

# Gaussian mixture models

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# Method of moments

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$$X \sim \frac{1}{k}\mathcal{N}(\mu_1, 1) + \dots + \frac{1}{k}\mathcal{N}(\mu_k, 1)$$

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- Do this via the *method of moments*: i.e. for  $\ell = 1, \dots, k$  solve

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- When  $k = 4$  the moment equations are

$$\bar{m}_1 = \frac{1}{4} \mu_1 + \frac{1}{4} \mu_2 + \frac{1}{4} \mu_3 + \frac{1}{4} \mu_4$$

$$\bar{m}_2 = \frac{1}{4} \mu_1^2 + \frac{1}{4} \mu_2^2 + \frac{1}{4} \mu_3^2 + \frac{1}{4} \mu_4^2 + 1$$

$$\bar{m}_3 = \frac{1}{4} (\mu_1^3 + 3\mu_1) + \frac{1}{4} (\mu_2^3 + 3\mu_2) + \frac{1}{4} (\mu_3^3 + 3\mu_3) + \frac{1}{4} (\mu_4^3 + 3\mu_4)$$

$$\bar{m}_4 = \frac{1}{4} (\mu_1^4 + 6\mu_1^2) + \frac{1}{4} (\mu_2^4 + 6\mu_2^2) + \frac{1}{4} (\mu_3^4 + 6\mu_3^2) + \frac{1}{4} (\mu_4^4 + 6\mu_4^2) + 3$$

We are interested in multiple aspects of the polynomial moment equations:

- ① Find closed form solution for unique solution (up to symmetry) to moment equations
- ② Compute/understand discriminant conditions for various  $k$ . i.e. find conditions where the moment equations don't have  $k!$  regular, isolated points
- ③ Compute/understand *statistically meaningful* regions in the moment space (moments that give rise to real solution  $\mu_1, \dots, \mu_k$ )
- ④ Investigate various projection schemes for projecting sample moments onto statistically meaningful ones
- ⑤ Investigate the problem of how to pick  $k$  from data