# SDLU simulation on different number of nodes with constraint

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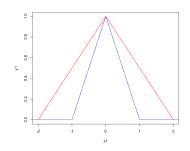
#### Background information about SDLU

ullet SDLU model for estimating true function f

$$f(x;\theta) = \beta_0 + \sum_{m=1}^{M} \beta_m \sigma(\alpha_{0m} + \alpha_{1m} x)$$

where  $x \in \mathbb{R}$ 

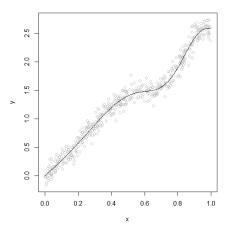
- *M*: the number of nodes
- ullet  $\frac{2}{|lpha_{1m}|}$ : width of B-spline activation function



#### Settings for simulation

- Size of data: 500
- x: 500 data randomly chosen from uniform distribution U(0,1)
- $f(\text{true function}): 2.5x + 0.2\sin((3x)^2)$
- y:  $f + 0.1\varepsilon$  where  $\varepsilon \sim N(0,1)$
- List of number of nodes: 4, 8, 16, 32, 64, 128
- Widths: Generate widths from uniform distribution U(1,2), U(0.25,0.5), U(0.0625,0.125)
- No Lasso penalty given

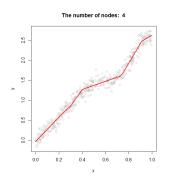
## $f(x) = 2.5x + 0.2\sin((3x)^2)$

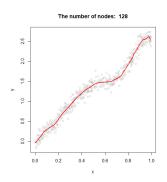


## Test with width [1,2]

- For all number of nodes(4, 8, 16, 32, 64, 128), estimated function captures global trend of data.
- There is no such number of nodes that captures locality of data.

Figure: Comparison of fit depending on the number of nodes

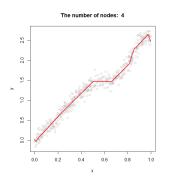


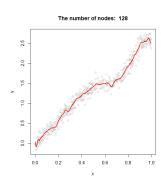


## Test with width [0.25, 0.5]

- Small number of nodes(4, 8, 16, 32) well captures global trend of data.
- Estimated function starts to overfit to data when the number of nodes exceeds 64.

Figure: Comparison of fit depending on the number of nodes



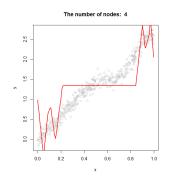


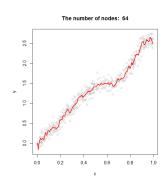
### Test with width [0.0625, 0.125] with animation

## Test with width [0.0625, 0.125]

- When the number of nodes are small(4, 8), estimated function fails to capture global trend of data. Rather, it only captures locality at certain points.
- As the number of nodes increases, there is a tendency of overfitting.

Figure: Comparison of fit depending on the number of nodes



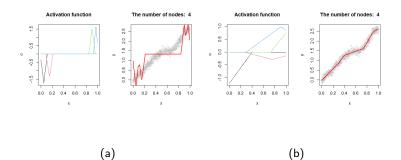


#### Conclusion

- Overall fit of data shows more interpretable patterns when constraint is given to width of B-spline activation function.
- If the width of activation function is small, we need sufficiently many nodes to capture global trend of data.
- If the width of activation function is relatively large, small number of nodes are enough to capture global trend.

#### Conclusion

Figure: (a) Narrow width (b) Large width



#### To-do

- Fit SDLU model with residual fitting method and analyze results.
- Compare the effect of different number of nodes when residual fitting is added.