

July 20 Update

Simulating a Modulated Signal

From Kian's thesis we have the following formula for signals with frequency modulation:

$$X_t = \sum_m \mu_m \cos(2\pi f_m t + 2\pi \int_0^t \phi_m(\tau) d\tau) + Z_t$$

Now let $\theta(t) = 2\pi f_m t + 2\pi \int_0^t \phi_m(\tau) d\tau$. Since ϕ and θ are both polynomials, we can simplify to:

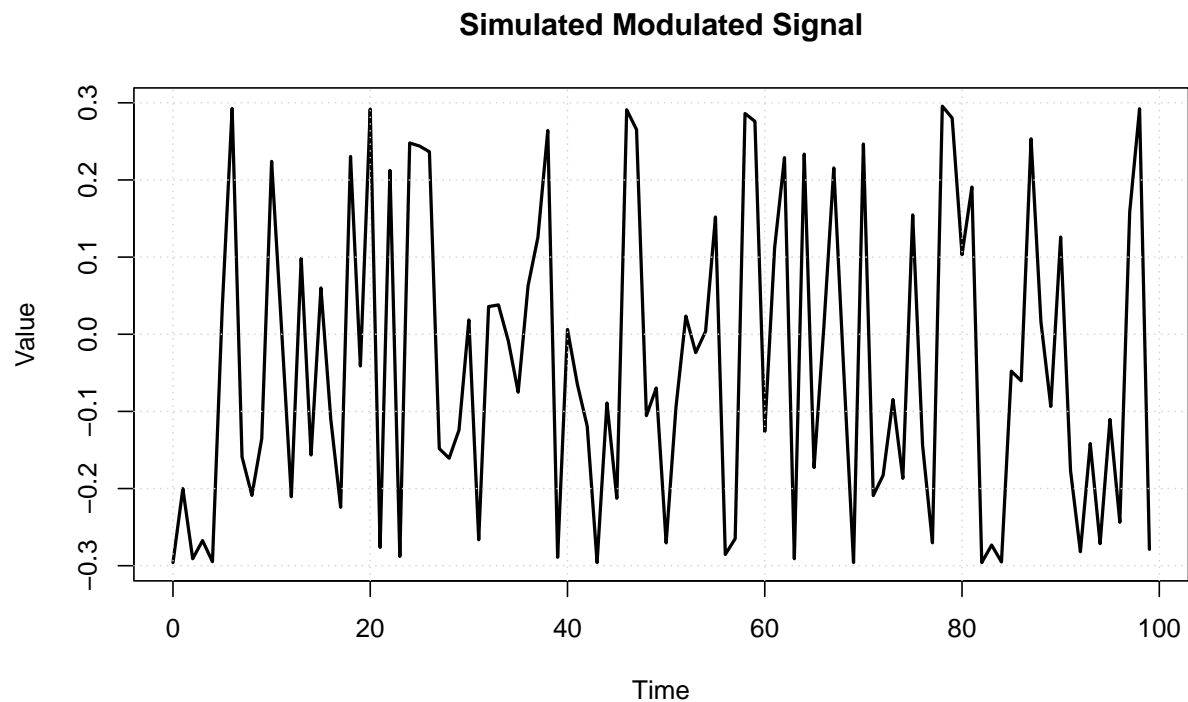
$$\begin{aligned}\phi(\tau) &= \sum_{p=0}^P a_p \tau^p \\ \theta(t) &= 2\pi f_m t + 2\pi \int_0^t \sum_{p=0}^P a_p \tau^p d\tau \\ \theta(t) &= 2\pi f_m t + 2\pi \sum_{p=0}^P \frac{a_p}{p+1} t^{p+1}\end{aligned}$$

Using these formulas I wrote a function to simulate a modulated signal given a length N , number of frequencies m , and the maximum degree of polynomial to consider P .

```
## Simulating modulated signal
set.seed(11)
Tt = simTt_mod(N = 100, numFreq = 1, P = 2)

## Printing the corresponding summation
print(Tt$fn)
```

```
## [1] "(-0.296)*cos(2*pi*(1.635*t + (-1.098)*t^1 + (0.4235)*t^2 + (-0.09233333333333333)*t^3 + 0))"
```



To generate values for μ , f , and a we use a similar idea to the `interpTools` “`simTt`” function:

```
## Defining "Fourier Frequency"
# fourierFreq = 2*pi/N

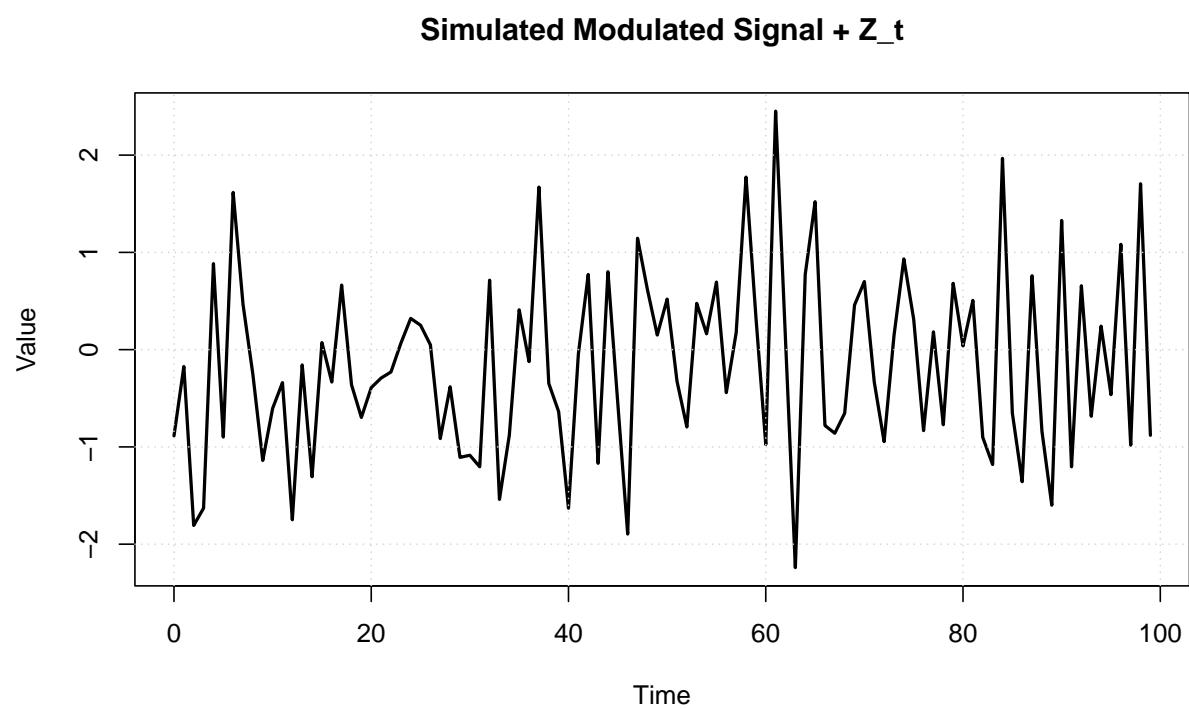
## Generating cosine coefficients
# mu = rnorm(1, mean = 0, sd = N/200)

## Generating frequencies
# f = runif(1, fourierFreq, pi)

## Generating "a" coefficients
# a = rnorm(P, mean = 0, sd = N/200)
```

Now using the `interpTools` “`simWt`” function, we can add some weakly-stationary noise to account for Z_t :

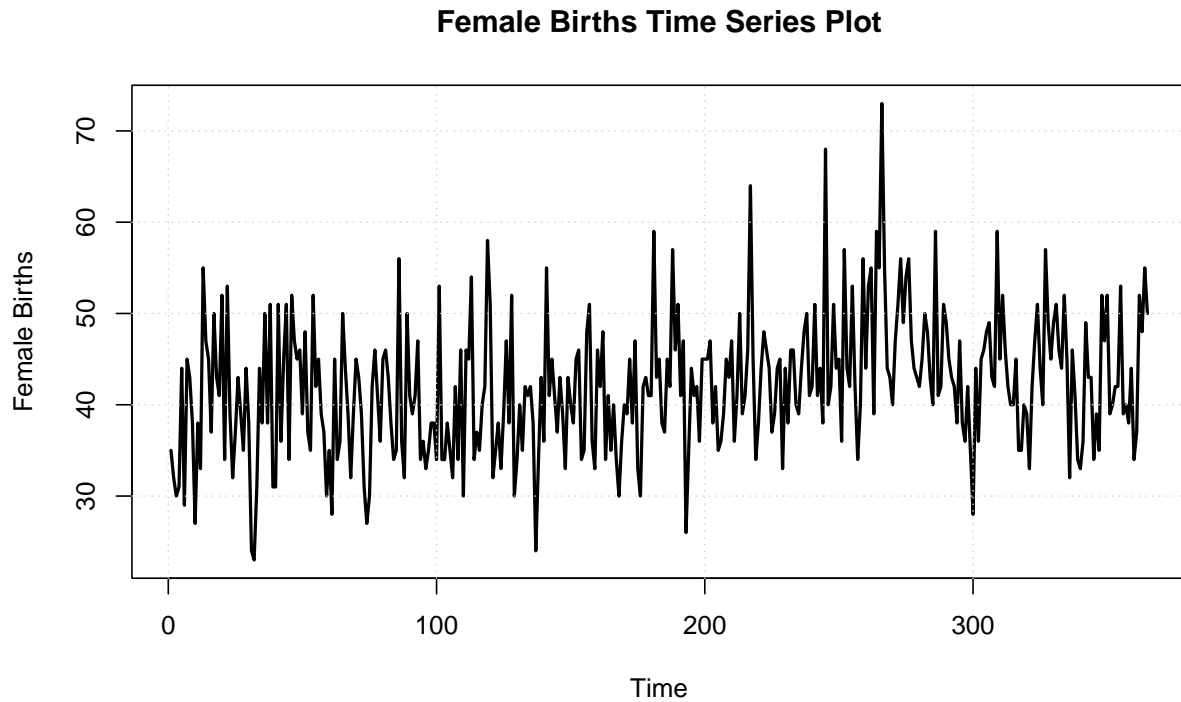
```
## Simulating noise from interpTools
set.seed(11)
Wt = interpTools::simWt(N = 100)$value
```



Simulation Results

- Used three real-world time series
- Used one time series tailored for the Hybrid Weiner method
- Used one time series with a modulated signal
- Considered missing proportion (P) in [10%, 20%, 30%]
- Considered gap width (G) in [10, 25, 50]
- Each method is tested with 25 unique combinations of P and G
- Displayed results are averaged over the 25 iterations
- Considered log-cosh loss (LCL), MAE, and RMSE for evaluation
- Note: Neural nets are designed to minimize RMSE in training

Data Set 1:



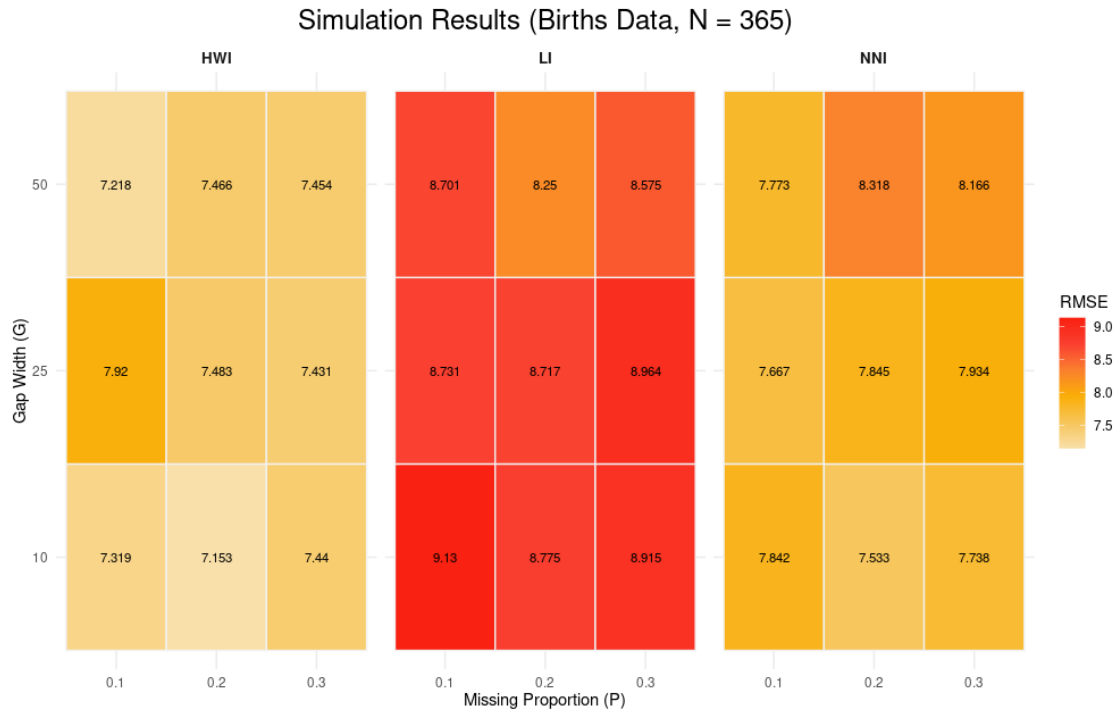


Figure 1: Births data RMSE performance

Data Set 2:

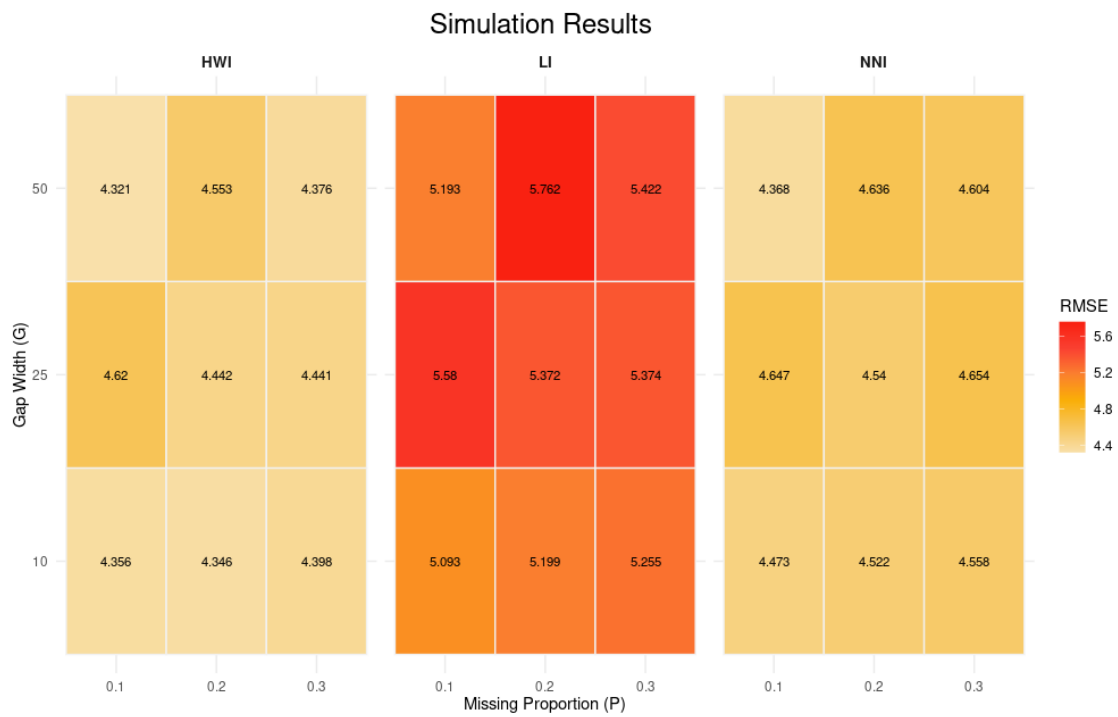
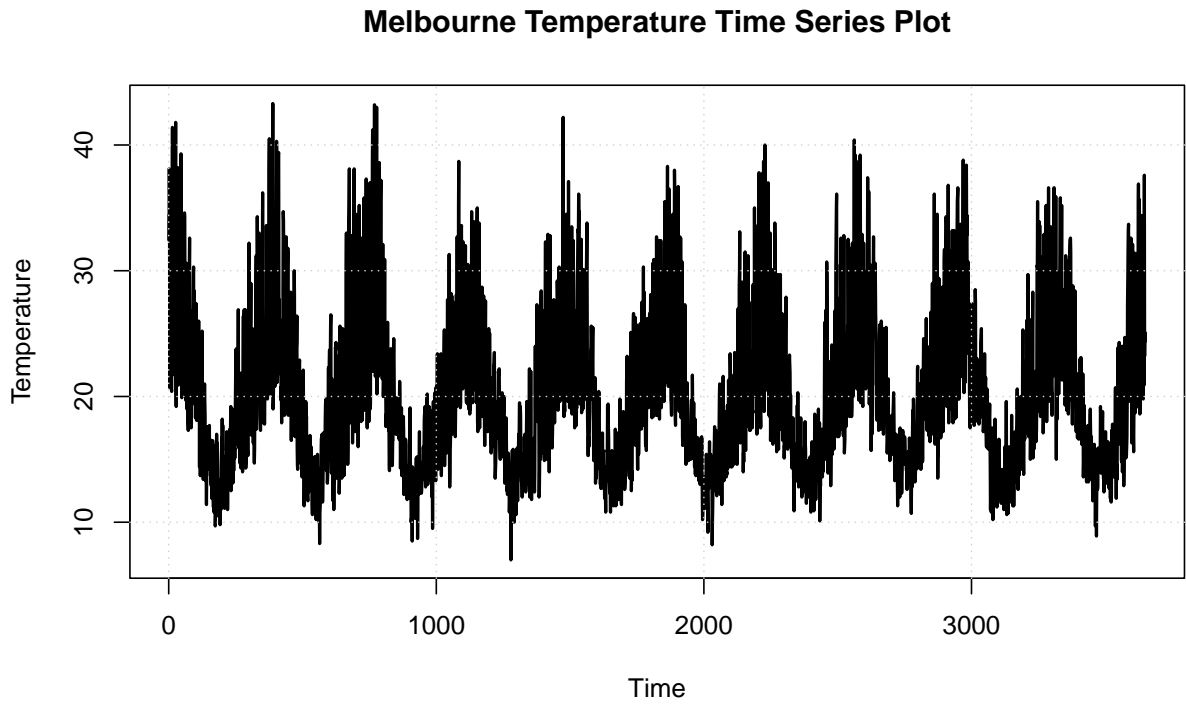


Figure 2: Temperature data RMSE performance

Data Set 3:

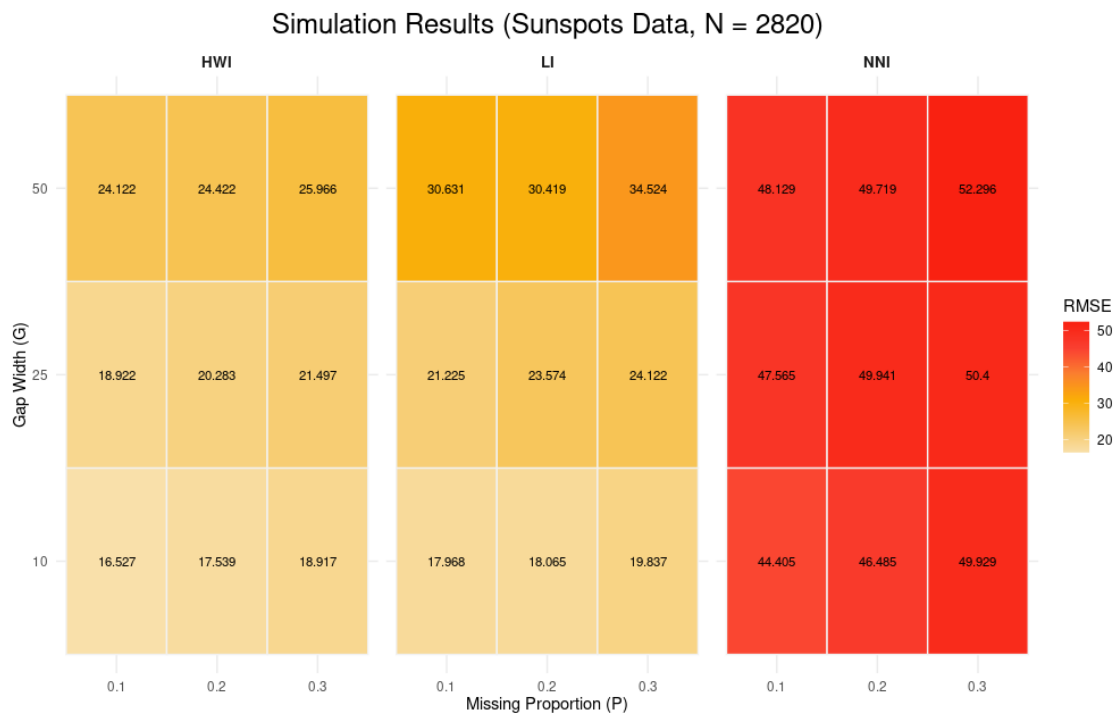
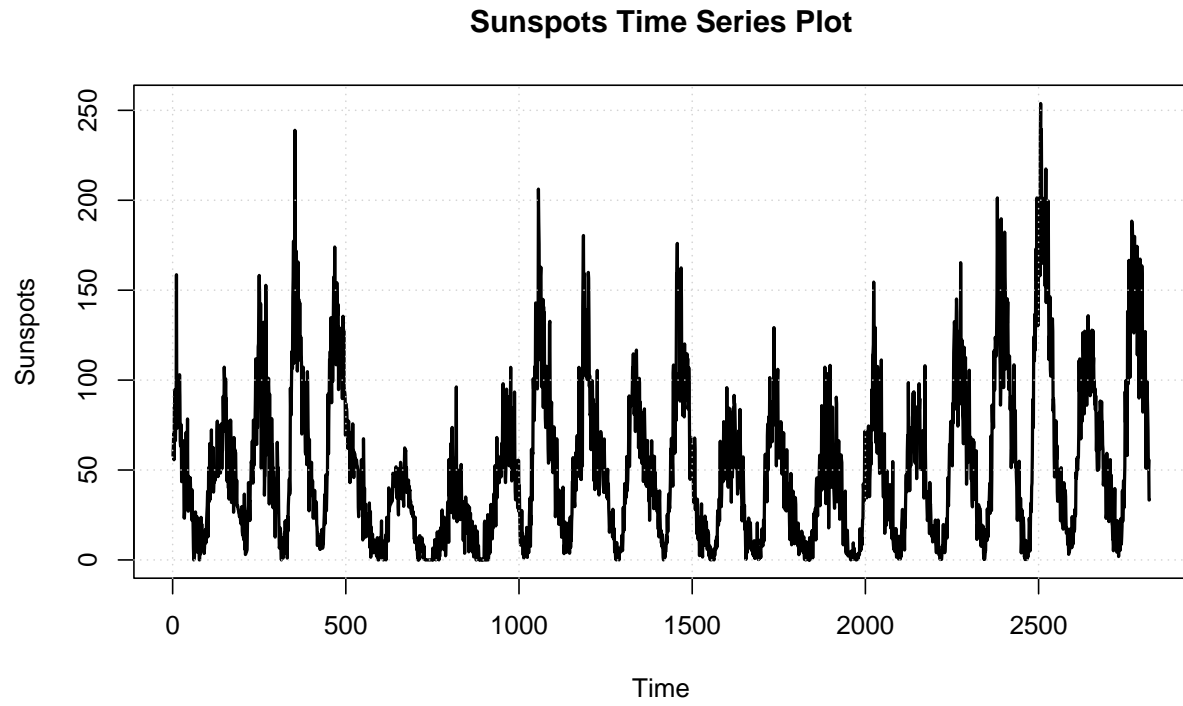


Figure 3: Sunspots data RMSE performance

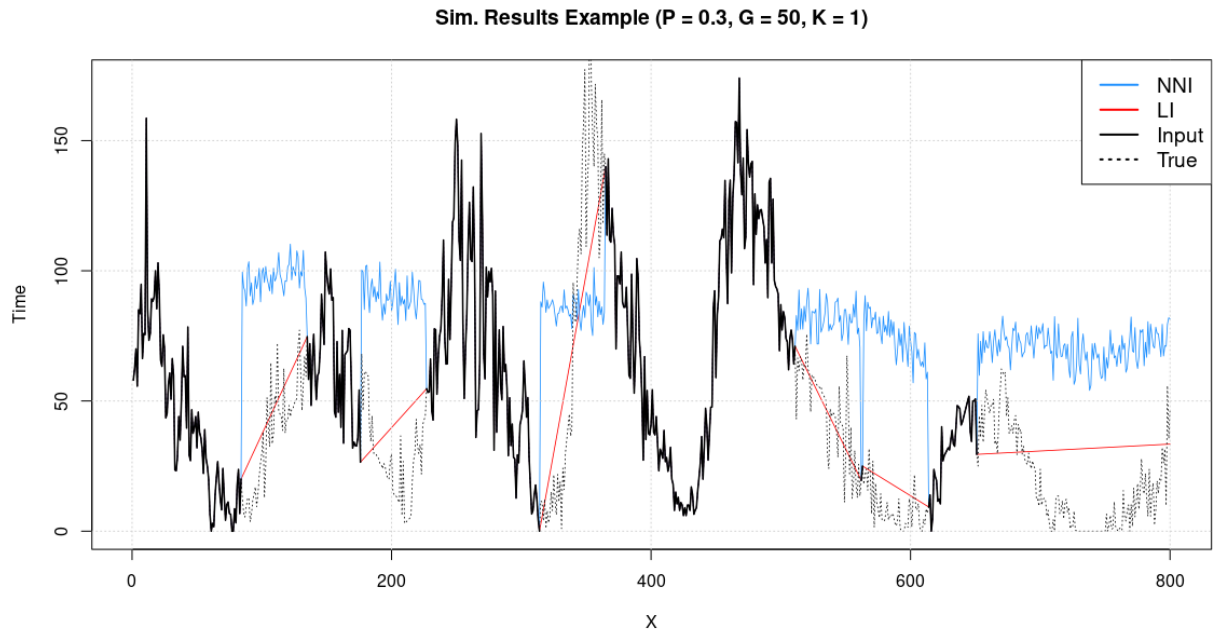


Figure 4: Sunspots performance example

Data Set 4:

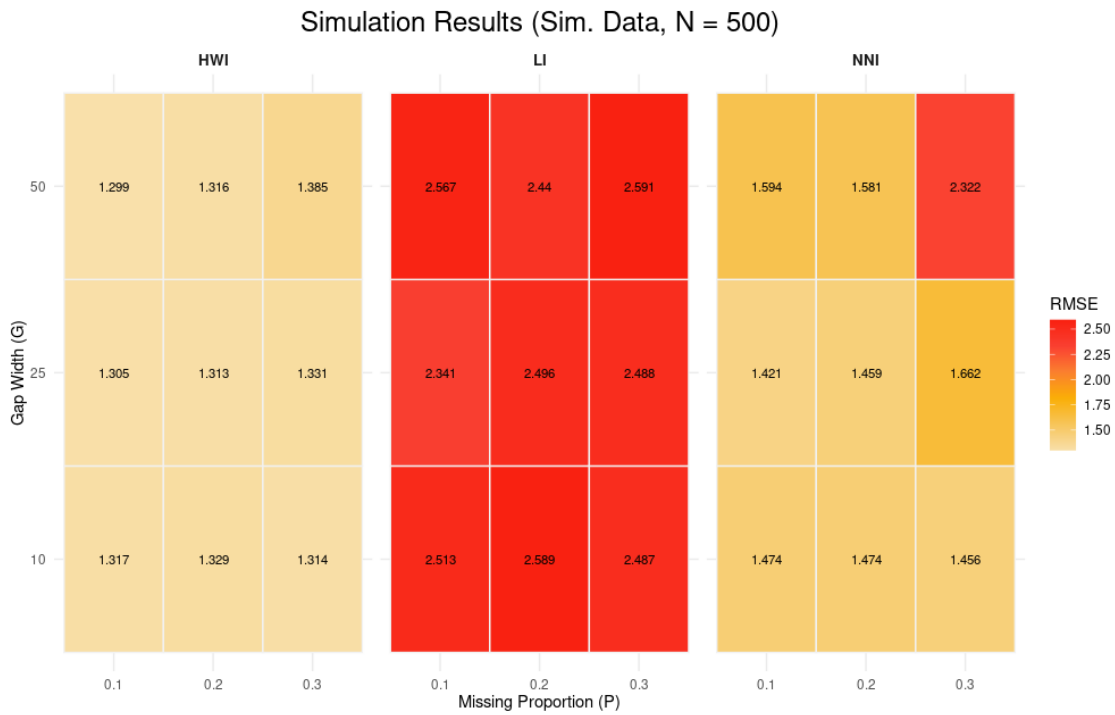
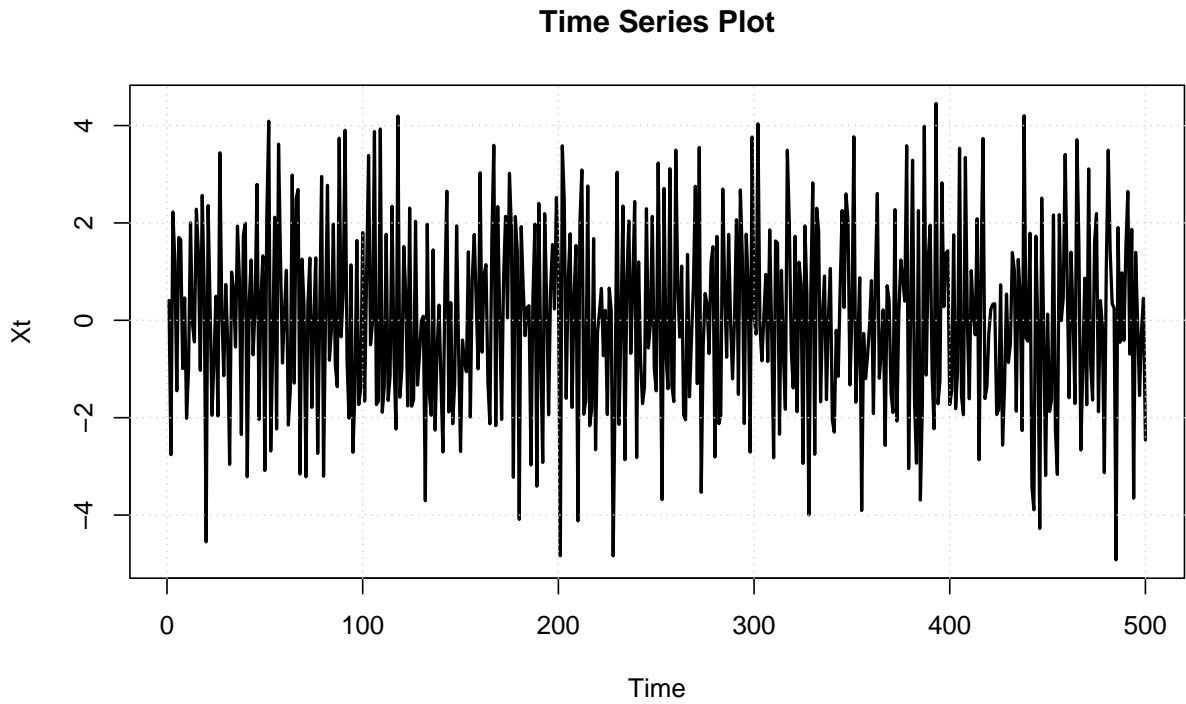


Figure 5: Simulated data RMSE performance

Data Set 5:

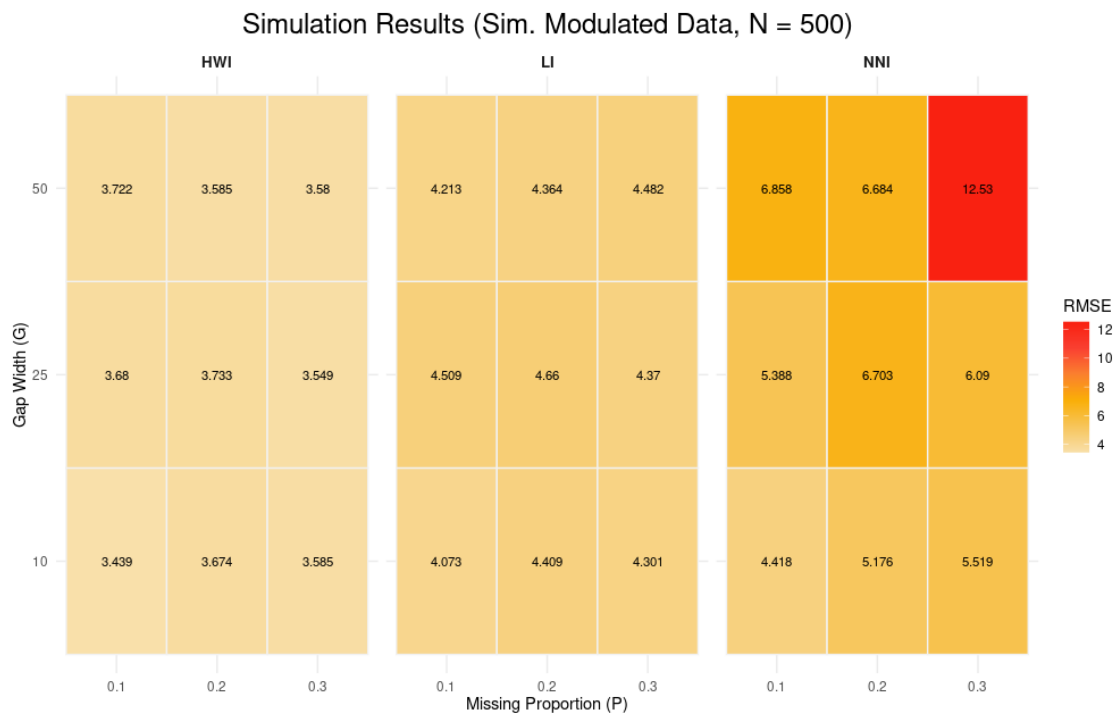
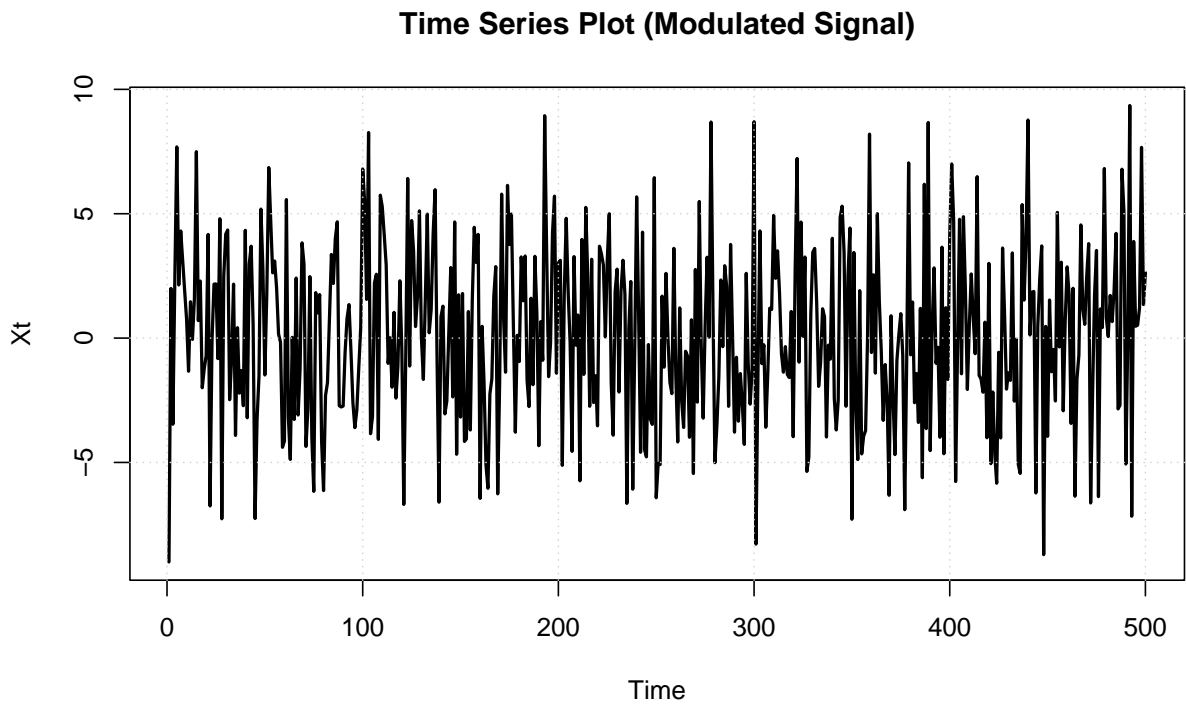


Figure 6: Simulated modulated data RMSE performance