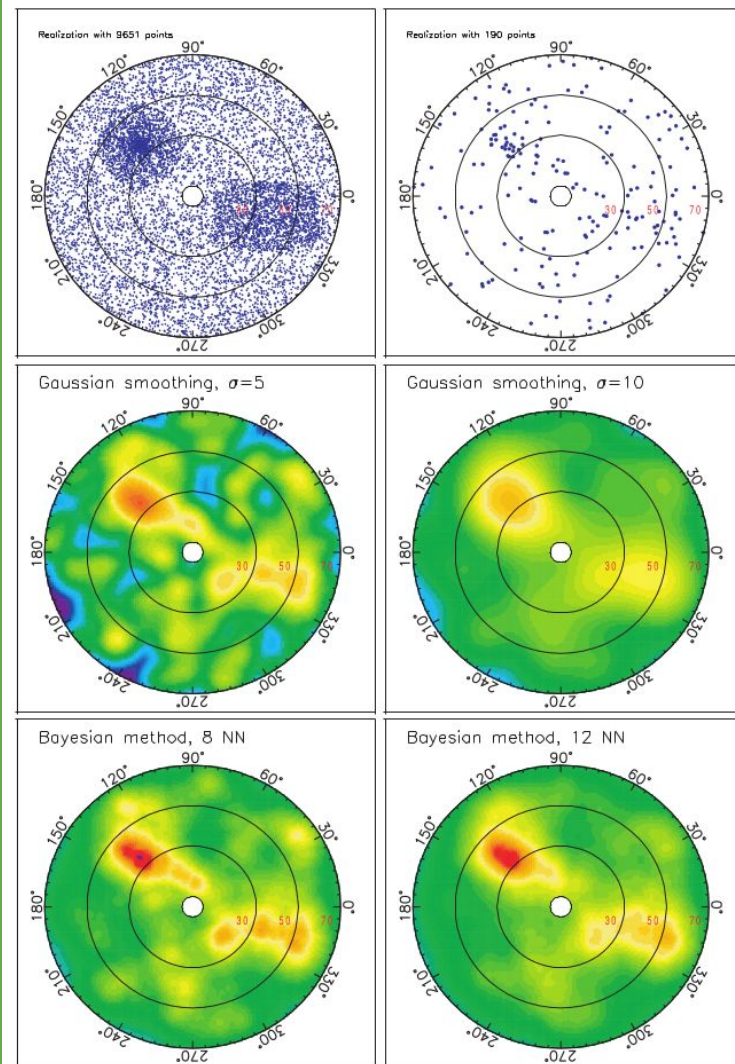


"On Estimating Non-Uniform  
Density Distributions Using  
N Nearest Neighbors"  
Woźniak & Kruszewski 2012

Radek Poleski  
29 Oct 2025

# What problem is discussed?

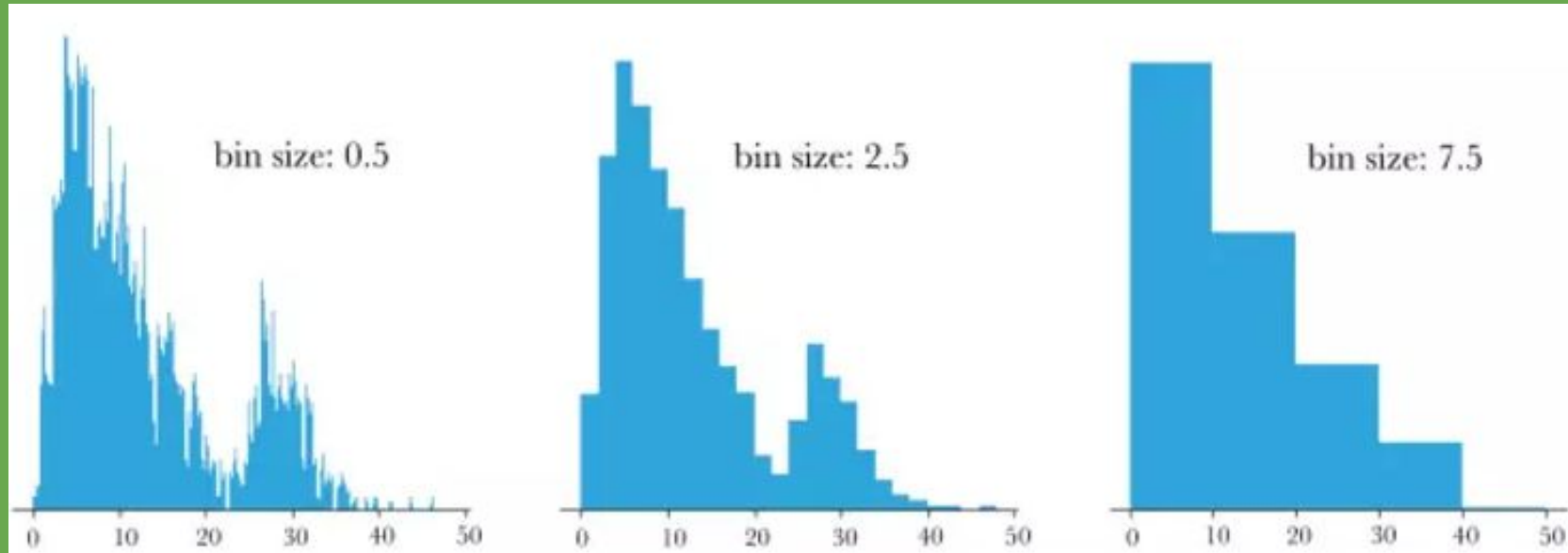
- Spatial studies of distributions of objects: RR Lyr, galaxies, GRBs, stars in globular clusters, finding star clusters...
- Simulations
- Color-magnitude diagrams of globular clusters



# Solution 1 - binning

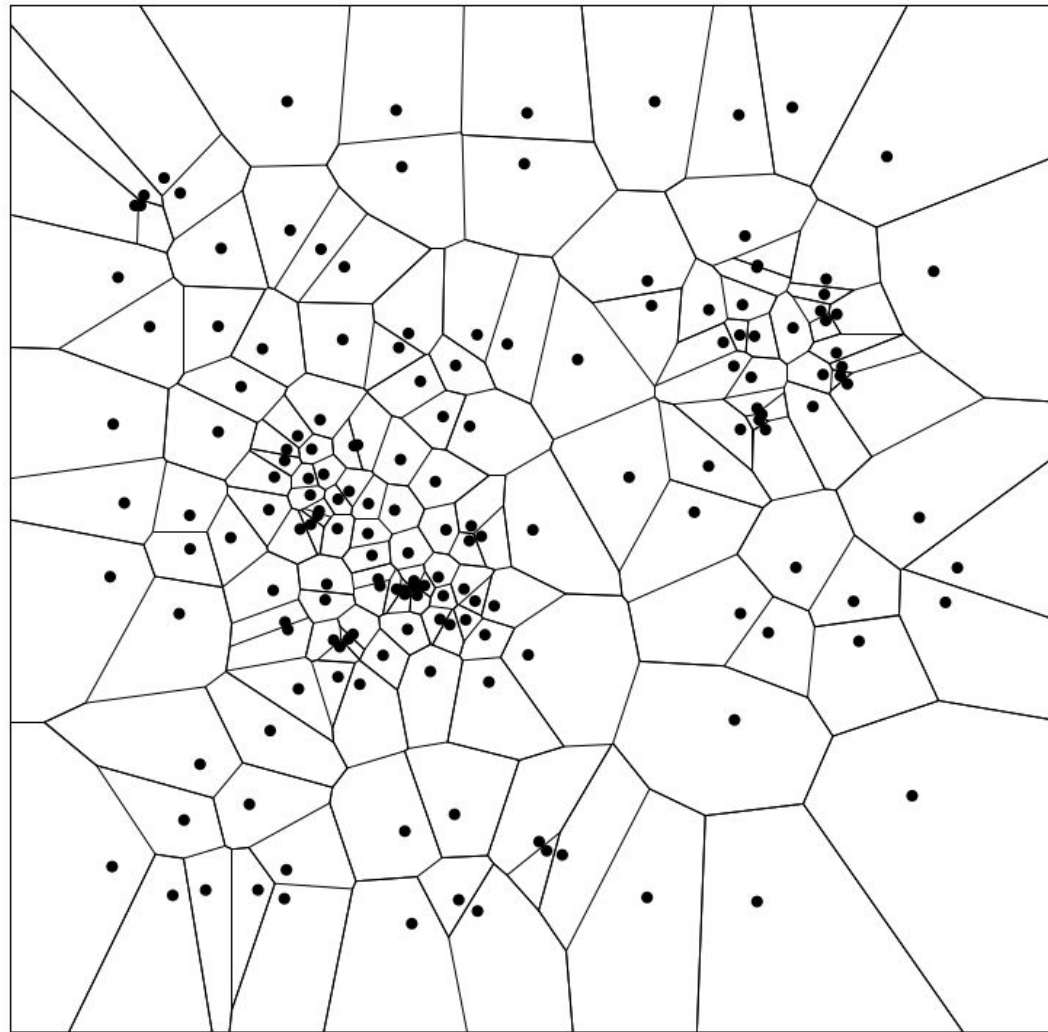
Obvious issues with bin size and phase.

See SJC on 14 Dec. 2021 by P. Szewczyk



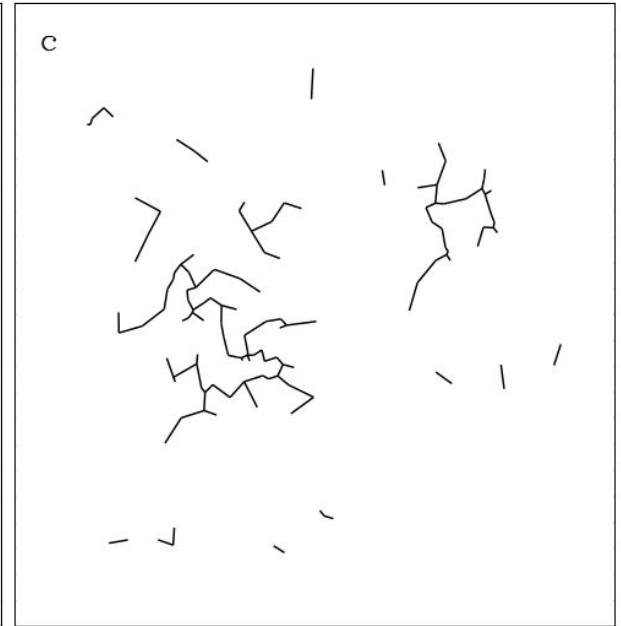
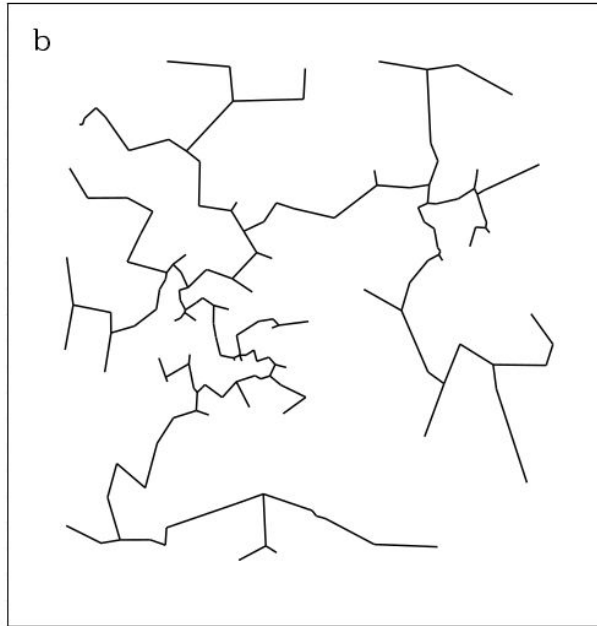
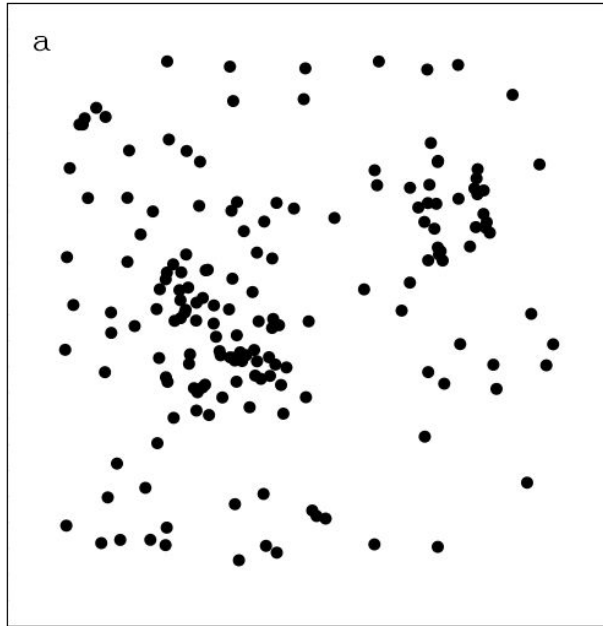
# Solution 2 - Voronoi tessellation

Very sensitive to small-scale fluctuations -> smoothing

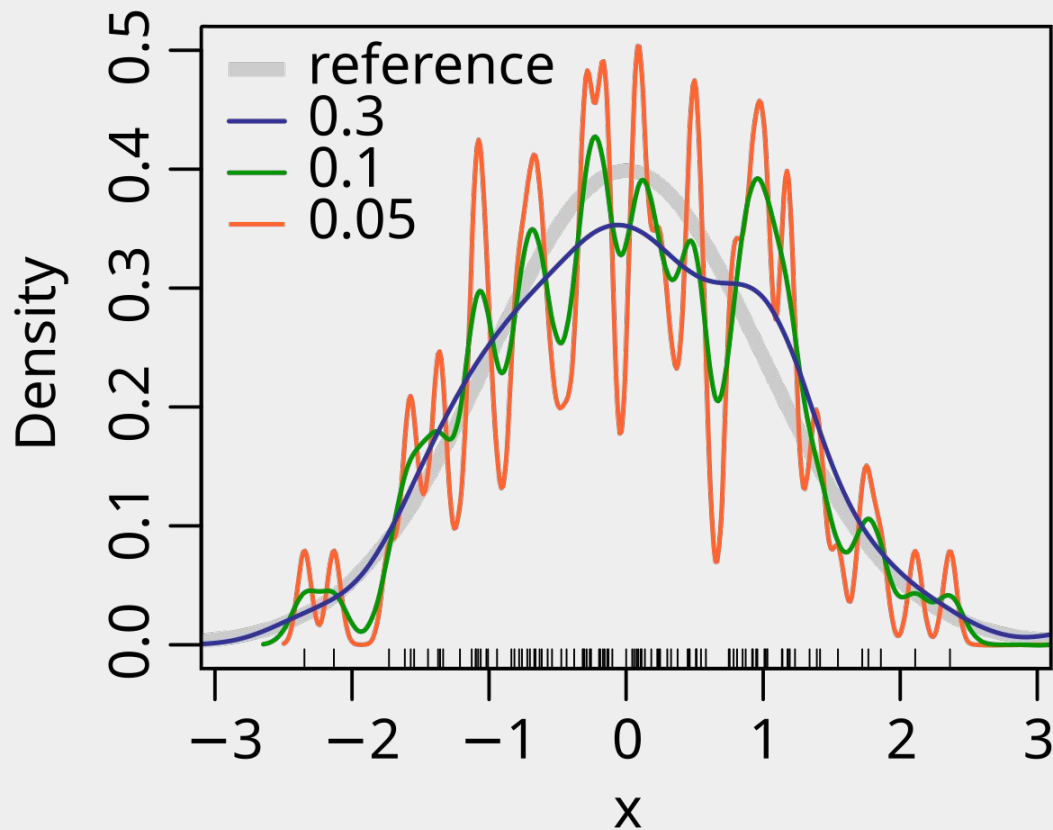


# Solution 3 - minimum spanning tree separation

MST is the unique set of straight lines (“edges”) connecting a given set of points (“vortices”) without closed loops, such that the sum of the edge lengths is minimum.



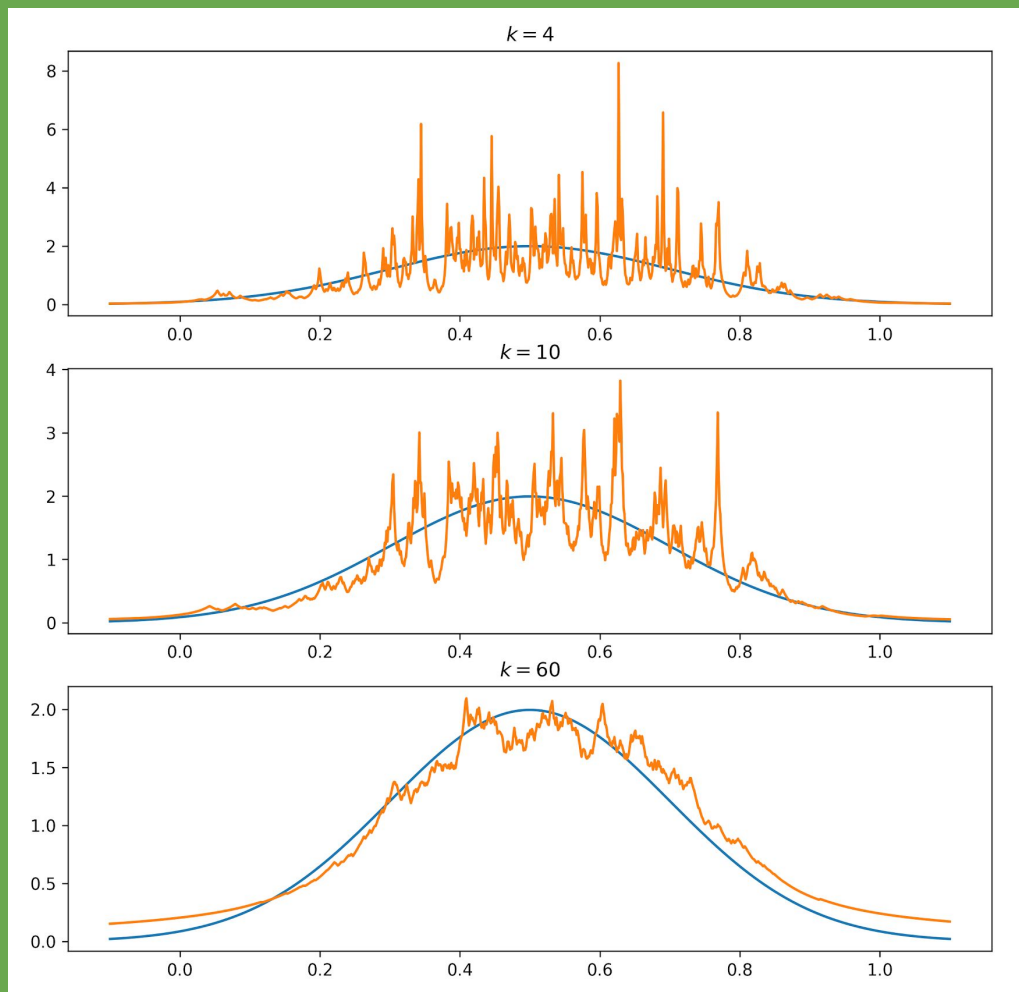
## Solution 4 - kernel density estimation



# Solution 5 - nearest neighbour density

Depends only on a single parameter.

Slow, but faster than MST.



# 5a - Bayesian approach to Nearest Neighbours

Ivezić et al. 2005

$$p(n_0|\{d_k; k = 1, N\}, I) \\ \propto p(d_N|n_0; I)p(n_0|\{d_k; k = 1, N - 1\}, I), \quad (\text{B2})$$

$$p(n_0|\{d_k; k = 1, N\}, I) \propto p(n_0|I) \prod_{k=1}^N p(d_k|n_0, I). \quad (\text{B3})$$

$$p(k|\mu) = \frac{\mu^k e^{-\mu}}{k!}. \quad (\text{B4})$$

The number of expected points,  $\mu = n_0 V_D(d)$ , can be conveniently parametrized as  $(d/d_0)^D$ , where  $d_0$  is the characteristic (mean) distance between two points [determined from  $n_0 V_D(d_0) = 1$ ]. The probability that the distance to the  $k$ th neighbor is between  $d$  and  $d + \delta d$  is the same as the probability that there are exactly  $k - 1$  neighbors enclosed by  $d$  and follows from equation (B4). (For a treatment of non-Poisson distributions, see White 1979.) With the change of variables from  $\mu$  to  $d$ , the probability density distribution for the distance to the  $k$ th nearest neighbor becomes

$$p(d_k|d_0) = \frac{D e^{-(d_k/d_0)^D}}{d_0 (k-1)!} \left( \frac{d_k}{d_0} \right)^{Dk-1}. \quad (\text{B5})$$



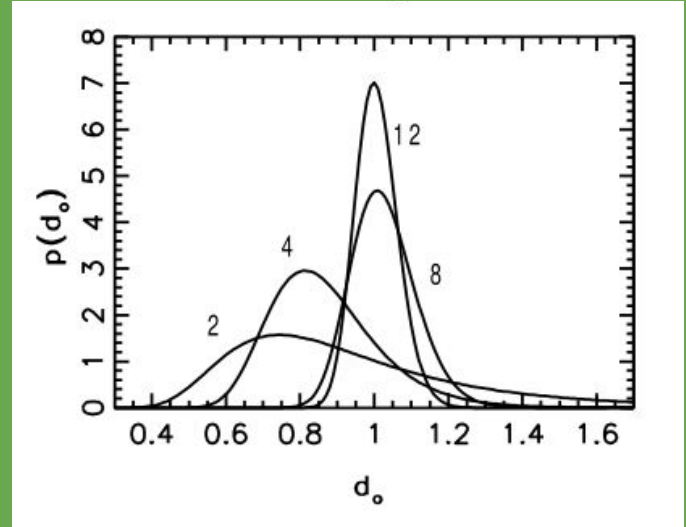
In the two-dimensional case adopted here to analyze the distribution of candidate RR Lyrae stars (because the SDSS Data Release 1 footprint consists of two elongated strips; see Fig. 5), the probability density distribution for  $d_0$ , which is related to the local density via  $n_0 = \pi d_0^2$ , is thus determined from<sup>9</sup>

$$p(d_0|\{d_k; k = 1, N\}) \propto \prod_{k=1}^N \frac{2e^{-(d_k/d_0)^2}}{d_0(k-1)!} \left(\frac{d_k}{d_0}\right)^{2k-1}. \quad (\text{B6})$$

A few steps in a typical realization of this computation are shown in the middle panel in Fig. 9, where  $p(d_0)$  is evaluated for  $N = 2, 4, 8$ , and  $12$ , for a random sample with true  $d_0 = 1$ .

<sup>9</sup> Eq. (B6) can be recast in a form that allows much simplified computation of the expectation value for the local density and its uncertainty (such that the dependences on  $d_k$  and  $d_0$  are separated and the final expressions involve only a summation of  $d_k^D$ , without a need to evaluate the full posterior probability distribution [P. Wozniak 2004, private communication].

## 5a - Bayesian approach to Nearest Neighbours



## 5a - Bayesian approach to Nearest Neighbours

Comments by Woźniak & Kruszewski 2012:

Another way of diminishing the smoothing bias was introduced by Ivezić *et al.* (2005) and Cowan and Ivezić (2008) who take a “Bayesian” approach to combine contributions from all  $N$  nearest neighbors. The net effect is, again, lower bias at the cost of increased variance. Here we propose a new method of dealing with the smoothing bias that captures the information on density variations contained in distances to all  $N$  nearest neighbors using the Legendre series expansion.

# 5b - use normalised volumes and Legendre polynomials

$$\hat{\rho}_{N,k} = \frac{1}{v_N} \sum_{i=1}^{N-1} \sum_{l=0}^k (-1)^l (2l+1) P_l(2y_i - 1)$$

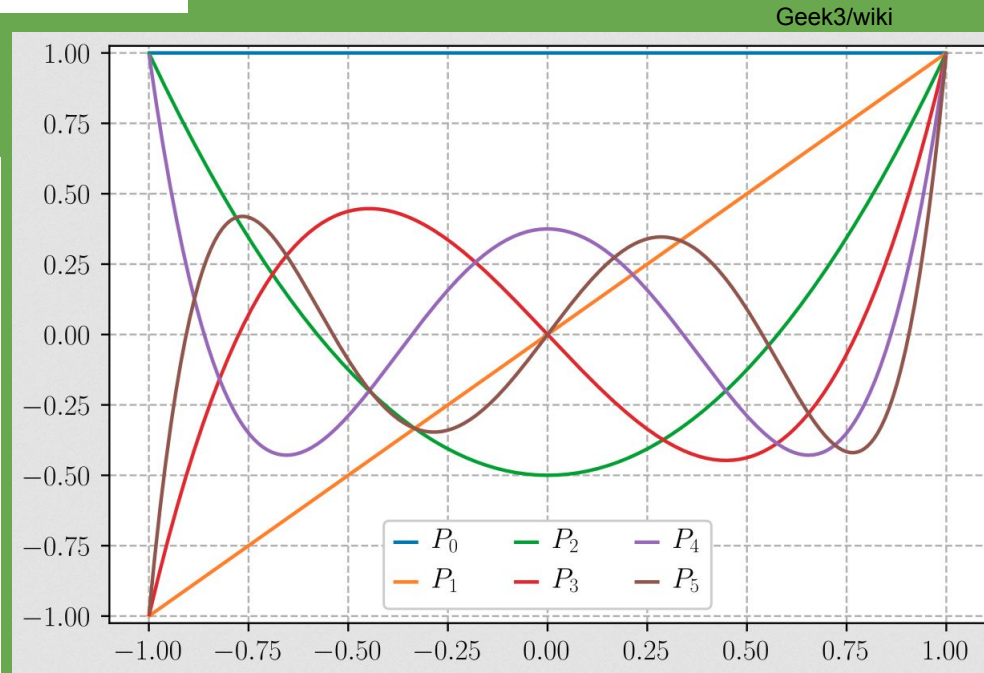
Woźniak & Kruszewski 2012 (though the formula was known years earlier)



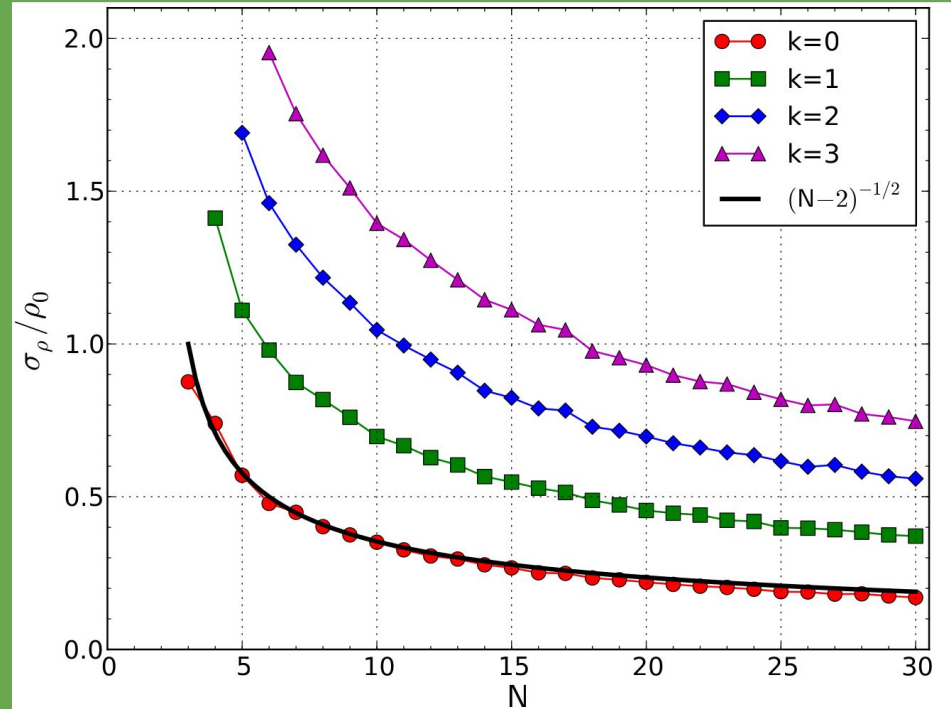
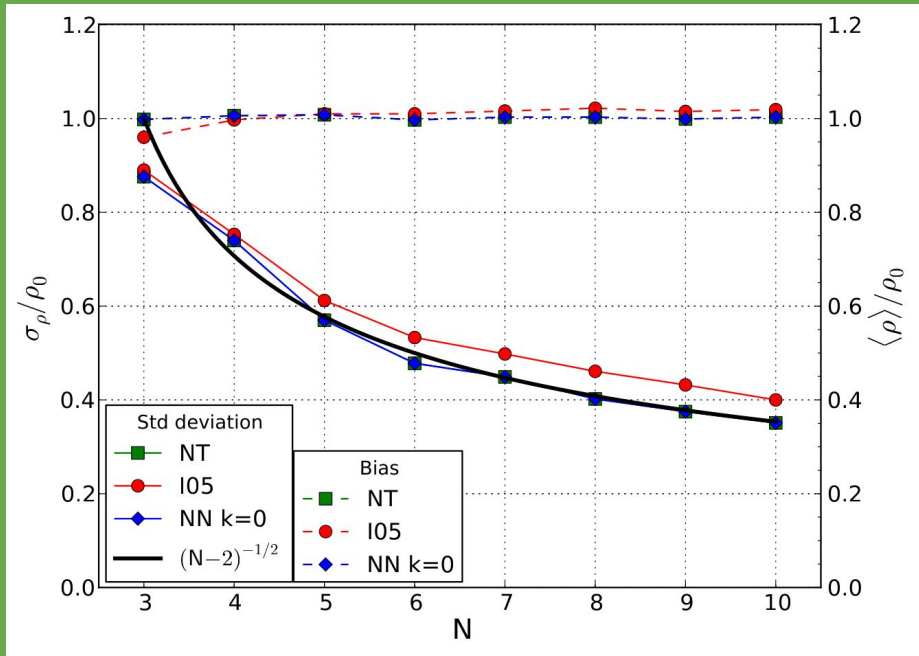
Watercolor caricature by Julien-Léopold Boilly (see [§ Mistaken portrait](#)), the only Authenticated portrait of Legendre<sup>[2]</sup>



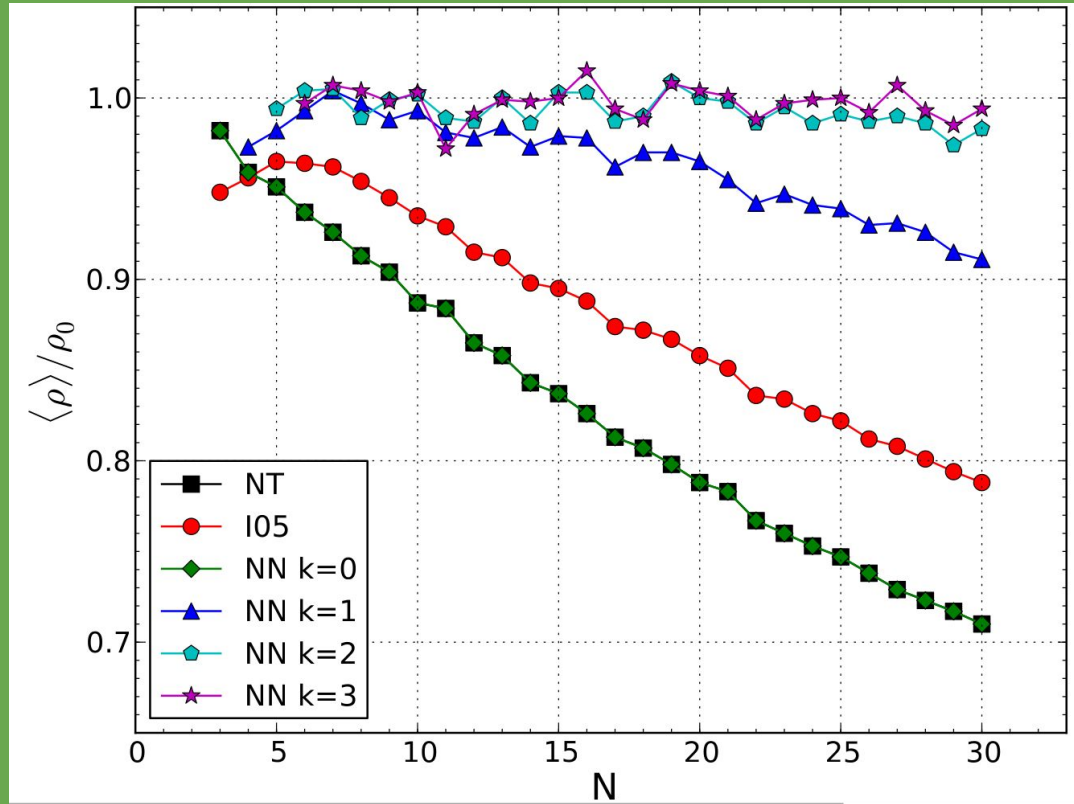
Portrait based on an anonymous and undated sketch, allegedly given by Legendre to [François Arago](#) in 1829, according to Arago's son.



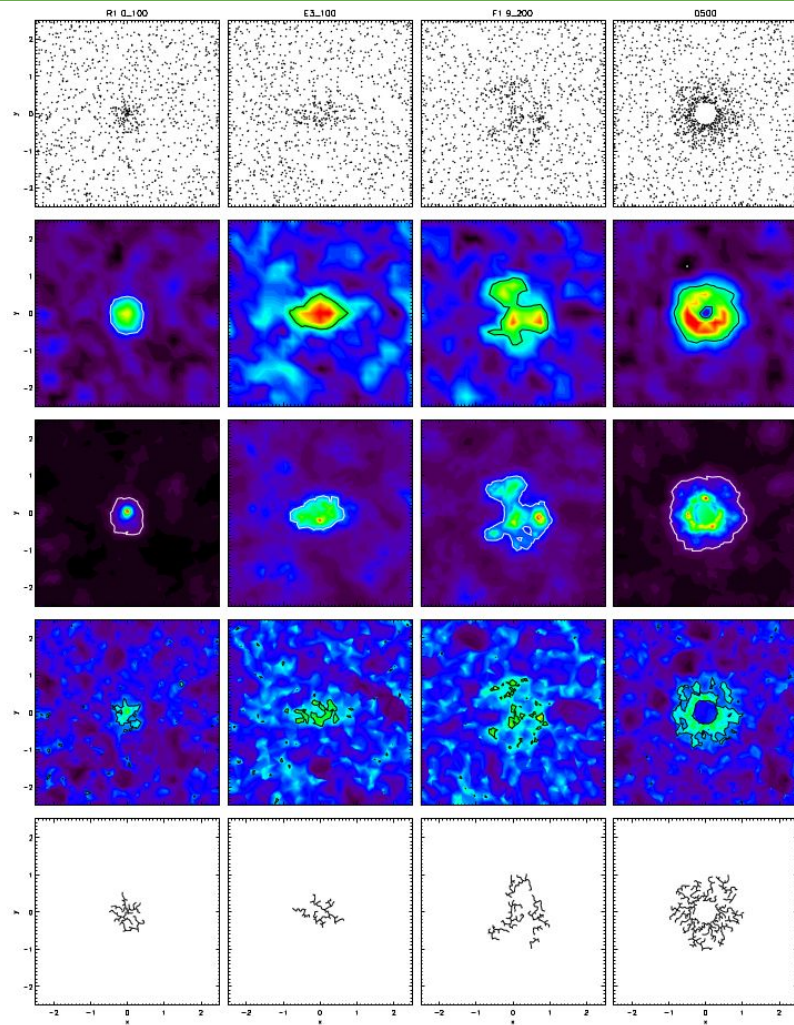
# 5b - use normalised volumes and Legendre polynomials



## 5b - use normalised volumes and Legendre polynomials







| Model    | SC | NN | VT | sVT | MST |
|----------|----|----|----|-----|-----|
| R0.1_50  | —  | —  | —  | —   | —   |
| R0.1_100 | ○  | ○  | —  | —   | —   |
| R0.1_200 | ○  | ○  | —  | ○   | ○   |
| R0.1_500 | ●  | ●  | ○  | ●   | ●   |
| R1.0_50  | ○  | ●  | —  | ○   | ○   |
| R1.0_100 | ●  | ●  | ○  | ○   | ●   |
| R1.0_200 | ●  | ●  | ○  | ●   | ●   |
| R1.5_50  | ●  | ●  | ●  | ●   | ●   |
| R1.5_100 | ●  | ●  | ●  | ●   | ●   |
| R1.5_200 | ●  | ●  | ●  | ●   | ●   |
| E2_50    | —  | ○  | —  | —   | —   |
| E2_100   | ○  | ●  | —  | —   | ○   |
| E2_200   | ●  | ●  | ○  | ○   | ○   |
| E3_50    | ○  | ○  | —  | —   | ○   |
| E3_100   | ●  | ●  | —  | —   | ○   |
| E3_200   | ●  | ●  | ○  | ○   | ●   |
| F1.9_100 | ○  | ○  | —  | —   | ○   |
| F1.9_200 | ○  | ●  | —  | ○   | ○   |
| F1.9_500 | ●  | ●  | ○  | ○   | ○   |
| D100     | ○  | ○  | —  | —   | —   |
| D200     | ○  | ○  | —  | —   | ○   |
| D500     | ●  | ●  | ○  | ○   | ○   |