

SDLU simulation on different number of nodes with residual fitting

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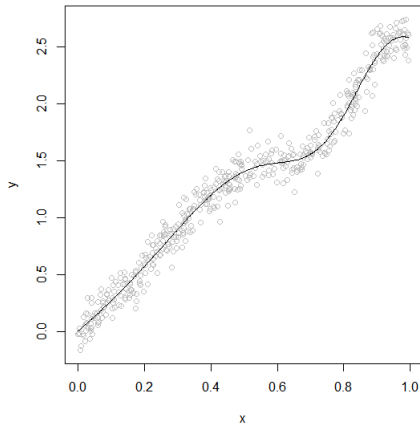
Review and Goal

- Last meeting
 - Conducted SDLU simulation on different number of nodes with constraint on width of activation function.
- Goal for today's meeting
 - SDLU simulation with residual fitting on different number of nodes and compare results.

Settings for simulation

- Size of data: 500
- x : 500 data randomly chosen from uniform distribution $U(0,1)$
- f (true function): $2.5x + 0.2\sin((3x)^2)$
- y : $f + 0.1\varepsilon$ where $\varepsilon \sim N(0,1)$
- Widths: Gradually decrease the width of activation function.
($[1,2] \rightarrow [0.5,1] \rightarrow [0.25,1] \rightarrow \dots$)
- No Lasso penalty given

$$f(x) = 2.5x + 0.2\sin((3x)^2)$$



Residual fitting process

- Sequentially fit model to residuals of previous model in decreasing order with respect to the width of nodes

$$y \sim x \rightarrow f_1$$

$$y - f_1 \sim x \rightarrow f_2$$

$$(y - f_1) - f_2 \sim x \rightarrow f_3$$

$$\vdots$$

$$(y - f_1 - f_2 - f_3 - f_4) - f_5 \sim x \rightarrow f_6$$

$$f = f_1 + \dots + f_6$$

First Simulation

The number of nodes assigned to each step:

- Baseline: 4 constraint : $[1, 2]$
- 1st residual fitting: 8 constraint : $[1, 2]/2$
- 2nd residual fitting: 16 constraint : $[1, 2]/4$
- 3rd residual fitting: 32 constraint : $[1, 2]/8$
- 4th residual fitting: 64 constraint : $[1, 2]/16$
- 5th residual fitting: 128 constraint : $[1, 2]/32$

Total: 252 nodes

Simulation animation

Second Simulation

The number of nodes assigned to each step:

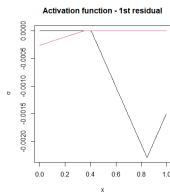
- Baseline: 42 constraint : $[1, 2]$
- 1st residual fitting: 42 constraint : $[1, 2]/2$
- 2nd residual fitting: 42 constraint : $[1, 2]/4$
- 3rd residual fitting: 42 constraint : $[1, 2]/8$
- 4th residual fitting: 42 constraint : $[1, 2]/16$
- 5th residual fitting: 42 constraint : $[1, 2]/32$

Total: 252 nodes

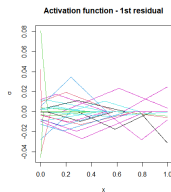
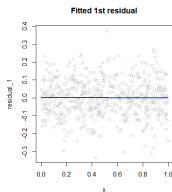
Simulation animation

Comparison of residual activation function - width $[1,2]/2$

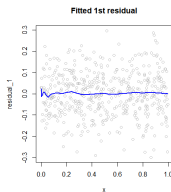
Figure: (a) Total 12 nodes(8 nodes) (b) Total 84 nodes(42 nodes)



(a)

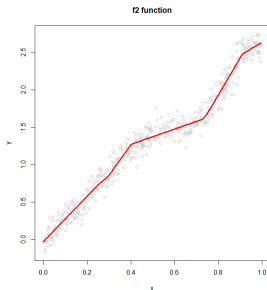


(b)

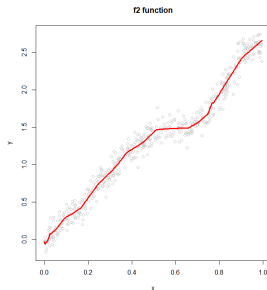


Comparison of fit - width $[1,2]/2$

Figure: (a) Total 12 nodes(8 nodes) (b) Total 84 nodes(42 nodes)



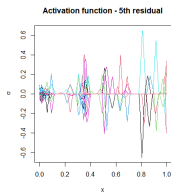
(a)



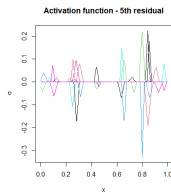
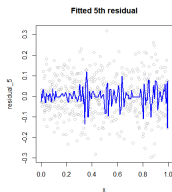
(b)

Comparison of residual activation function - width $[1, 2]/32$

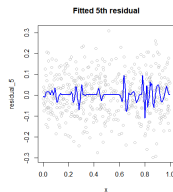
Figure: (a) Total 256 nodes(128 nodes) (b) Total 256 nodes(42 nodes)



(a)

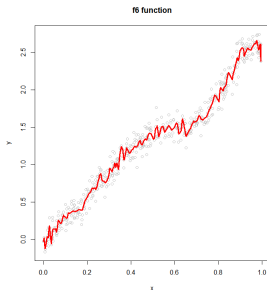


(b)

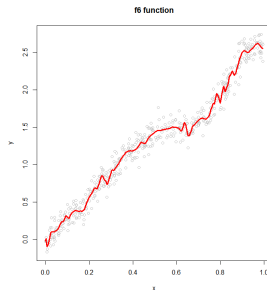


Comparison of fit - width $[1,2]/32$

Figure: (a) Total 256 nodes(128 nodes) (b) Total 256 nodes(42 nodes)



(a)



(b)

Results summary

- Large width
 - No need to use large number of nodes to fit residuals when the width of activation function is big.
- Narrow width
 - Well captured overall locality of data when many nodes are used to fit residuals.