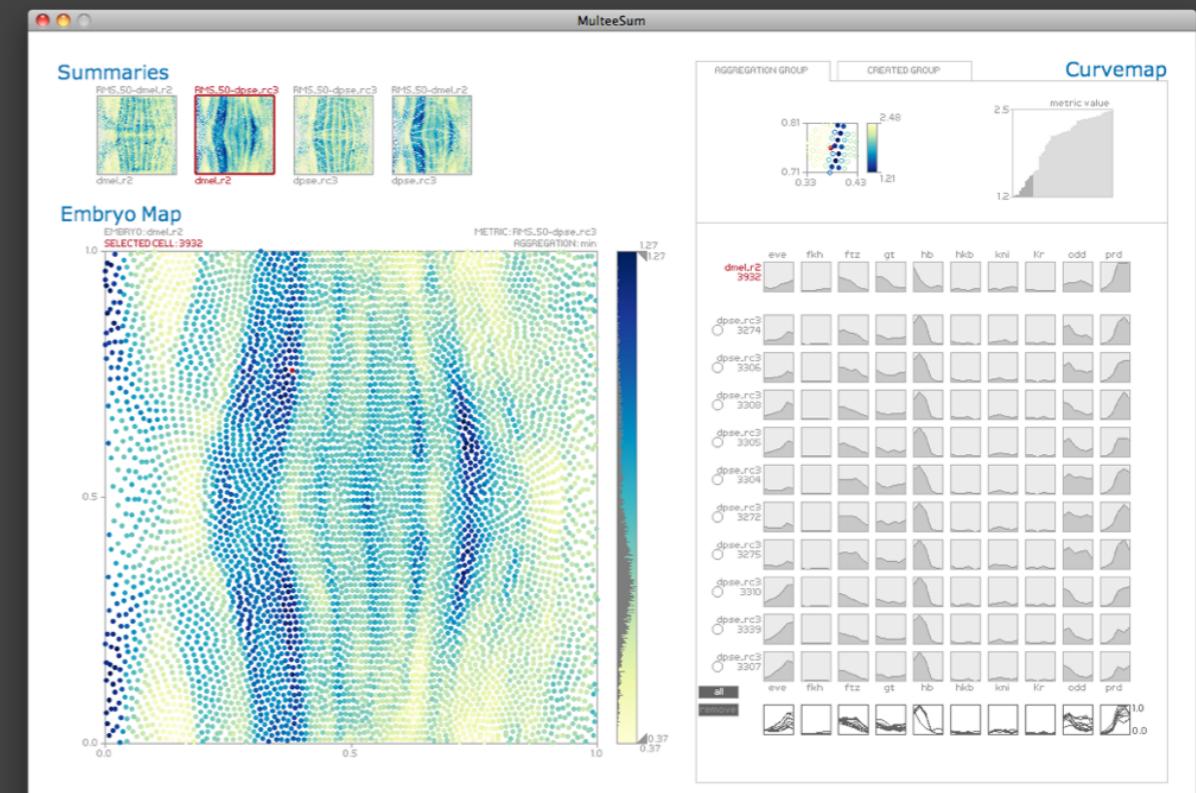
encodings & interaction

conclusions

contributions

MulteeSum

spatial and temporal gene expression data from multiple species



workflow

visualization supports upstream computation via summaries

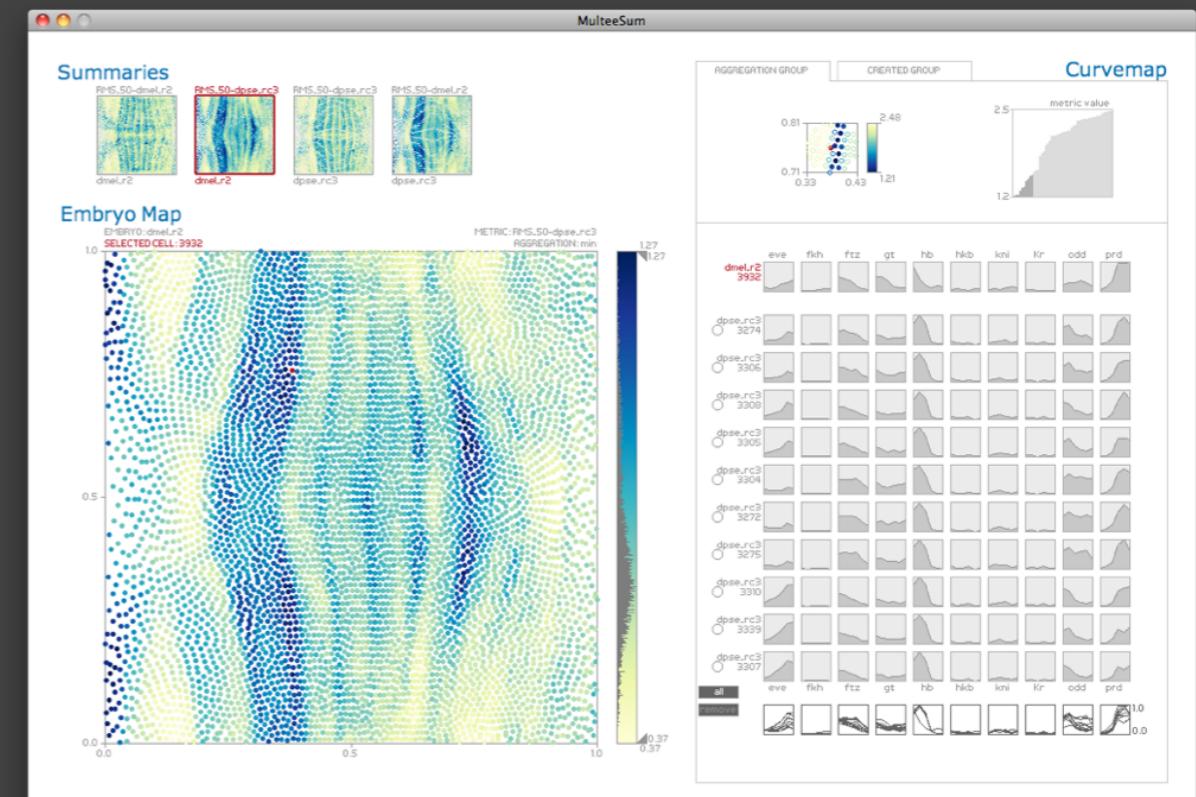
validation

case studies, deployment

contributions

MulteeSum

spatial and temporal gene expression data from multiple species



workflow

visualization supports upstream computation via summaries

validation

case studies, deployment

Cellular resolution comparison of gene expression in Drosophila reveals coordinated shifts in the segmentation network.

DePace et. al, in preparation.

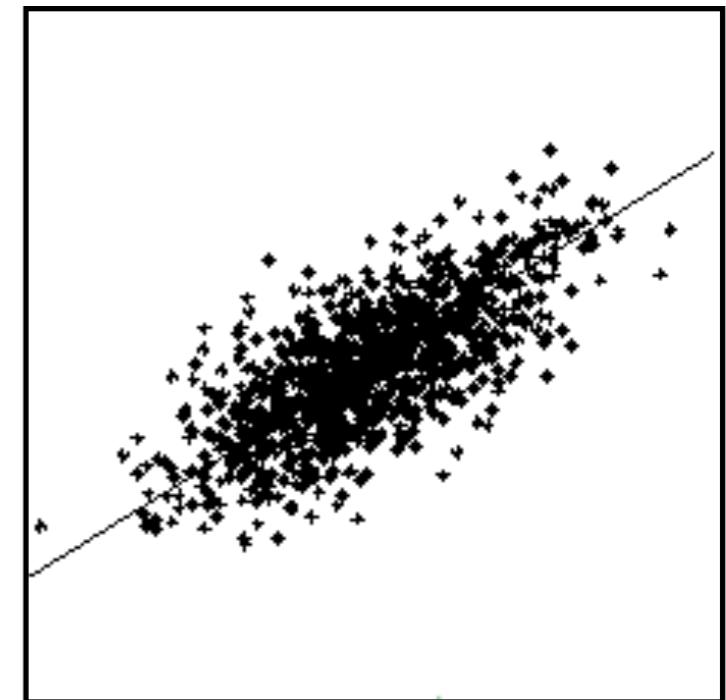
attribute aggregation

- 1) group attributes and compute a similarity score across the set
- 2) dimensionality reduction, to preserve meaningful structure**

dimensionality reduction

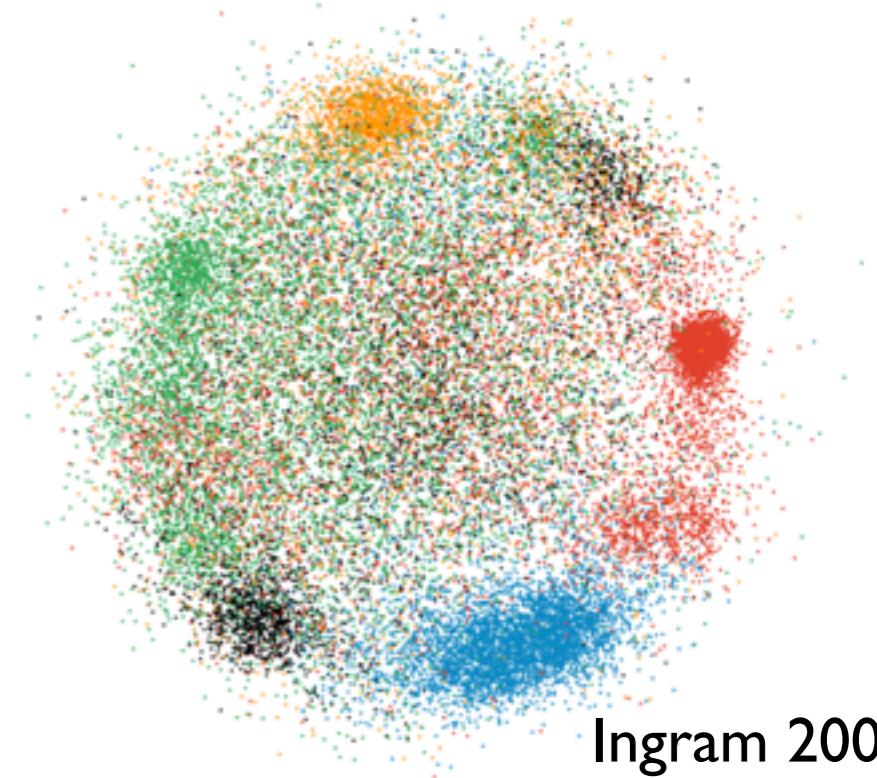
-PCA

- linear approach
- new dimensions are weighted combinations of original ones
- new dimensions created in order of maximum variance



-MDS

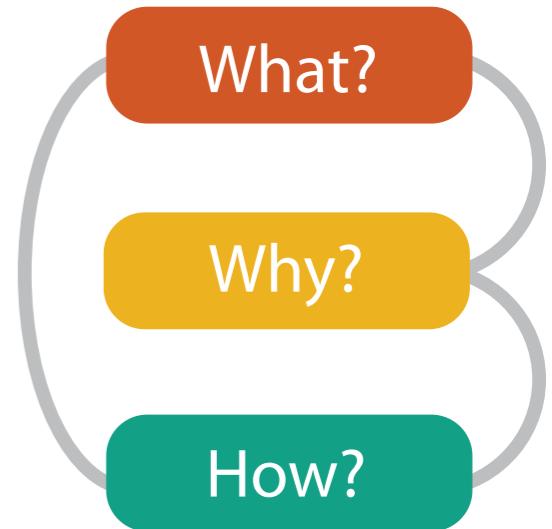
- nonlinear class of approaches
- maximize differences in distances from high dim space in the low dim space



end of foundations...

Analysis: What, why, and how

- **what** is shown?
 - data abstraction
- **why** is the user looking at it?
 - task abstraction
- **how** is it shown?
 - idiom: visual encoding and interaction
- abstract vocabulary avoids domain-specific terms
 - translation process iterative, tricky
- what-why-how analysis framework as scaffold to think systematically about design space



What?

Why?

How?

What?

Datasets

→ Data Types

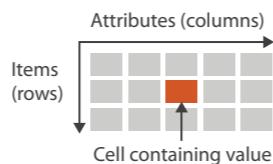
- Items → Attributes → Links → Positions → Grids

→ Data and Dataset Types

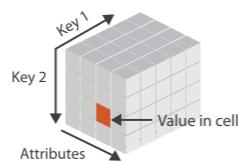
Tables	Networks & Trees	Fields	Geometry	Clusters, sets, lists
Items	Items (nodes)	Grids	Items	Clusters, sets, lists
Attributes	Links	Positions	Positions	Items

→ Dataset Types

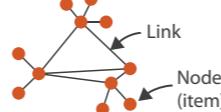
- Tables



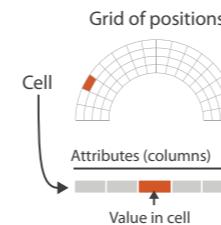
- Multidimensional Table



- Networks



- Fields (Continuous)



- Trees



- Geometry (Spatial)



→ Dataset Availability

- Static



- Dynamic



Attributes

→ Attribute Types

- Categorical



- Ordered

- Ordinal



- Quantitative



→ Ordering Direction

- Sequential



- Diverging



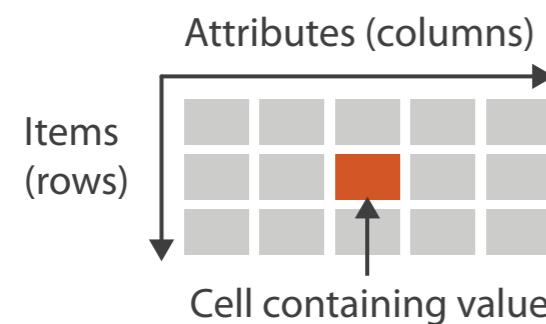
- Cyclic



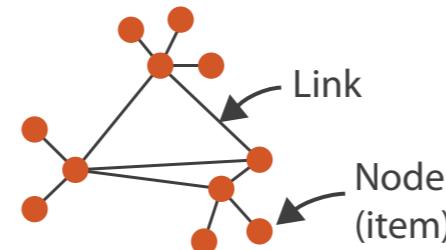
Dataset types

→ Dataset Types

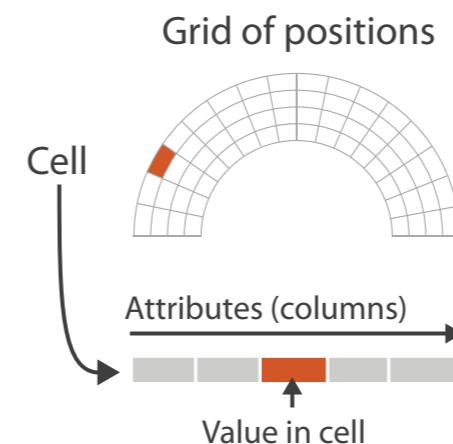
→ Tables



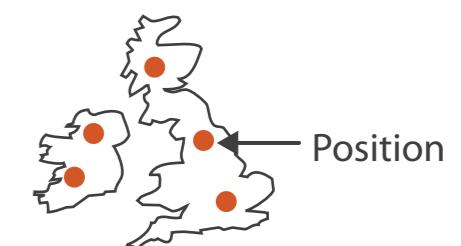
→ Networks



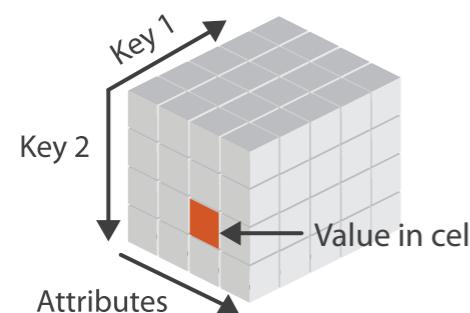
→ Fields (Continuous)



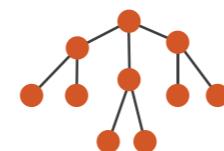
→ Geometry (Spatial)



→ Multidimensional Table



→ Trees



Dataset and data types

→ Data and Dataset Types

Tables	Networks & Trees	Fields	Geometry	Clusters, sets, lists
Items	Items (nodes)	Grids	Items	Items
Attributes	Links	Positions	Positions	

→ Data Types

→ Items → Attributes → Links → Positions → Grids

→ Dataset Availability

→ Static → Dynamic



Attribute types

→ Attribute Types

→ Categorical



→ Ordered

→ *Ordinal*



→ *Quantitative*



→ Ordering Direction

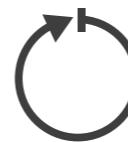
→ Sequential

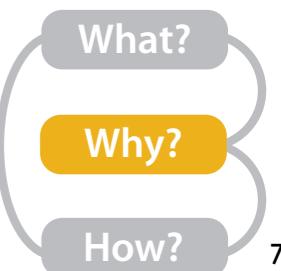
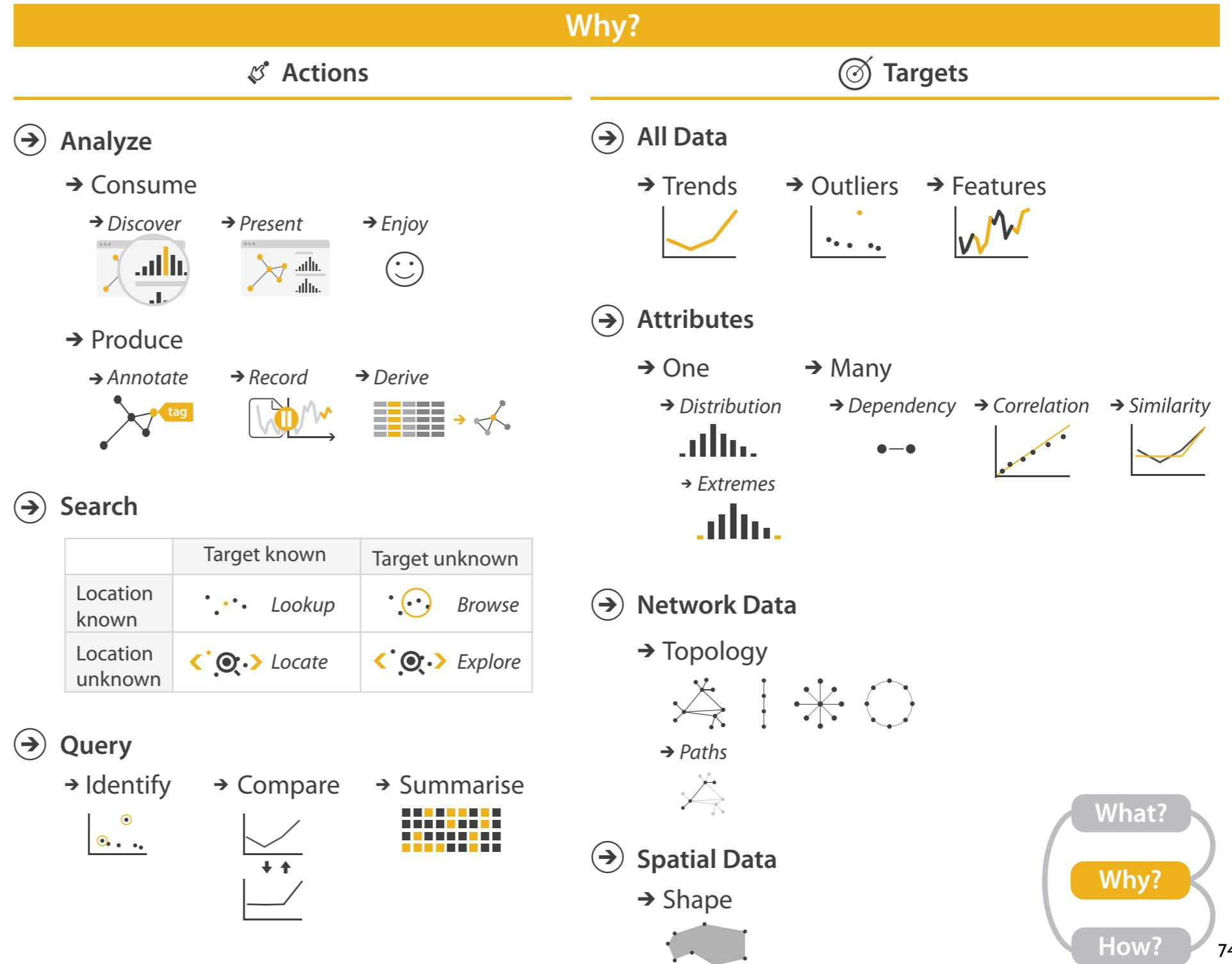
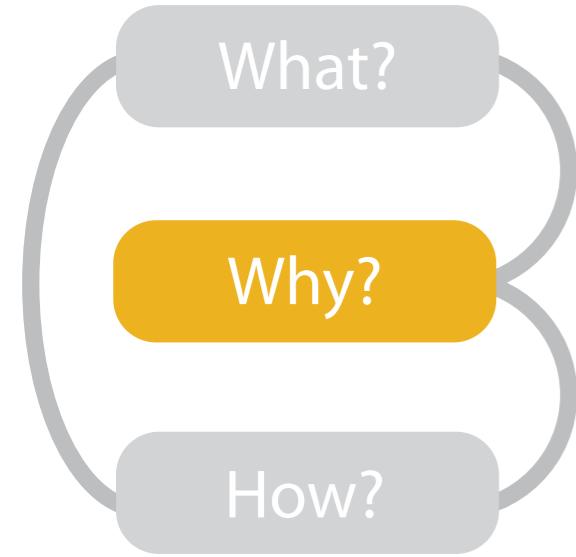


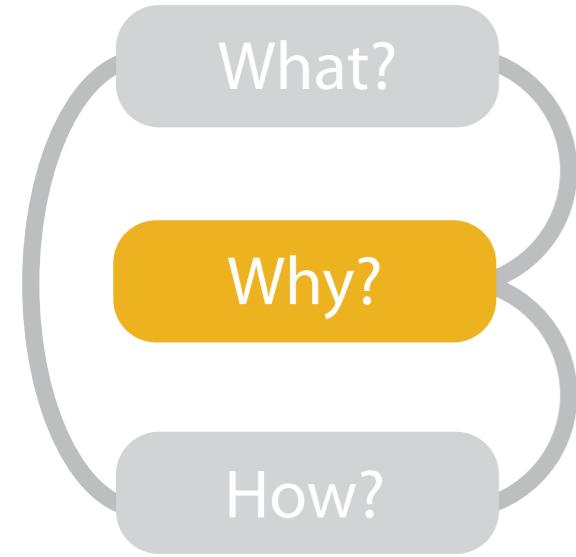
→ Diverging



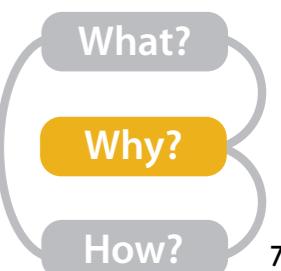
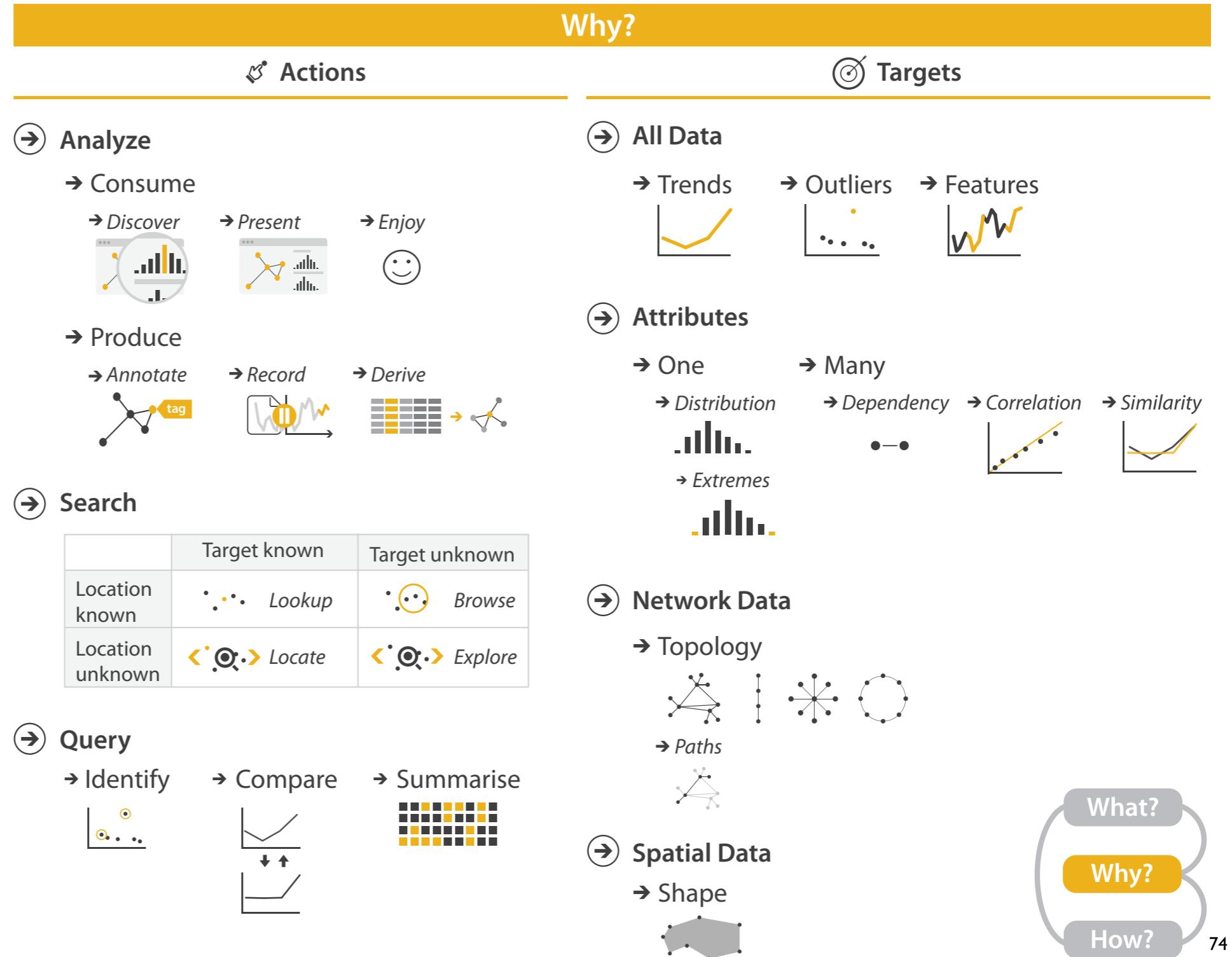
→ Cyclic





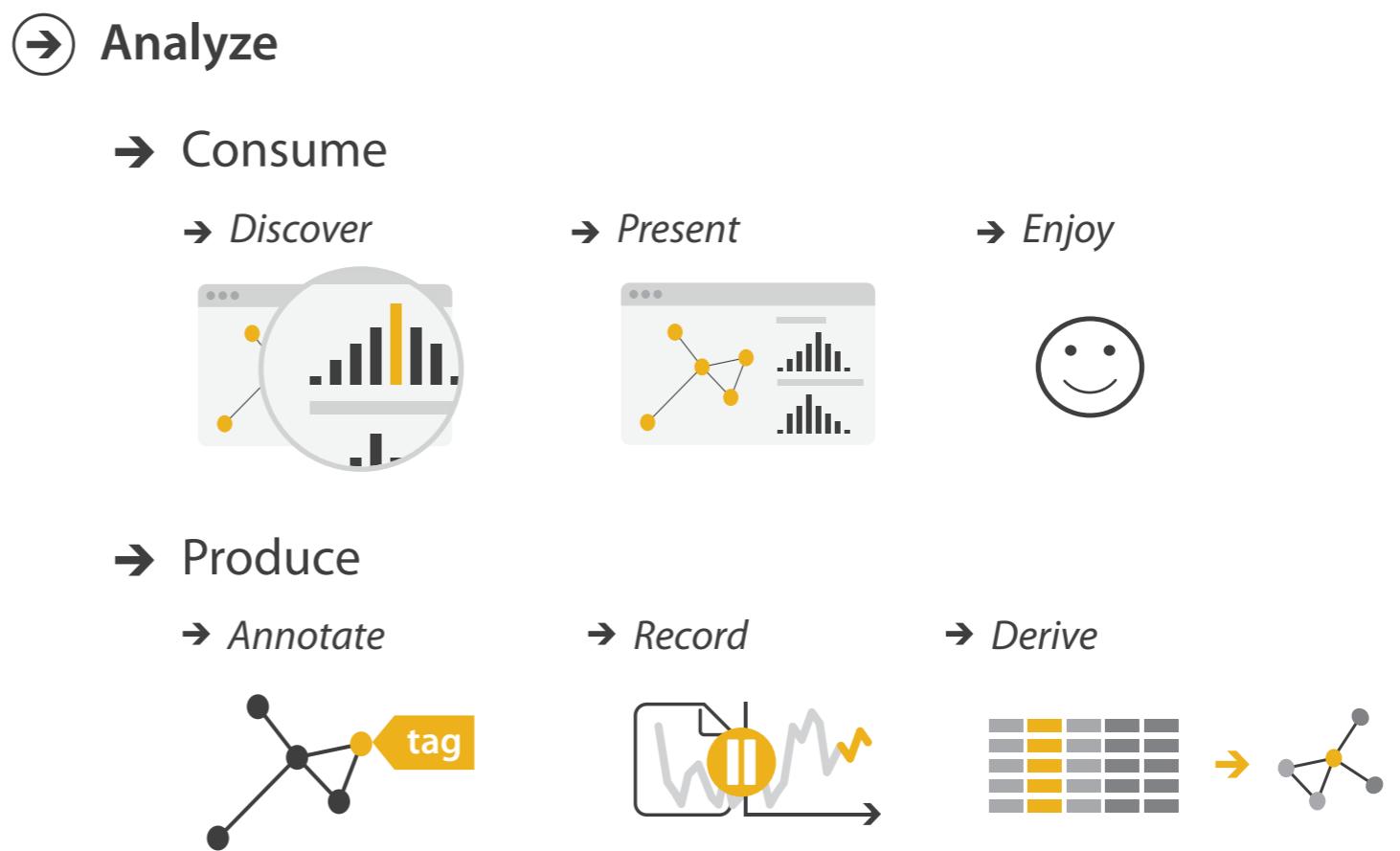


- {action, target} pairs
 - *discover distribution*
 - *compare trends*
 - *locate outliers*
 - *browse topology*



High-level actions: Analyze

- consume
 - discover vs present
 - classic split
 - aka explore vs explain
 - enjoy
 - newcomer
 - aka casual, social
- produce
 - annotate, record
 - derive
 - crucial design choice



Actions: Mid-level search, low-level query

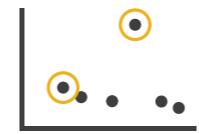
- what does user know?
 - target, location
- how much of the data matters?
 - one, some, all

➔ Search

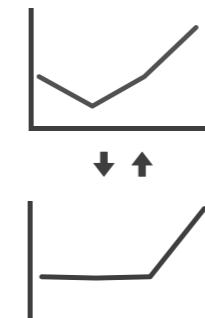
	Target known	Target unknown
Location known	 <i>Lookup</i>	 <i>Browse</i>
Location unknown	 <i>Locate</i>	 <i>Explore</i>

➔ Query

→ Identify



→ Compare



→ Summarise



Why: Targets

→ ALL DATA

→ Trends



→ Outliers

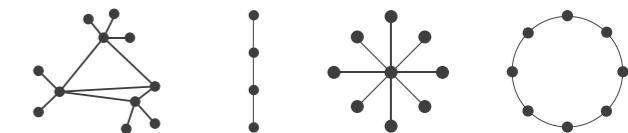


→ Features



→ NETWORK DATA

→ Topology



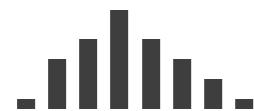
→ Paths



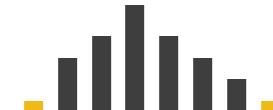
→ ATTRIBUTES

→ One

→ Distribution



↓ Extremes



→ Many

→ Dependency



→ Correlation

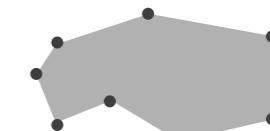


→ Similarity

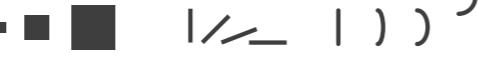
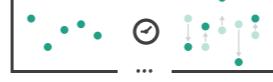


→ SPATIAL DATA

→ Shape



How?

Encode	Manipulate	Facet	Reduce
<p>→ Arrange</p> <p>→ Express</p>  <p>→ Separate</p>  <p>→ Order</p>  <p>→ Align</p>  <p>→ Use</p> 	<p>→ Map from categorical and ordered attributes</p> <p>→ Color</p>    <p>→ Size, Angle, Curvature, ...</p>  <p>→ Shape</p>  <p>→ Motion <i>Direction, Rate, Frequency, ...</i></p> 	<p>→ Change</p>  <p>→ Juxtapose</p>  <p>→ Select</p>  <p>→ Partition</p>   <p>→ Navigate</p>  <p>→ Superimpose</p>  	<p>→ Filter</p>  <p>→ Aggregate</p>  <p>→ Embed</p> 

What?

Why?

How?

+ perception and design guidelines...

L12: Tables

REQUIRED READING

Chapter 7

Arrange Tables

7.1 The Big Picture

Figure 7.1 shows the four visual encoding design choices for how to arrange tabular data spatially. One is to express values. The other three are to separate, order, and align regions. The spatial orientation of axes can be rectilinear, parallel, or radial. Spatial layouts may be dense, and they may be space-filling.

► A fifth arrangement choice, to use a given spatial layout, is not an option for nonspatial information; it is covered in Chapter 8.

7.2 Why Arrange?

The **arrange** design choice covers all aspects of the use of spatial channels for visual encoding. It is the most crucial visual encoding choice because the use of space dominates the user's mental model of the dataset. The three highest ranked effectiveness channels for quantitative and ordered attributes are all related to spatial position: planar position along an aligned scale, planar position against a common scale, and length. The highest ranked effectiveness channel for categorical attributes, grouping items within the same region, is also about the use of space. Moreover, there are no nonspatial channels that are highly effective for all attribute types: the others are split into being suitable for either ordered or categorical attributes, but not both, because of the principle of expressiveness.

► The primacy of the spatial position channels is discussed at length in Chapter 5, as are the principles of effectiveness and expressiveness.

7.3 Arrange by Keys and Values