

Statical manifold

Gaussians form a manifold. You can use differential geometry to create a distance function between gaussians. Euclidean distance doesn't work for this.

3 approaches to big data problems

- 1) Supervised learning: Assume a parametric model and then find parameters that minimize some loss function that measures the error rate on some labeled data set.
- 2) Probabalistic aprooach: Maximize likelihood of the data
- 3) Geometric method: Geometric methods let us use visual intuition. For the linear regression problem want to make $y \approx X\theta$ where y is the correct outputs vector, X is an input matrix and θ is the vector with the parameters. $X\theta$ is some linear combination of the columns of X . Suppose that X is n by m . The possible outputs for $X\theta$ is $B = span(\text{columns of } X)$. That is an m dimensional subspace of R^n with the columns of X as it's basis. But y lives in R^n . So we have to find the point in B closest to y . You can do that by projecting y onto B . Look at the rate of change of The frame $\{x_u, x_v, N\}$ to detect how the manifold is curved.

Christoffel symbols

$$x_{uu} = \Gamma_1^{11}x_u + \Gamma_2^{11}x_v + L_1N$$

$$x_{uv} = \Gamma_1^{12}x_u + \Gamma_2^{12}x_v + L_2N$$

$$x_{vu} = \Gamma_1^{21}x_u + \Gamma_2^{21}x_v + L_2N$$

$$x_{vv} = \Gamma_1^{22}x_u + \Gamma_2^{22}x_v + L_2N$$

$$N_u = a_{11}x_u + a_{21}x_v$$

$$N_v = a_{12}x_u + a_{22}x_v$$

Then $e = L_1, f = L_2, g = L_3$.

How to find christoffel symbols

Take inner product of both sides with x_u and x_v for the first 4 equations for the christoffel symbols. This gives you a system of equations.

$$\langle x_{uu}, x_u \rangle = \Gamma_1^{11} \langle x_u, x_u \rangle + \Gamma_2^{11} \langle x_v, x_u \rangle$$

$$\langle x_{uu}, x_v \rangle = \Gamma_1^{11} \langle x_u, x_v \rangle + \Gamma_2^{11} \langle x_v, x_v \rangle$$

⋮

Solve that system of equations to find the christoffeel symbols.

Theorem

The gaussian curvature ofa surface is invariattnt by local isometries.