



Universidad de Oviedo

MorphingProjections: Interactive Visualization of Electric Power Demand Time Series

Ignacio Díaz¹, Manuel Domínguez², Abel A. Cuadrado¹
Alberto B. Diez¹, Juan J. Fuertes²

¹ Universidad de Oviedo

² Universidad de León

Eurovis 2012, Wien, Austria



Analysis of power demand time series

Requirements

- **Temporal behavior**

- ▶ Regular periodicities (daily, weekly, yearly...)
- ▶ Social time granularities (months, feasts, exams, holidays,...)

- **Prototypes**

- ▶ Find & classify typical demand profiles
→ exploit prior knowledge → refine → interaction

- **Global view**

- ▶ Find connections in different domains, e.g.:
daily, weekly, monthly & day-pattern



Basic idea: multiple views

weekdays

**develop
several views
showing different aspects**

year

day
hours

months
(calendar)

day
patterns



Basic idea: multiple views

weekdays

¿how to connect
all these views?

year

¿how to get
a global understanding?

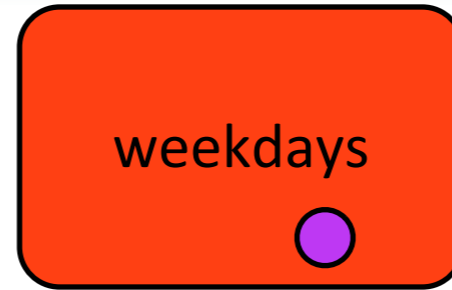
day
hours

months
(calendar)

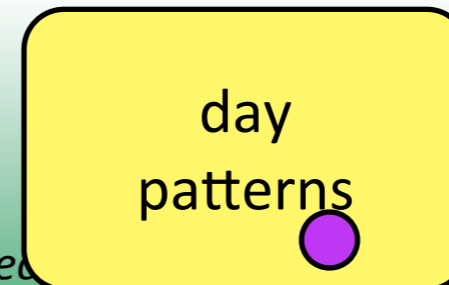
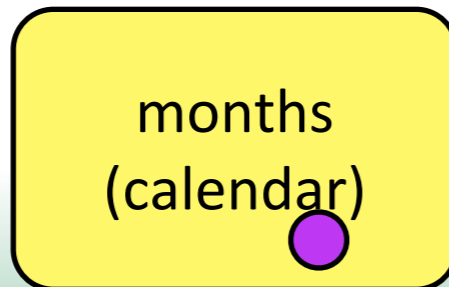
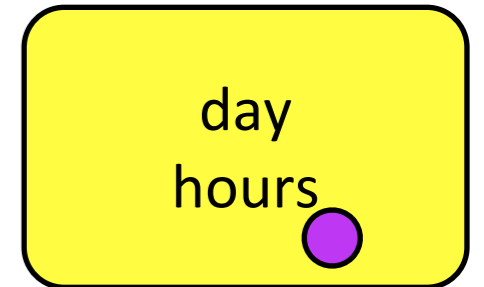
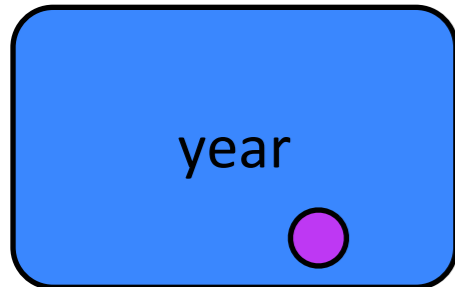
day
patterns



Basic idea: multiple views

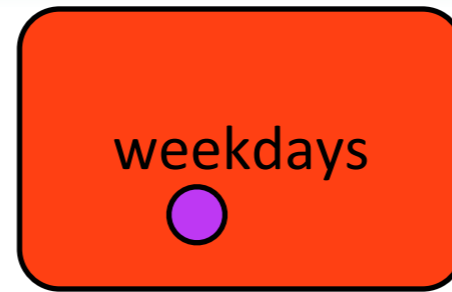


Coordinated views

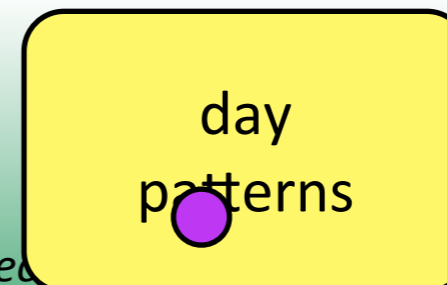
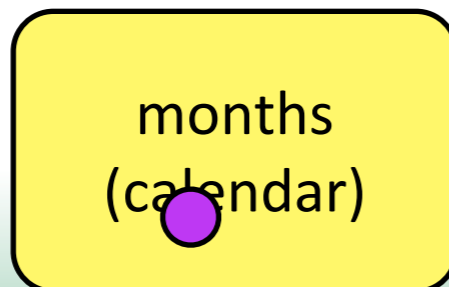
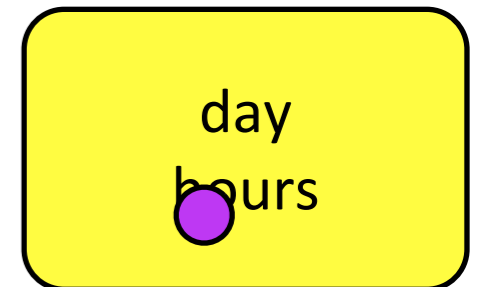
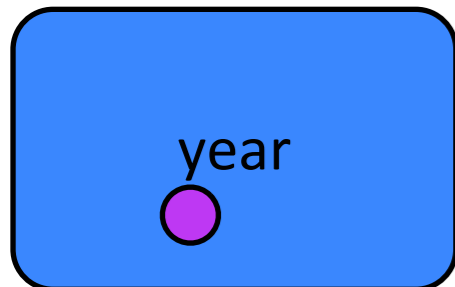




Basic idea: multiple views



Coordinated views





Basic idea: multiple views

Animated transitions

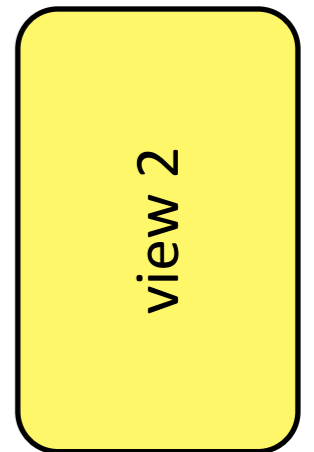
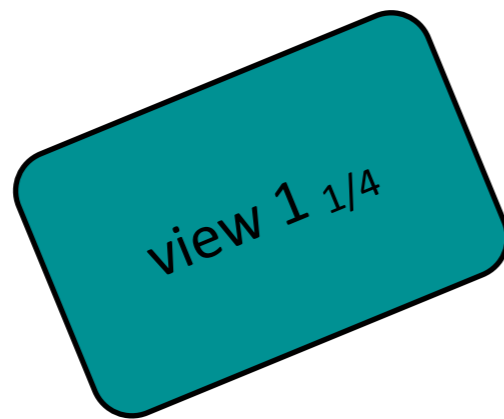
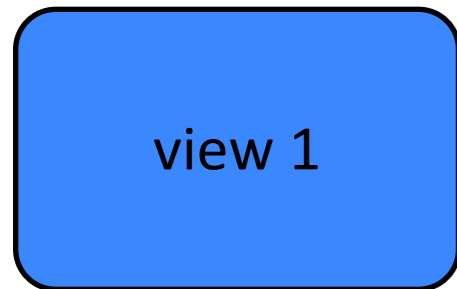
view 1

view 2



Basic idea: multiple views

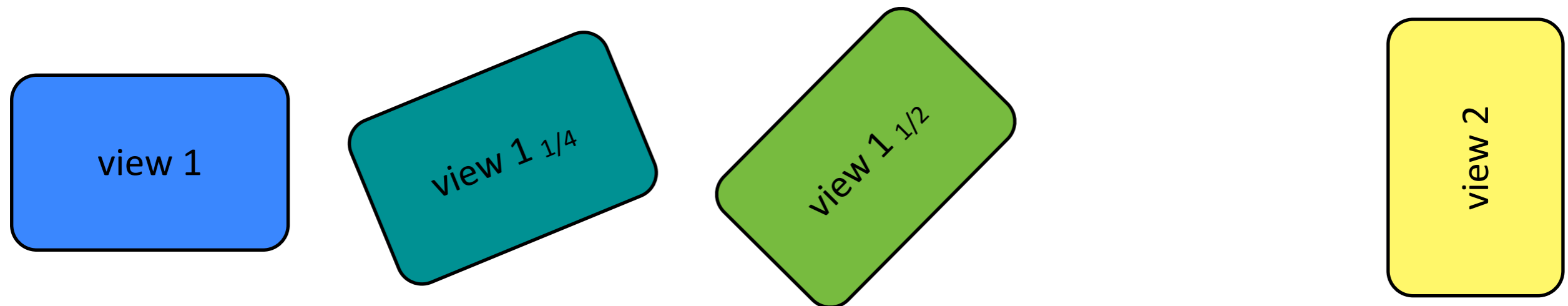
Animated transitions





Basic idea: multiple views

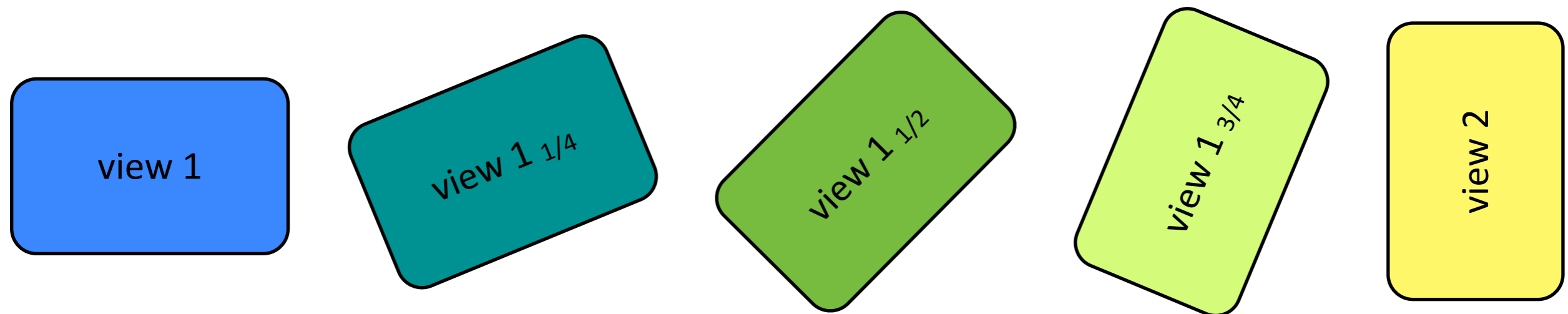
Animated transitions





Basic idea: multiple views

Animated transitions



- ✓ helps to keep a **mental model** of data between views
- ✗ **intermediate views** are usually meaningless



Basic idea: multiple views

**Animated transitions
on scatterplots**



morphingProjections

demand
each weekday

demand
each day hour



Basic idea: multiple views

**Animated transitions
on scatterplots**



morphingProjections

demand
each weekday

hourly demand
each weekday

demand
each day hour



Basic idea: multiple views

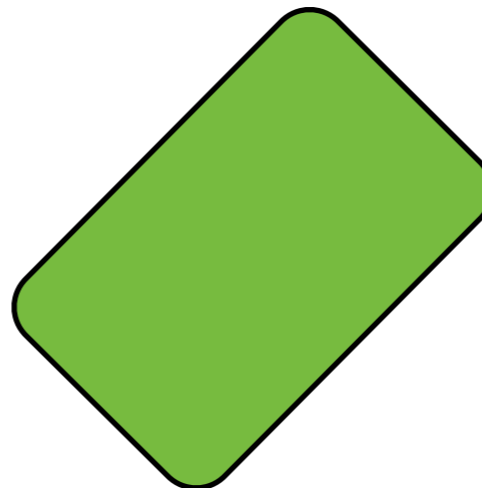
Animated transitions
on scatterplots



morphingProjections

demand
each weekday

hourly demand
each weekday



demand
each day hour

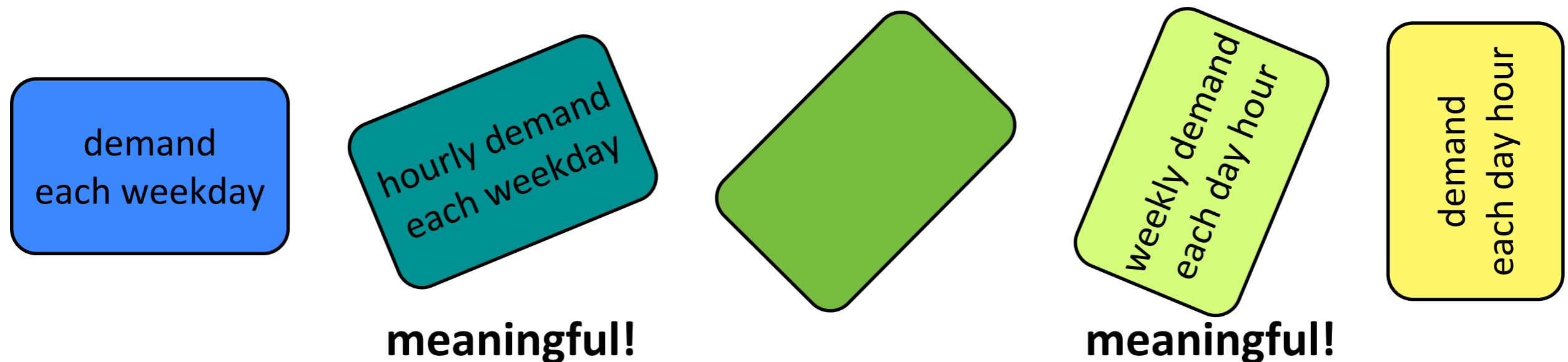


Basic idea: multiple views

Animated transitions
on scatterplots



morphingProjections

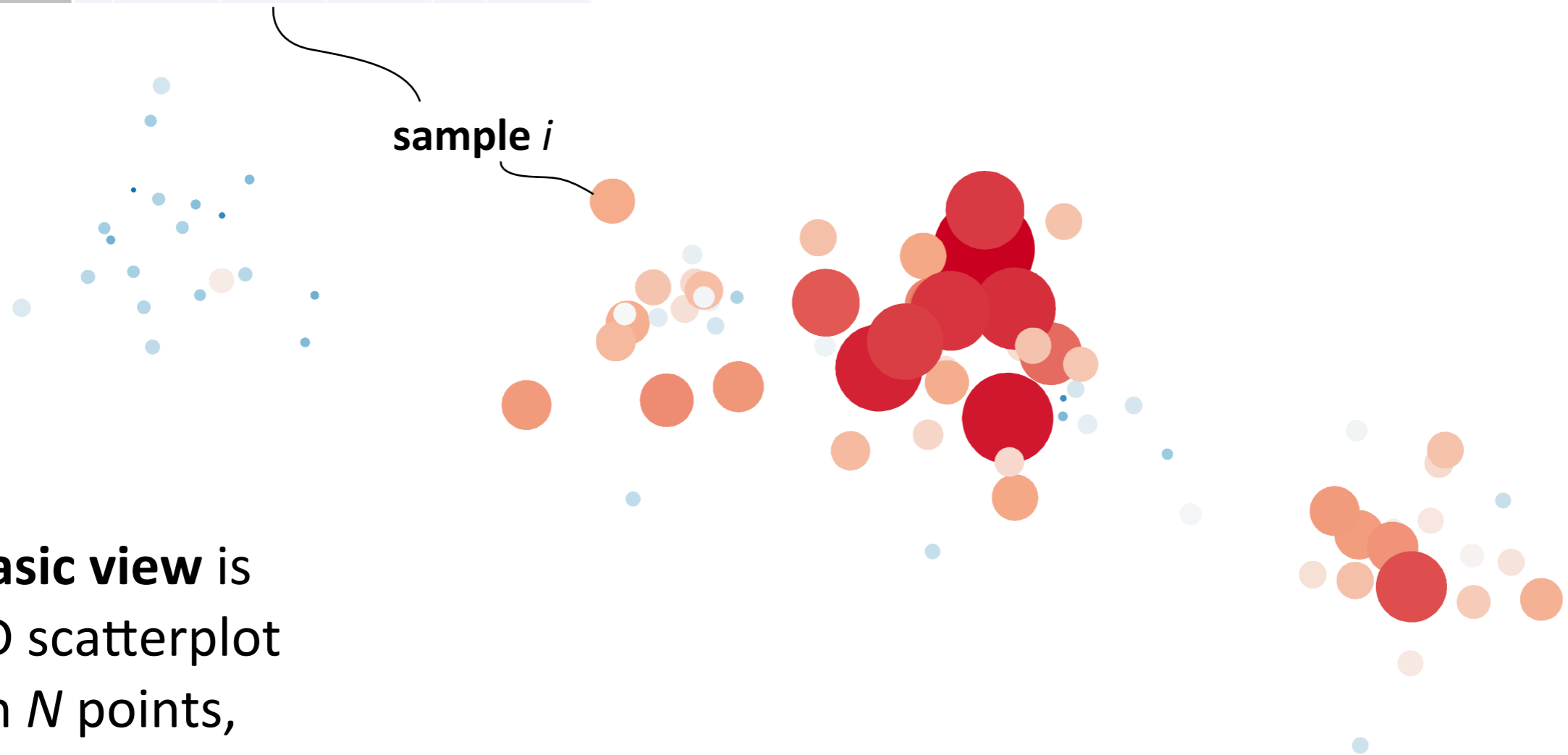


- ✓ helps to keep a **mental model** of data between views
- ✓ yields **meaningful intermediate views!**



Basic view elements

		sample $i-1$	sample i	sample $i+1$		sample N
absolute hour $h(i)$...	47	48	49	...	8760
absolute day $d(i)$...	2	2	3	...	365
active power	...	21,232	22,232	22,352	...	55,333
reactive power	...	1,112	1,456	1,234	...	2,444



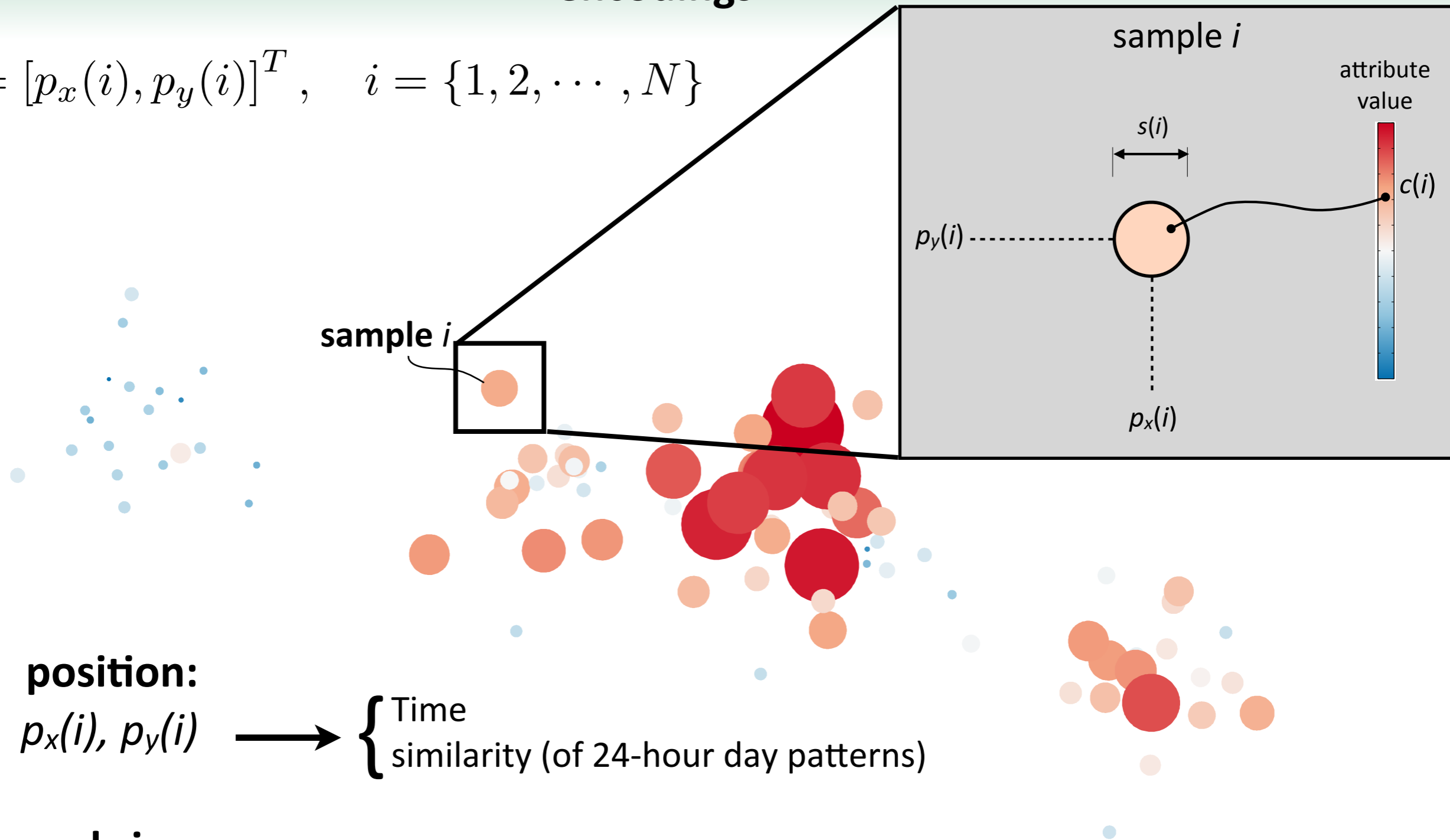
A **basic view** is
a 2D scatterplot
with N points,
representing N samples



Basic view

encodings

$$\mathbf{p}(i) = [p_x(i), p_y(i)]^T, \quad i = \{1, 2, \dots, N\}$$



position:

$p_x(i), p_y(i)$

→ { Time
similarity (of 24-hour day patterns)

color and size:

$c(i), s(i)$

→ { attribute value (e.g. active power demand)

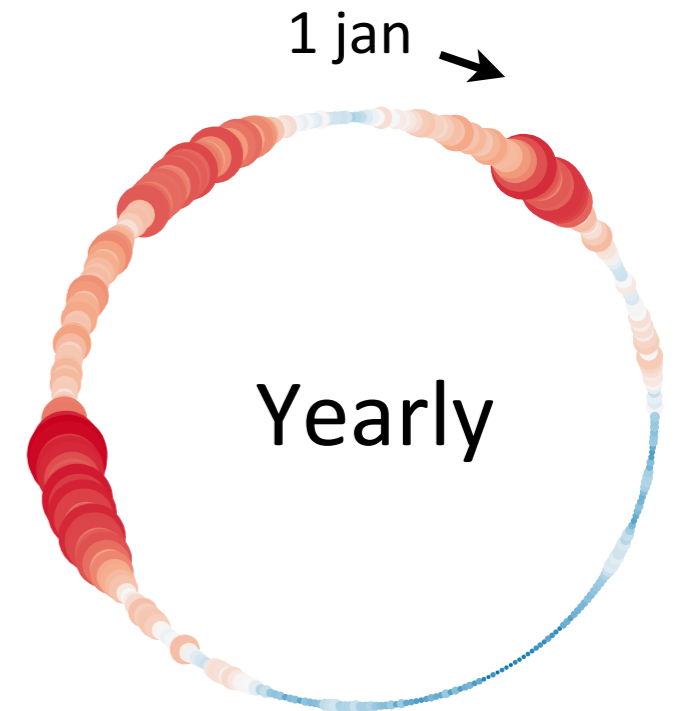
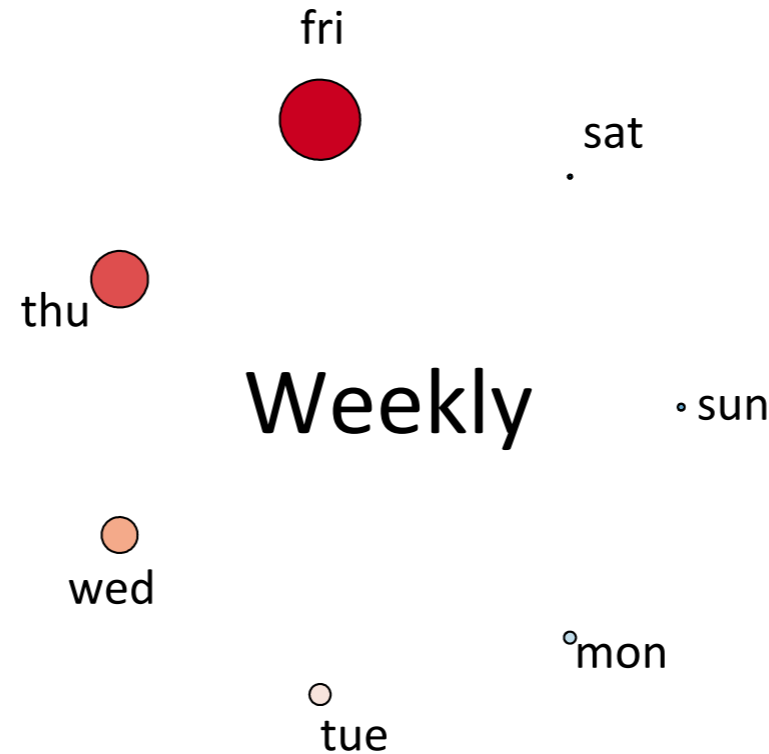
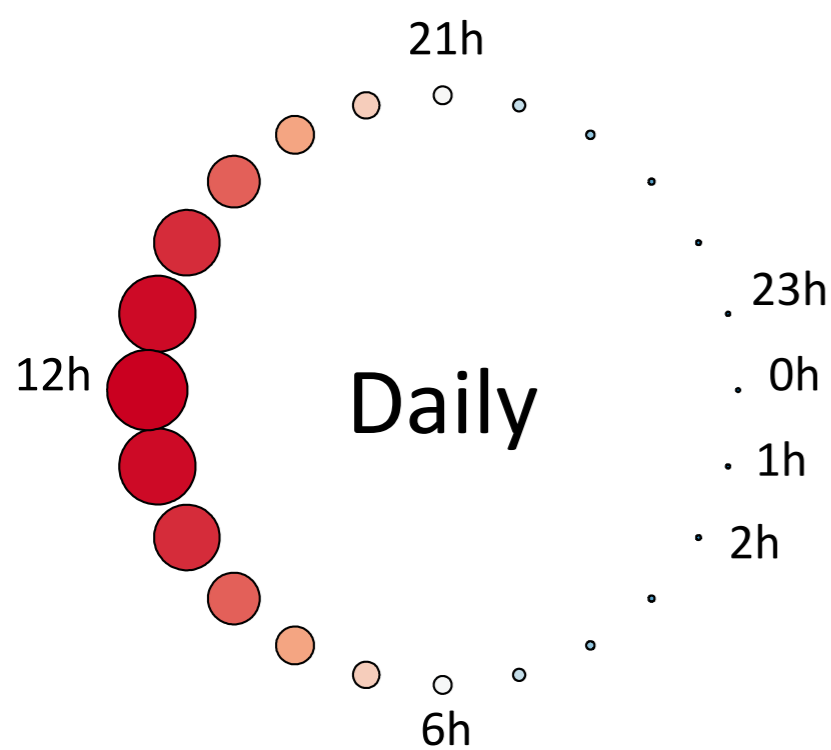


Clock encodings

describing regular periodicities

Rationales:

- clocks are widely accepted conventions
- congruence: periodical nature of hours, weekdays and years



$$\mathbf{p}_D(i) = \left[\cos \left(2\pi \frac{h(i)}{24} \right), \sin \left(2\pi \frac{h(i)}{24} \right) \right]$$

$$\mathbf{p}_Y(i) = \left[\cos \left(2\pi \frac{h(i)}{365 \cdot 24} \right), \sin \left(2\pi \frac{h(i)}{365 \cdot 24} \right) \right]$$

$$\mathbf{p}_W(i) = \left[\cos \left(2\pi \frac{d(i)}{7} \right), \sin \left(2\pi \frac{d(i)}{7} \right) \right]$$



Calendar encodings

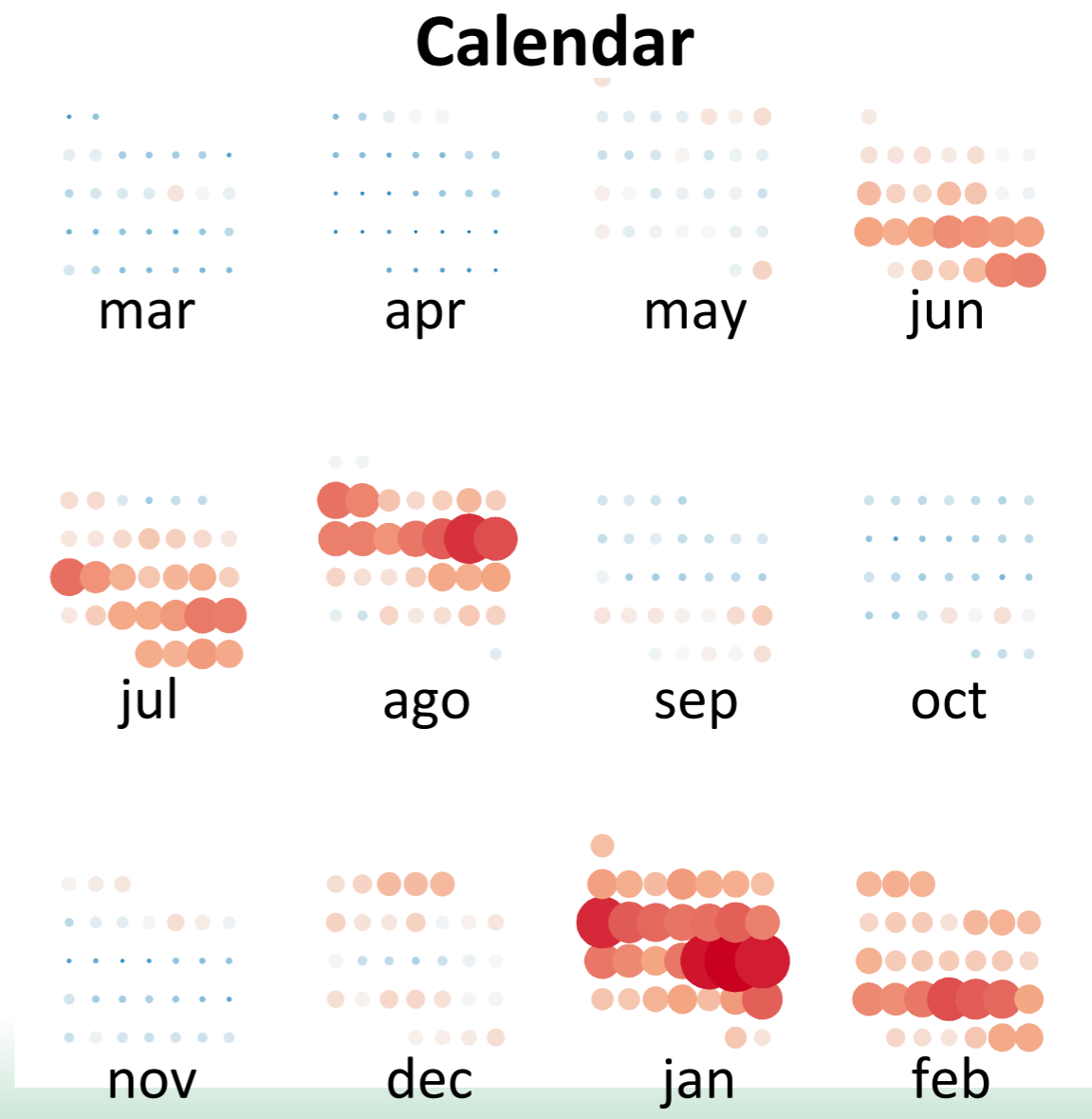
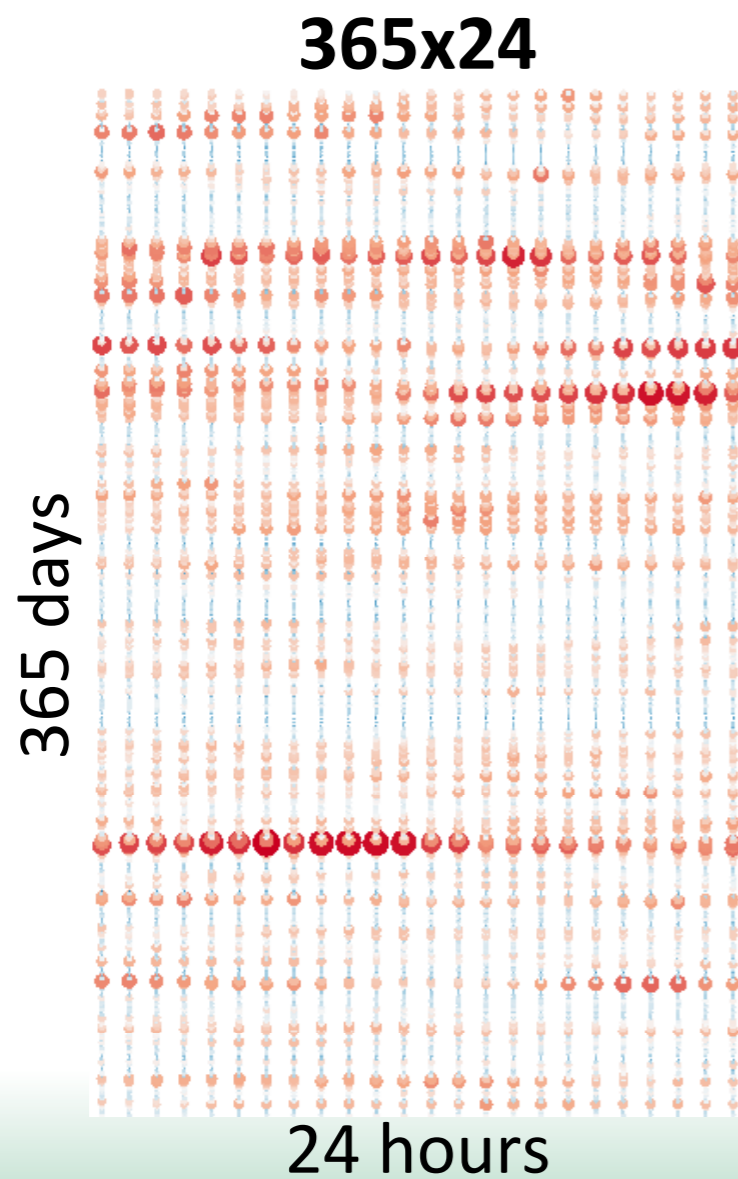
seasonal and social time granularities

Rationales:

- seasonal periodicities
- relative position in the year

Rationales:

- social time granularities
- common mental models for time



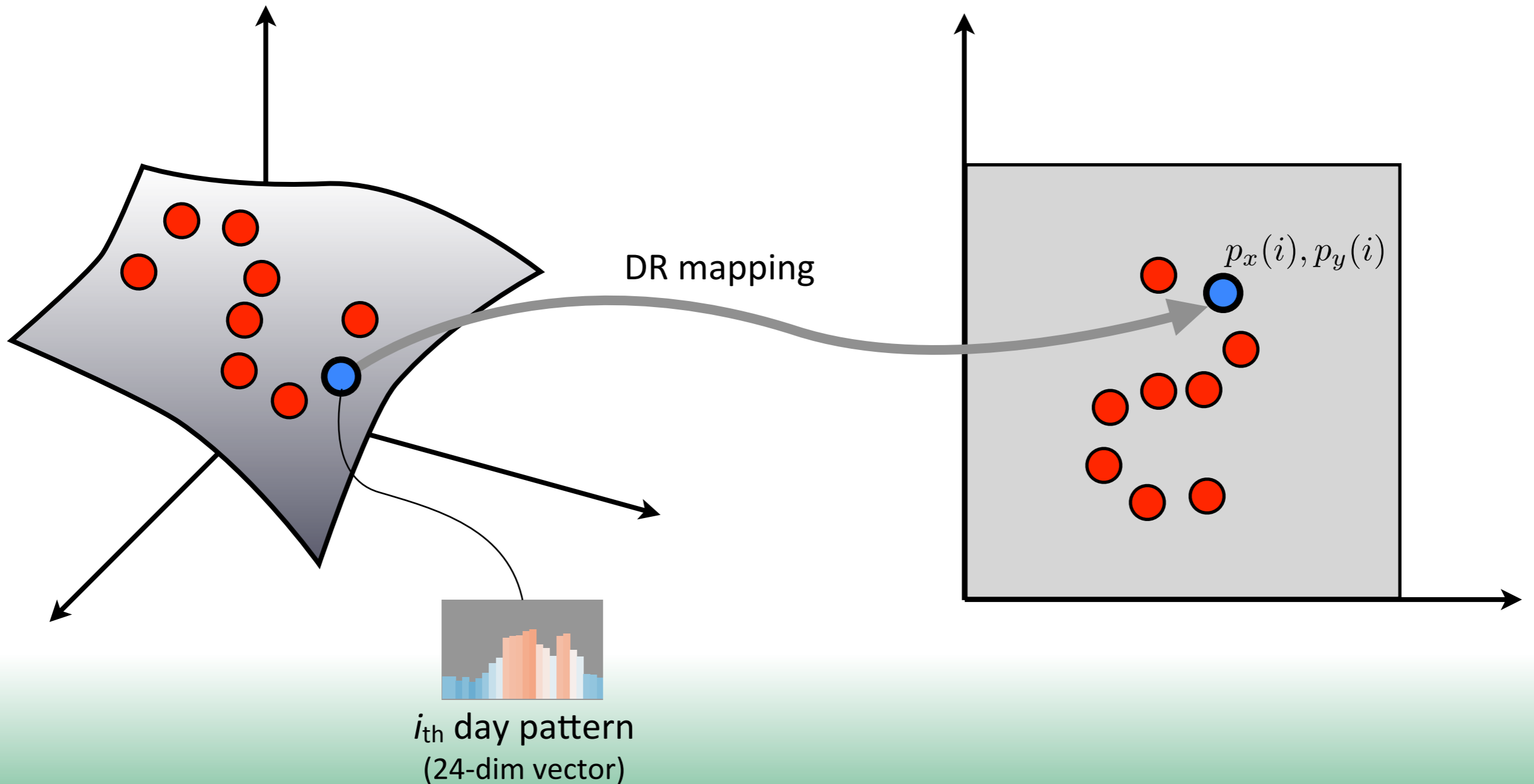


Dimensionality reduction

describing similarity

high dimensional space of
Day patterns R^{24}

2D space
for visualization





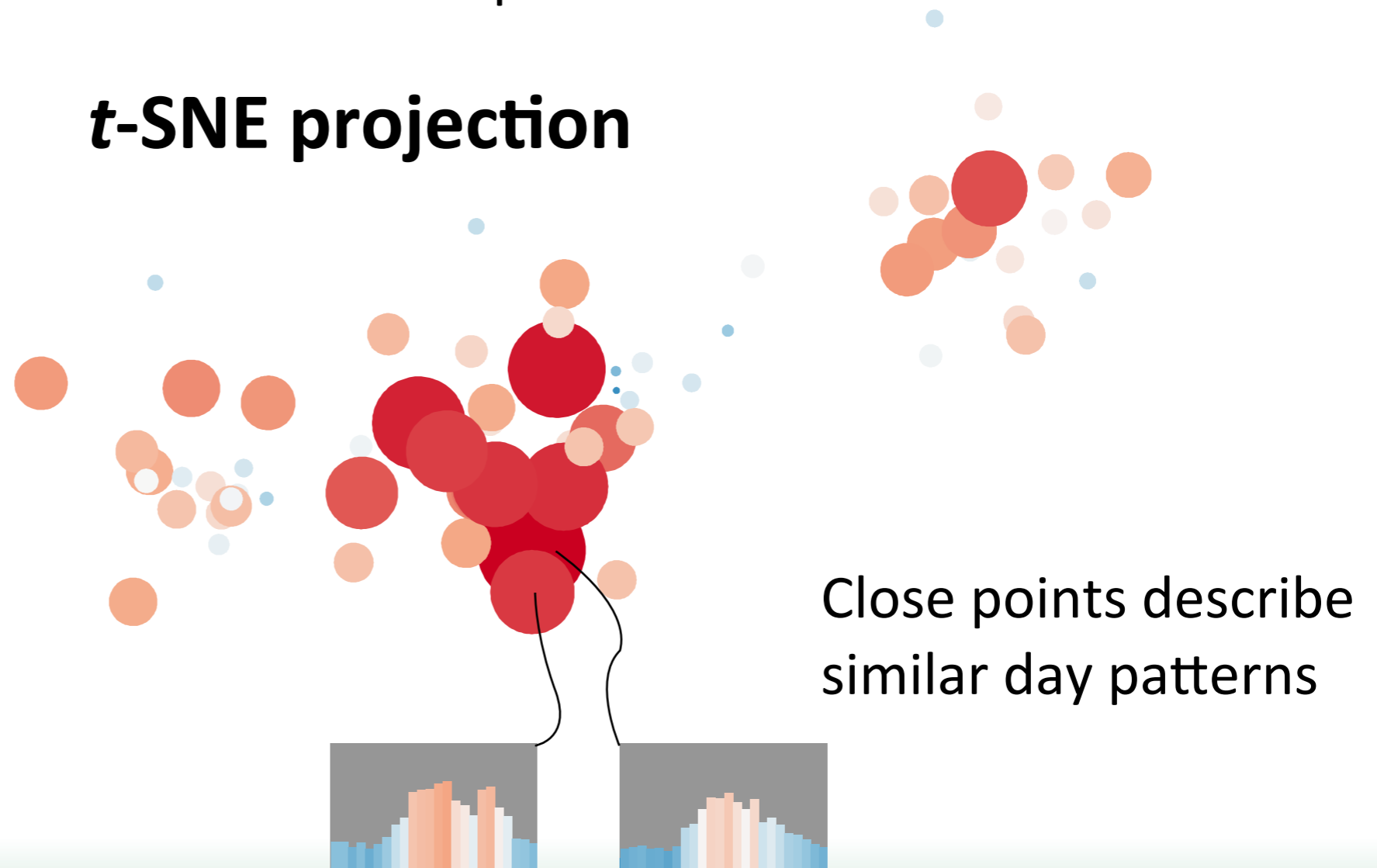
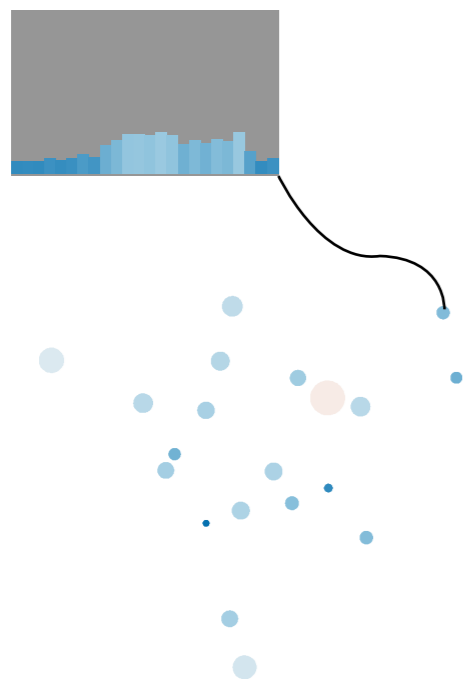
Dimensionality reduction

describing similarity

Rationales:

- associate spatial proximity to similarity
- clusters reveal days with similar demand patterns

t-SNE projection





Color and size encodings

describing attributes

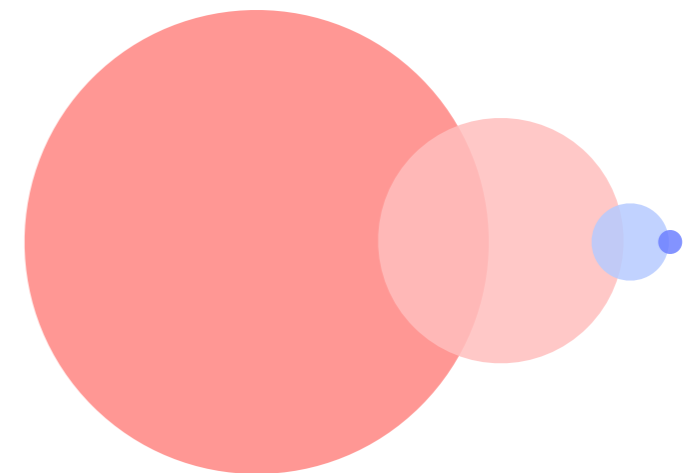
Color

- ▶ Scale ColorBrewer 2.0 tool
- ▶ Multihue blue/white/red
- ▶ divergent
- ▶ colorblind safe



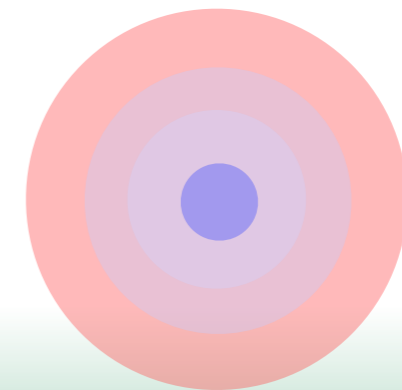
Size

- ▶ Exponential w.r.t. attribute value
- ▶ Highlight peaks of demand



Transparency

- ▶ conveys aggregation info
- ▶ concentric circles reveal patterns



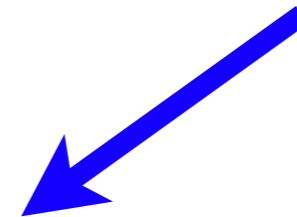
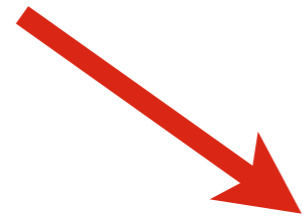
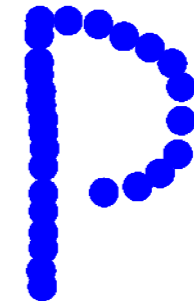
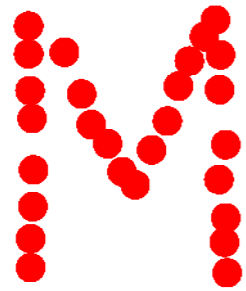


Morphing

basic concept

$$\mathbf{p}(i), \quad i = \{1, 2, \dots, N\}$$

$$\mathbf{q}(i), \quad i = \{1, 2, \dots, N\}$$



$$\mathbf{z}(i) = \lambda \mathbf{p}(i) + (1 - \lambda) \mathbf{q}(i)$$



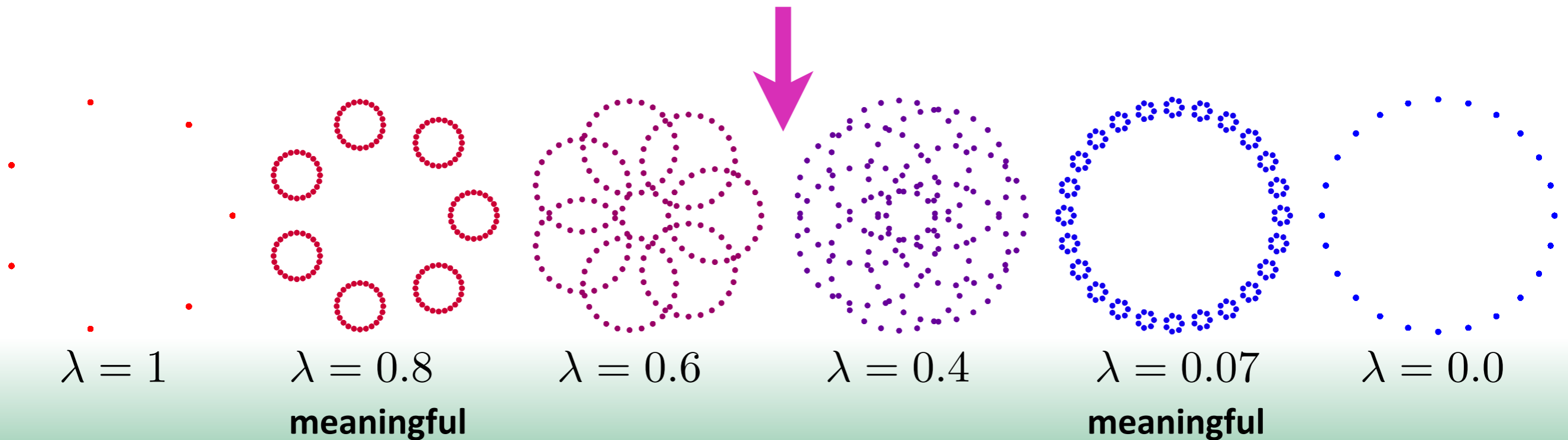


Morphing

meaningful transition views



$$\mathbf{z}(i) = \lambda \mathbf{p}(i) + (1 - \lambda) \mathbf{q}(i)$$



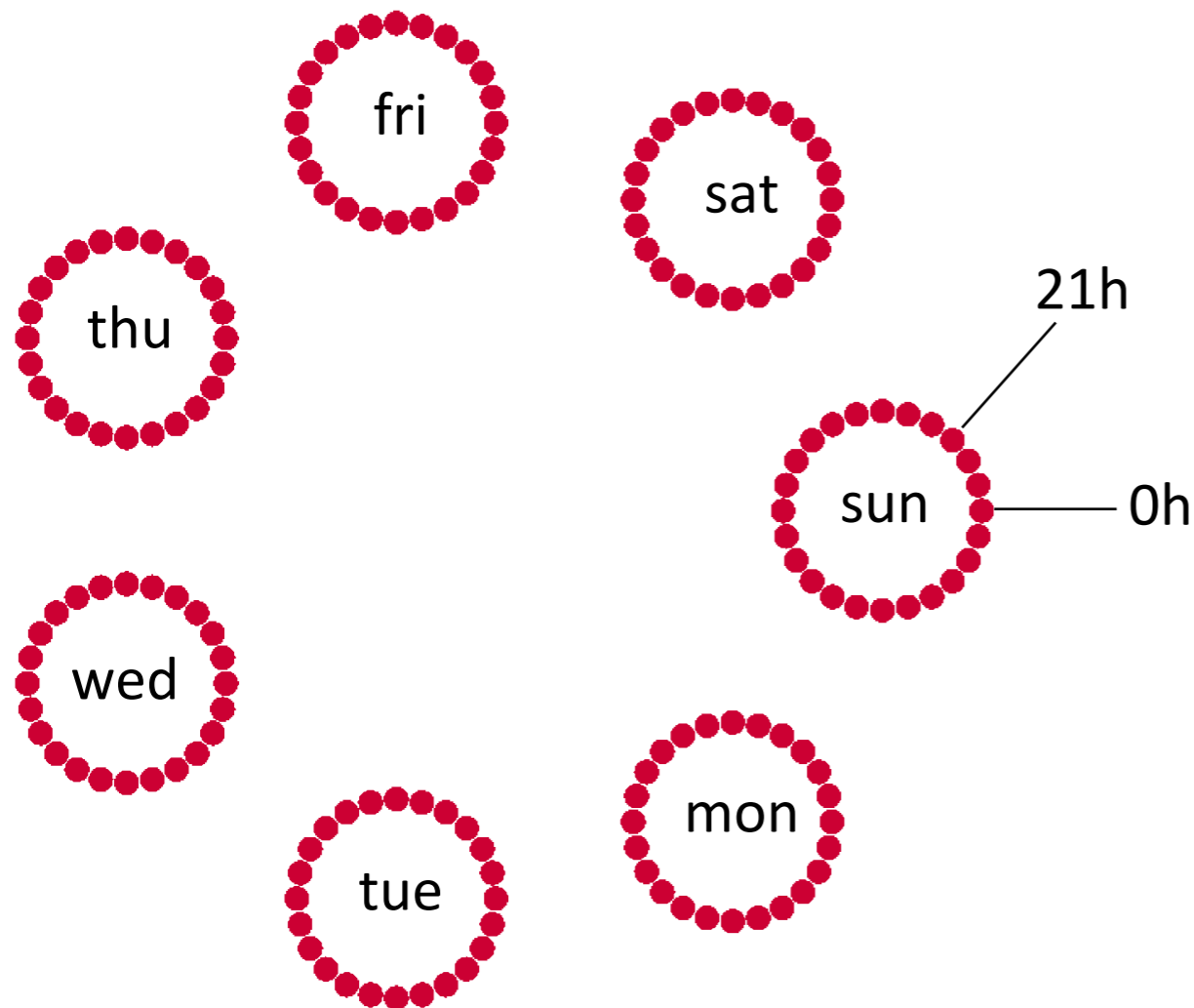


Morphing

meaningful transition views

$$\lambda = 0.8$$

7 days, 24 hours each



“daily evolving
of the demand
for each weekday”

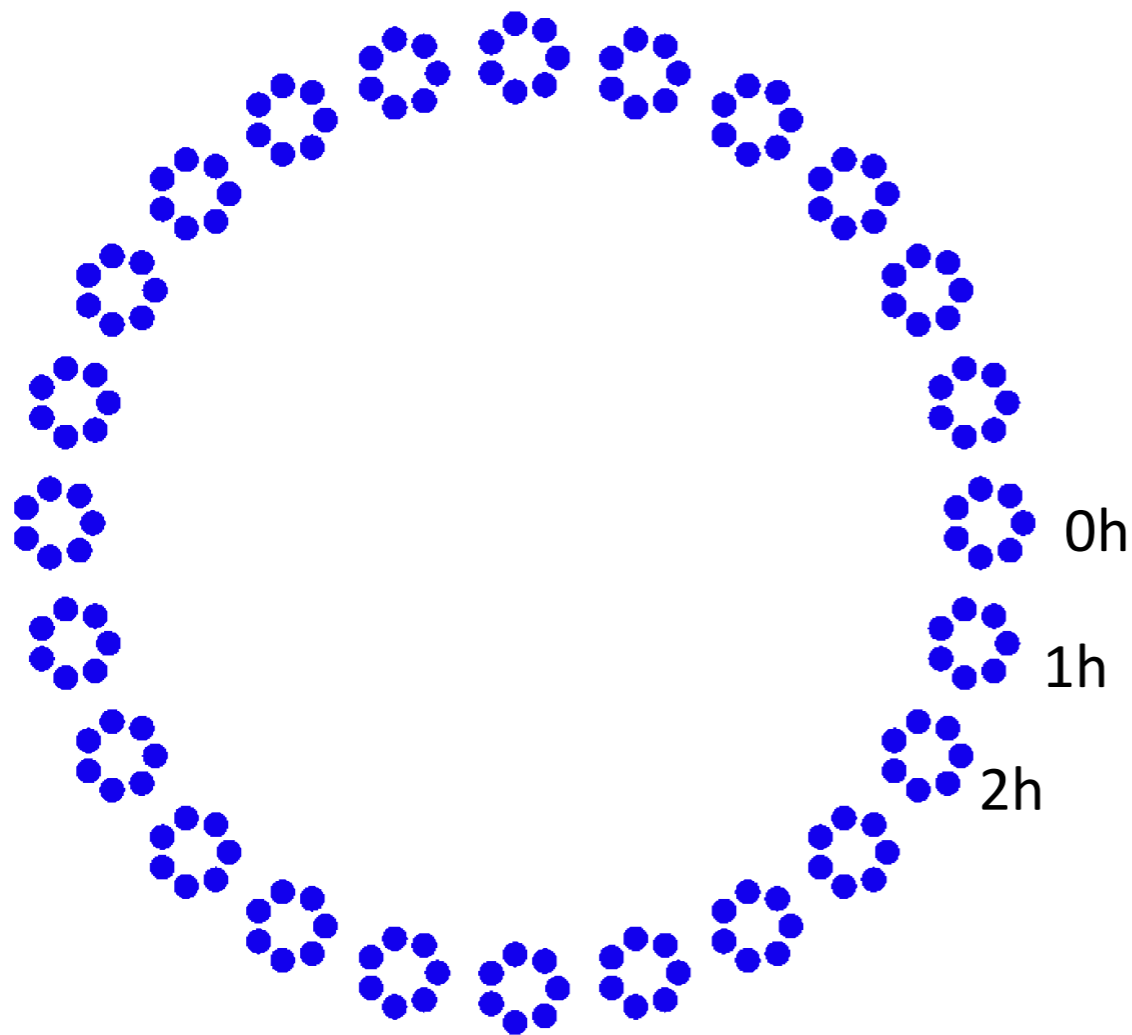


Morphing

meaningful transition views

$$\lambda = 0.07$$

24 hours, 7 days each



weekly distribution
of the demand
for every day hour

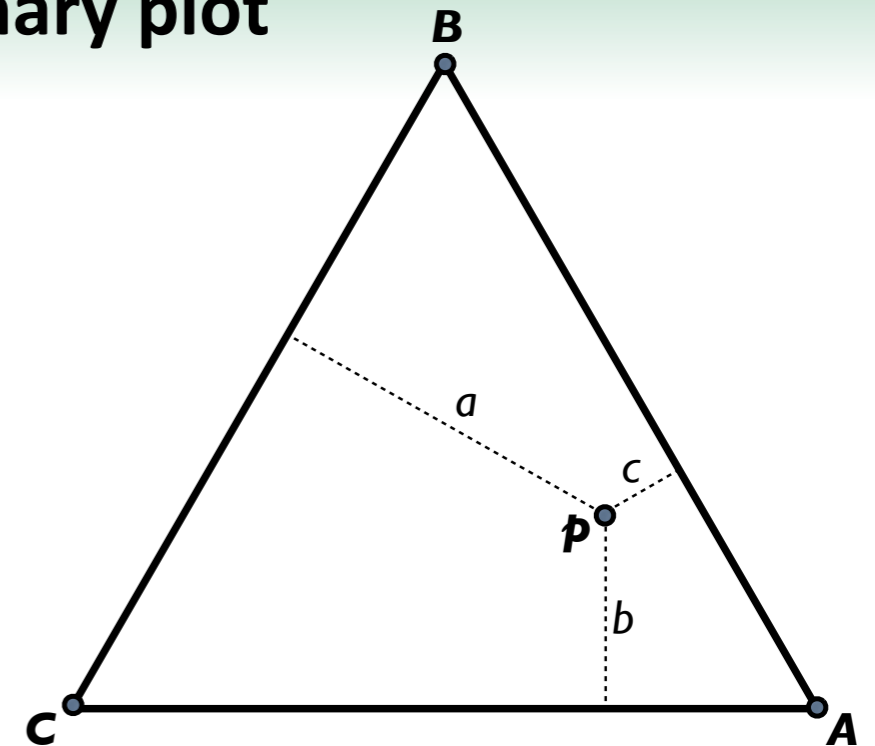


Morphing

blending three views: the ternary plot

Mixing coefficient
for encoding A

coordinates
of encoding A



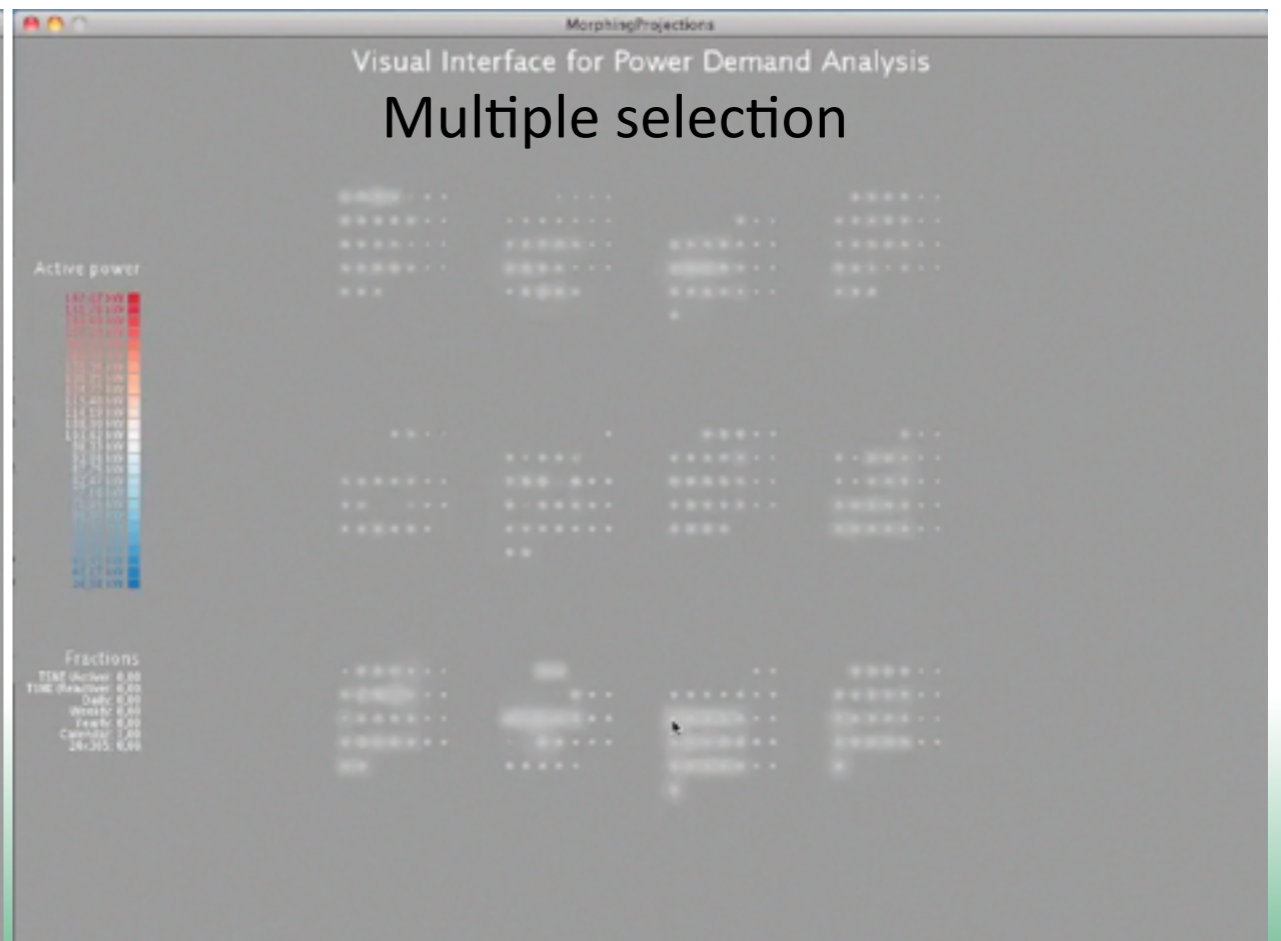
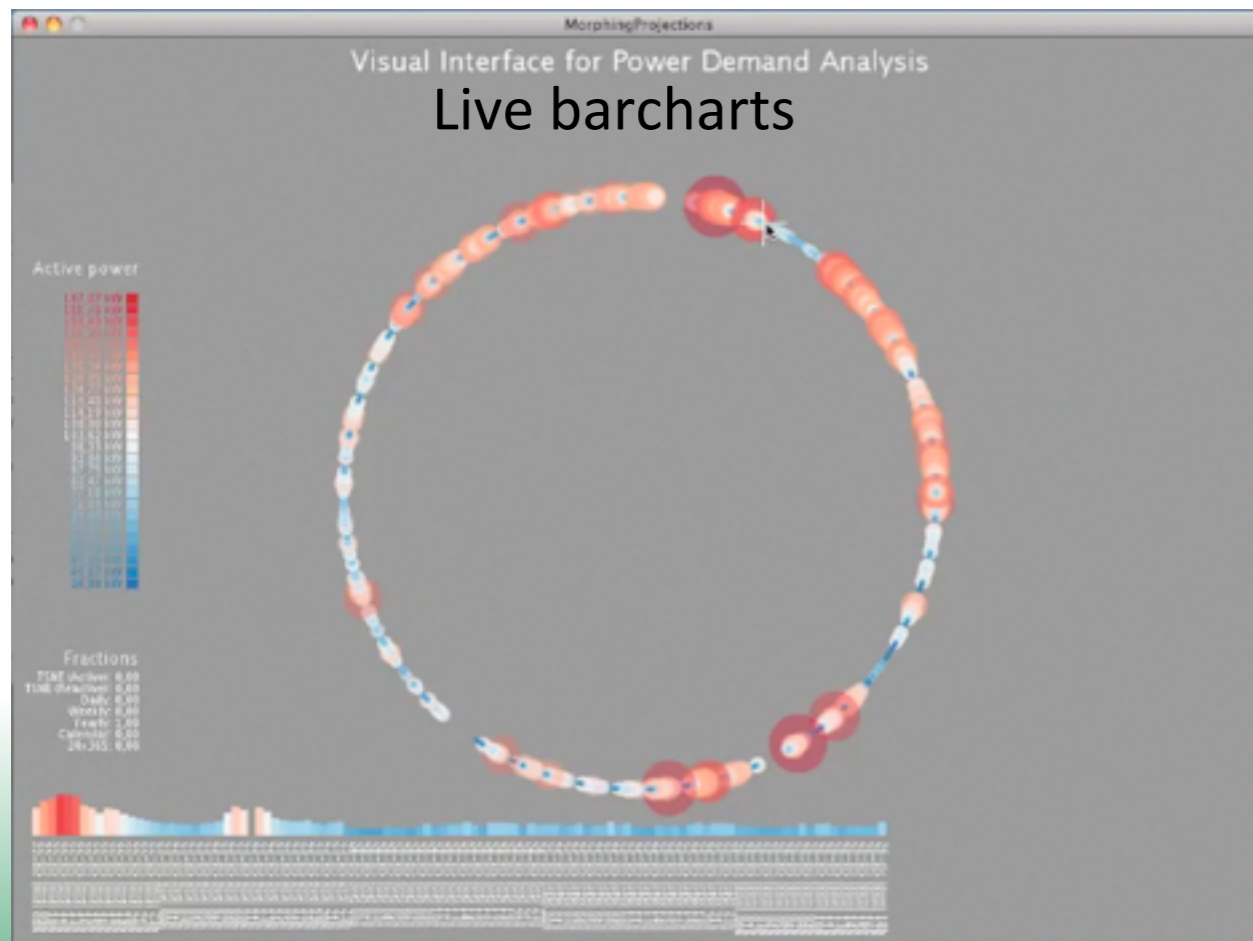
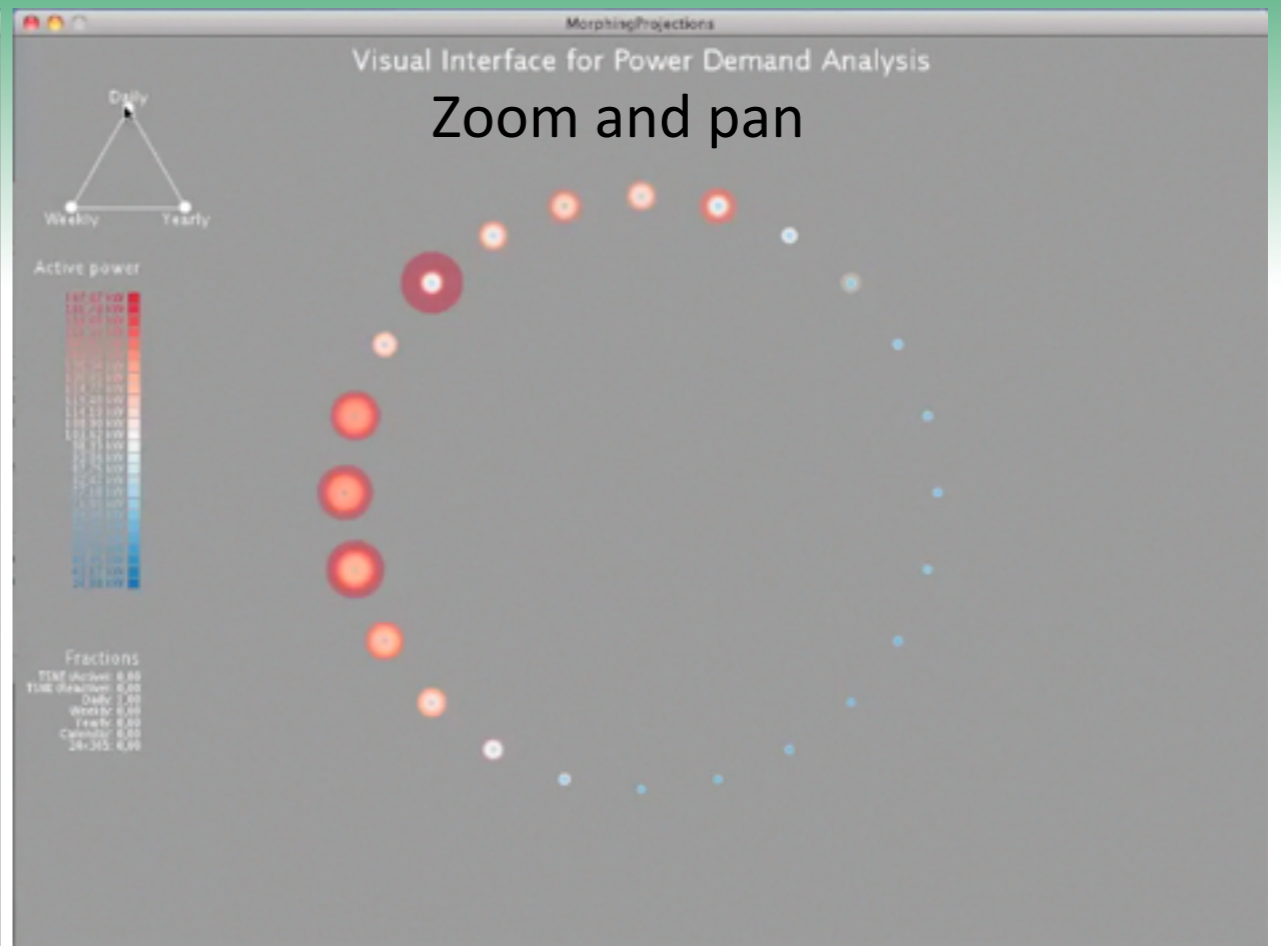
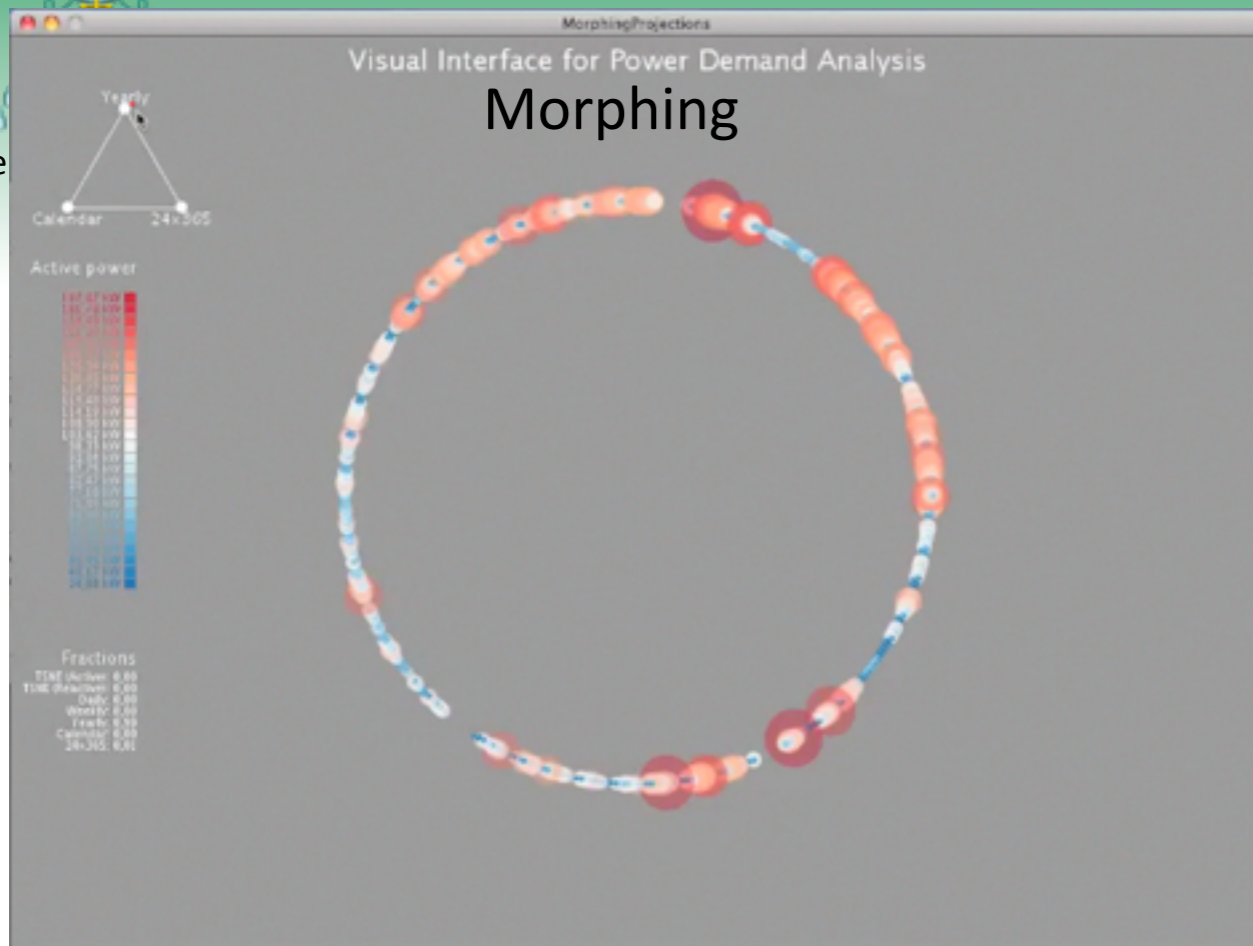
$$\mathbf{p}(i, t) = a(t)\mathbf{p}_A(i) + b(t)\mathbf{p}_B(i) + c(t)\mathbf{p}_C(i)$$

The mixture proportions sum up to 1 (100%):

$$a(t) + b(t) + c(t) = 1$$



Interface and Interaction elements





Conclusions

- **Spatial encodings**

- ▶ clocks → regular periodicities
- ▶ calendars → social time granularities
- ▶ dimension reduction → group by similarity → interactive clustering

- **Morphing**

- ▶ advantages of animated transitions → global view
- ▶ meaningful results → combined views

- **Interaction mechanisms**

- ▶ ternary plot, multiple selection, brushing, zoom and pan, context information retrieval

- **Future research**

- ▶ user study
- ▶ morphing: quite general idea
- ▶ spatio-temporal analysis (geo-encodings)
- ▶ Multiway analysis