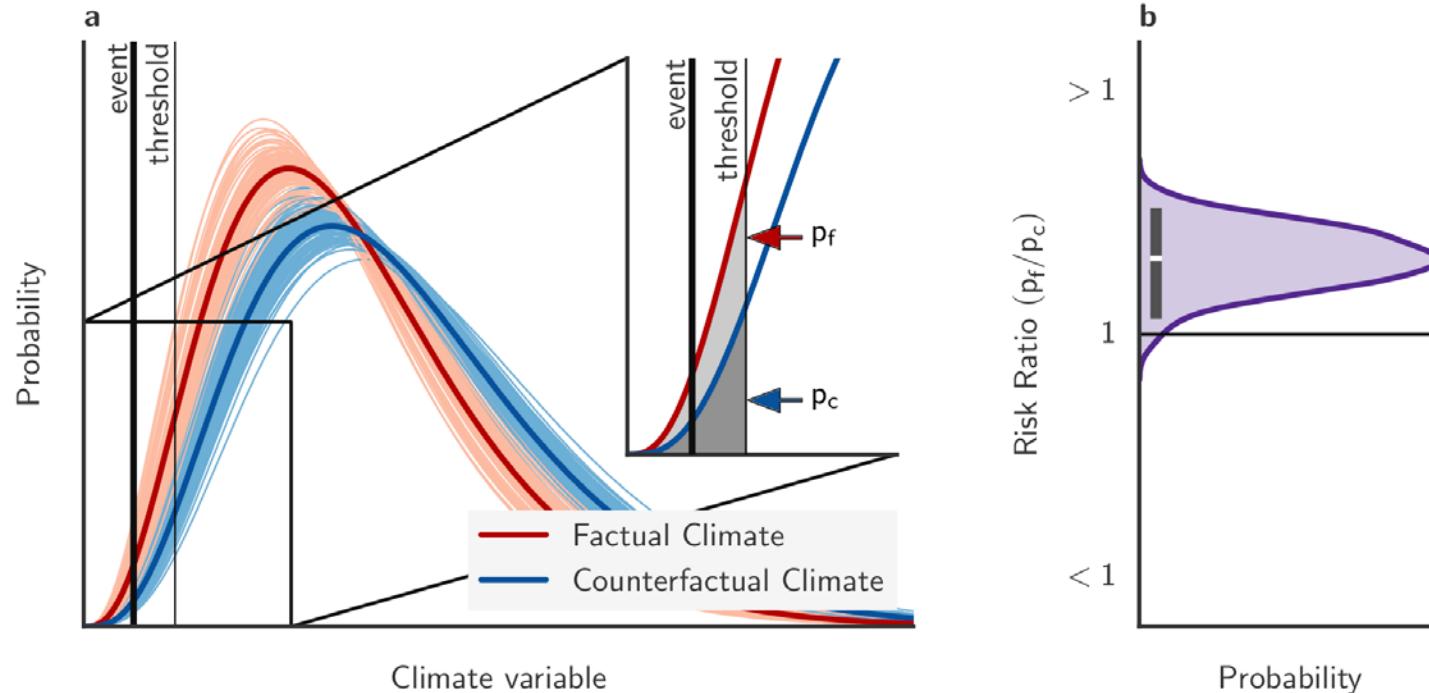


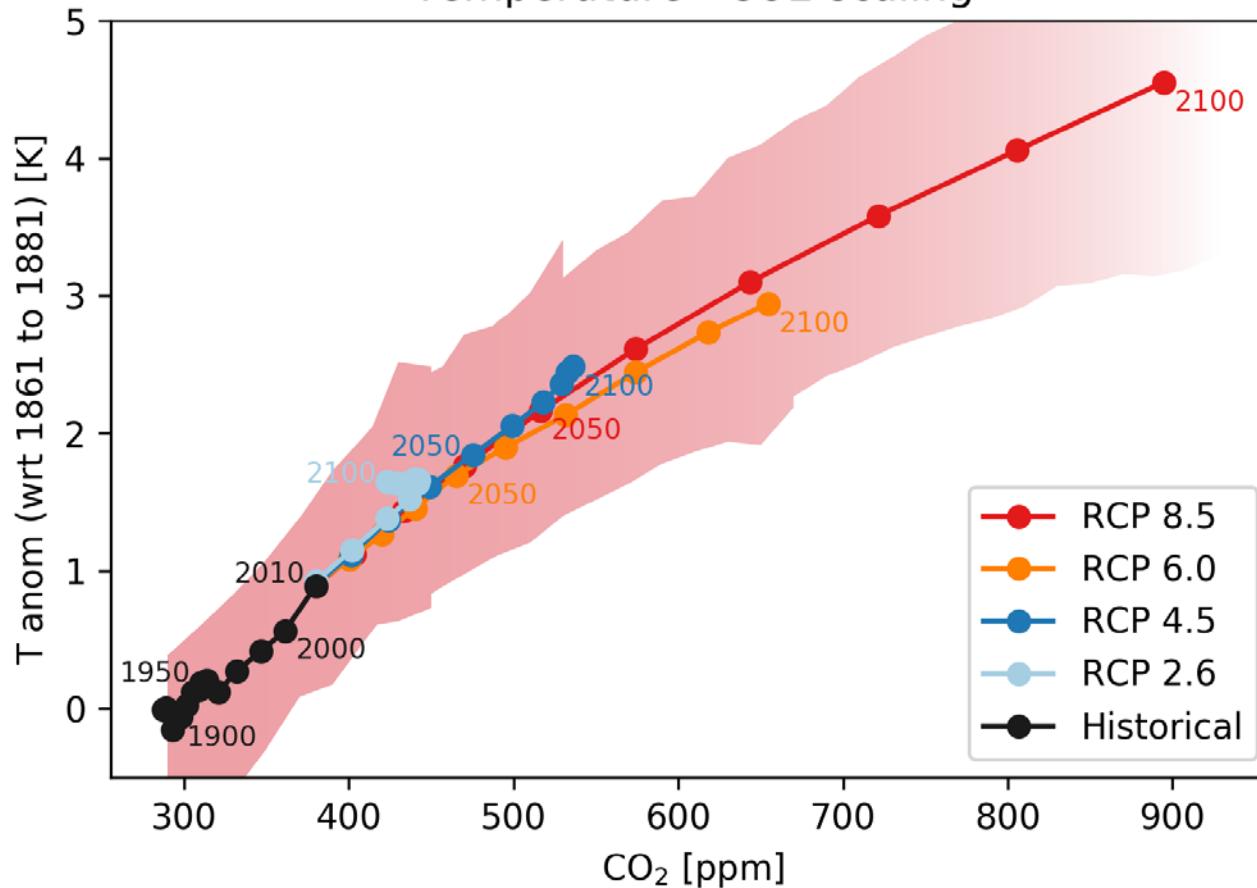
Python Visualization Workshop



C2SM – Mathias Hauser, Tarun Chadha

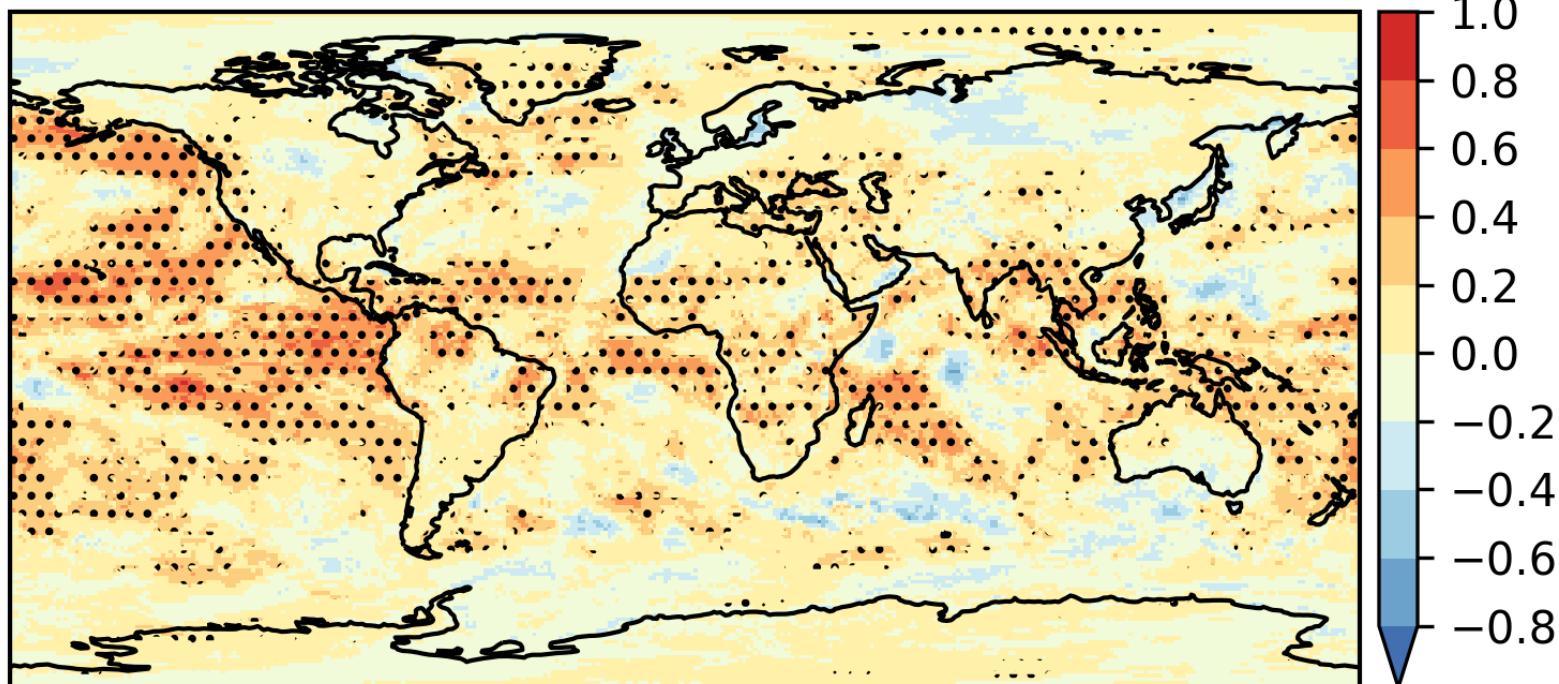
Teaser

Temperature - CO₂ scaling

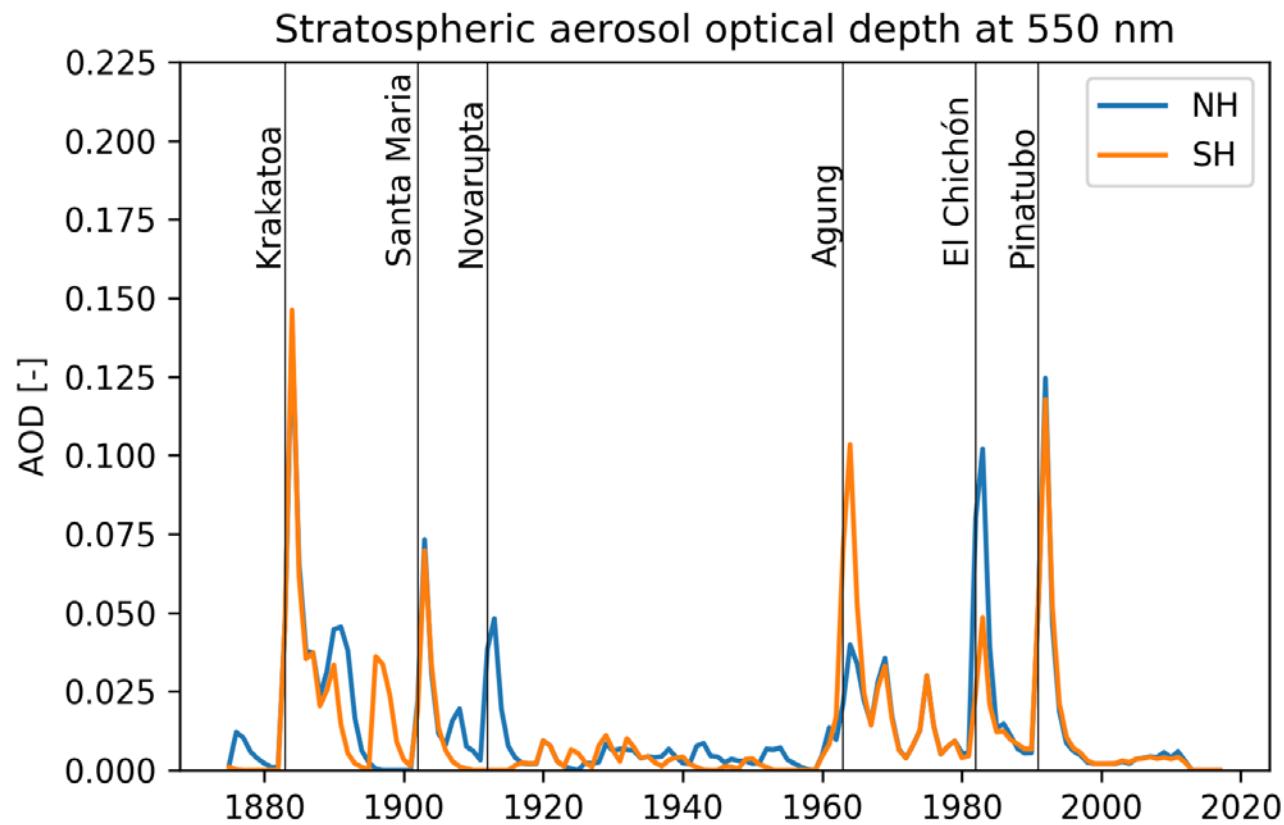


Teaser

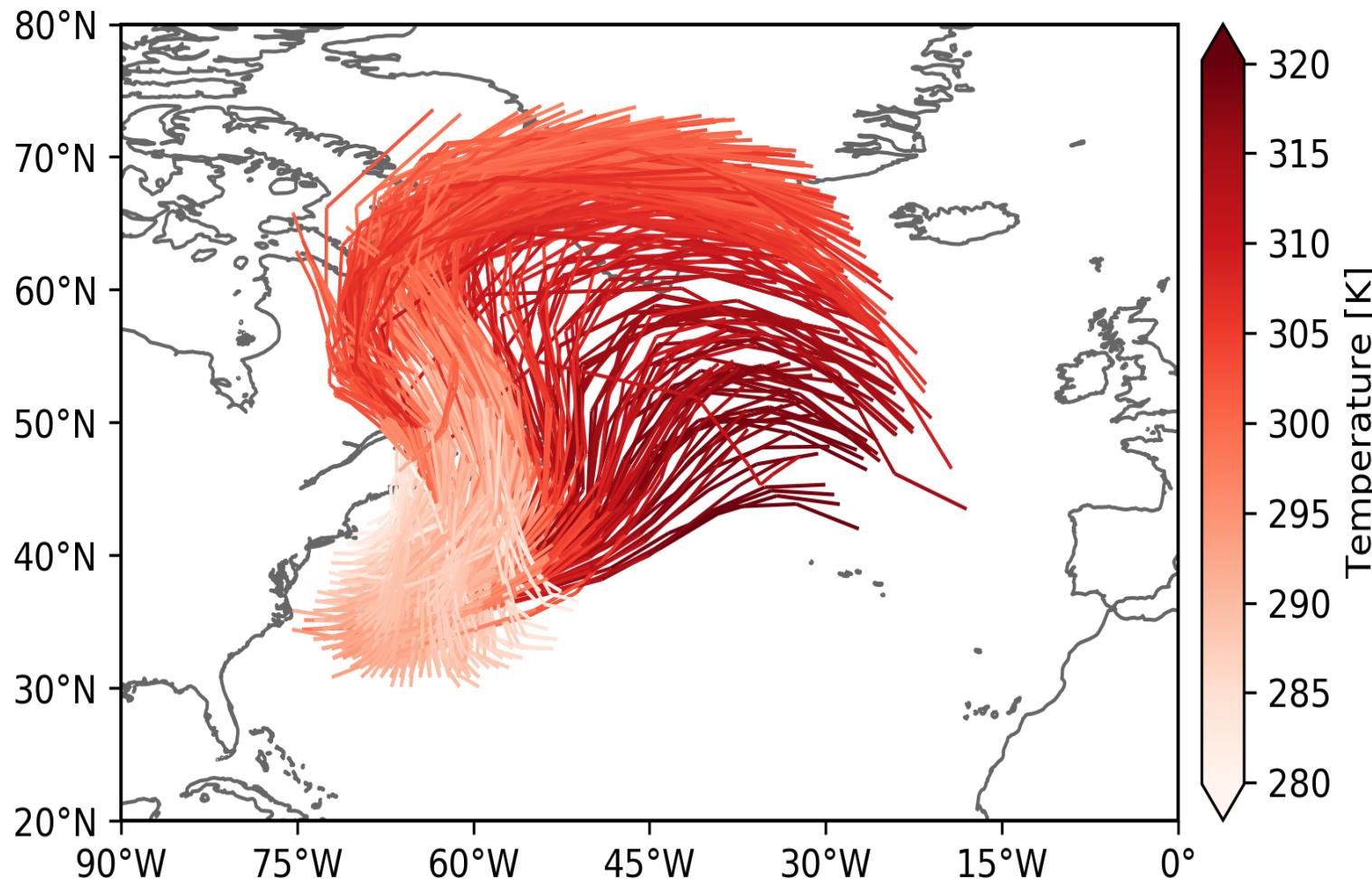
T2M: Ranked probability skill score, JJA



Teaser

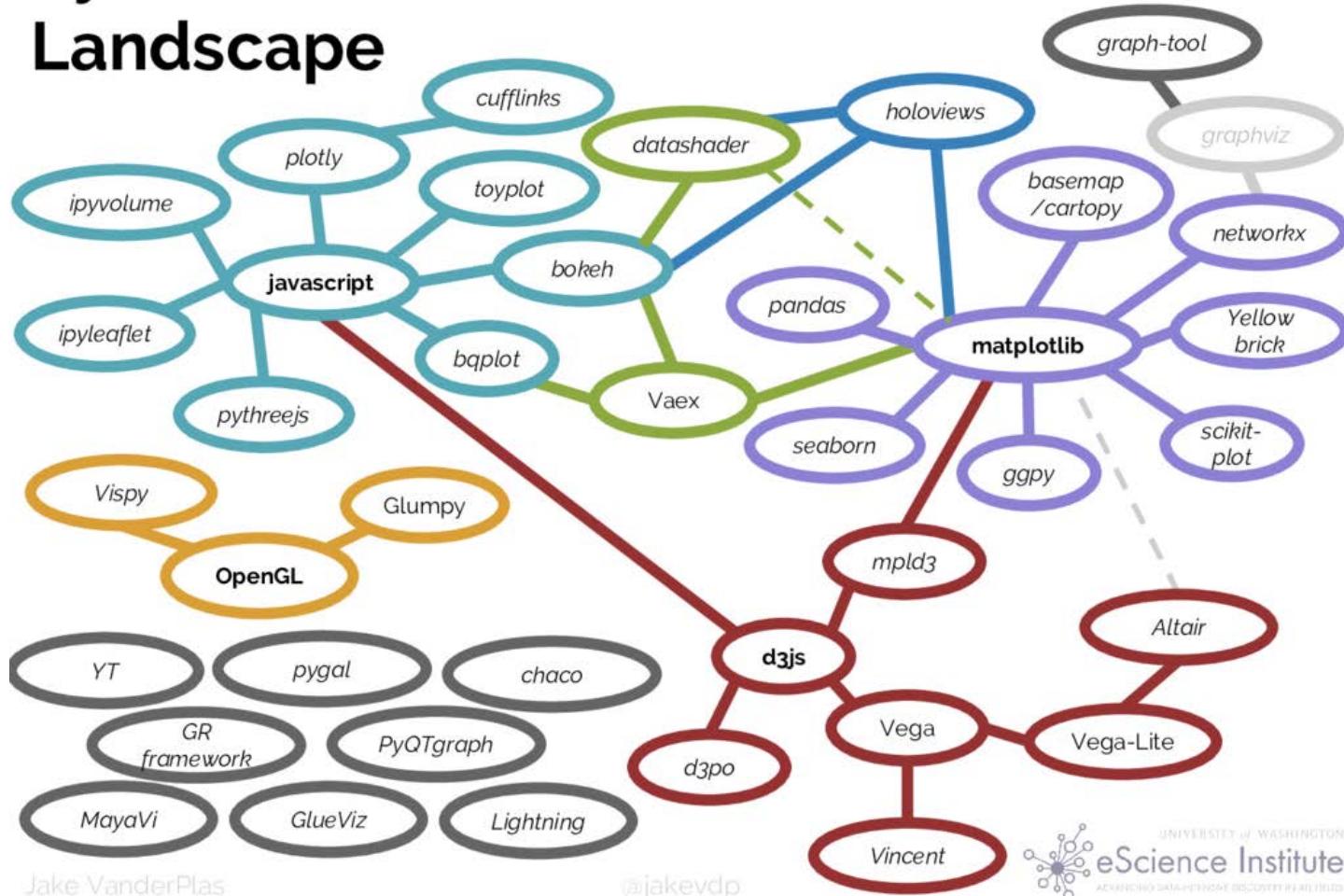


Teaser



Python Plotting Libraries

Python's Visualization Landscape



Jake VanderPlas

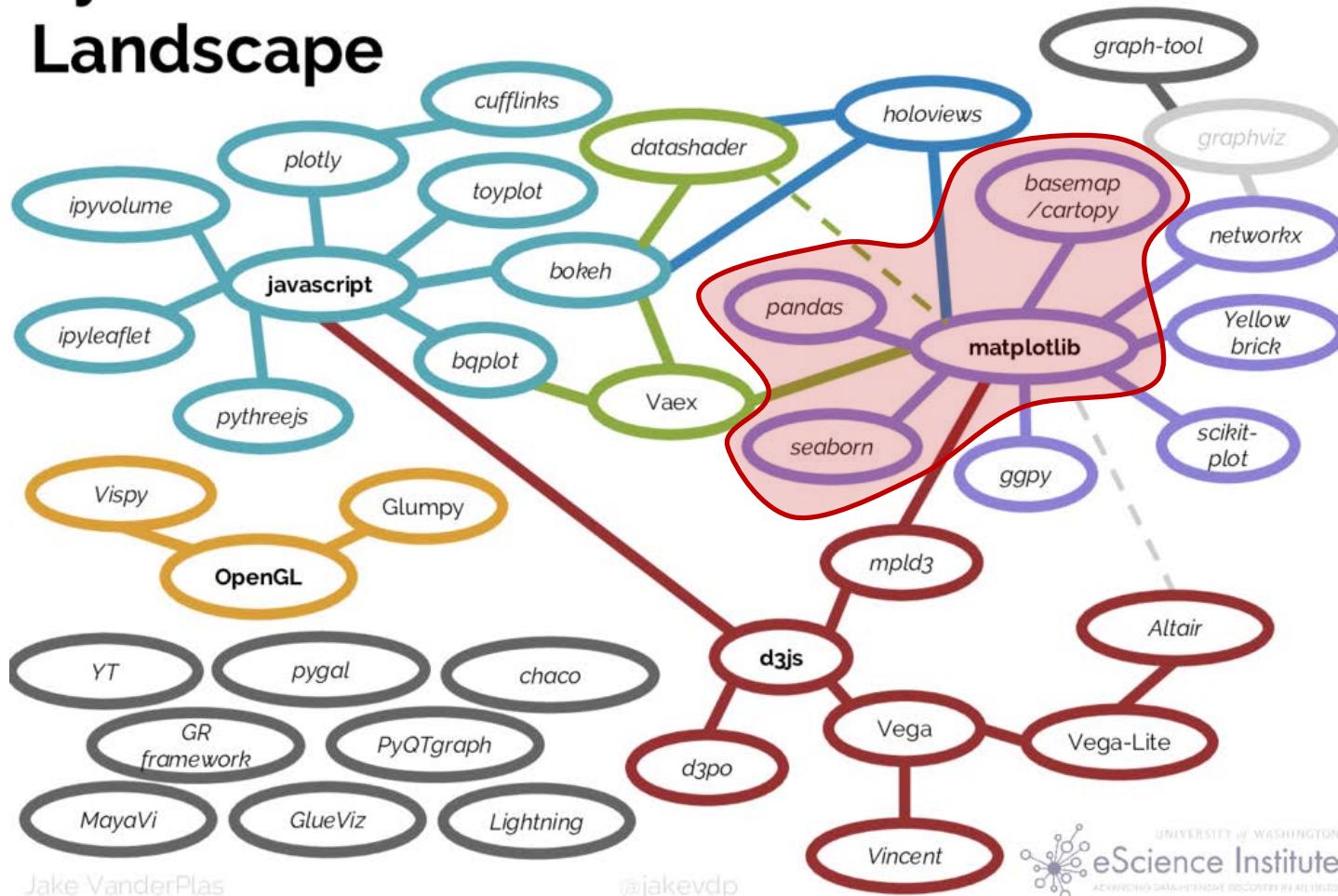
@jakevdp

UNIVERSITY OF WASHINGTON
eScience Institute
ADVANCED DATA-INTENSIVE DISCOVERY IN SCIENCE

www.matplotlib.org

Python Plotting Libraries

Python's Visualization Landscape



Jake VanderPlas

@jakevdp



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Part 0: Introduction

- No plotting, yet (sorry)
- integrated development environment
 - ipython notebooks (Exercise 0.1)
- Data handling libraries
 - numpy (Exercise 0.2)
 - netCDF4 (Exercise 0.3)
 - xarray (Exercise 0.4)
 - pandas (Exercise 0.5)

Part 0: numpy

- <http://www.numpy.org/>
- N-dimensional array

```
>>> import numpy as np
>>> a = np.arange(15).reshape(3, 5)
>>> a
array([[ 0,  1,  2,  3,  4],
       [ 5,  6,  7,  8,  9],
       [10, 11, 12, 13, 14]])
>>> a.shape
(3, 5)
```

Part 0: pandas

- Statistical package for “labeled” data
- Provides an R-like DataFrame

```
In [6]: dates = pd.date_range('20130101', periods=6)

In [7]: dates
Out[7]:
DatetimeIndex(['2013-01-01', '2013-01-02', '2013-01-03', '2013-01-04',
               '2013-01-05', '2013-01-06'],
              dtype='datetime64[ns]', freq='D')

In [8]: df = pd.DataFrame(np.random.randn(6,4), index=dates, columns=list('ABCD'))

In [9]: df
Out[9]:
          A         B         C         D
2013-01-01  0.469112 -0.282863 -1.509059 -1.135632
2013-01-02  1.212112 -0.173215  0.119209 -1.044236
2013-01-03 -0.861849 -2.104569 -0.494929  1.071804
2013-01-04  0.721555 -0.706771 -1.039575  0.271860
2013-01-05 -0.424972  0.567020  0.276232 -1.087401
2013-01-06 -0.673690  0.113648 -1.478427  0.524988
```

Part 0: netCDF4

- Most common module to read & write netCDFs

```
In [5]: fN = '../data/HadEX2_GSL.nc'
```

```
In [6]: ncf = nc.Dataset(fN)

print(ncf)
```

```
Out[6]: <class 'netCDF4._netCDF4.Dataset'>
root group (NETCDF4_CLASSIC data model, file format HDF5):
    data: Growing season length
    source: HadEX2 (http://www.climdex.org/)
    reference: Donat et al., 2013
    dimensions(sizes): lon(96), lat(73), time(50)
    variables(dimensions): float64 lon(lon), float64 lat(lat), int32 time(time), float32 GSL(time,lat,lon),
    (lat,lon), float64 p_val(lat,lon)
    groups:
```

```
In [11]: # get data of lon from the file
lon = ncf.variables['lon'][:]
# this is a numpy array
lon
```

```
Out[11]: array([  0. ,   3.75,   7.5 ,  11.25,  15. ,  18.75,  22.5 ,
 26.25,  30. ,  33.75,  37.5 ,  41.25,  45. ,  48.75,
```

Part 0: xarray

- Multidimensional labelled arrays
- combines a netCDF-like data model with capabilities of pandas
- ‘editors pick’

Part 0: xarray

```
In [2]: fN = './data/HadEX2_GSL.nc'
```

```
In [3]: ds = xr.open_dataset(fN)

ds
```

```
Out[3]: <xarray.Dataset>
Dimensions: (lat: 73, lon: 96, time: 50)
Coordinates:
 * lon      (lon) float64 0.0 3.75 7.5 11.25 15.0 18.75 22.5 26.25 30.0 ...
 * lat      (lat) float64 -90.0 -87.5 -85.0 -82.5 -80.0 -77.5 -75.0 -72.5 ...
 * time     (time) datetime64[ns] 1956-01-01 1957-01-01 1958-01-01 ...
Data variables:
    GSL      (time, lat, lon) float64 ...
    trend    (lat, lon) float64 ...
    p_val    (lat, lon) float64 ...
Attributes:
    data:      Growing season length
    source:    HadEX2 (http://www.climdex.org/)
    reference: Donat et al., 2013
```

```
In [4]: ds.trend
```

```
Out[4]: <xarray.DataArray 'trend' (lat: 73, lon: 96)>
array([[ nan,  nan,  nan, ...,  nan,  nan,  nan],
       [ nan,  nan,  nan, ...,  nan,  nan,  nan],
       [ nan,  nan,  nan, ...,  nan,  nan,  nan],
       ...,
       [ nan,  nan,  nan, ...,  nan,  nan,  nan],
       [ nan,  nan,  nan, ...,  nan,  nan,  nan],
       [ nan,  nan,  nan, ...,  nan,  nan,  nan]])
Coordinates:
 * lon      (lon) float64 0.0 3.75 7.5 11.25 15.0 18.75 22.5 26.25 30.0 ...
 * lat      (lat) float64 -90.0 -87.5 -85.0 -82.5 -80.0 -77.5 -75.0 -72.5 ...
```

Part 0: Setup

- www.github.com/c2sm/pyvis
- Setup in HG D12
 - git clone <https://github.com/c2sm/pyvis.git>
 - export CONDA_ENVS_PATH=/opt/kunden/hauser/conda/envs
 - source activate pyvis
 - jupyter notebook

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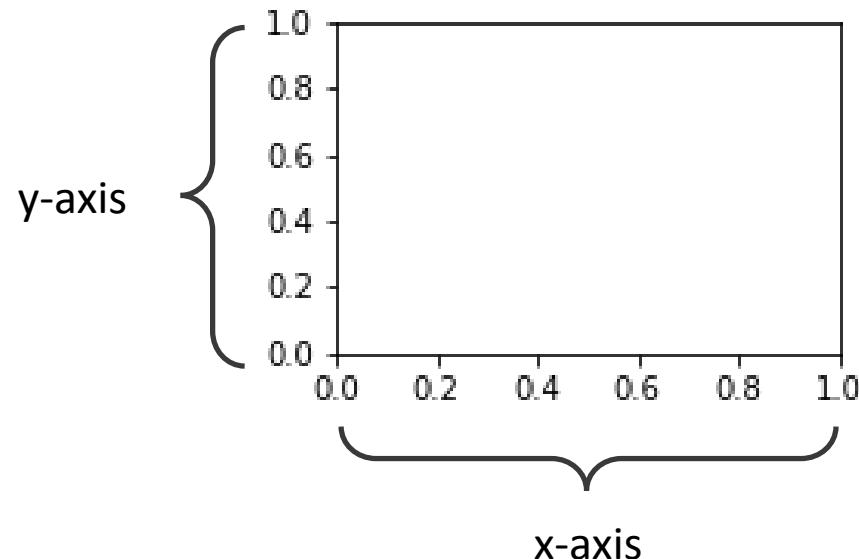
Part 1: matplotlib

- Matplotlib is a Python 2D plotting library
- Similar syntax as MATLAB
- Current version 2.2

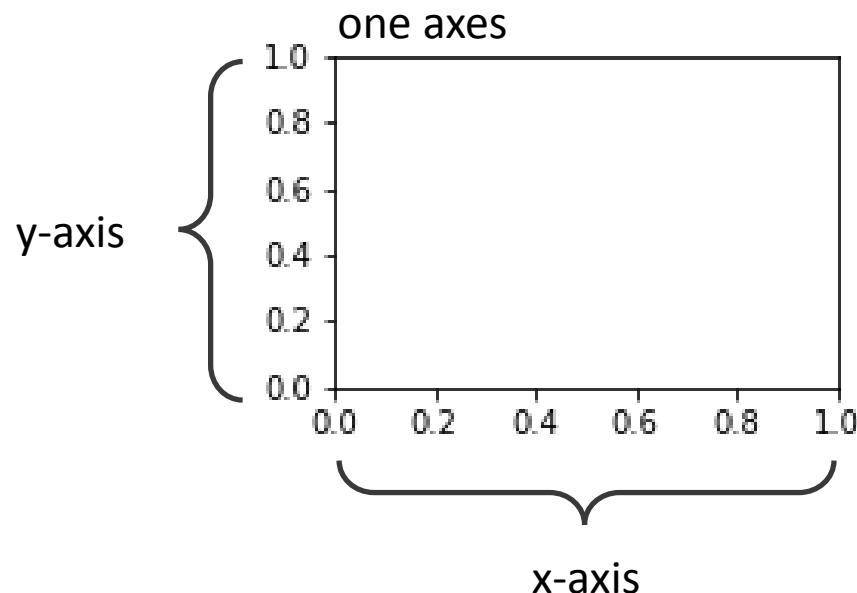
Part 1: Plotting in Python with matplotlib

- line plots
- uncertainty
- scatter plots
- subplots
- histograms
- annotations

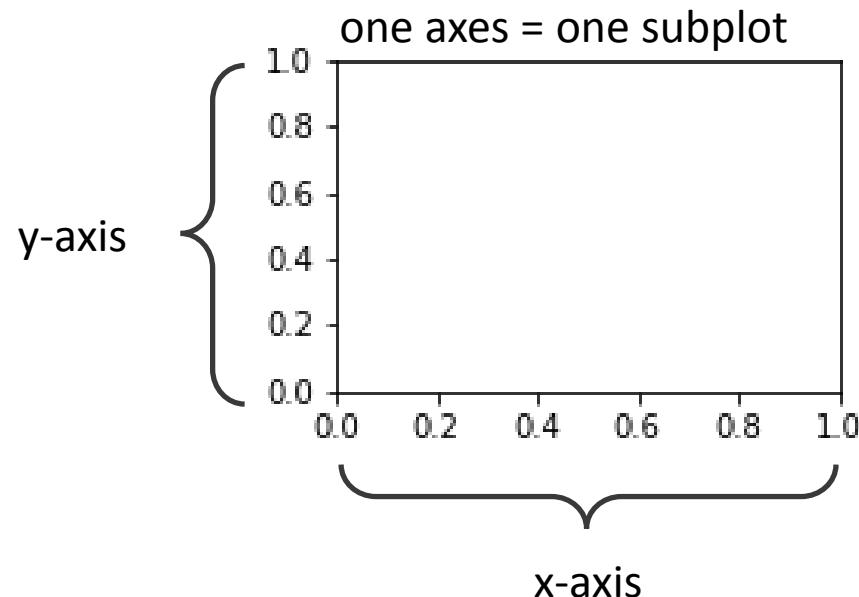
axis, axes, subplots, figure, spines, ...



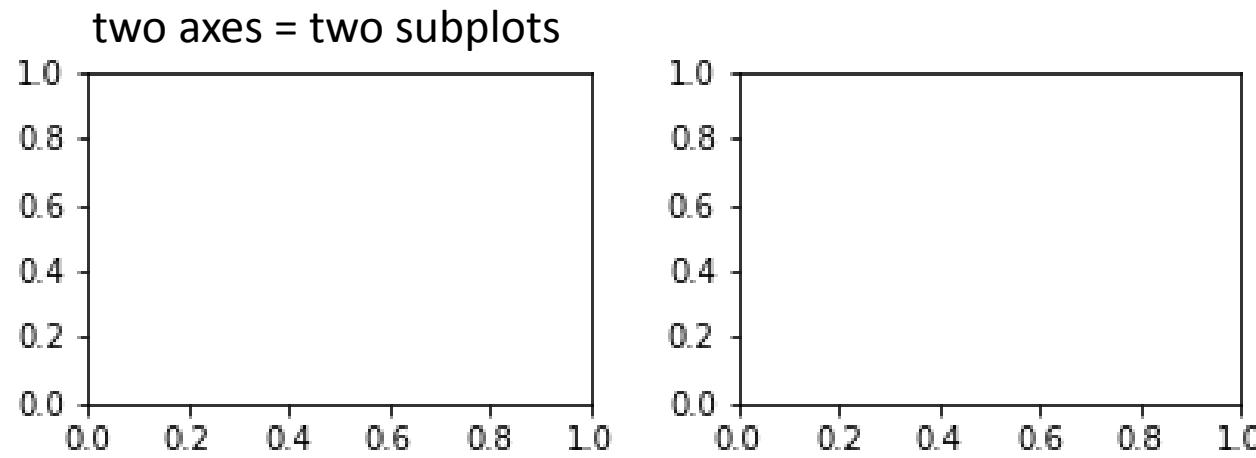
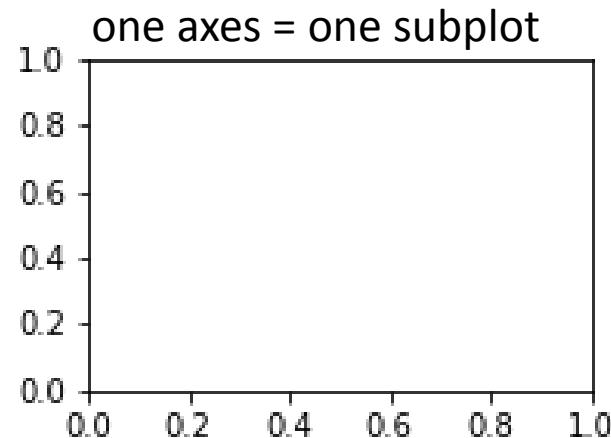
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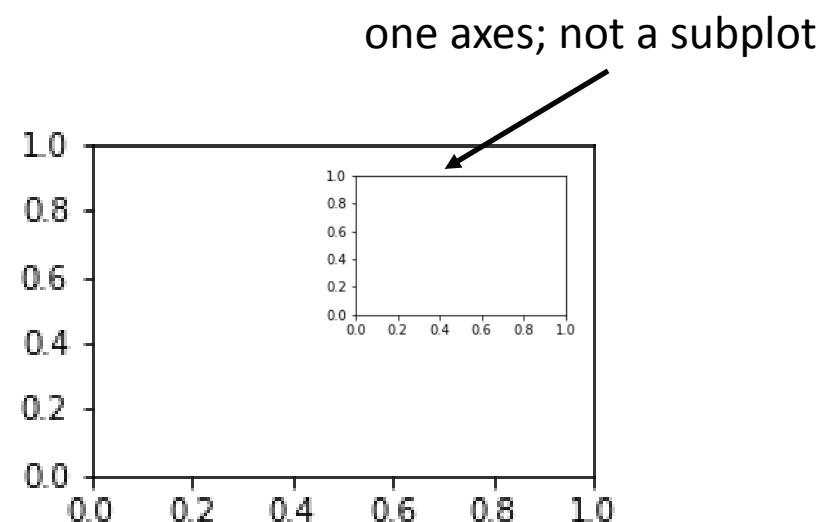
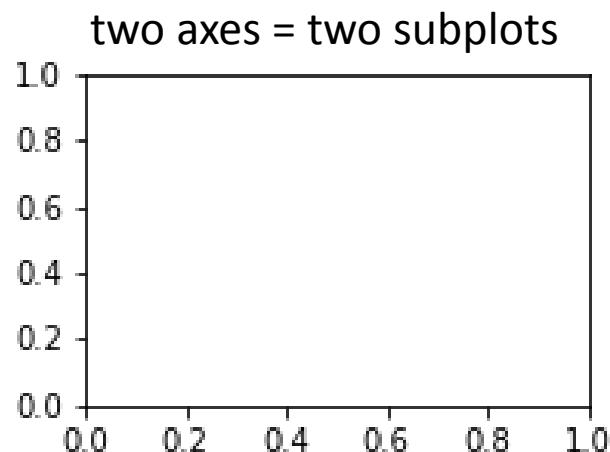
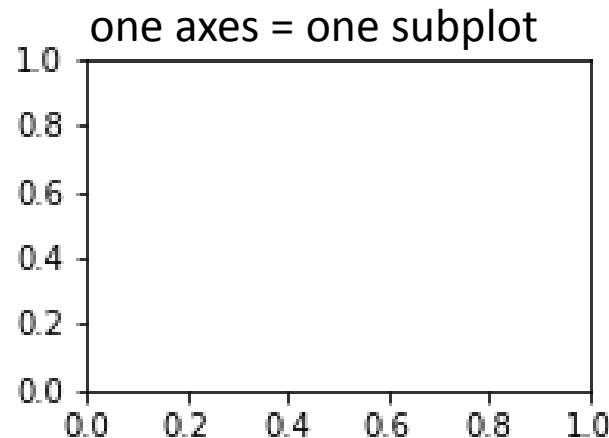
axis, axes, subplots, figure, spines, ...



axis, axes, subplots, figure, spines, ...

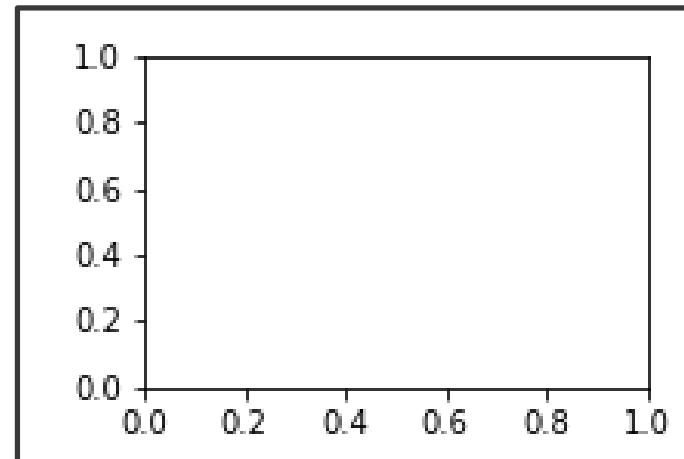


axis, axes, subplots, figure, spines, ...

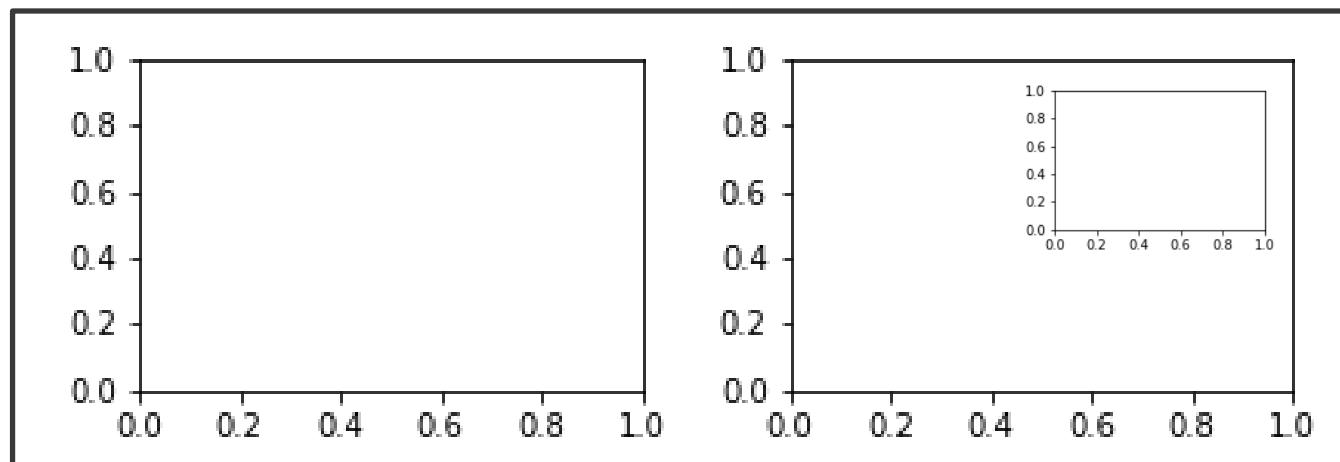


axis, axes, subplots, figure, spines, ...

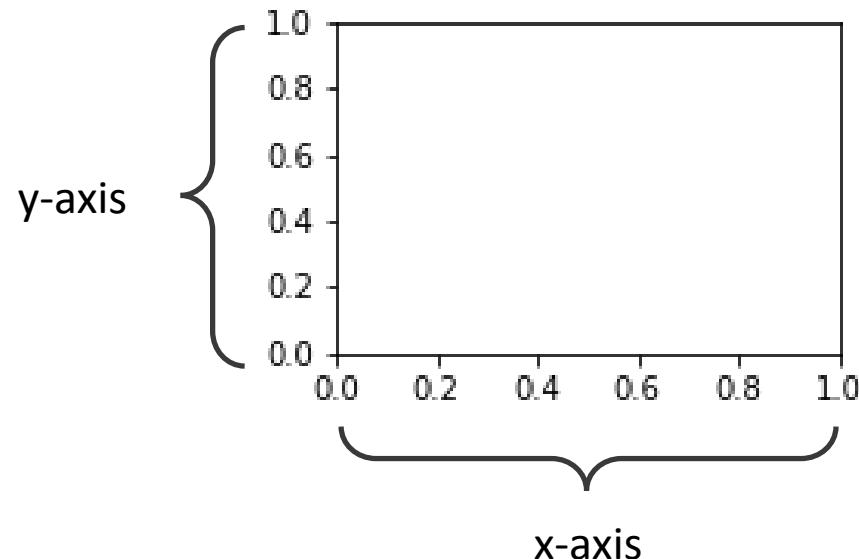
figure



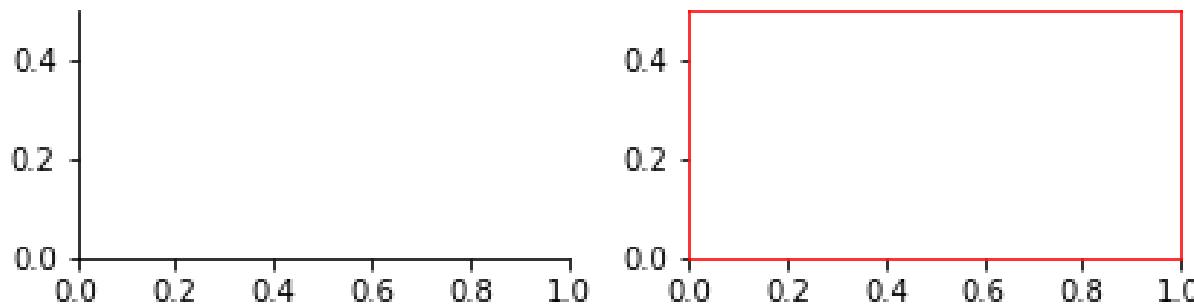
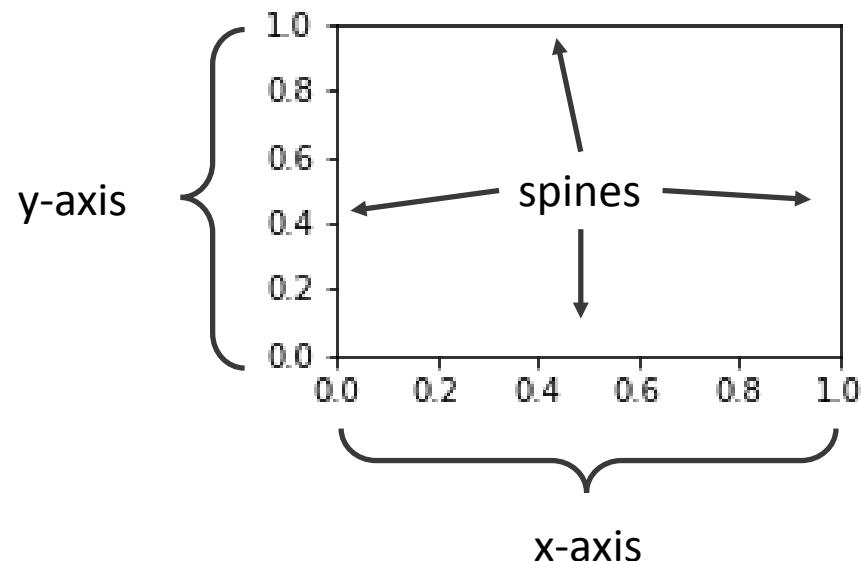
figure



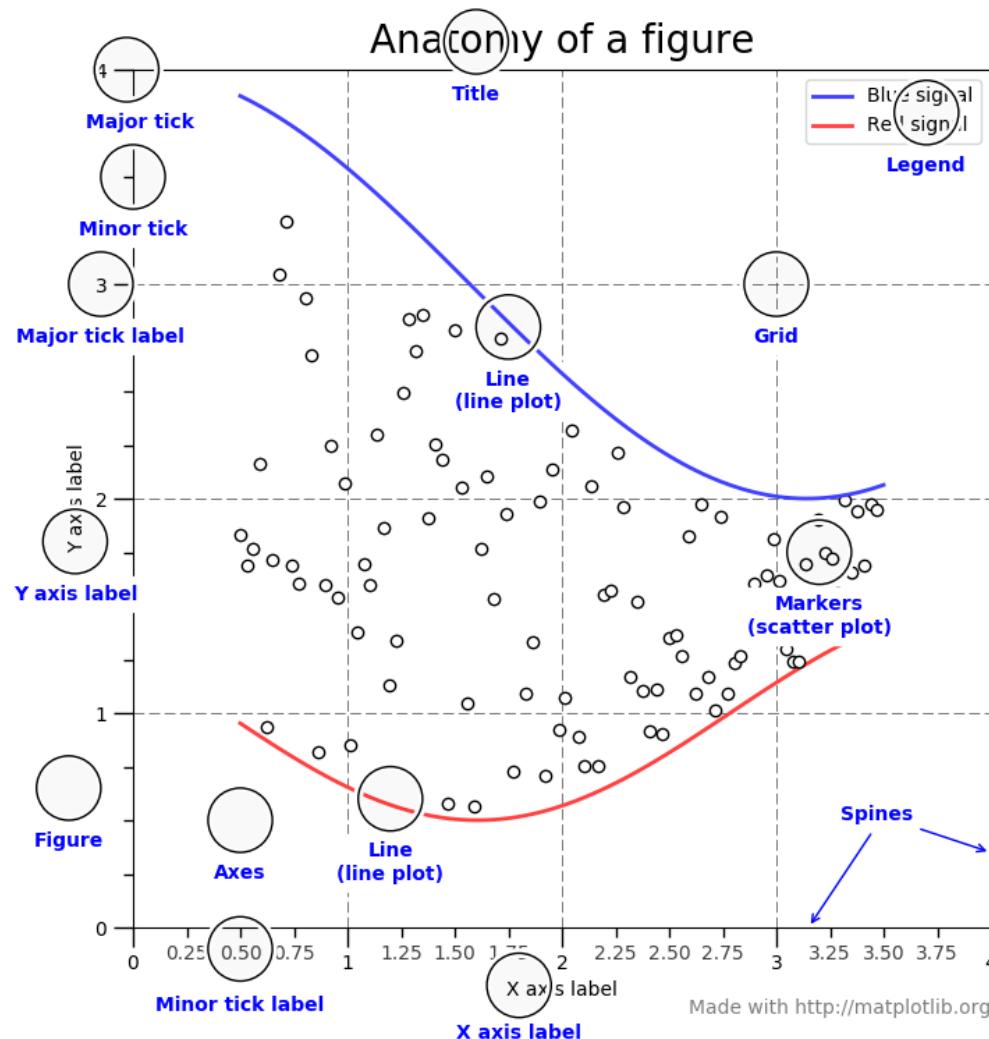
axis, axes, subplots, figure, spines, ...



axis, axes, subplots, figure, spines, ...



Part 1: Anatomy of a Figure



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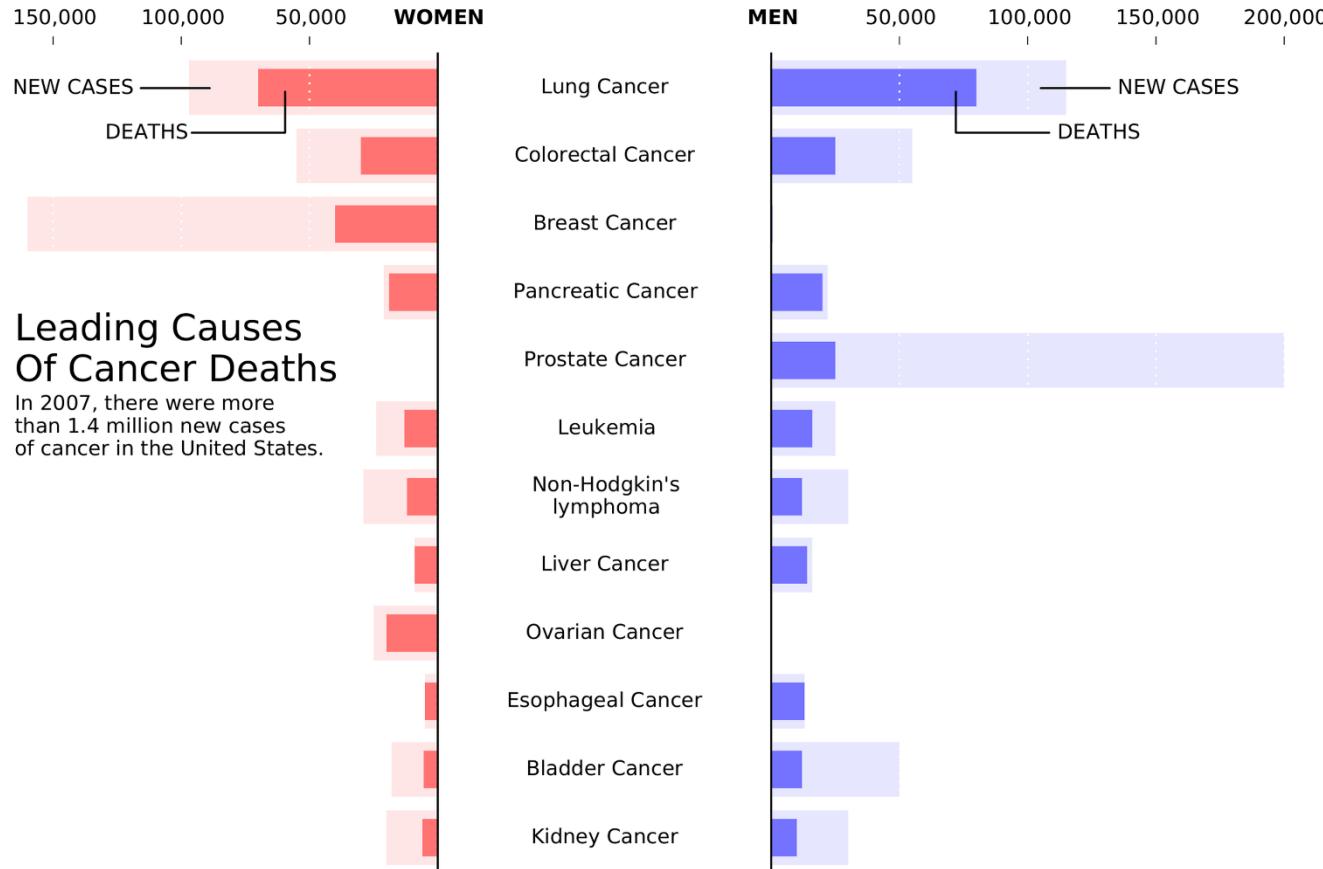
Advanced Plotting

- styles
- helper functions
- seaborn (statistical plots)

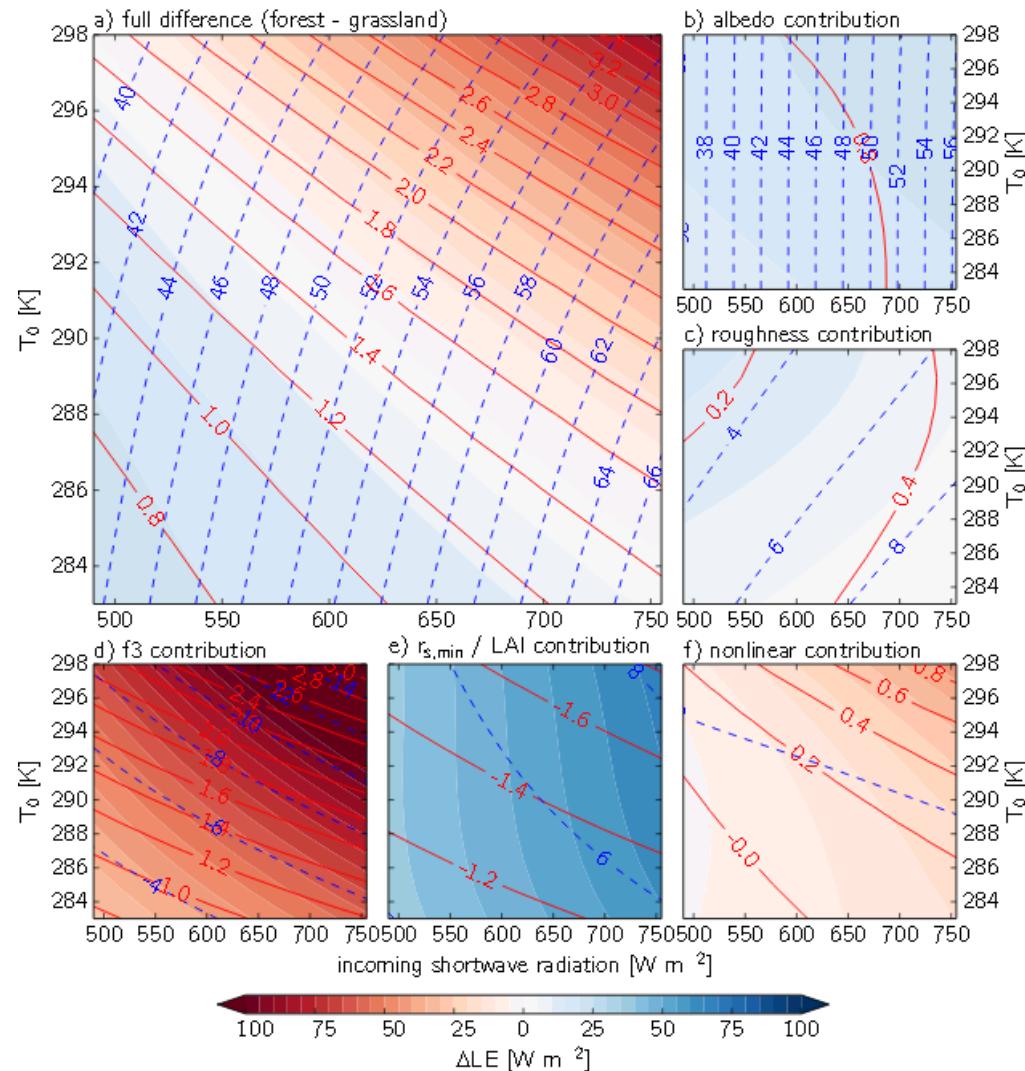
Ten Simple Rules for Better Figures

Rougier et al., 2014 PLOS

Rule 1: Know your audience

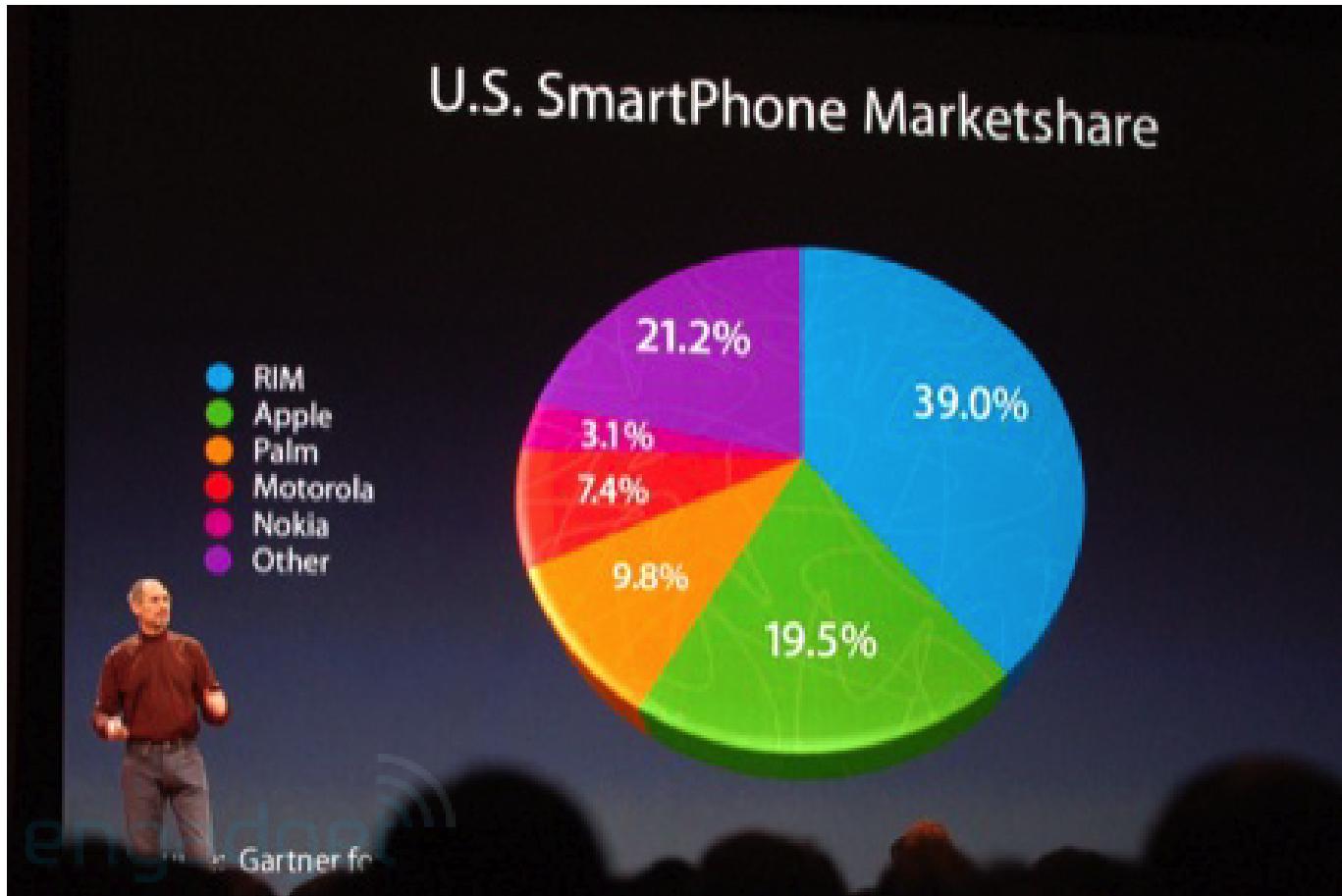


Rule 1: Know your audience

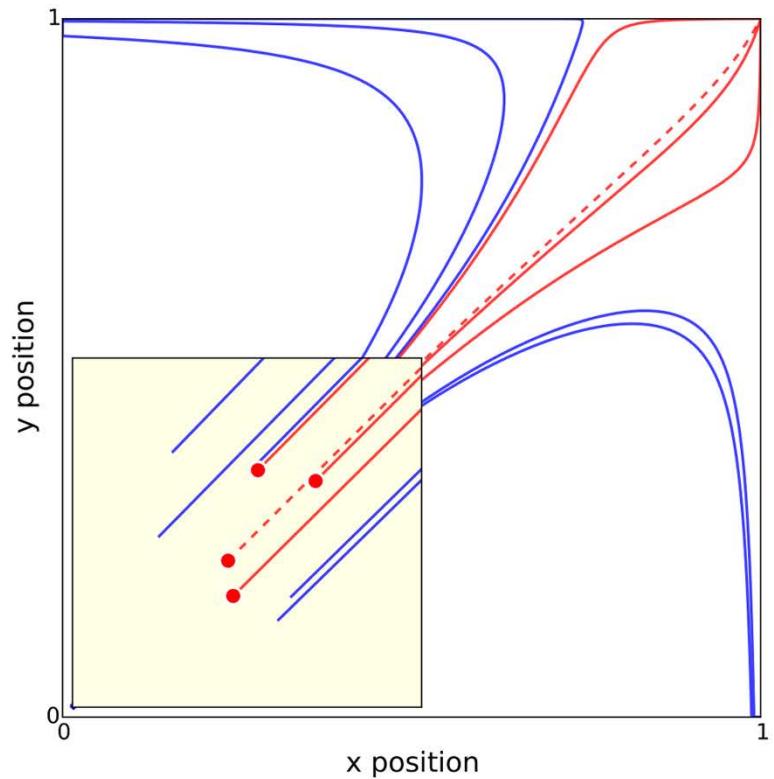
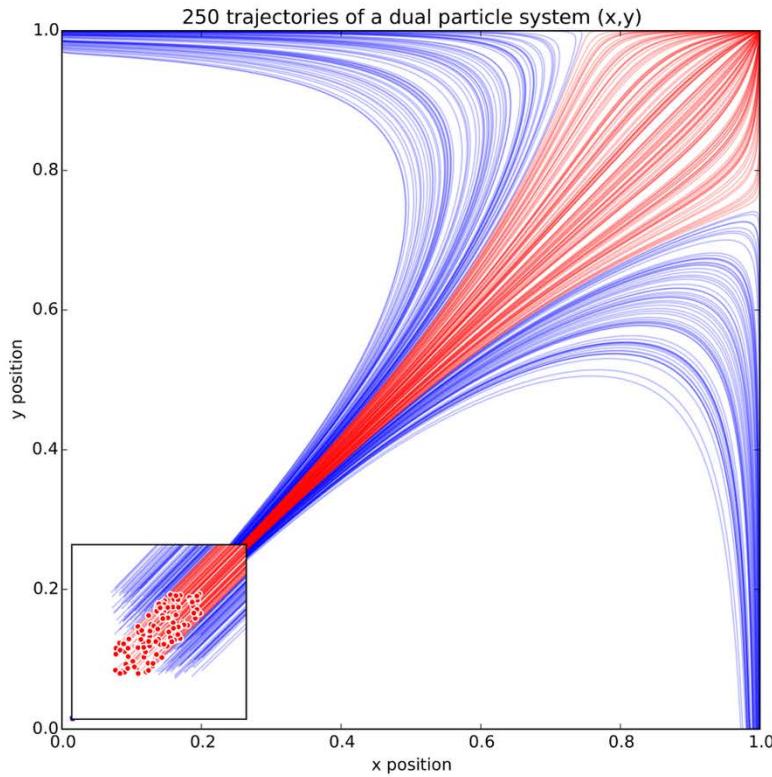


Van Heerwaarden et al., 2014

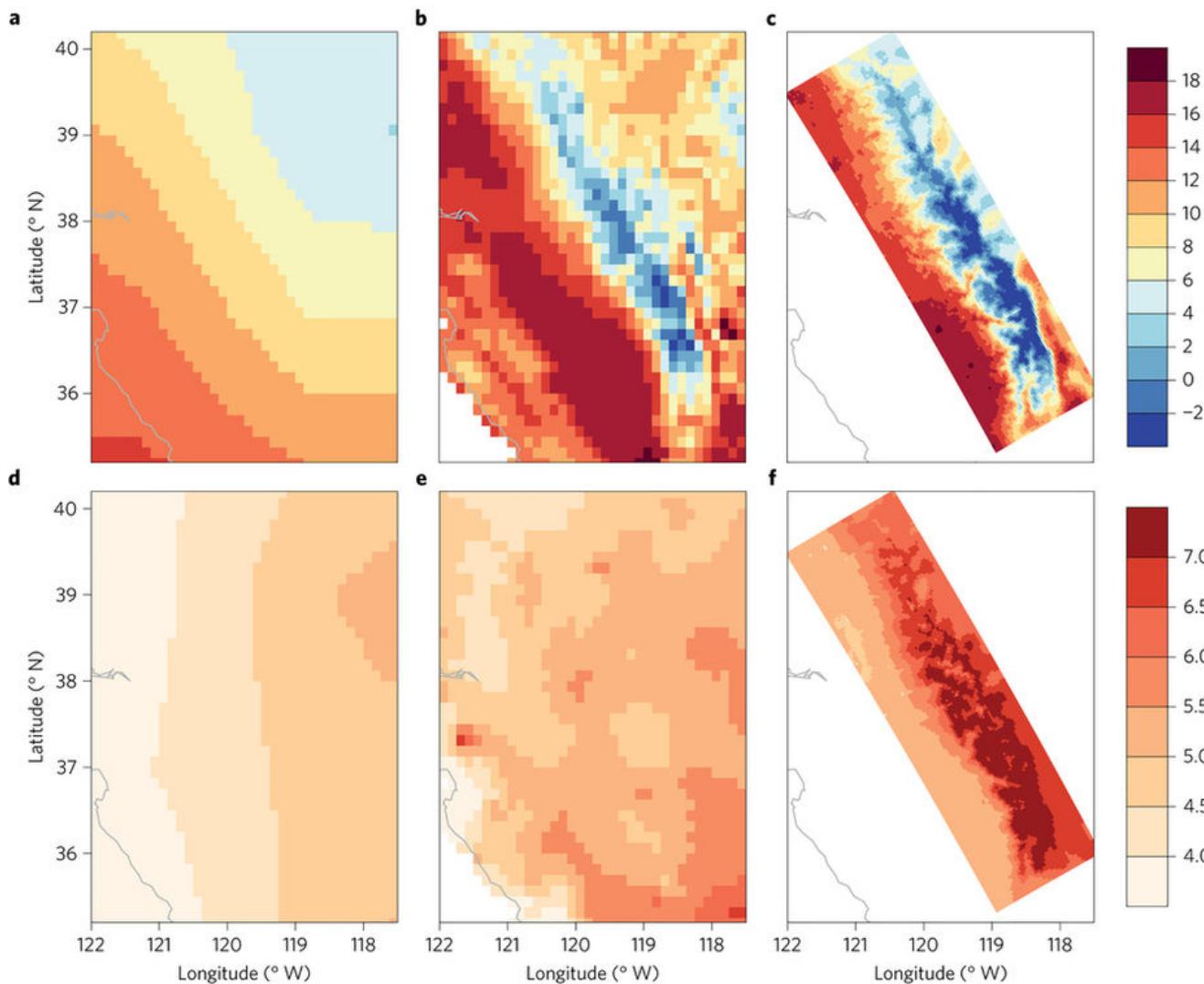
Rule 2: Identify Your Message



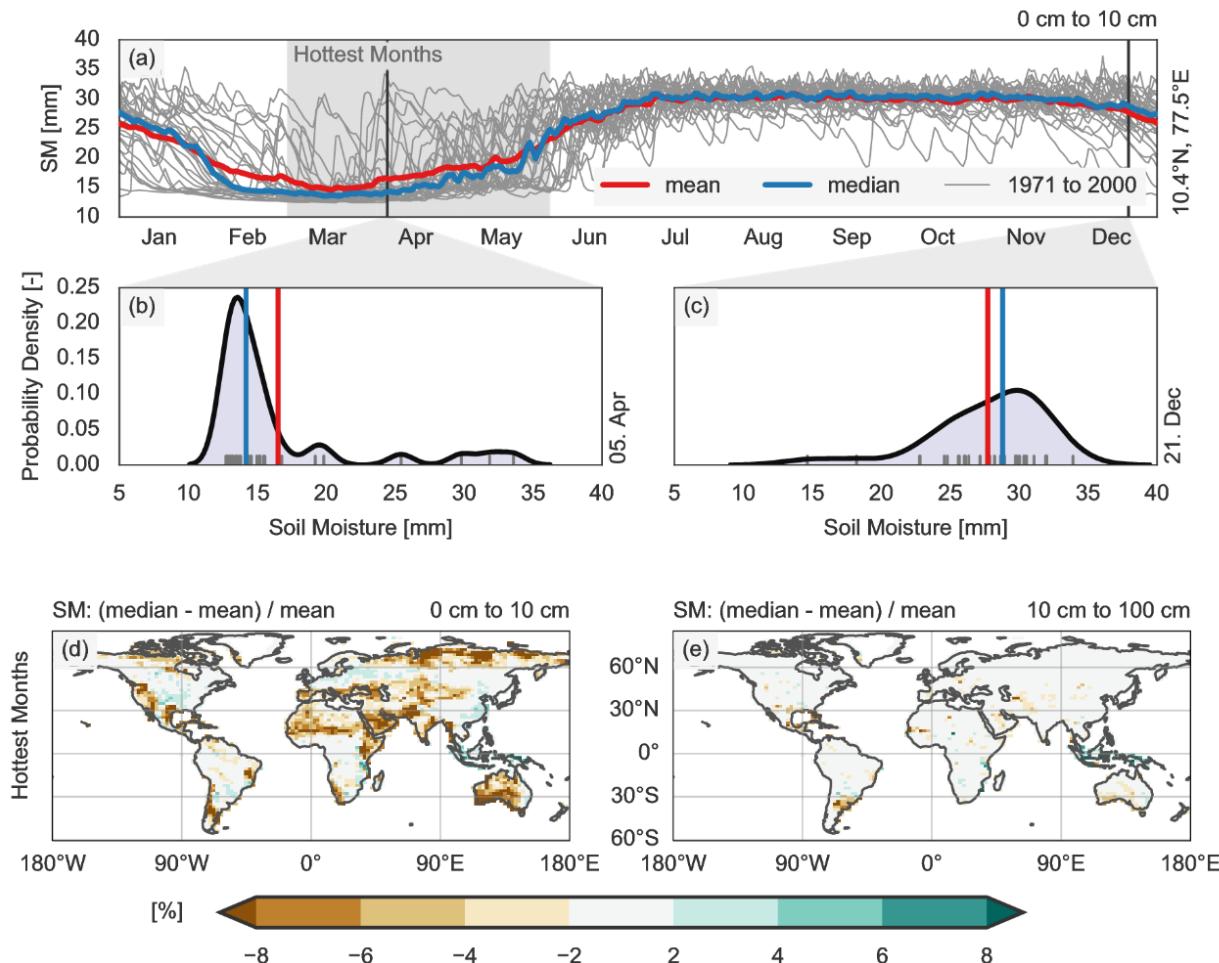
Rule 3: Adapt the Figure to the Support Medium



Rule 4: Labels, Titles, ... Are Not Optional

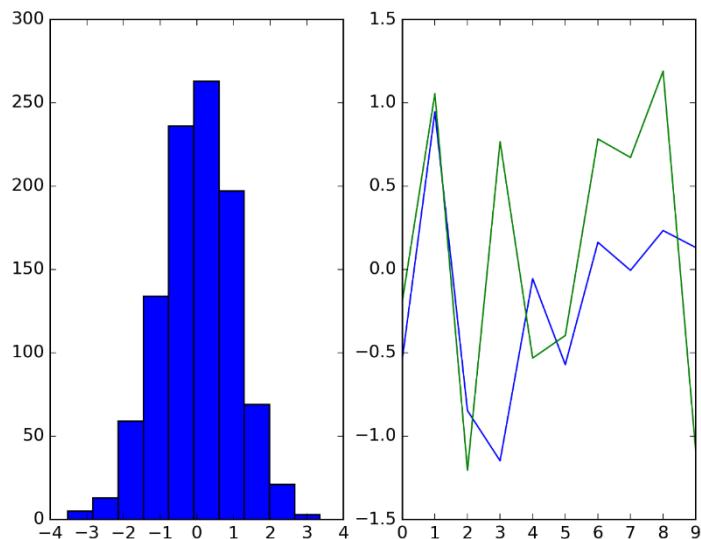


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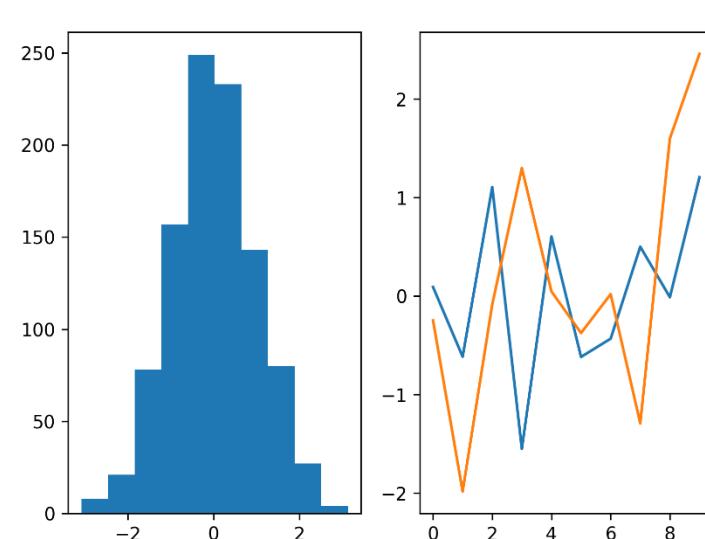


Rule 5: Do Not Trust the Defaults

Version 1.5



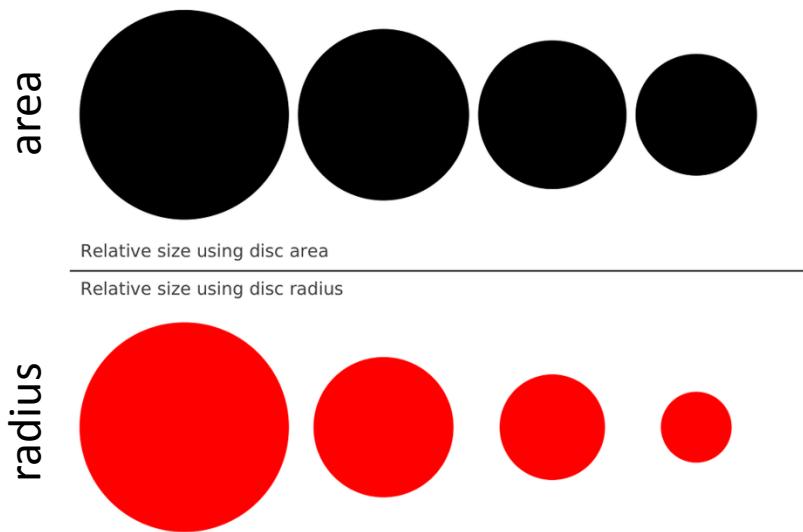
Version 2.1



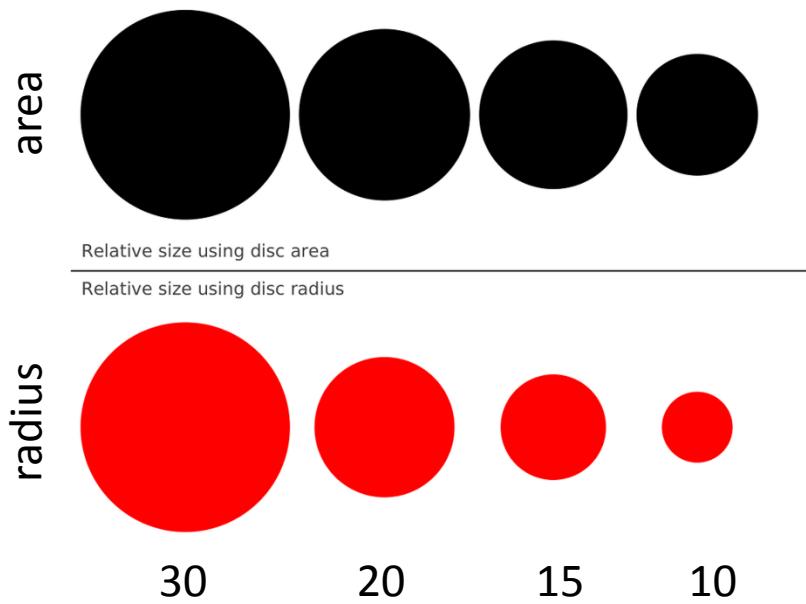
Rule 6: Use Color Effectively

- See later

Rule 7: Do Not Mislead the Reader

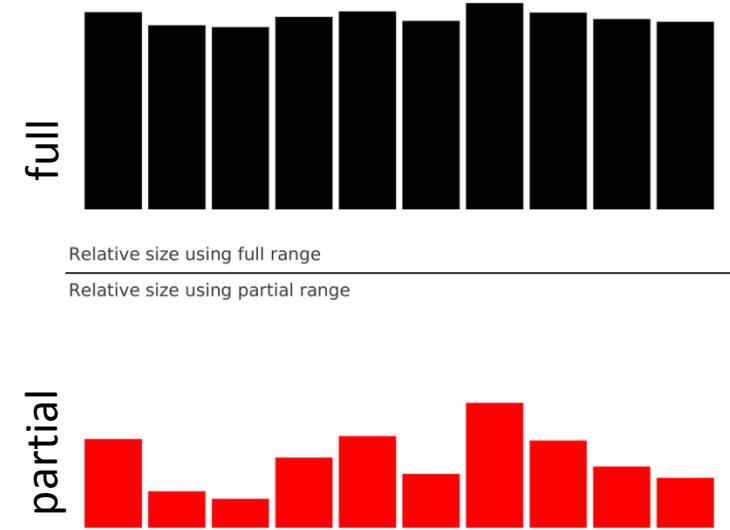
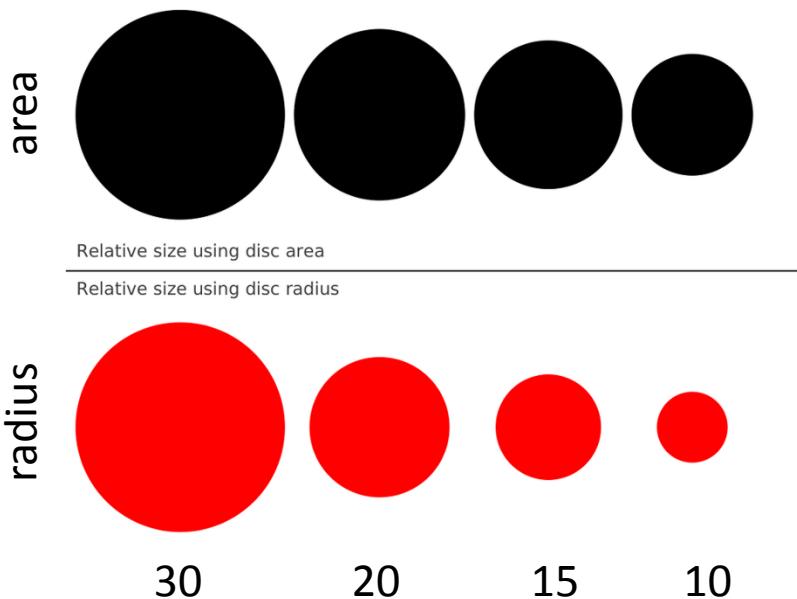


Rule 7: Do Not Mislead the Reader

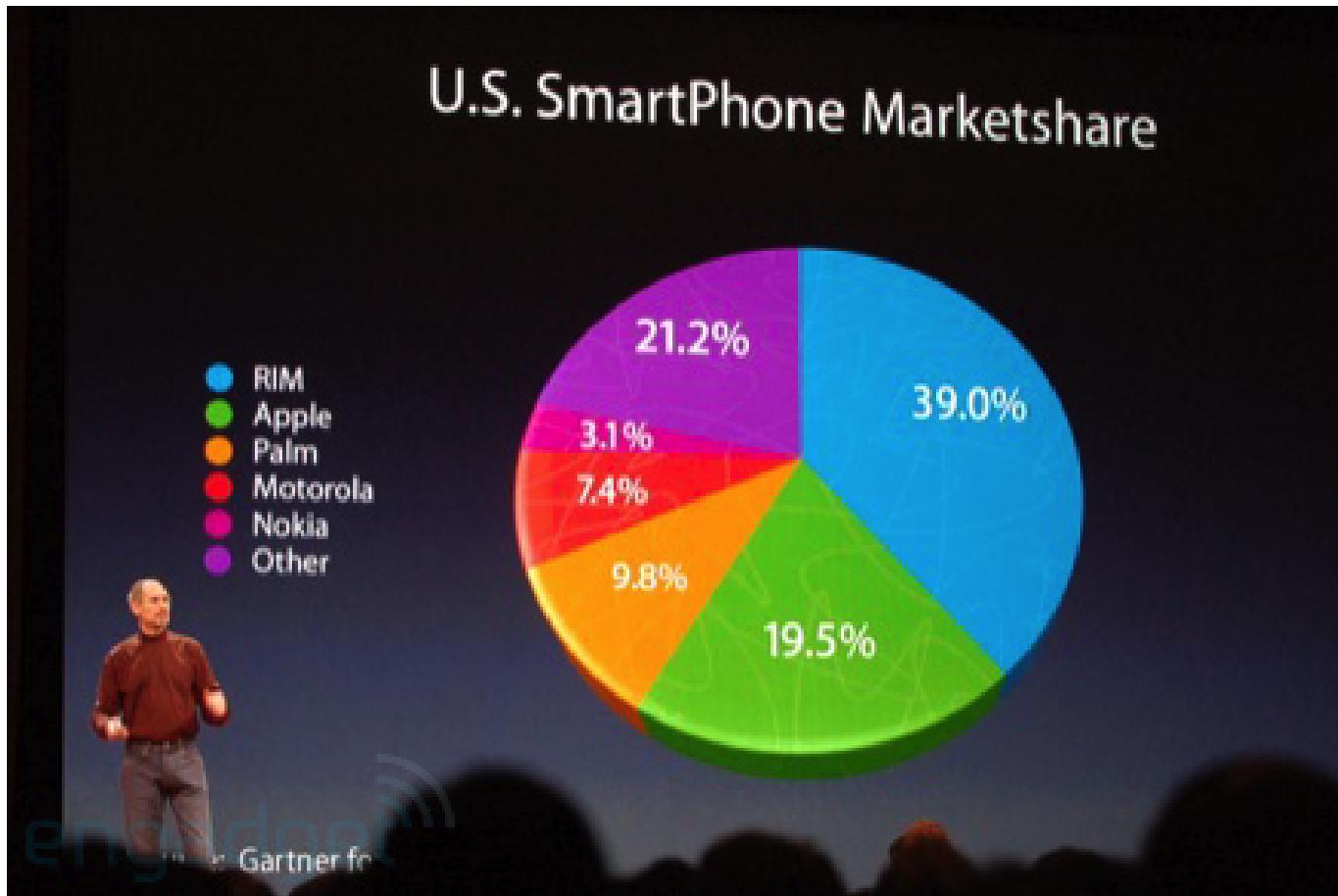


Rougier et al., 2014

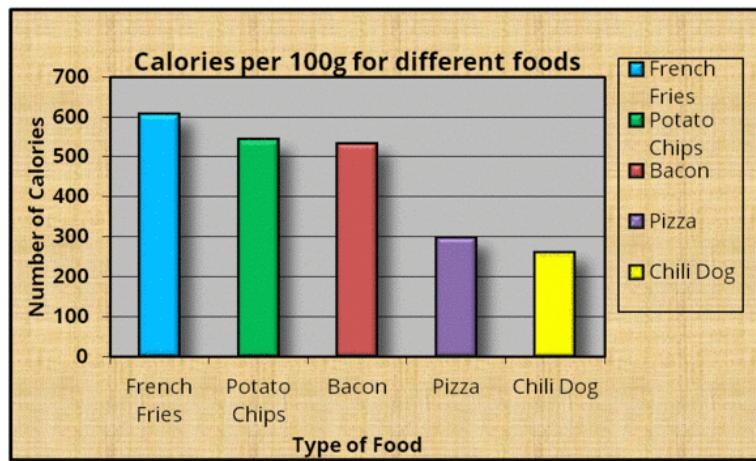
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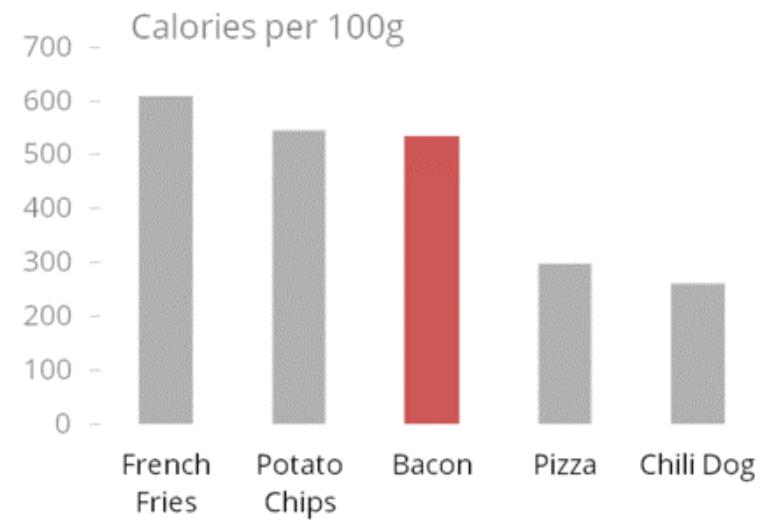
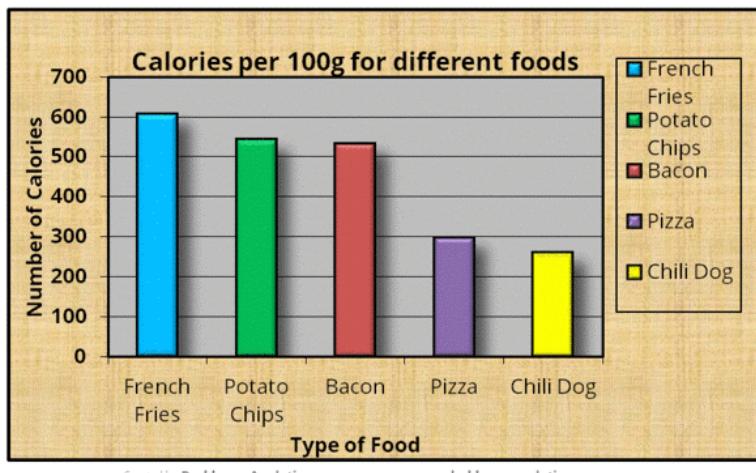
Rule 8: Avoid “Chartjunk”



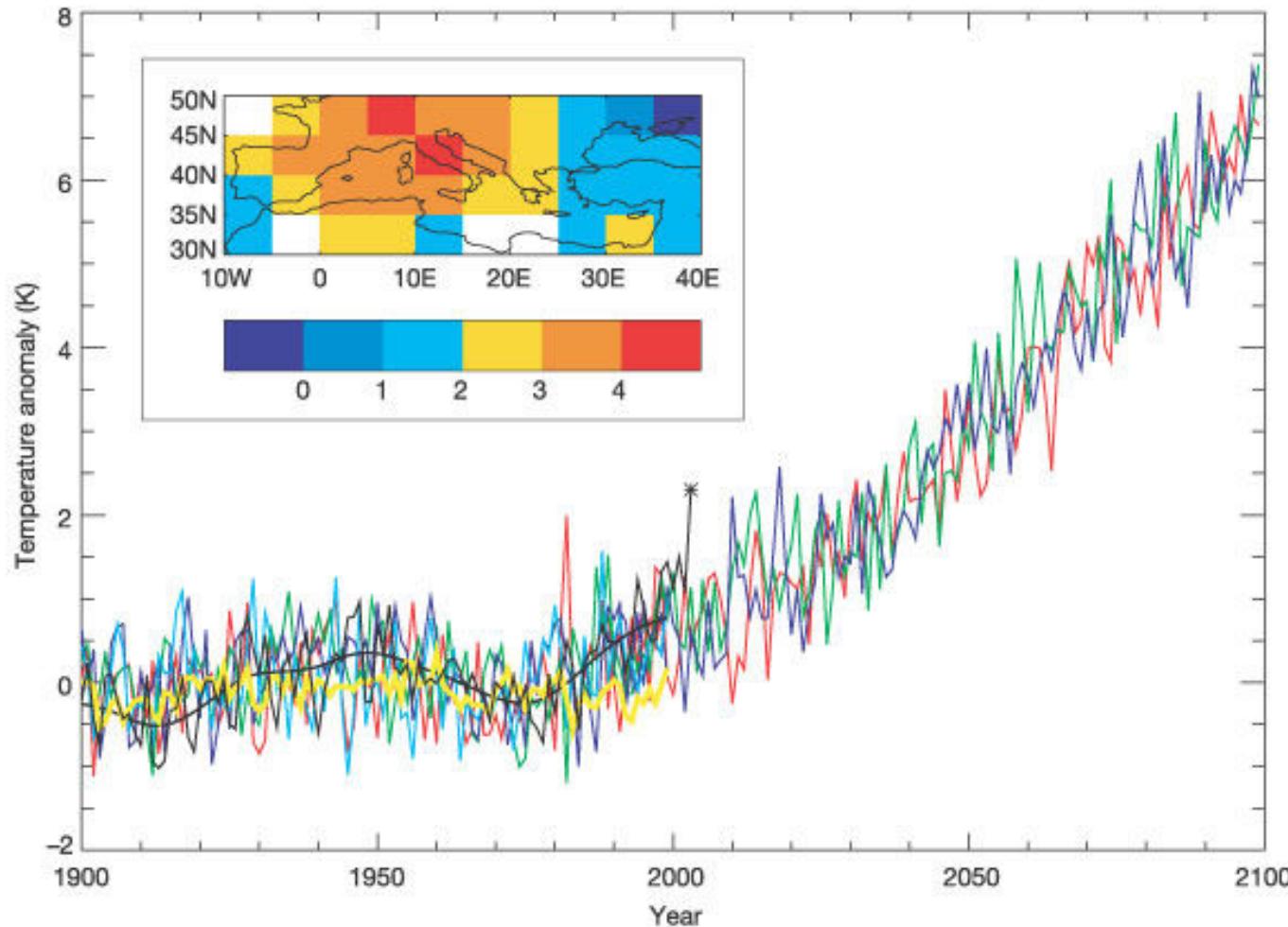
Created by Darkhorse Analytics

www.darkhorseanalytics.com

Rule 8: Avoid “Chartjunk”

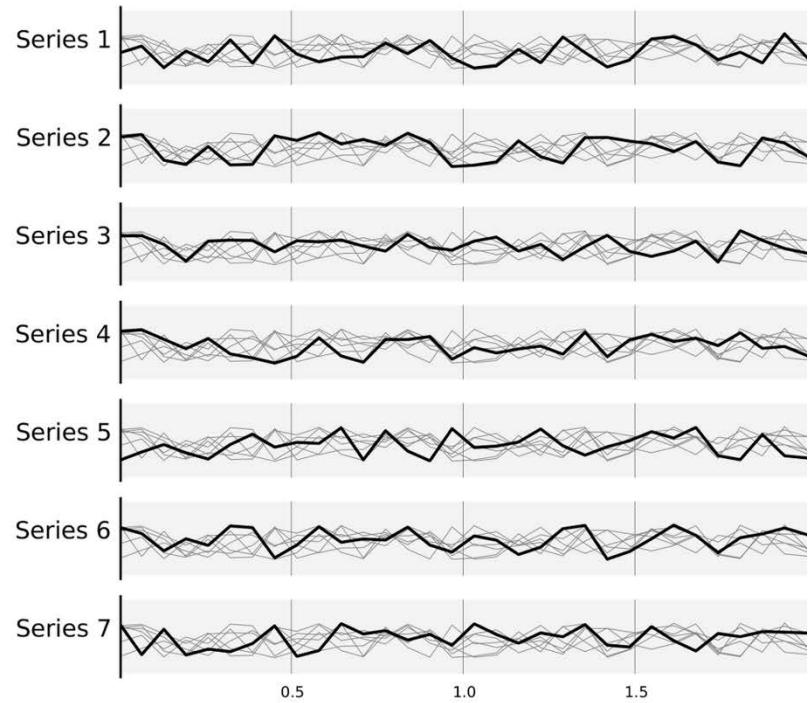
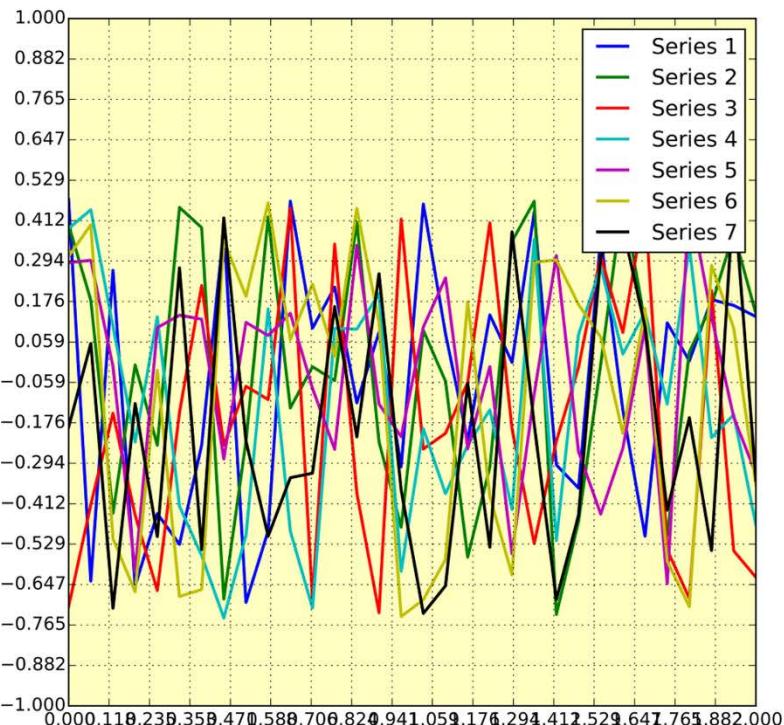


Rule 8: Avoid “Chartjunk”



Stott et al., 2004, Nature

Rule 8: Avoid “Chartjunk”

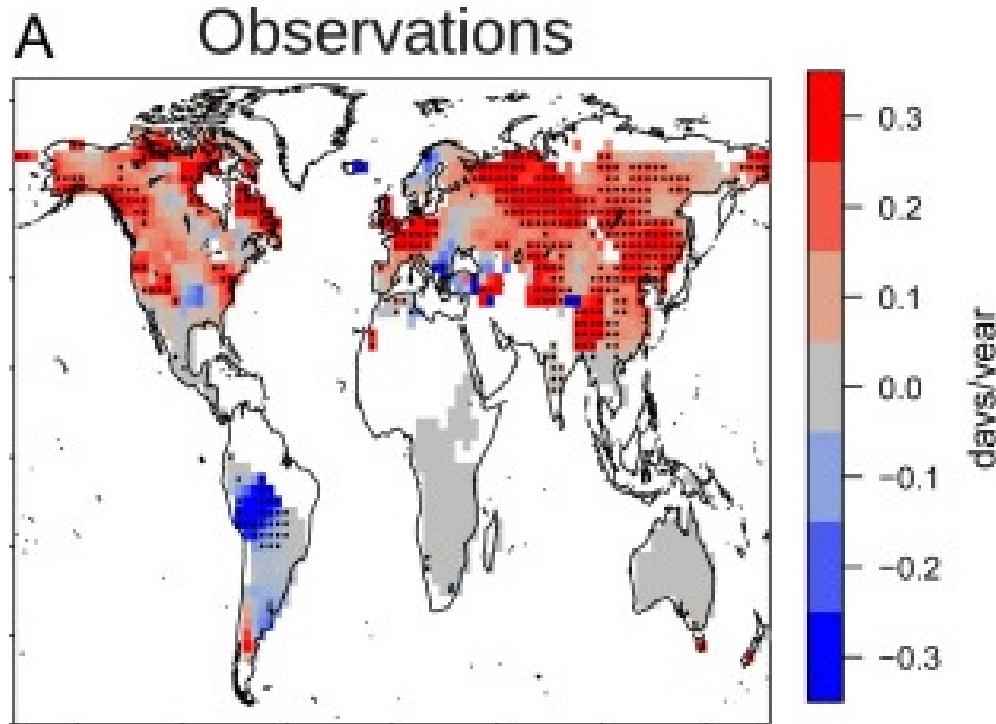


Rougier et al., 2014

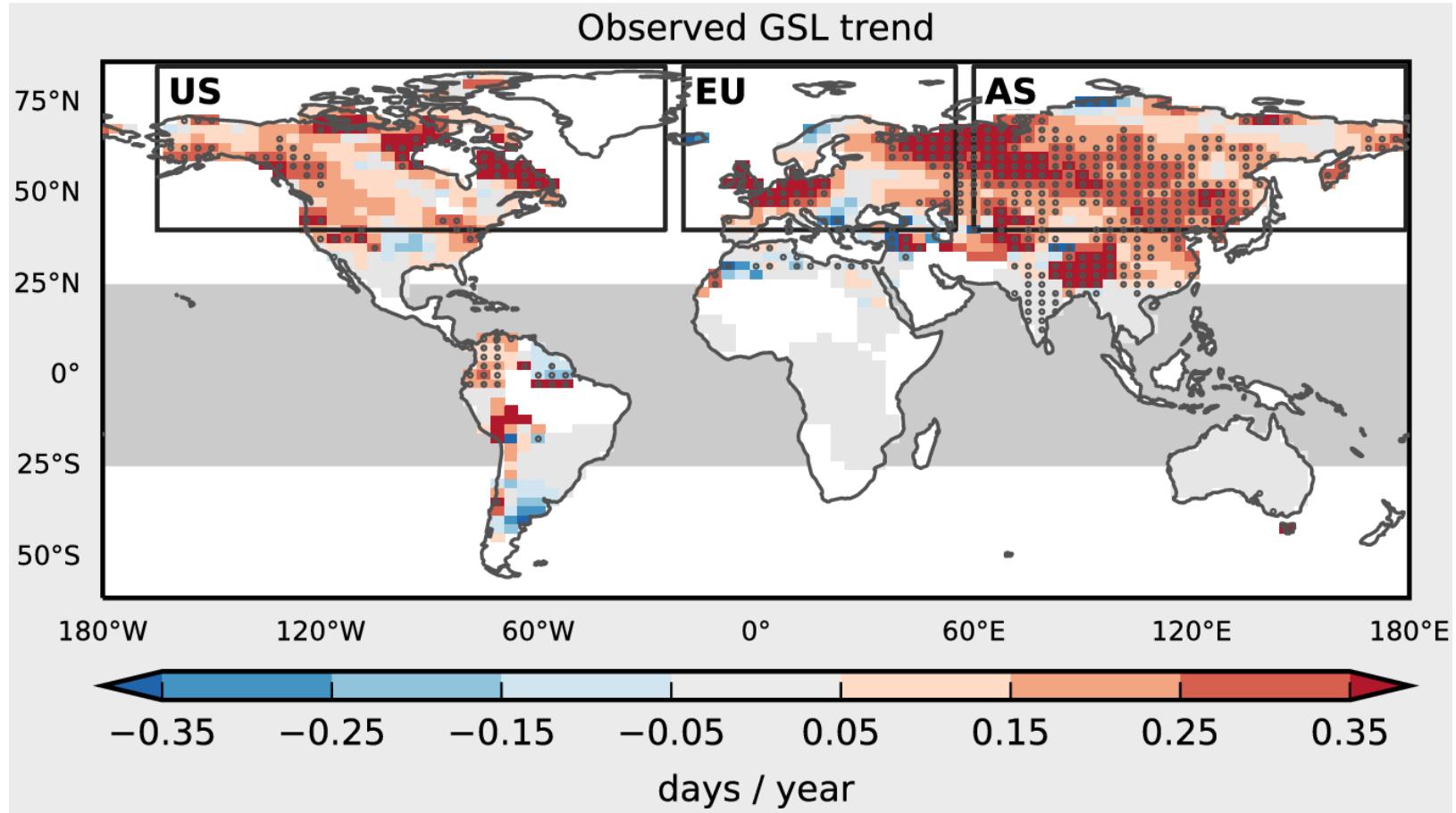
Rule 9: Message Trumps Beauty

Rule 10: Get the Right Tool

Can you do better?



Can you do better?



SCHEDULE – PYTHON VISUALISATION COURSE

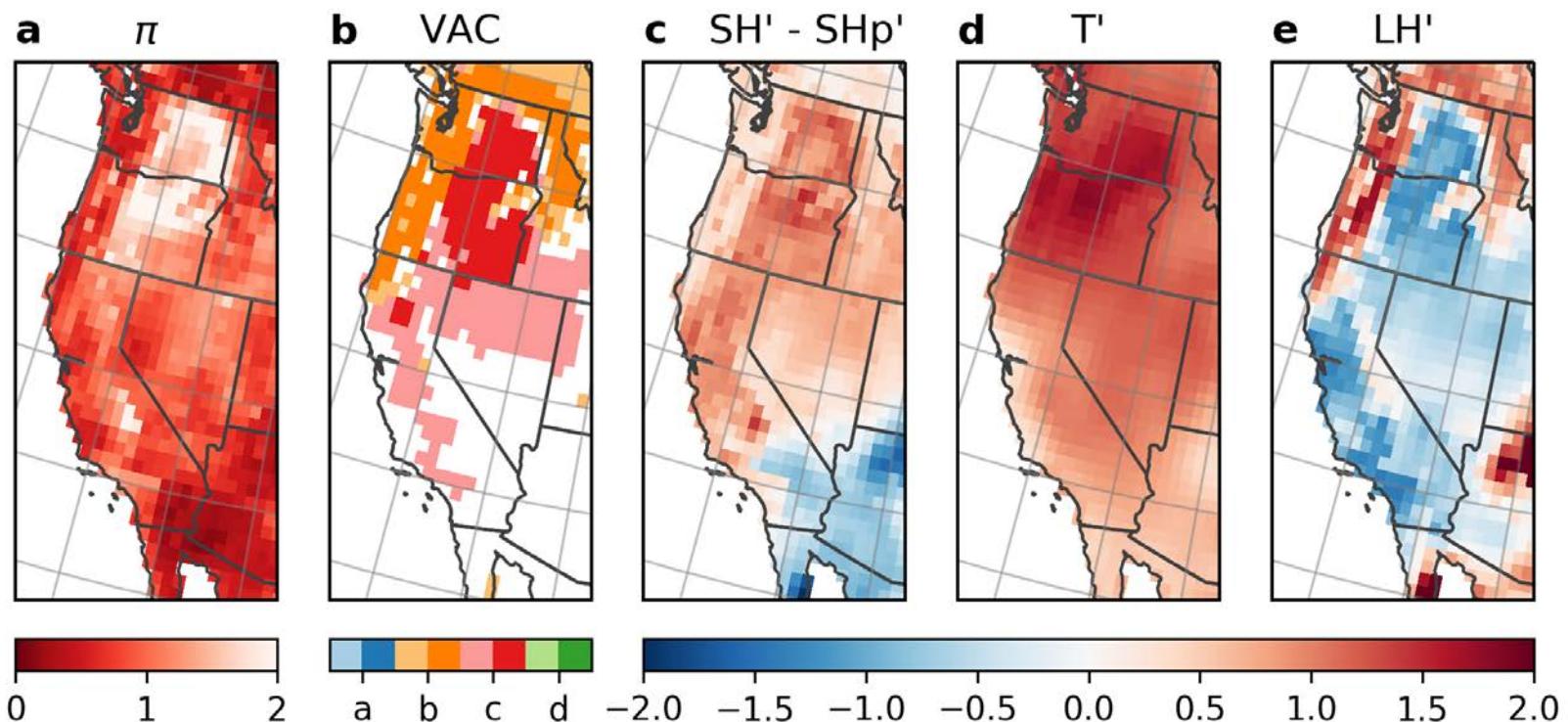
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Georeferenced data with cartopy



Georeferenced data with cartopy

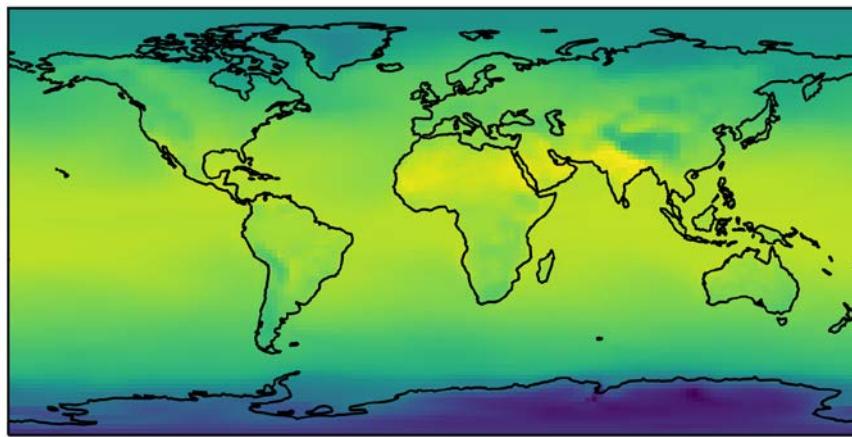
- cartopy
- basemap

Georeferenced data with cartopy

- cartopy
- ~~basemap~~

Georeferenced data with cartopy

```
ax = plt.axes(projection=ccrs.PlateCarree())
ax.coastlines()
h = ax.pcolormesh(LON, LAT, cesm.temp - 273.15,
                   transform=ccrs.PlateCarree())
ax.set_global()
```



Georeferenced data with cartopy

```
ax = plt.axes(projection=ccrs.PlateCarree())

ax.coastlines()

h = ax.pcolormesh(LON, LAT, cesm.temp - 273.15,
                    transform=ccrs.PlateCarree())
```

- projection: type of map
 - can change
- transform: coordinate system of data
 - Hardly ever changes
 - PlateCarree -> coordinates in lon/ lat

Plotting georeferenced data

- scatter points data on maps
- contour and pcolormesh
- colorbars
- stippling
- multiple map plots
- trajectories
- rotated pole data

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13:30 – 15:00	Part 3: Plotting georeferenced data
15:00 – 15:30	Coffee Break
15:30 – 17:00	Part 4: User contributions

Colormaps

```
([[ 0.40784314,  0.40784314,  0.40784314,  ...,  0.42745098,
  0.42745098,  0.42745098],
 [ 0.41176471,  0.41176471,  0.41176471,  ...,  0.42745098,
  0.42745098,  0.42745098],
 [ 0.41960785,  0.41568628,  0.41568628,  ...,  0.43137255,
  0.43137255],
 ...,
 [ 0.43921569,  0.43529412,  0.43137255,  ...,  0.45490196,
  0.4509804 ,  0.4509804 ],
 [ 0.44313726,  0.44313726,  0.43921569,  ...,  0.4509804 ,
  0.44705883,  0.44705883],
 [ 0.44313726,  0.4509804 ,  0.4509804 ,  ...,  0.44705883,
  0.44705883,  0.44313726]], dtype=float32)
```

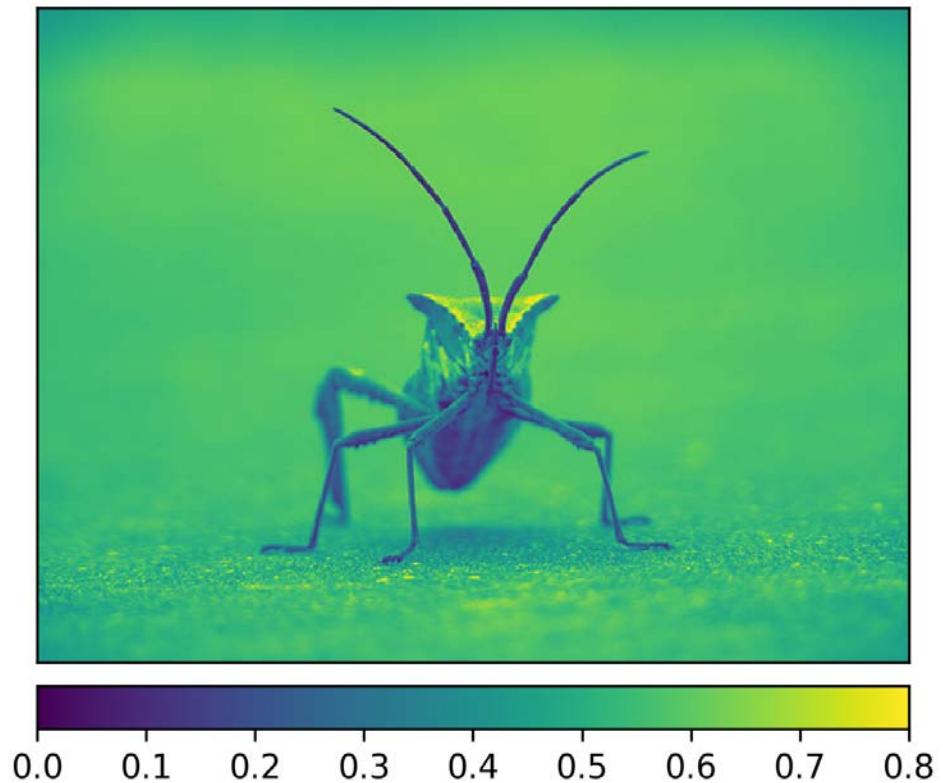
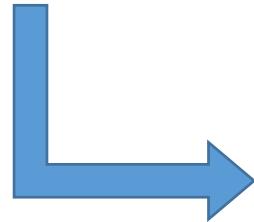
Colormaps

```
([[ 0.40784314,  0.40784314,  0.40784314,  ...,  0.42745098,
  0.42745098,  0.42745098],
 [ 0.41176471,  0.41176471,  0.41176471,  ...,  0.42745098,
  0.42745098,  0.42745098],
 [ 0.41960785,  0.41568628,  0.41568628,  ...,  0.43137255,
  0.43137255],
 ...,
 [ 0.43921569,  0.43529412,  0.43137255,  ...,  0.45490196,
  0.4509804 ,  0.4509804 ],
 [ 0.44313726,  0.44313726,  0.43921569,  ...,  0.4509804 ,
  0.44705883,  0.44705883],
 [ 0.44313726,  0.4509804 ,  0.4509804 ,  ...,  0.44705883,
  0.44705883,  0.44313726]], dtype=float32)
```



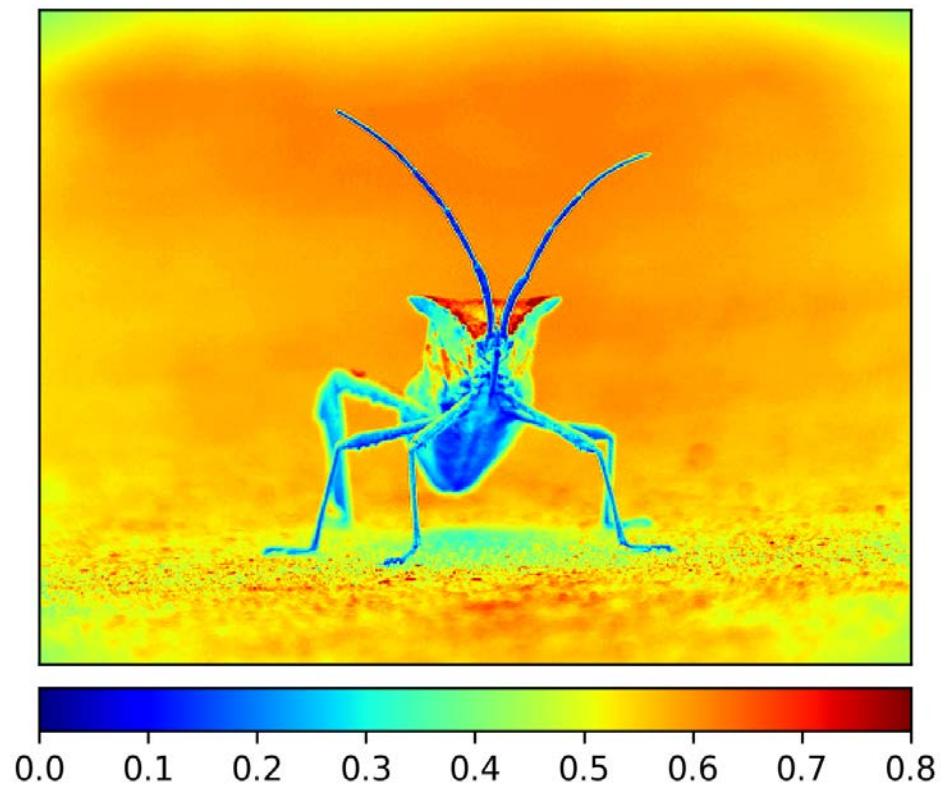
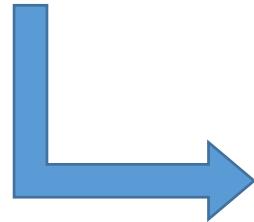
Colormaps

```
[[[ 0.40784314,  0.40784314,  0.40784314,  ...,  0.42745098,
    0.42745098,  0.42745098],
   [ 0.41176471,  0.41176471,  0.41176471,  ...,  0.42745098,
    0.42745098,  0.42745098],
   [ 0.41960785,  0.41568628,  0.41568628,  ...,  0.43137255,
    0.43137255,  0.43137255],
   ...,
   [ 0.43921569,  0.43529412,  0.43137255,  ...,  0.4509804 ,
    0.4509804 ,  0.4509804 ],
   [ 0.44313726,  0.44313726,  0.43921569,  ...,  0.44705883,
    0.44705883,  0.44705883],
   [ 0.44313726,  0.4509804 ,  0.4509804 ,  ...,  0.44705883,
    0.44313726]], dtype=float32)
```

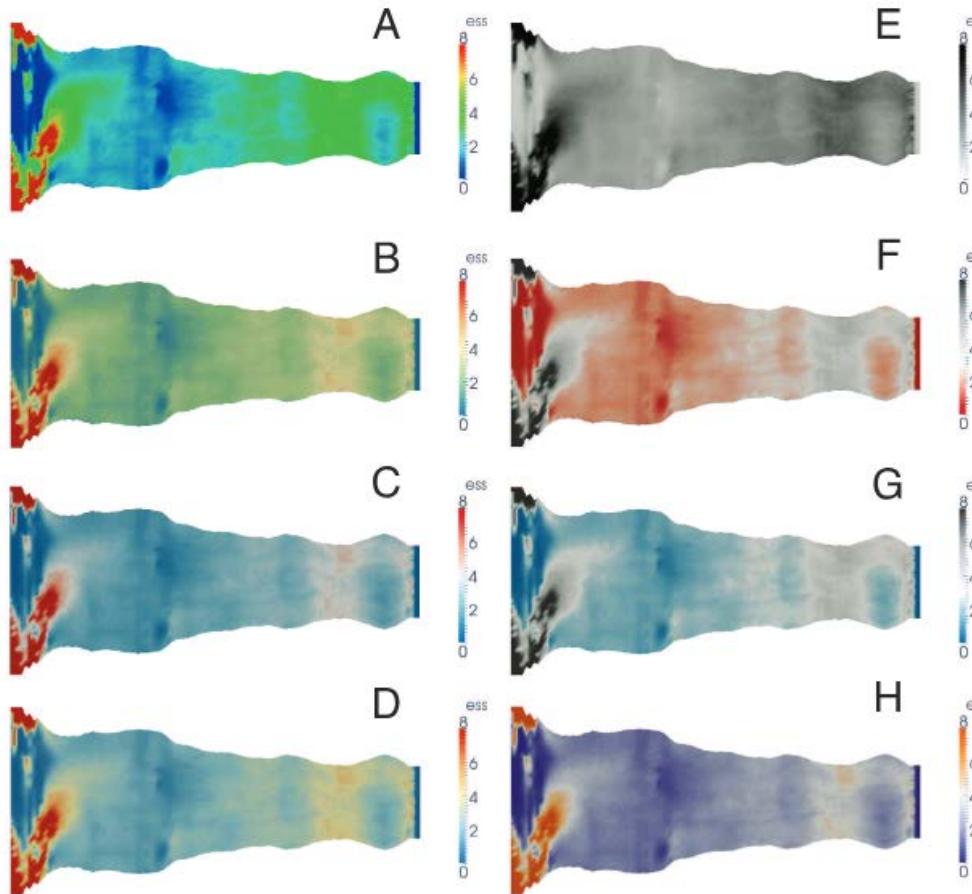


Colormaps: choice matters

```
[[[ 0.40784314,  0.40784314,  0.40784314,  ...,  0.42745098,
    0.42745098,  0.42745098],
   [ 0.41176471,  0.41176471,  0.41176471,  ...,  0.42745098,
    0.42745098,  0.42745098],
   [ 0.41960785,  0.41568628,  0.41568628,  ...,  0.43137255,
    0.43137255,  0.43137255],
   ...,
   [ 0.43921569,  0.43529412,  0.43137255,  ...,  0.4509804 ,
    0.4509804 ,  0.4509804 ],
   [ 0.44313726,  0.44313726,  0.43921569,  ...,  0.44705883,
    0.44705883,  0.44705883],
   [ 0.44313726,  0.4509804 ,  0.4509804 ,  ...,  0.44705883,
    0.44313726]], dtype=float32)
```



Colormaps: choice matters



Borkin et al., 2011

Colormaps: choice matters

Jet

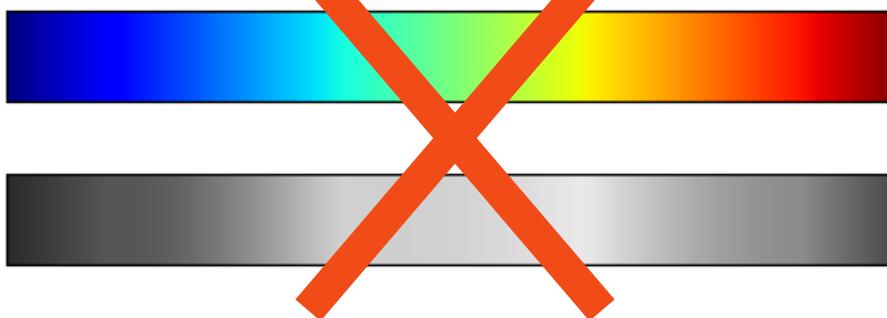


Viridis

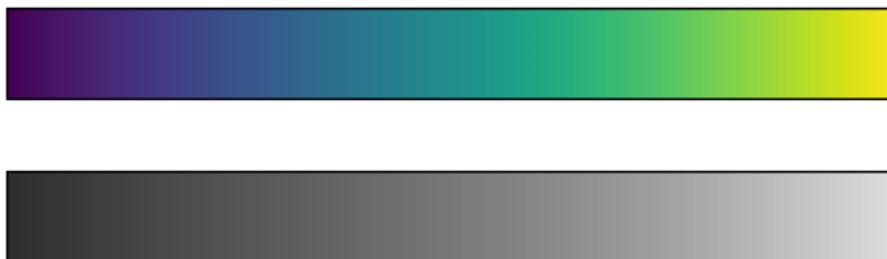


Colormaps: choice matters

Jet

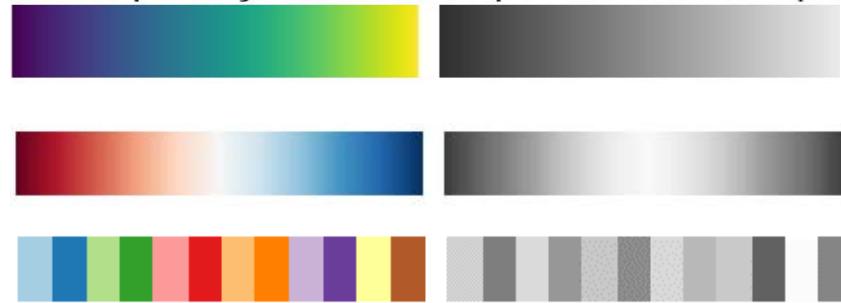


Viridis



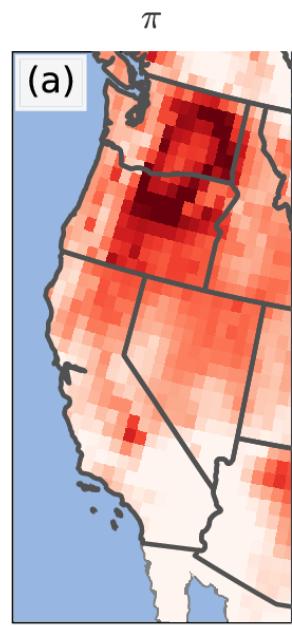
Colormaps: type of data

- Sequential
- Diverging
- Qualitative

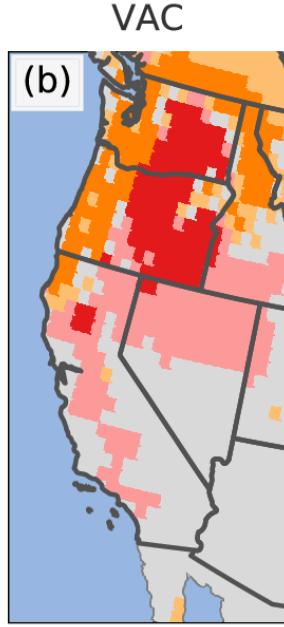


Colormaps: type of data

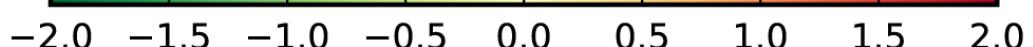
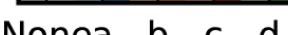
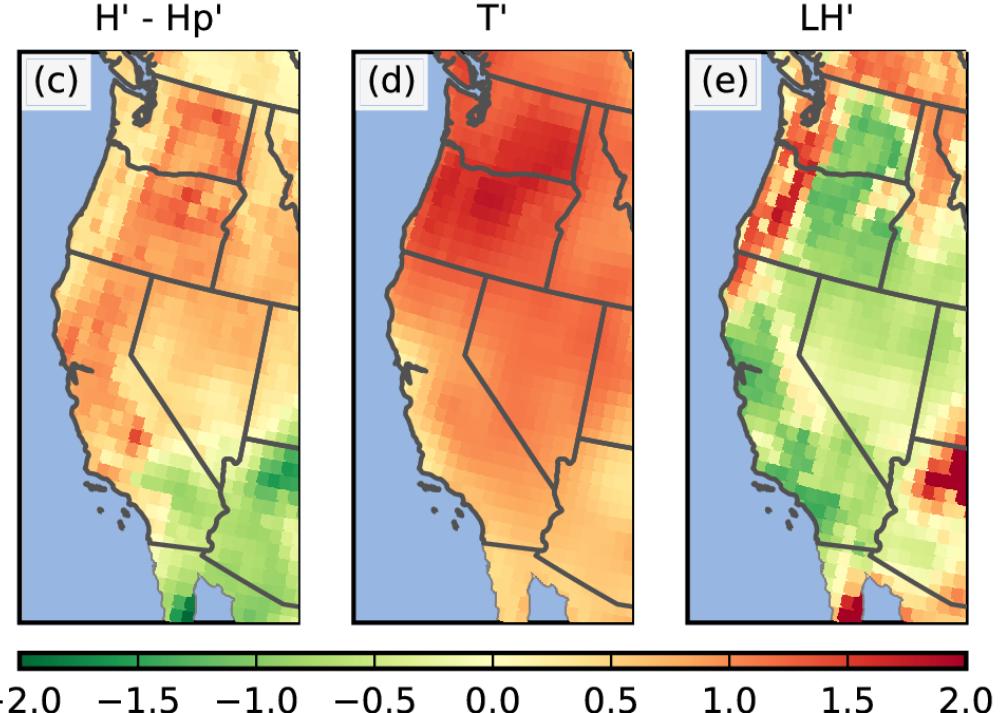
Sequential



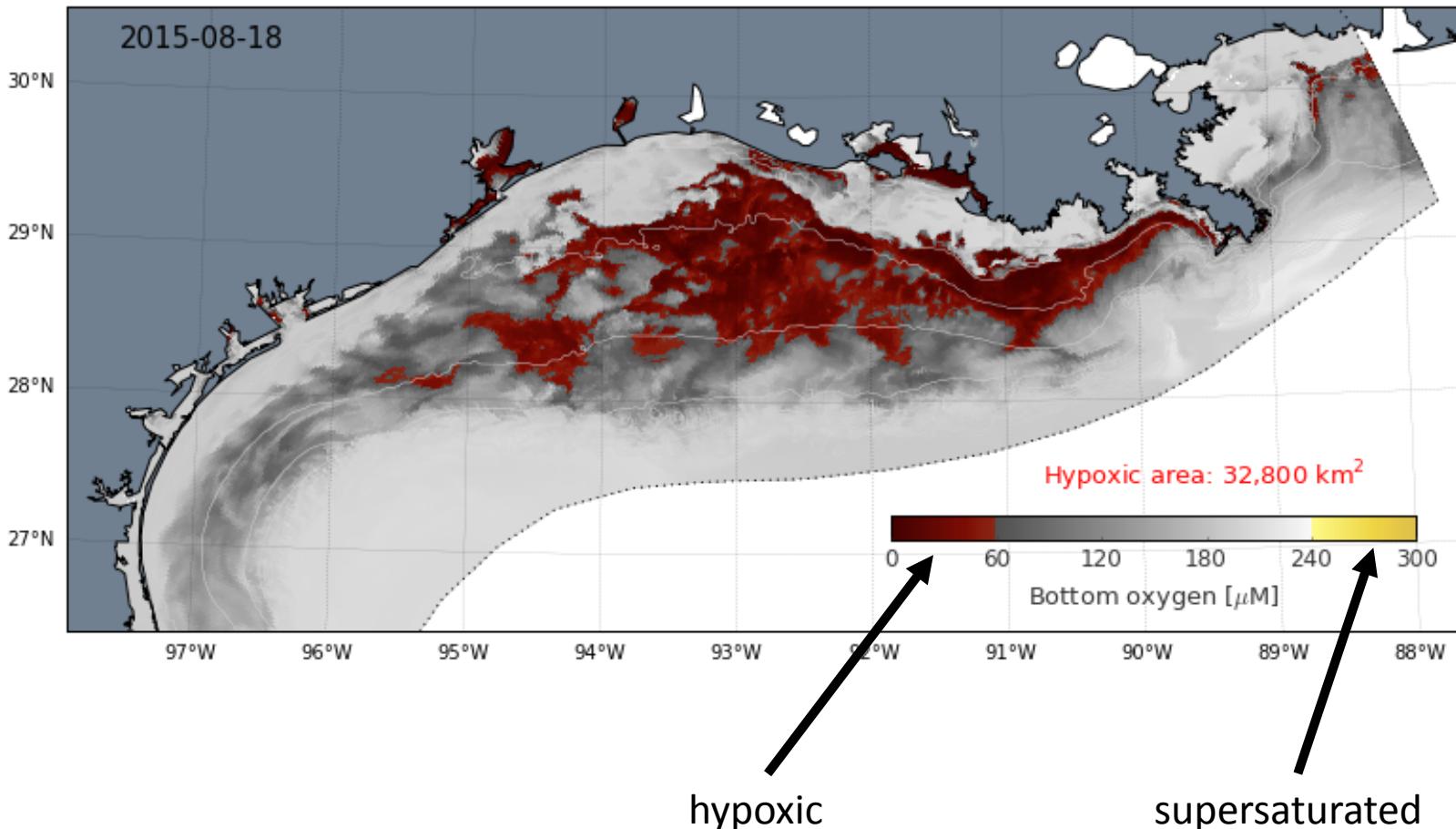
Qualitative



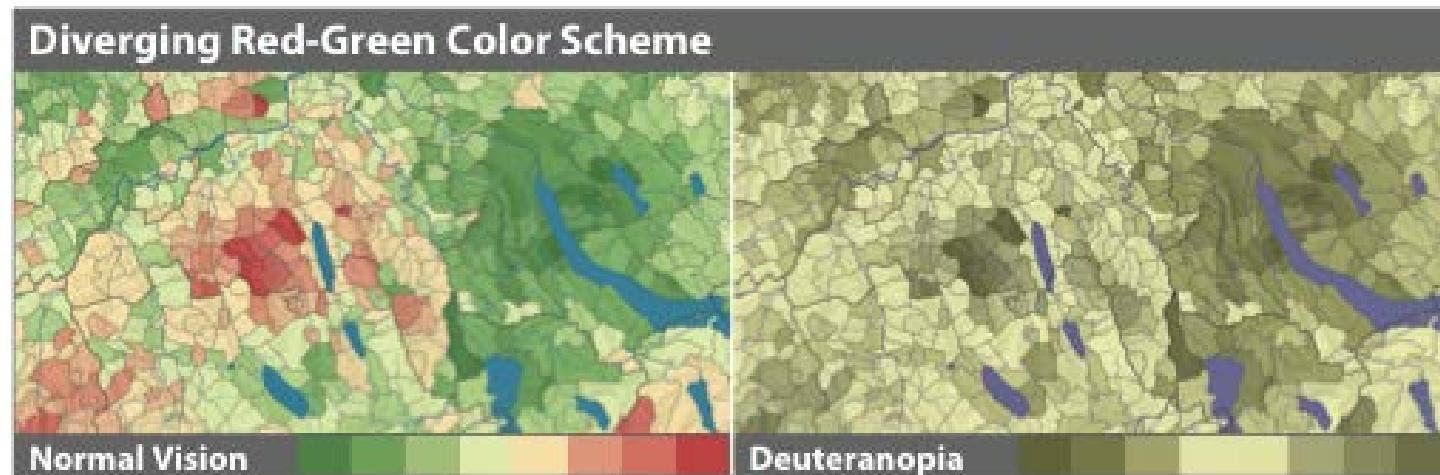
Diverging



Colormaps: non-diverging breakpoints

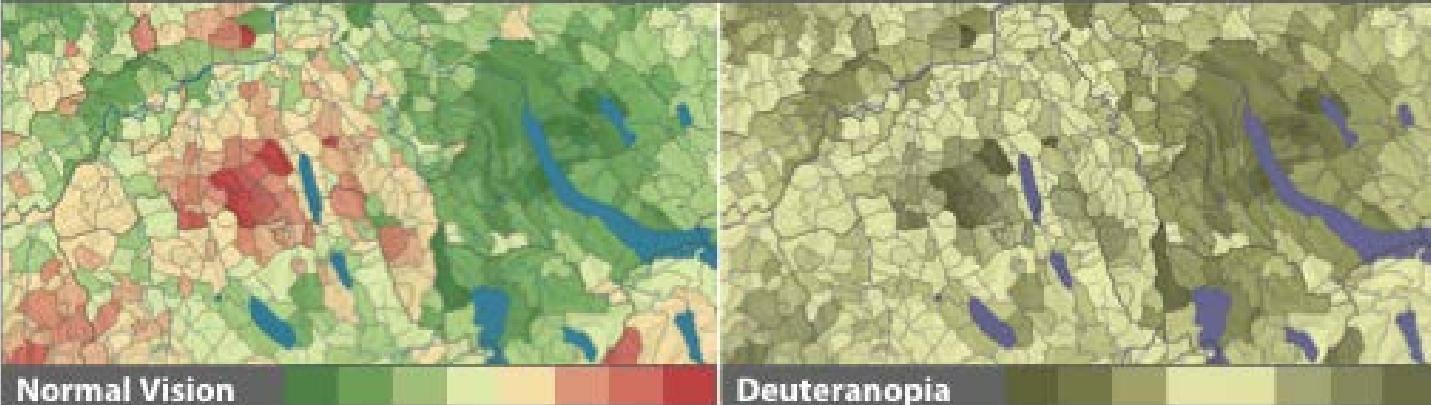


Colormaps: color blindness

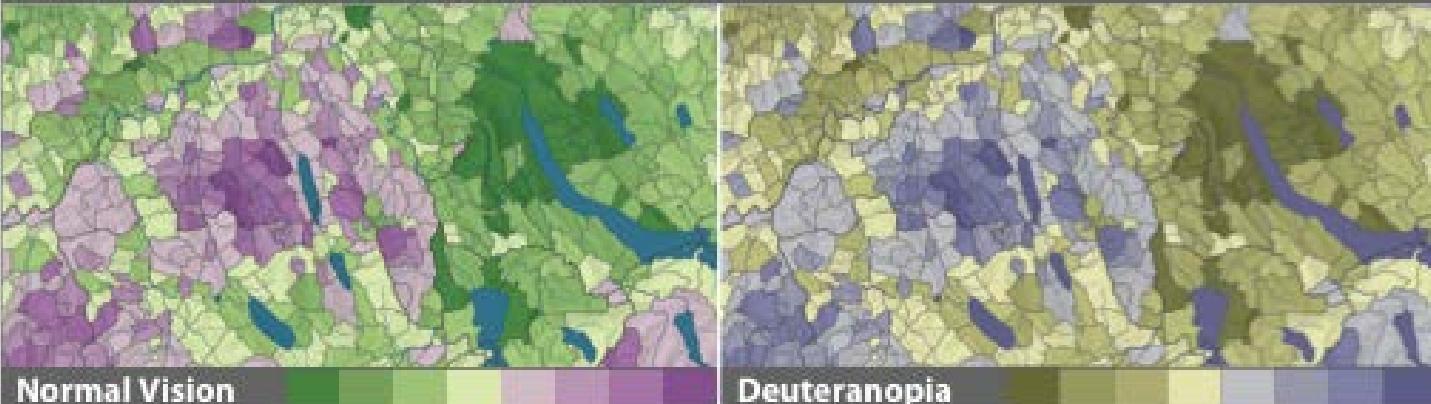


Colormaps: color blindness

Diverging Red-Green Color Scheme



Diverging Purple-Green Color Scheme



Ressources

- Choosing Colormaps
 - [Color Brewer](#)
 - [Cmocean](#)
- Color blindness
 - [vischeck.com](#)

SCHEDULE – PYTHON VISUALISATION COURSE

Day 1

09:00 – 10:30	Part 0: Introduction
10:30 – 11:00	Coffee Break
11:00 – 12:30	Part 1: Plotting in Python with matplotlib
12:30 – 13:30	Lunch Break
13:30 – 15:00	Part 1: Plotting in Python with matplotlib
15:00 – 15:30	Coffee Break
15:30 – 17:00	Part 2: Advanced plotting

Day 2

09:00 – 10:30	Part 3: Plotting georeferenced data
10:30 – 11:00	Coffee Break
11:00 – 12:30	Part 3: Plotting georeferenced data
12:30 – 13:30	Lunch Break
13:30 – 15:00	Part 3: Plotting georeferenced data
15:00 – 15:30	Coffee Break
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15:30 – 17:00	Part 4: User contributions

User contributions

- Advanced plots from the community

References

- Hauser et al., 2016, GRL (<https://doi.org/10.1002/2016GL068036>)
- Hauser et al., 2017, GMD (<https://doi.org/10.5194/gmd-10-1665-2017>, 2017)
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- Rougier et al., 2014, PLOS (<https://doi.org/10.1371/journal.pcbi.1003833>)
- Stott et al., 2004, Nature (<https://doi.org/10.1038/nature03089>)
- Van Heerwaarden et al., 2014, Biogeosciences (<https://doi.org/doi:10.5194/bg-11-6159-2014>)