

Principles of Spatial Analysis

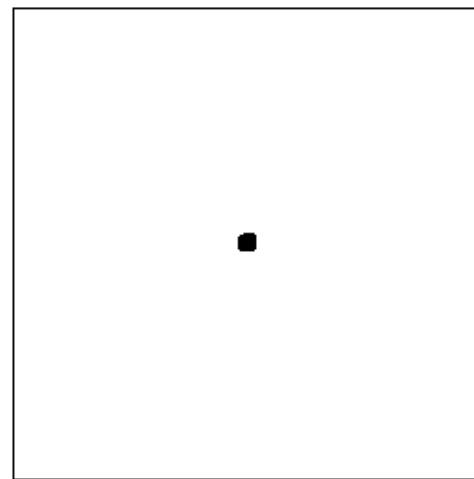
SHORT LECTURE 01, WEEK 06: POINT PATTERN ANALYSIS



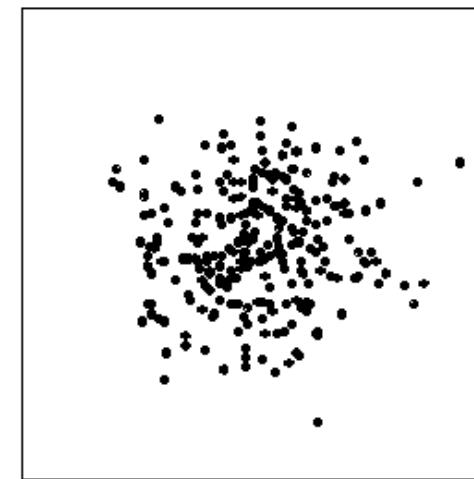
point pattern analysis

what is it all about

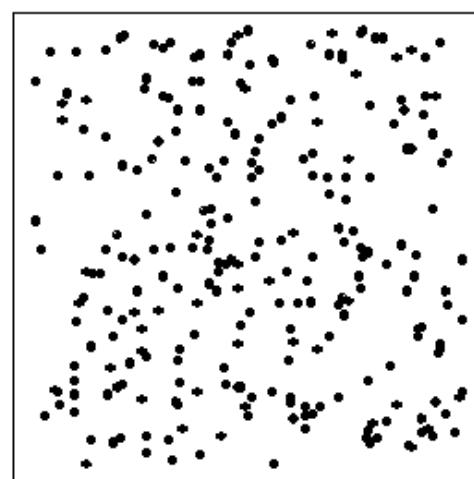
clustered



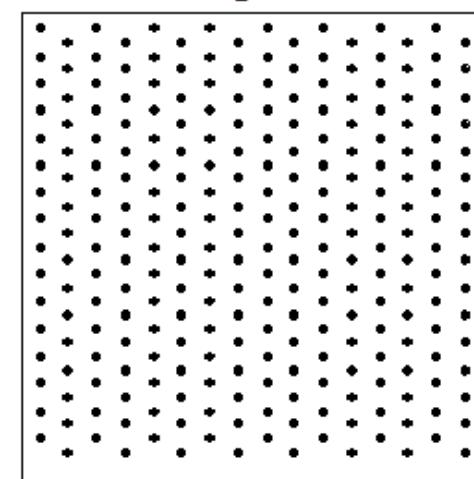
normal



random

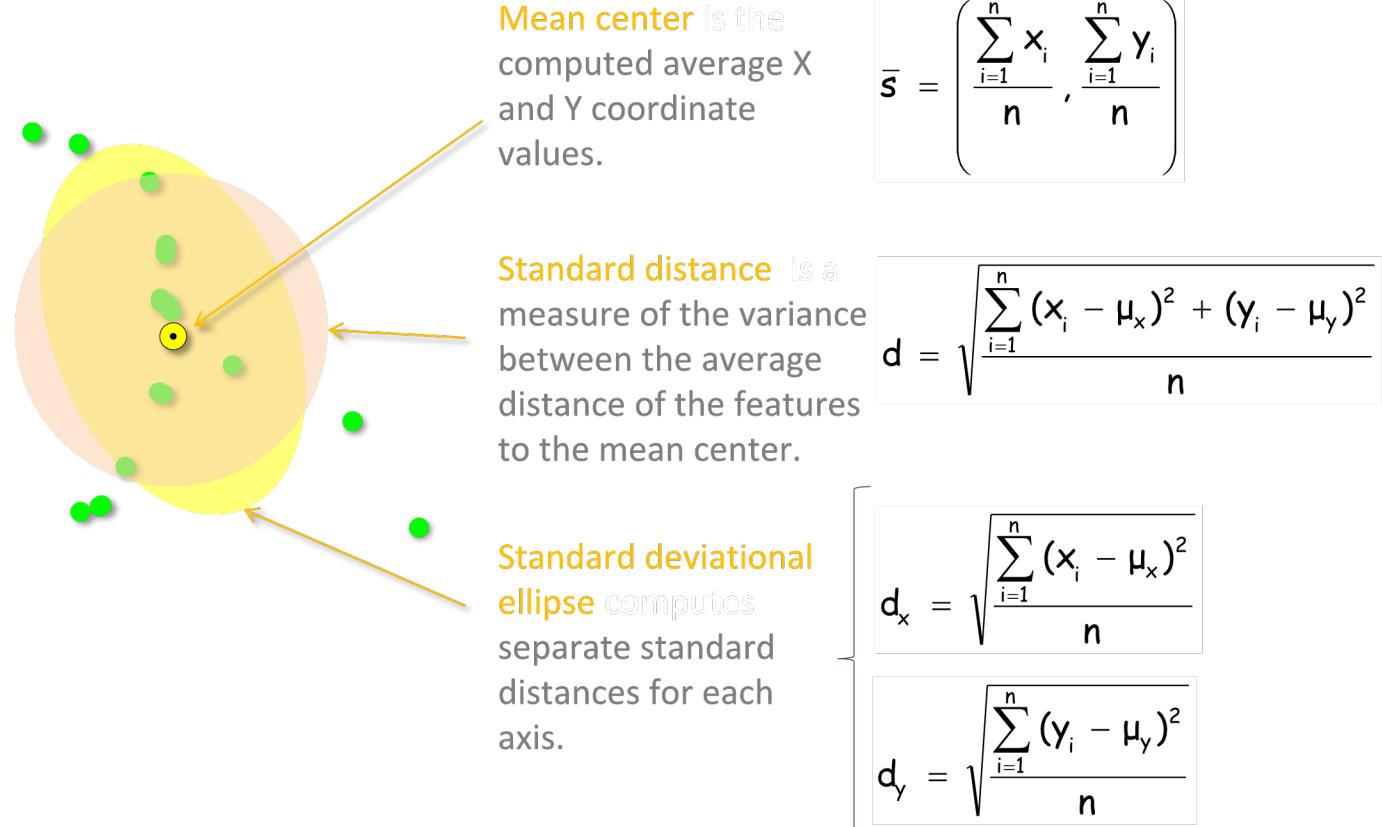


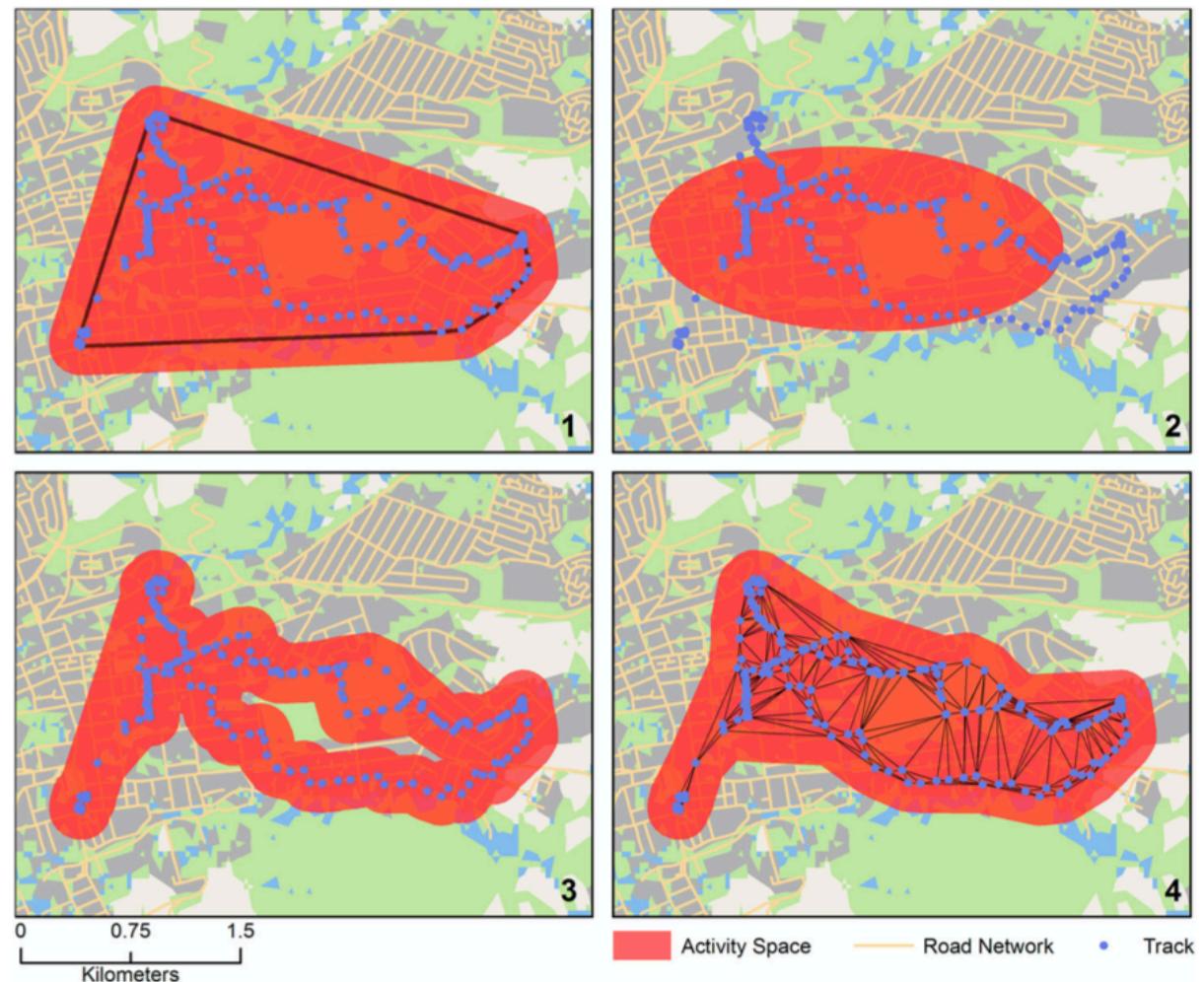
regular



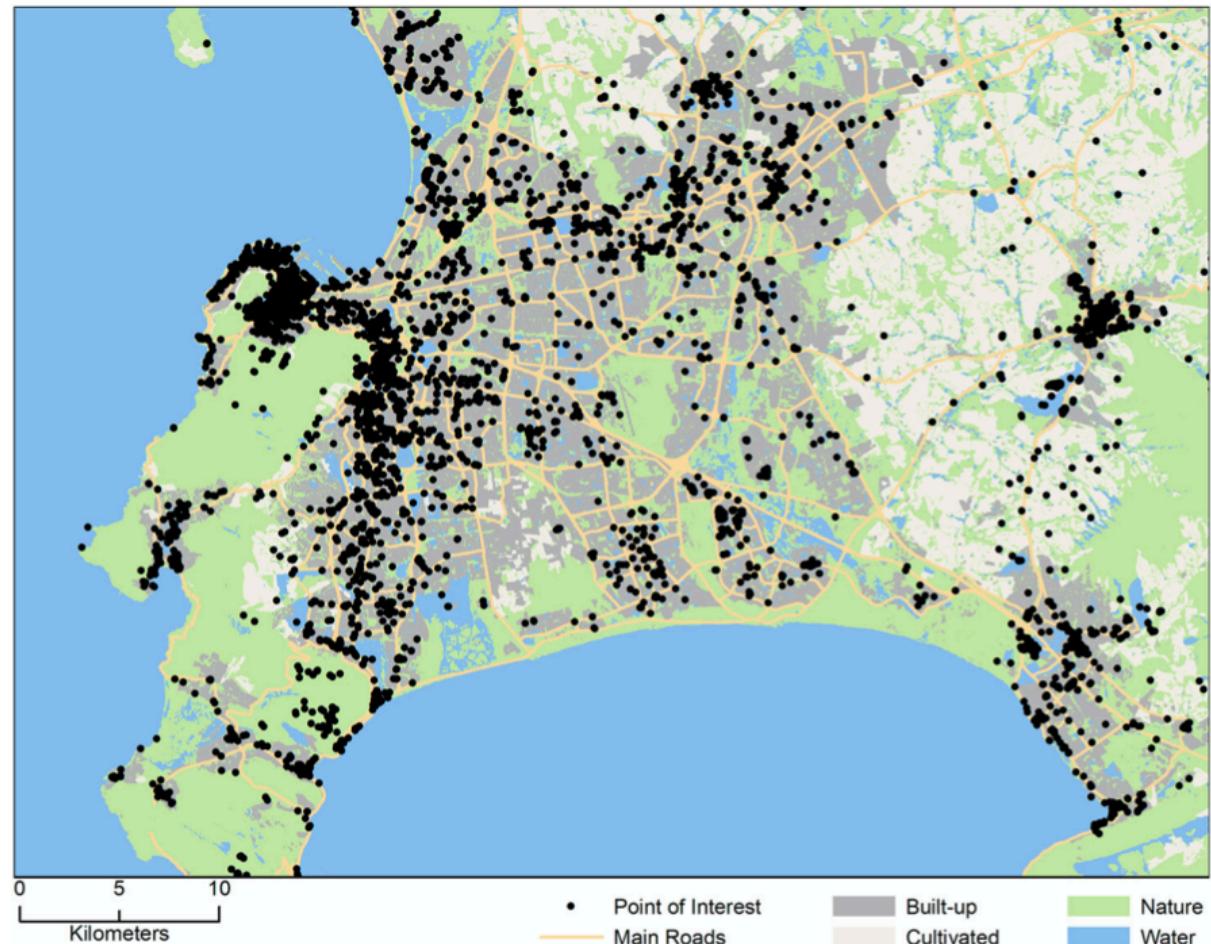
3 ways
of thinking about patterns in your points

descriptive statistics





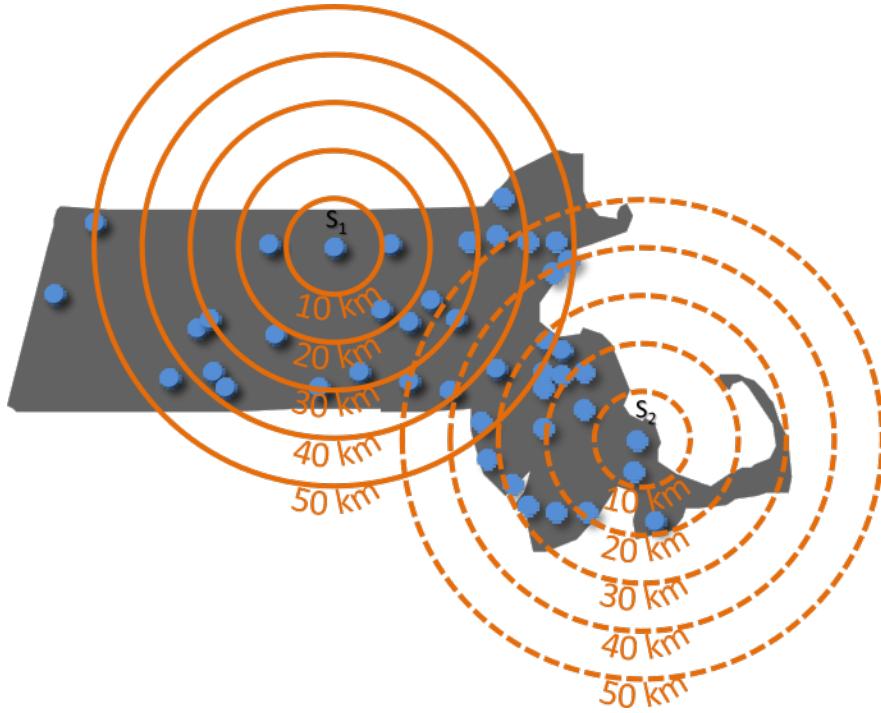
Van Dijk, J. T. & S. Krygsman. 2017. Analyzing Travel Behavior by Using GPS-Based Activity Spaces and Opportunity Indicators. *Journal of Urban Technology* 25(2): 102-124



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distance based methods

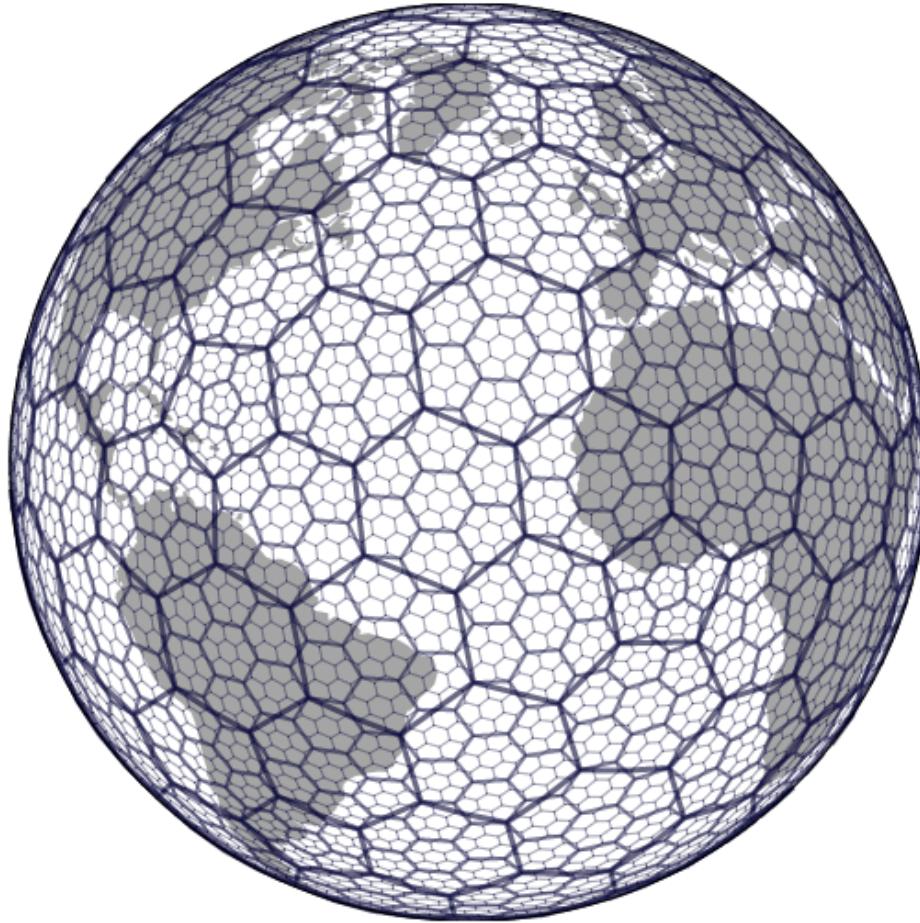
average nearest neighbour



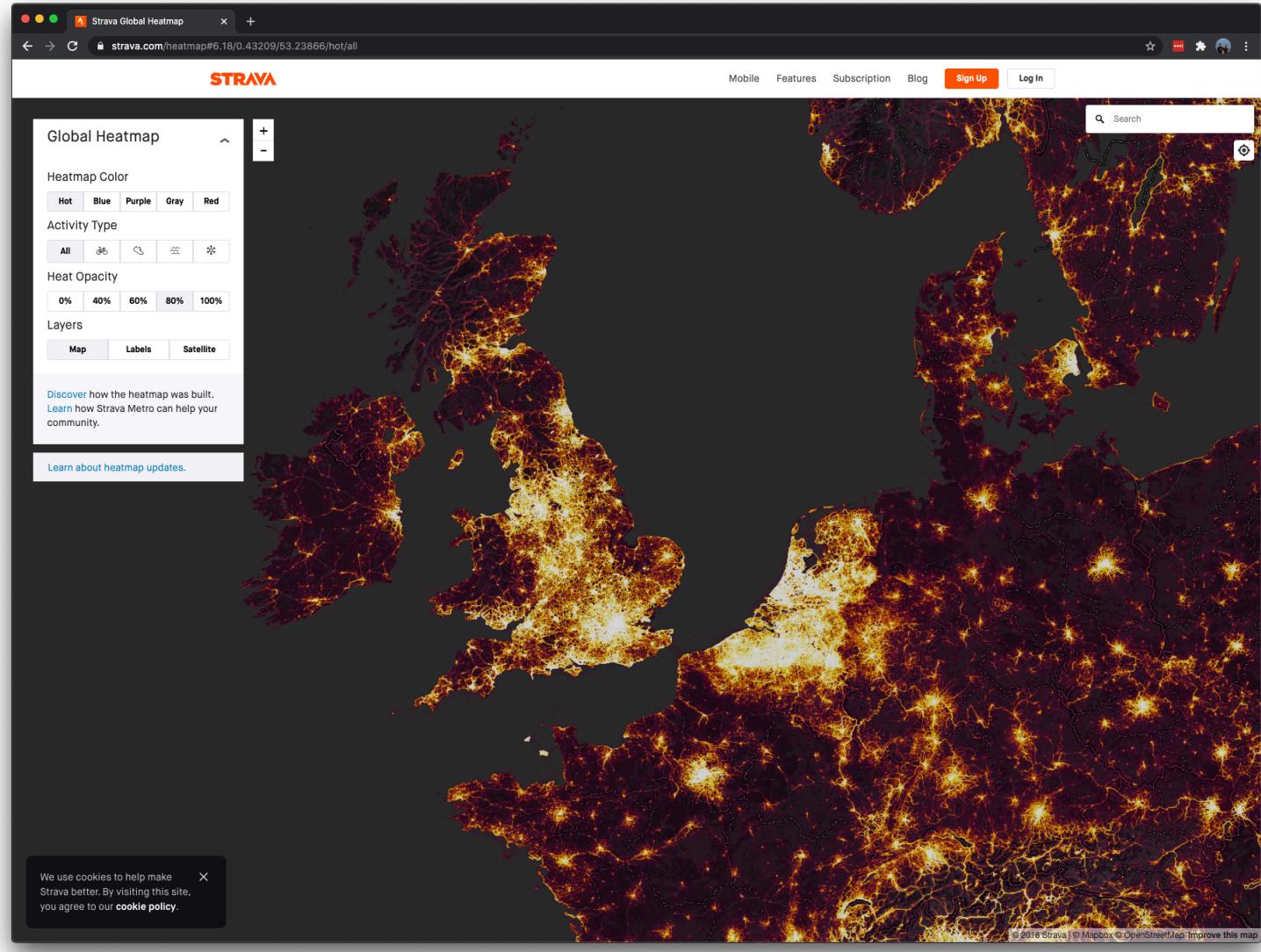
density based methods



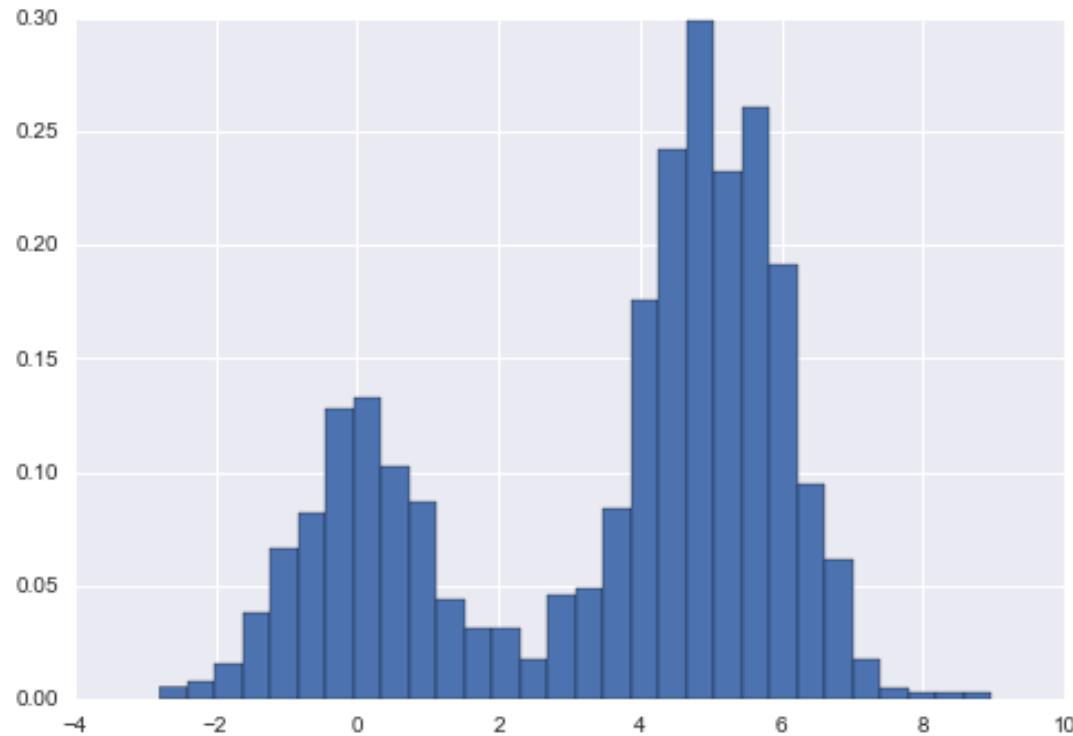
Gimond, M. 2020. *Geodesic geometry*. [online] <https://mgimond.github.io/Spatial/index.html>



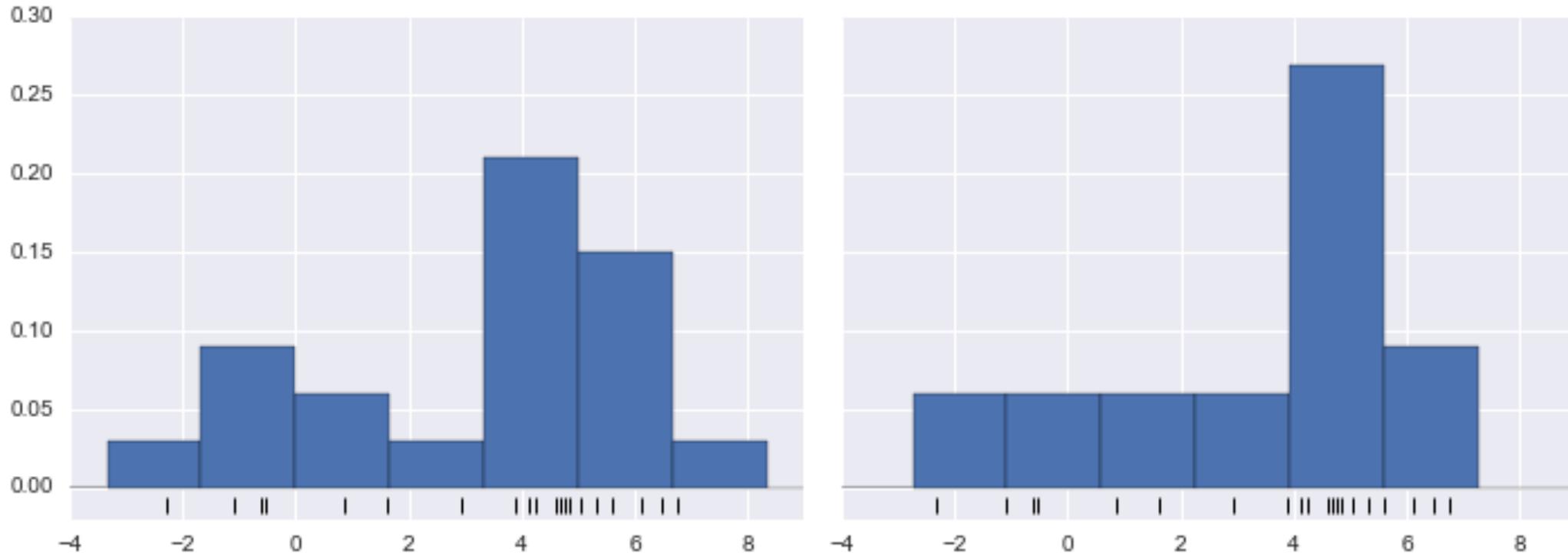
Uber. 2018. *H3: Uber's Hexagonal Hierarchical Spatial Index*. [online] <https://eng.uber.com/h3/>



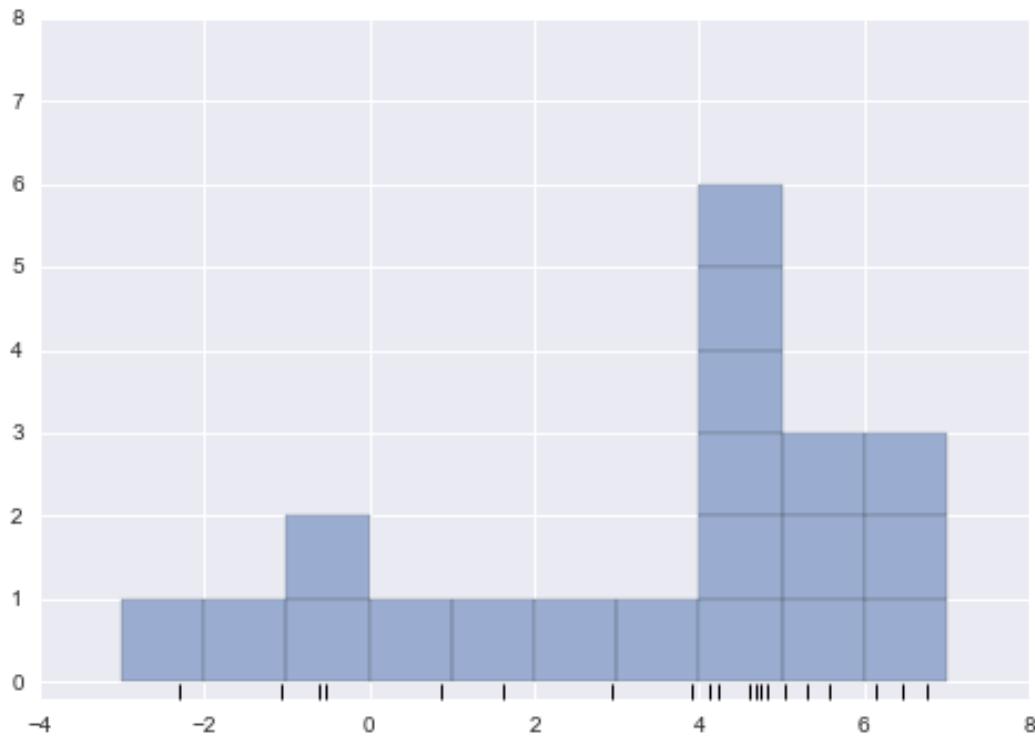
```
Tycho — python — 85x25
>>> def make_data(N, f=0.3, rseed=1):
...     rand = np.random.RandomState(rseed)
...     x = rand.randn(N)
...     x[int(f * N):] += 5
...     return x
...
[>>> x = make_data(1000)
[>>> x
array([-1.62434536e+00, -6.11756414e-01, -5.28171752e-01, -1.07296862e+00,
       8.65407629e-01, -2.30153870e+00,  1.74481176e+00, -7.61206901e-01,
       3.19039096e-01, -2.49370375e-01,  1.46210794e+00, -2.06014071e+00,
      -3.22417204e-01, -3.84054355e-01,  1.13376944e+00, -1.09989127e+00,
      -1.72428208e-01, -8.77858418e-01,  4.22137467e-02,  5.82815214e-01,
      -1.10061918e+00,  1.14472371e+00,  9.01590721e-01,  5.02494339e-01,
       9.008555949e-01, -6.83727859e-01, -1.22890226e-01, -9.35769434e-01,
      -2.67888080e-01,  5.30355467e-01, -6.91660752e-01, -3.96753527e-01,
      -6.87172700e-01, -8.45205641e-01, -6.71246131e-01, -1.26645989e-02,
      -1.11731035e+00,  2.34415698e-01,  1.65980218e+00,  7.42044161e-01,
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      -7.54397941e-01,  1.25286816e+00,  5.12929820e-01, -2.98092835e-01,
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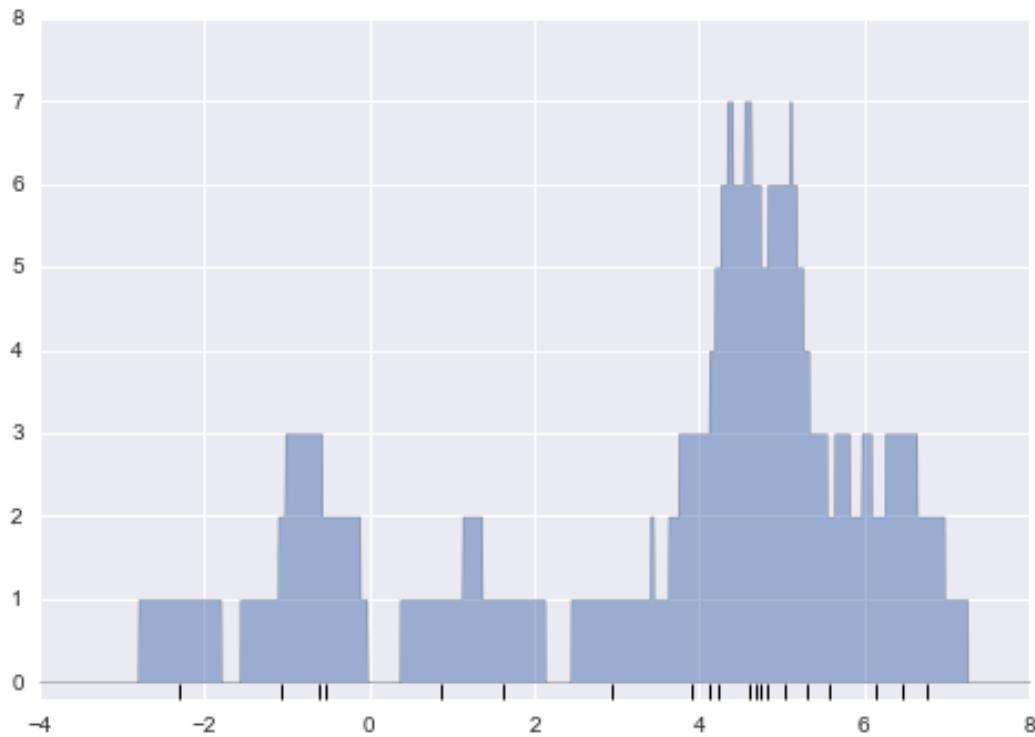
VanderPlas, J. 2016. *Python data science handbook: essential tools for working with data*.
O'Reilly Media, Inc.



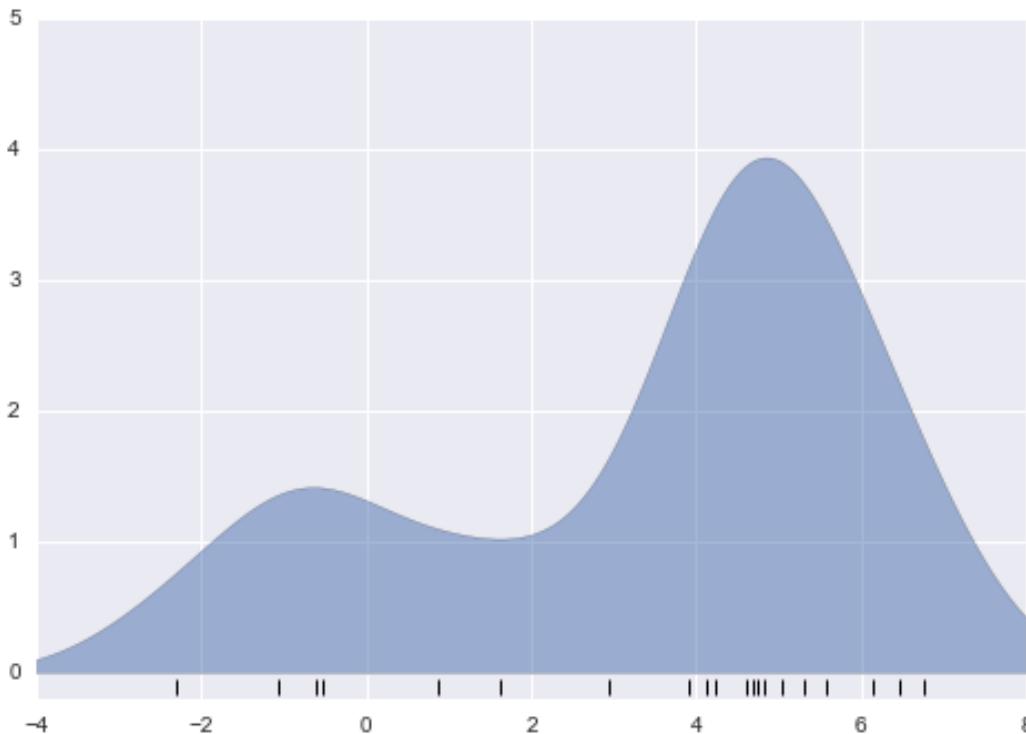
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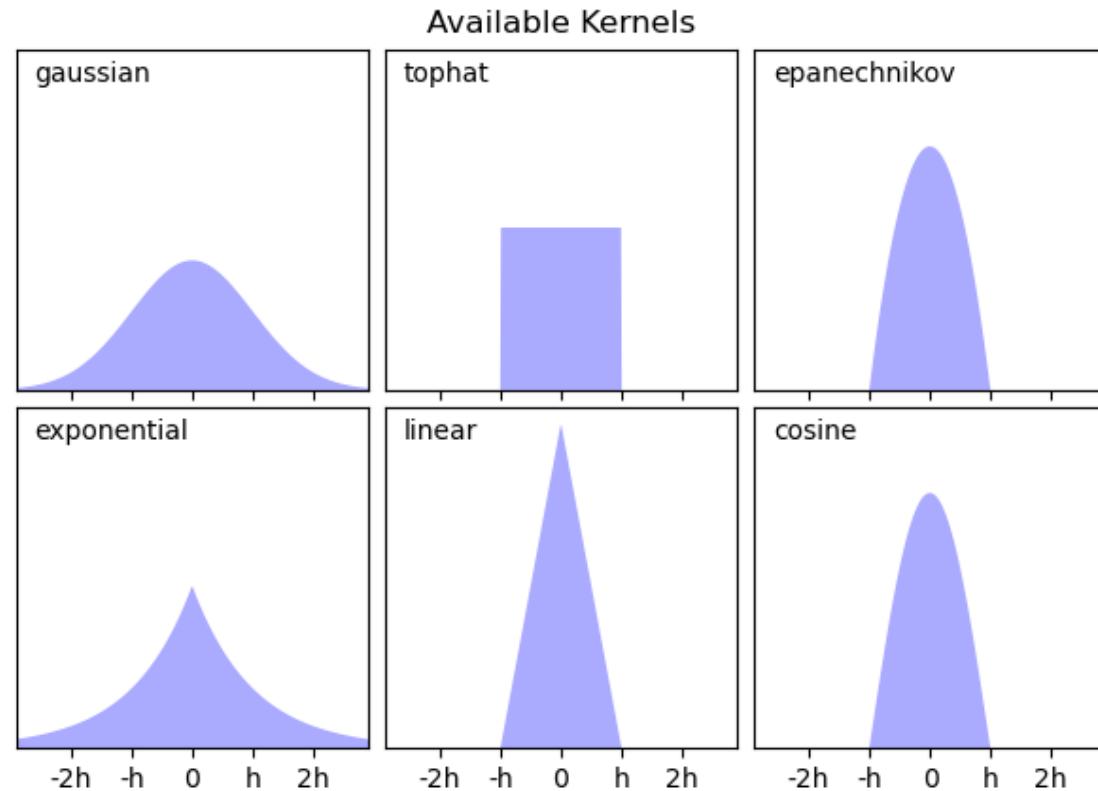
VanderPlas, J. 2016. *Python data science handbook: essential tools for working with data*.
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O'Reilly Media, Inc.



Scikit-learn. 2020. *Density estimation*. [online] <https://scikit-learn.org/stable/modules/density.html>

Shi, X. 2010. Selection of bandwidth type and adjustment side in kernel density estimation over inhomogeneous backgrounds.
International Journal of Geographical Information Science 24(5): 643-660

$$\hat{f}(x, y) = \frac{1}{nh^2} \sum_{i=1}^n K\left(\frac{d_{i,(x,y)}}{h}\right)$$

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$$\hat{f}(x, y) = \frac{1}{nh^2} \sum_{i=1}^n K\left(\frac{d_{i,(x,y)}}{h}\right)$$

Intensity at point (x,y)

Shi, X. 2010. Selection of bandwidth type and adjustment side in kernel density estimation over inhomogeneous backgrounds.
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$$\hat{f}(x, y) = \frac{1}{nh^2} \sum_{i=1}^n K\left(\frac{d_{i,(x,y)}}{h}\right)$$

Points within bandwidth

Shi, X. 2010. Selection of bandwidth type and adjustment side in kernel density estimation over inhomogeneous backgrounds.
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$$\hat{f}(x, y) = \frac{1}{nh^2} \sum_{i=1}^n K\left(\frac{d_{i,(x,y)}}{h}\right)$$

Kernel function

let's put it into practice