

# Data Visualisation II

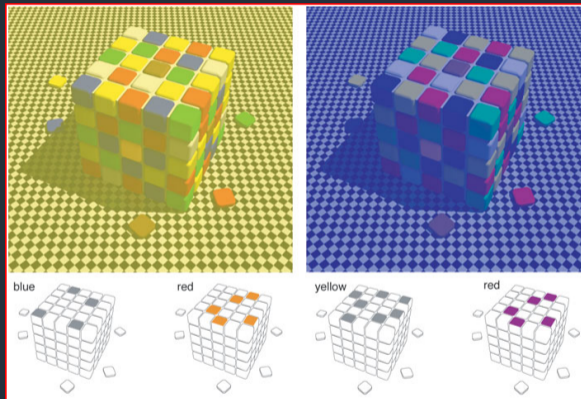
Gotchas

Peter Hill

# Data Visualisation II: Gotchas

- Visualisation can be tricky: Human visual perception system is full of hacks built on top of poorly-designed hardware
- What colour is the sky?
- Our visual system does not perceive all colours equally

# Our brains try to be clever



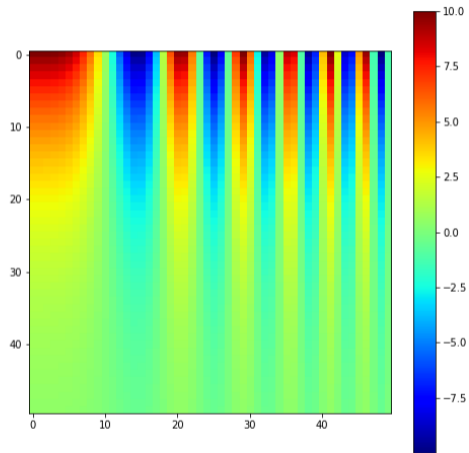
<http://www.mattnewport.com/pics/colour-constancy.png>

# Plotting 2D data

Need to plot something with structure of various scales

```
x = np.linspace(0, 6)
y = np.linspace(0, 3)[: , np.newaxis]
z = 10 * np.cos(x ** 2) * np.exp(-y)
fig, ax = plt.subplots(1, figsize=(8, 8))
ax.imshow(z, cmap='jet')
plt.colorbar(ax.imshow(z));
```

# Jet colourmap



# Convert to grey-scale

```
import matplotlib.colors as mpl_colors
```

```
def grayify_cmap(cmap):
```

```
    """Return a grayscale version of the colormap"""
```

```
    cmap = plt.cm.get_cmap(cmap)
```

```
    colors = cmap(np.arange(cmap.N))
```

```
    # convert RGBA to perceived greyscale luminance
```

```
    # cf. http://alienryderflex.com/hsp.html
```

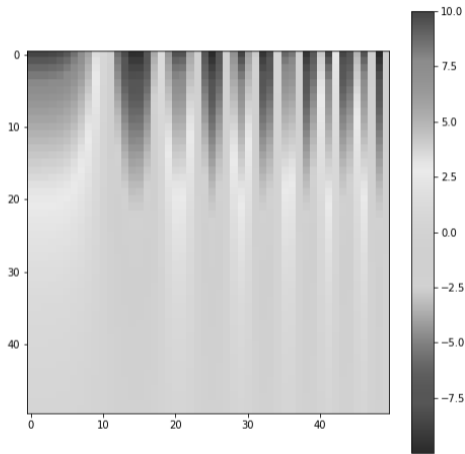
```
    RGB_weight = [0.299, 0.587, 0.114]
```

```
    luminance = np.sqrt(np.dot(colors[:, :3] ** 2, RGB_weight))
```

```
    colors[:, :3] = luminance[:, np.newaxis]
```

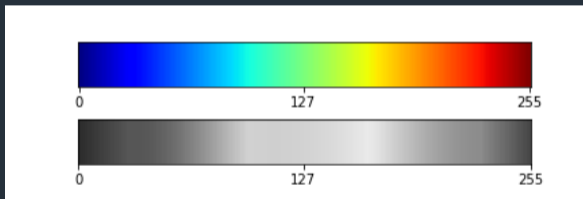
```
    return mpl_colors.LinearSegmentedColormap.from_list(cmap.name +  
        "_grayscale", colors, cmap.N)
```

# Jet colourmap converted to grey-scale



# Jet colourmap

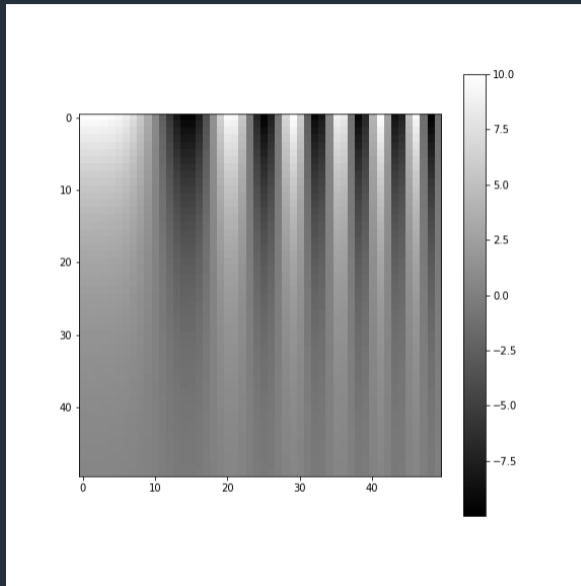
- Is this really what's happening?
- Look at colourmap converted to grey-scale:



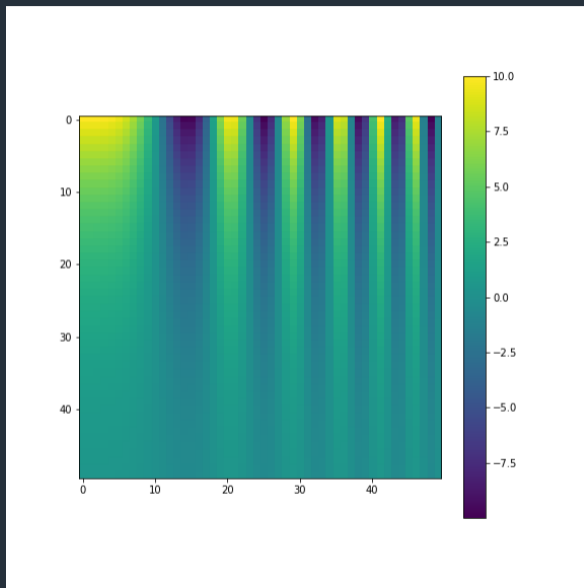
- Notice the banding?



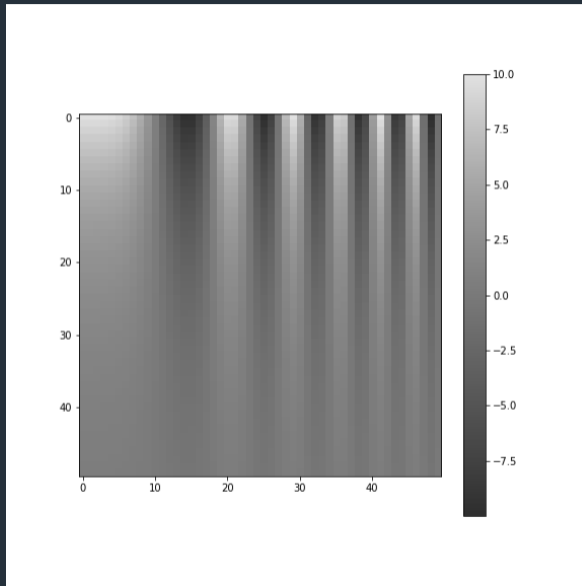
# Actual grey-scale colourmap



# Viridis colourmap

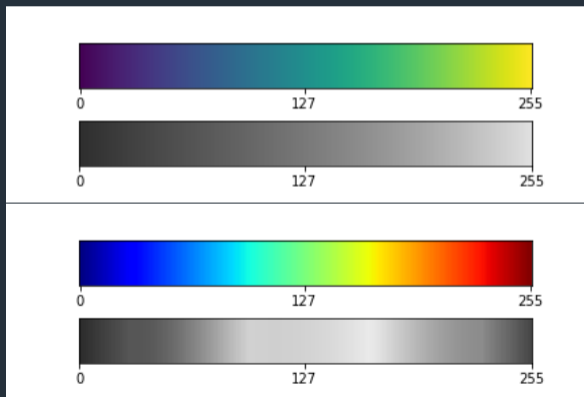


# Viridis colourmap converted to grey-scale

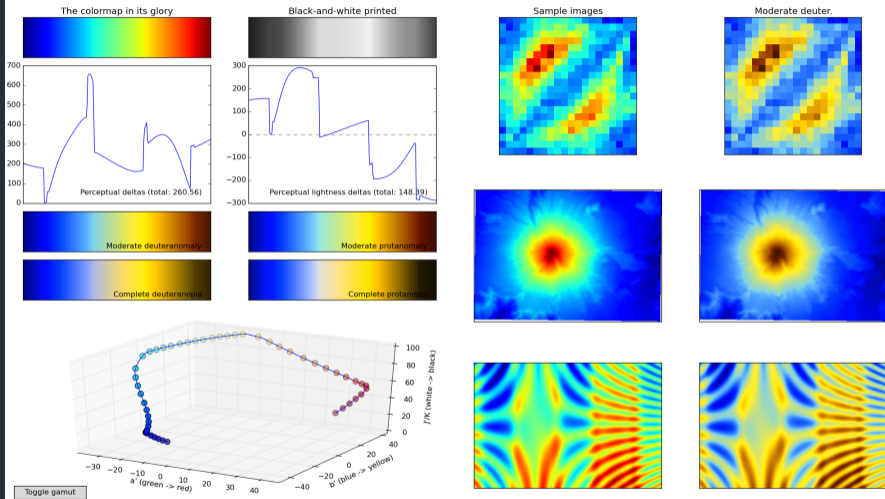


# Viridis colourmap

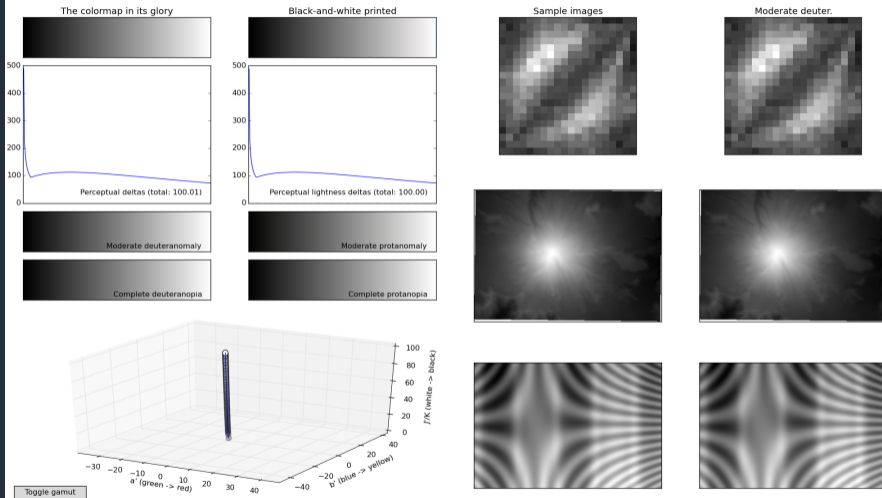
- Compare the grey-scale versions of viridis and jet
- Imperceptible banding in viridis! “Perceptually uniform”



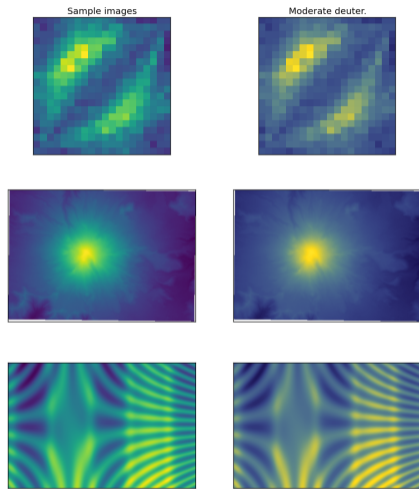
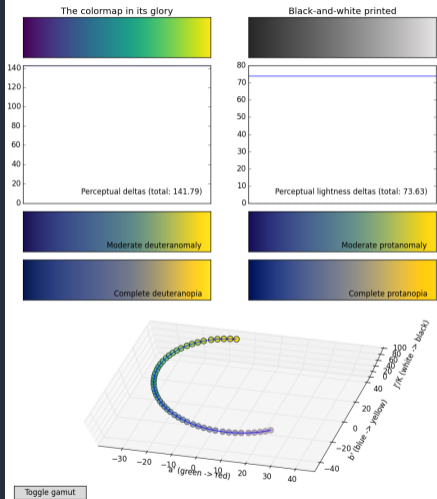
# Colormap evaluation: jet



## Colormap evaluation: gray



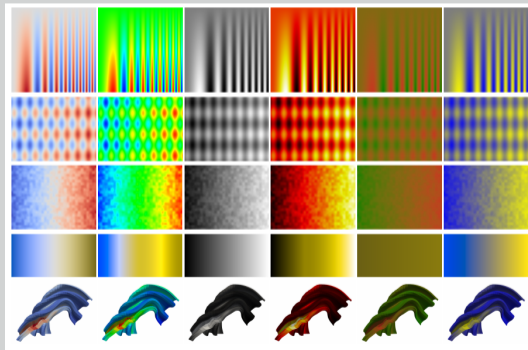
# Colormap evaluation: option\_d.py



# Other pitfalls

## Colour-blindness

- Affects about 8% of men, and 0.5% of women
- Various kinds, most common of which is red-green colour blindness
- Don't pick colour maps with both red and green





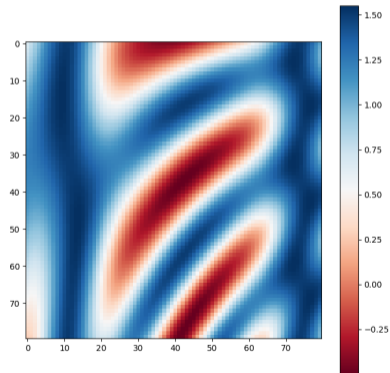
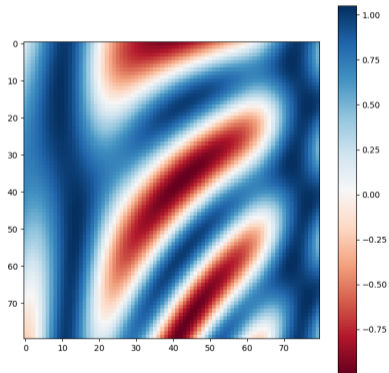
# Other pitfalls

## Diverging vs sequential vs qualitative datasets

- For diverging data (i.e. positive and negative values), need a good centre colour
- Paraview's default colour map is ideal and scientifically designed
- Matplotlib, "RdBu" is probably best (though has red as "negative")
- To use diverging colour schemes correctly, best to set min/max values to plus/minus the max absolute value
  - Careful with normalisations! Using the full dynamic range for both positive/negative makes the extreme values look equal in magnitude

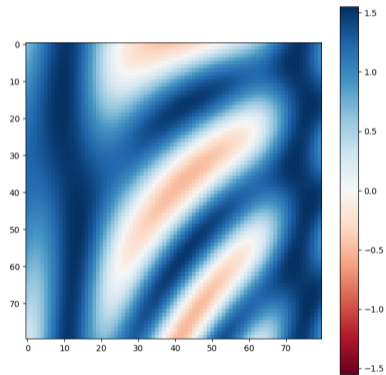
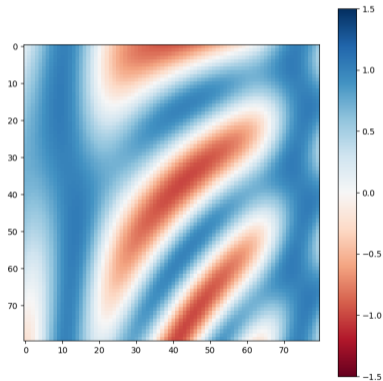
# Positive and negative data

- Just plotted



# Positive and negative data

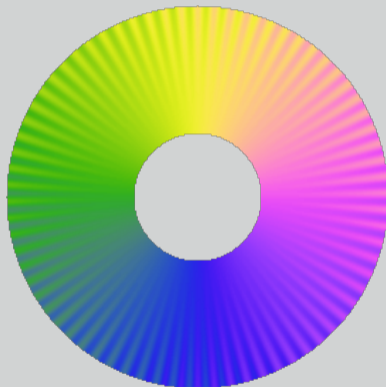
- Plotted with  $\pm \max(\text{abs}(f))$



# Other pitfalls

## Cyclic/phase colour maps

- Much trickier!
- Need to cover large area of gamut (colour space) whilst being periodic
- End up either being “washed out” or with banding
- Desirable to have “main” colours at cardinal directions
- Need to be 2D colourmap for complex plane (magnitude and phase)
- See <http://peterkovesi.com/projects/colourmaps/>

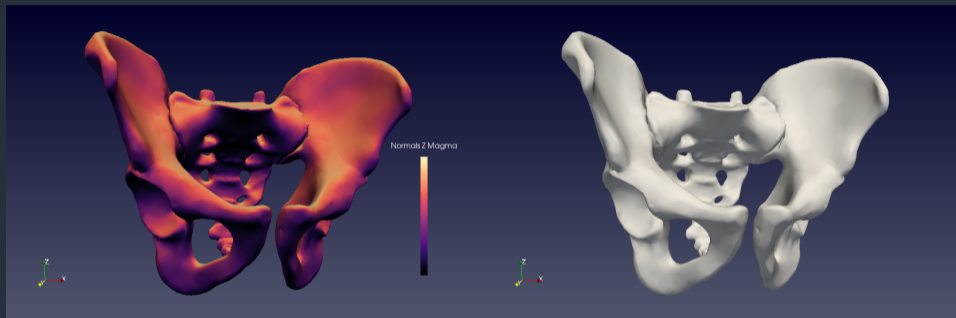


# Other pitfalls

## Colour maps for 3D images

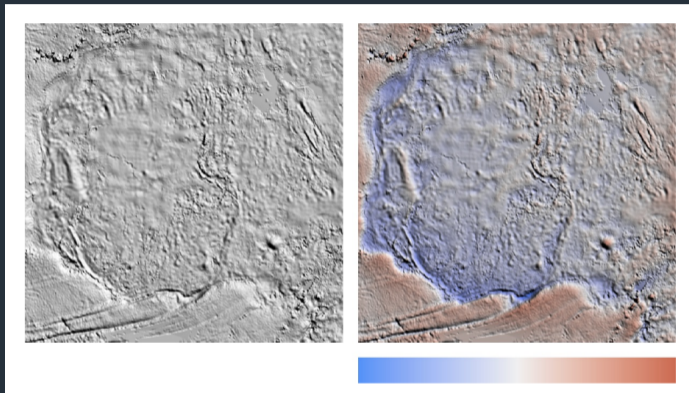
- 3D images with lighting and shading can interfere with colour maps
- Choose an “iso-luminant” colour mapping
  - Default colourmap in Paraview also designed for this
- Unfortunately, these tend to be “washed out” due to lack of saturated colours

# Colour maps for 3D images



<http://noeskasmit.com/colormaps-in-medical-visualization/>

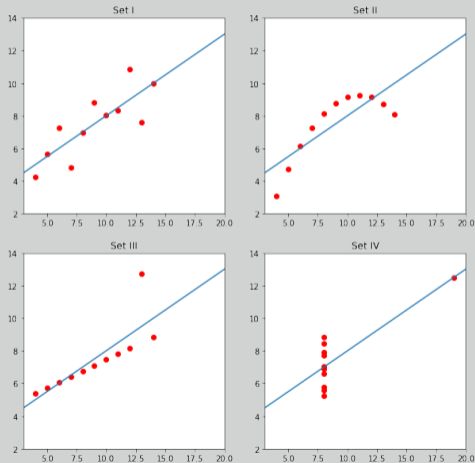
# Colour maps for 3D images



Good Colour Maps: How to Design Them, Peter Kovesi,  
<https://arxiv.org/abs/1509.03700>

# Plotting line graphs

## Anscombe's Quartet





# Plotting line graphs

## Data Channels

In order of effectiveness:

- Quantitative Data:
  - Position
    - On dependent scales
    - On independent/unaligned scales
  - Length
  - Angle
  - Area
  - Depth
  - Luminance
  - Saturation
  - Curvature
  - Volume

# Plotting line graphs

## Data Channels

In order of effectiveness:

- Categorical Data:
  - Spatial location
  - Hue
  - Motion
  - Shape (Glyph)

# Plotting line graphs

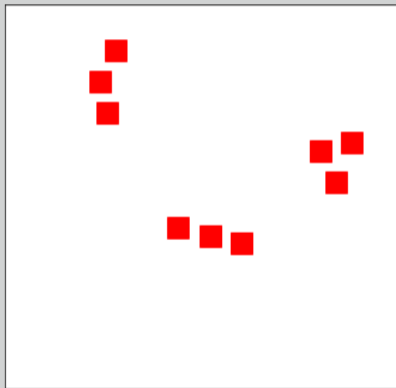
## Gestalt Principles of Perception

- Proximity - objects close to each other are seen as groups
- Similarity - objects that share channels (colour or shape for example) are seen as grouped
- Enclosure - objects enclosed by boundary and/or area are seen as grouped
- Closure - within limits, open objects are perceived as closed
- Continuity - objects that align/flow are seen as continuous objects
- Connection - objects that are connected are seen as grouped

# Gestalt Principles of Perception

## Proximity

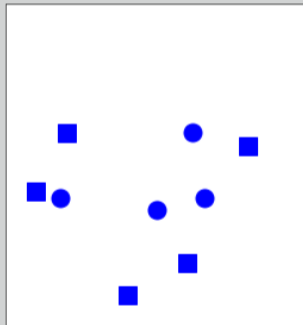
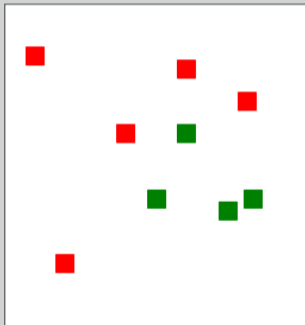
Proximity



# Gestalt Principles of Perception

## Similarity

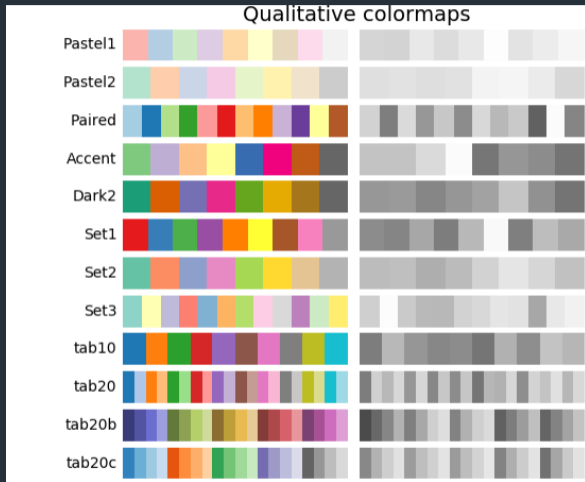
Similarity



# Colours for qualitative datasets

- Cynthia Brewer designed several sets of colours for plotting cartographic data
- More generally useful though
- Designed with B&W printing, colour blindness, perceptual uniformity in mind
- <https://colorbrewer2.org>

# Colours for qualitative datasets



<https://matplotlib.org/users/colormaps.html#colormaps>

# Conclusions

- Upgrade to matplotlib 2.0+ if you haven't already! Has sensible defaults
- Don't use Jet or rainbow color map!
- For data with positive/negative, use a diverging colourmap
- For qualitative data, use Brewer colours



# Resources

- Set of Jupyter notebooks on data visualisation: <https://github.com/UoMResearchIT/data-vis-truthiness-hurts>
- Brewer colours for qualitative data: <https://colorbrewer2.org>
- Theory of colourmaps: <http://peterkovesi.com/projects/colourmaps/>
- Colourmaps in Matplotlib: <https://matplotlib.org/users/colormaps.html#colormaps>