

Motion Lecture 2

In the previous lecture we described the slow-and-smooth model for motion. We now describe how these models can explain the distance fall-off effects which occur in motion capture. We analyze a special case. Note: that this relates to other topics in Machine Learning, such as Radial Basis Functions, Kernel Methods, and Gaussian Processes.

We will do the analysis in 1-dimension for simplicity.

$$E[\hat{v}] = \sum_{i=1}^n (v(x_i) - u_i)^2 + \lambda \int (v(x))'^2 dx + \mu \int \left(\frac{dv(x)}{dx}\right)'^2 dx + \nu \int \left(\frac{d^2v(x)}{dx^2}\right)'^2 dx$$

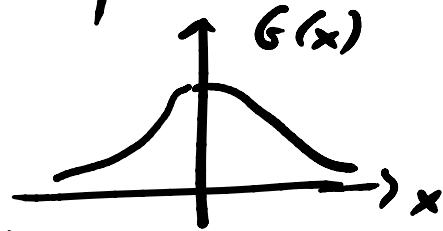
Claim: the solution $\hat{v} = \arg \min E[\hat{v}]$ can be expressed as:

$$\hat{v}(x) = \sum_{i=1}^n d_i G(x - x_i)$$

for a function $G(\cdot)$ and coefficients (d_i) . The function $G(x - x_i)$ is peaked at $x = x_i$.

The function $G(x-x_i)$ is peaked at $x=x_i$ and takes the form.

E.g. $G(\cdot)$ can be a



Gaussian $G(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-x^2/2\sigma^2}$

where the spatial fall-off depends on the standard deviation σ .

sub-claim: we can express the slow-and-smoothness terms as

$$\int v(x) L v(x) dx$$

where $L = \lambda - \mu \frac{d^2}{dx^2} + \nu \frac{d^4}{dx^4}$

is a differential operator.

The Green function G is the solution of the equation:

$$L G(x) = \delta(x)$$

↳ impulse function at $x=0$

Green functions are used to solve differential equations

$$L v(x) = p(x)$$

has a solution $v(x) = \int G(x-x') p(x') dx'$

How to obtain L ?

$$\mu \int \left(\frac{dv(x)}{dx} \right)^2 dx = \mu \int \frac{d}{dx} \left(v(x) \frac{dv(x)}{dx} \right) dx - \mu \int v(x) \frac{d^2 v(x)}{dx^2} dx$$

The term $\int \frac{d}{dx} \left(v(x) \frac{dv(x)}{dx} \right) dx$
 can be integrated $v(b) \frac{dv(b)}{dx} - v(a) \frac{dv(a)}{dx}$.

Assume that $v(x) = 0$ at the boundaries.

Hence we can rewrite

$$\mu \int \left(\frac{dv(x)}{dx} \right)^2 dx = - \mu \int v(x) \frac{d^2 v(x)}{dx^2} dx$$

Similarly

$$\nu \int \left(\frac{d^2 v(x)}{dx^2} \right)^2 dx = \nu \int v(x) \frac{d^4 v(x)}{dx^4} dx$$

provided we use boundary conditions
 $v(x) \rightarrow 0, \frac{dv(x)}{dx} \rightarrow 0$ on the boundaries.

To determine the α 's, substitute

$$v(x) = \sum_{i=1}^N \alpha_i G(x-x_i) \text{ into}$$

$$\sum_{i=1}^N (v(x_i) - u_i)^2 + \int v(x) L v(x) dx$$

$$v(x_i) = \sum_{j=1}^N \alpha_j G(x_i - x_j)$$

$$\dots = \int v(x) \sum_{i=1}^N \alpha_i G(x-x_i) dx$$

$$\int v(x) L v(x) dx = \int v(x) \sum_{i=1}^N d_i \delta(x-x_i) dx$$

$$= \sum_{i=1}^N d_i v(x_i) = \sum_{i,j=1}^N d_i d_j G(x_i-x_j)$$

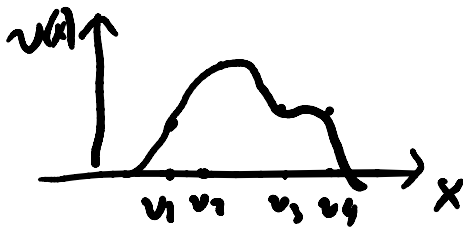
Hence we solve for \hat{v} by minimizing

$$E[\hat{v}] = \sum_{i=1}^N \left(\sum_{j=1}^N d_j G(x_i-x_j) - v_i \right)^2 + \sum_{i,j=1}^N d_i d_j G(x_i-x_j).$$

Summary

Solution $v(x) = \sum_{i=1}^N d_i G(x-x_i)$

where $G(\cdot)$ is the Green function of the differential operator L . The d 's are found by minimizing $E[\hat{v}]$.



The effect is to smooth the velocity measurements $\{v_1, \dots, v_N\}$, provided they are closer than ϵ .