#### An assessment of hydrological models commonly used in NZ

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## Scope

- 1. Develop a method of scoring the calibration fit of hydrological models
  - Need to assess the quality of fit
  - Also assess the **consistency of fit**
  - To do this we built an Excel tool
- 2. Assess how different models perform in NZ catchments
  - Used 10yrs of real data across three catchments to compare models to observed flow.



## **Hydrological Models Tested**

- Horton (linked to soil moisture deficit (SMD) similar to MIKE / ICM)
- Curve Number (SCS-CN)
- Fixed % run-off

## Methodology

- Spreadsheet model of the three hydrological models.
- Consistent run-off routing model and time of concentration (Tc) estimate.
- Fit scored consistently across three calibration events for each river system.
- Best fit obtained using a multi-start non-linear solver.
- Scoring and variance automated.

## **Assessing Visual Fit**

- R<sup>2</sup> can give misleading results explains variance from a linear regression model – not goodness of fit!
- RMSE based on the 'residual' so a better method
  0 = perfect fit. Measures peak error.
- NSE developed for this purpose 1 = perfect fit. Measures overall volume difference.
- Hwang et al recommend combining two methods e.g. NSE and RMSE.
- Above approach tested.

## **Scoring Methodology**

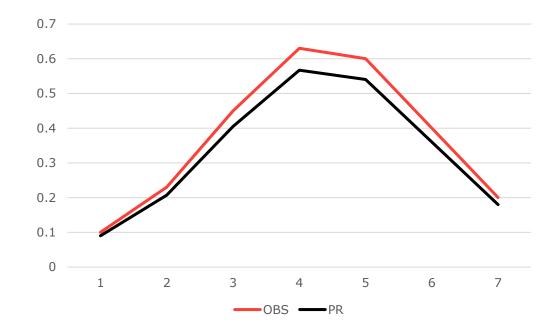
- Nash-Sutcliffe Efficiency (NSE) developed for hydrological calibration. Measures overall fit (and thus volume).
- RMSE as a measure of peak error.
- The two methods are combined for an overall score (volume and peak).
  - < 0 = unacceptable
  - 0.0 0.2 = poor calibration
  - 0.2 0.4 = moderate calibration
    - 0.4 0.6 = good calibration
    - > 0.6 = excellent calibration

 $NSE - (RMSE \div Q_M)$ 

 $Q_M = mean flow$ 

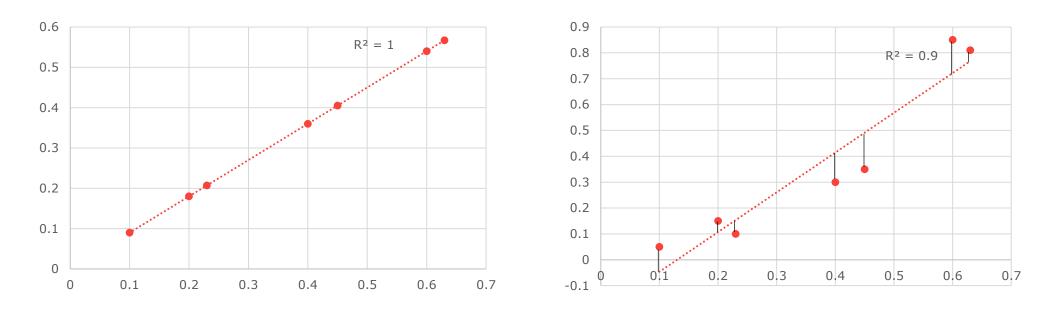
## **NSD**

### **Assessing Visual Fit – Off by Constant Factor**



R<sup>2</sup> = 1.0 NSE = 0.95 RMSE = 0.11 m<sup>3</sup>/s (10%) Score = 0.84 (excellent fit)

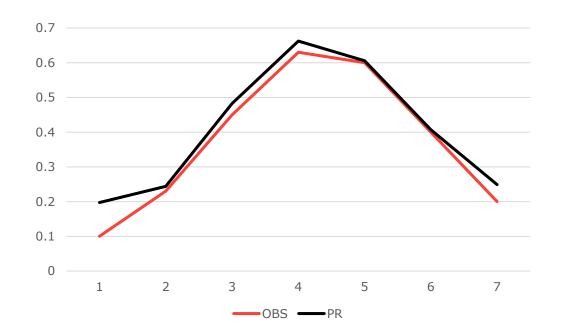
### **Assessing Visual Fit**



How does  $R^2 = 1$  when it is not a perfect model? Because it describes the variance explained by a simple linear regression model and does not need to mean y = x (i.e. observed = predicted)

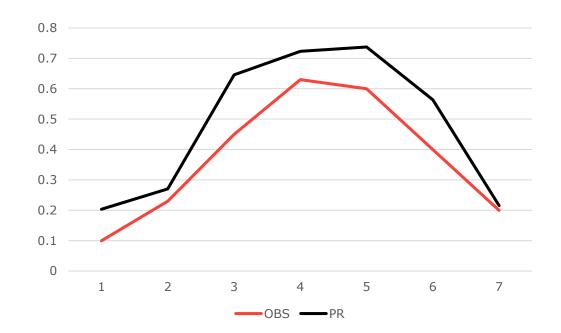
If a model is out massively, but by a constant % it will have  $R^2 = 1$  as the data can be 100% explained by a linear regression model of the form y = mx + c. Hence not a good criteria to use!

#### **Assessing Visual Fit – Good Fit**



 $R^{2} = 0.98$  NSE = 0.88  $RMSE = 0.18 \text{ m}^{3}/\text{s} (15\%)$ **Score = 0.71 (excellent fit)** 

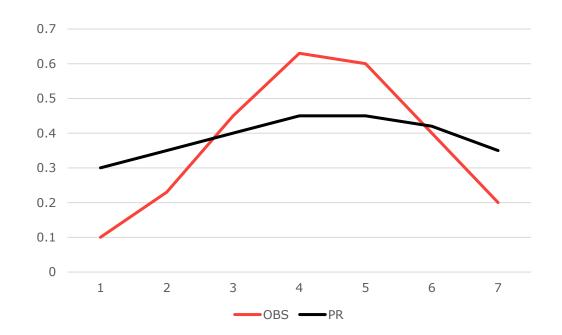
#### **Assessing Visual Fit – Poor Fit**



 $R^{2} = 0.94$  NSE = 0.62  $RMSE = 0.31 \text{ m}^{3}/\text{s} (24\%)$ **Score = 0.31 (poor fit)** 

R<sup>2</sup> is high, but model is a poor fit for the data.

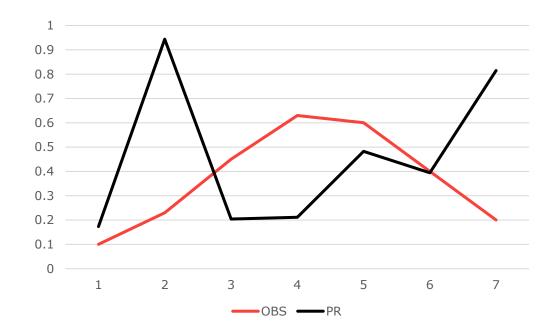
#### **Assessing Visual Fit – Poor Fit**



 $R^{2} = 0.95$  NSE = 0.46  $RMSE = 0.37 \text{ m}^{3}/\text{s} (29\%)$ **Score = 0.09 (poor fit)** 

R<sup>2</sup> is high, but model is a poor fit for the data.

#### **Assessing Visual Fit – Random**



R<sup>2</sup> = 0.07 NSE = -3.59 RMSE = 1.08 m<sup>3</sup>/s (46%) Score = -4.67 (no calibration)

### Limitations

- Spikes (e.g. pumped flow) smooth data to remove spikes
- Data drop outs could add logic to equation
- Not many "small" catchments have rain gauges (RG) and flow monitors (FM) with many years of data (for low AEP events need 10+ years of data in ideally 5 minute increments)
- Relying on one RG opens up issues. In practice RG often near FM so not in centre of catchment
- We analysed three events to avoid overfitting to a specific event.



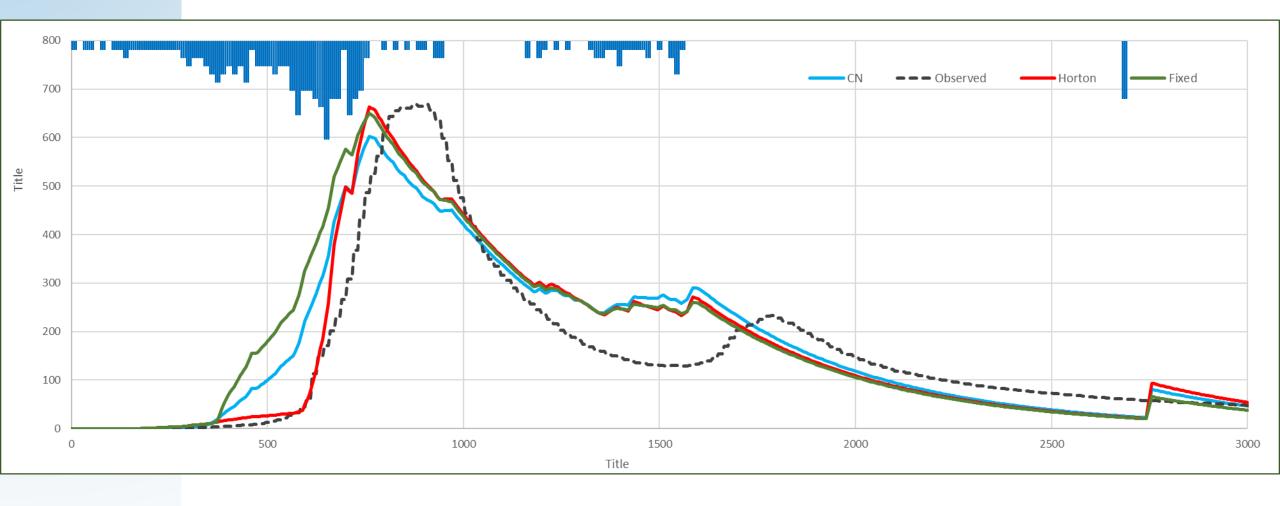
### **Hutt River**

- 450 km<sup>2</sup> (approx.)
- 450 minute Tc
- Flow gauged at Taita Gorge
- Rainfall gauge near flow gauge used

*Too large to be represented with one rain gauge and one catchment in practice* 

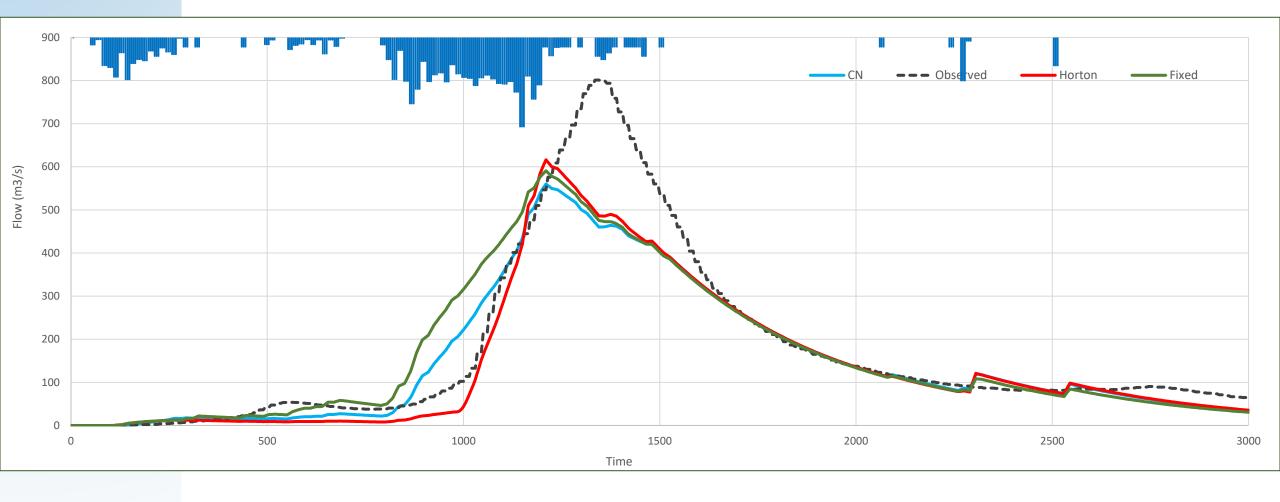


#### Hutt River – Event 1



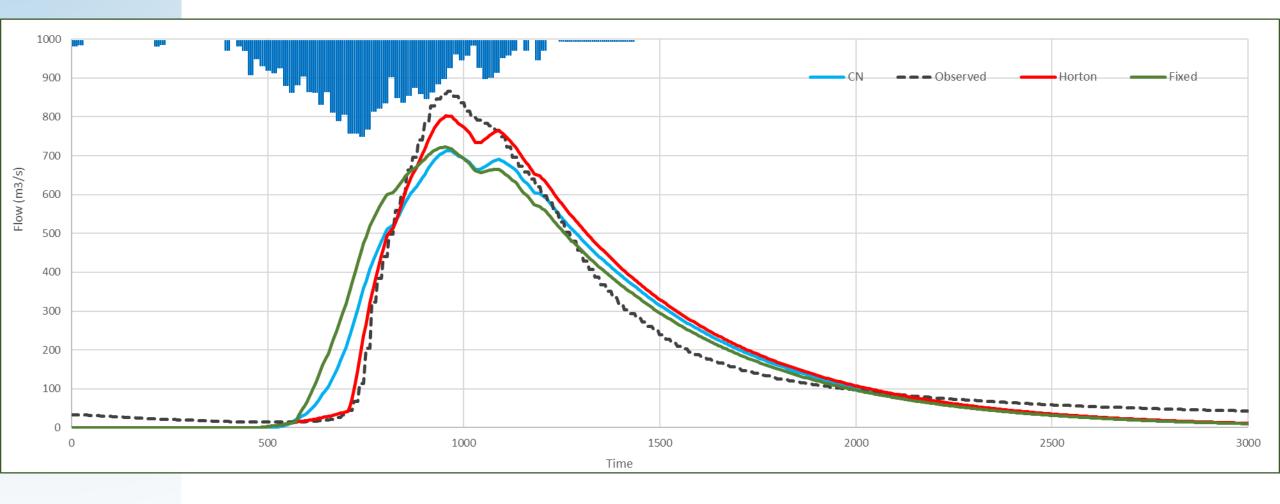


#### Hutt River – Event 2





#### Hutt River – Event 3





#### **Hutt River – Score**

Model Type	Mean Score	Range	Result
CN	0.44	0.35	Good
Horton	0.55	0.33	Good
Fixed	0.30	0.44	Poor

- < 0 = unacceptable
- 0.0 0.2 = poor calibration
- 0.2 0.4 = moderate calibration
- 0.4 0.6 = good calibration
- > 0.6 = excellent calibration

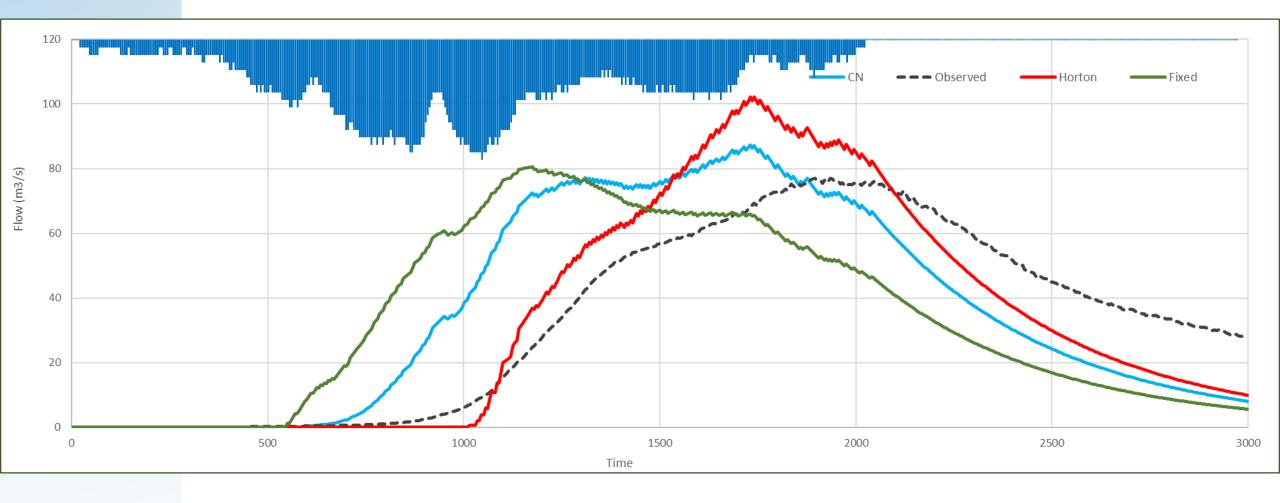


## **Selwyn River**

- 164 km<sup>2</sup> (approx.)
- 456 minute Tc
- Flow gauged at Whitecliffs
- Rainfall gauge near flow gauge used

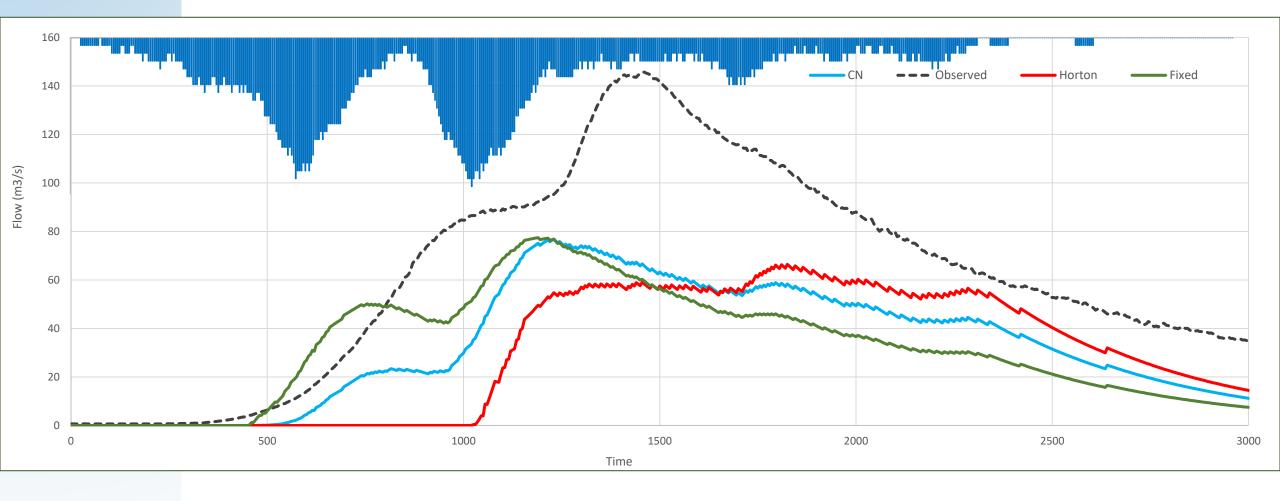
*Too large to be represented with one rain gauge and one catchment in practice* 

Selwyn at Whitecliffs – Event 1

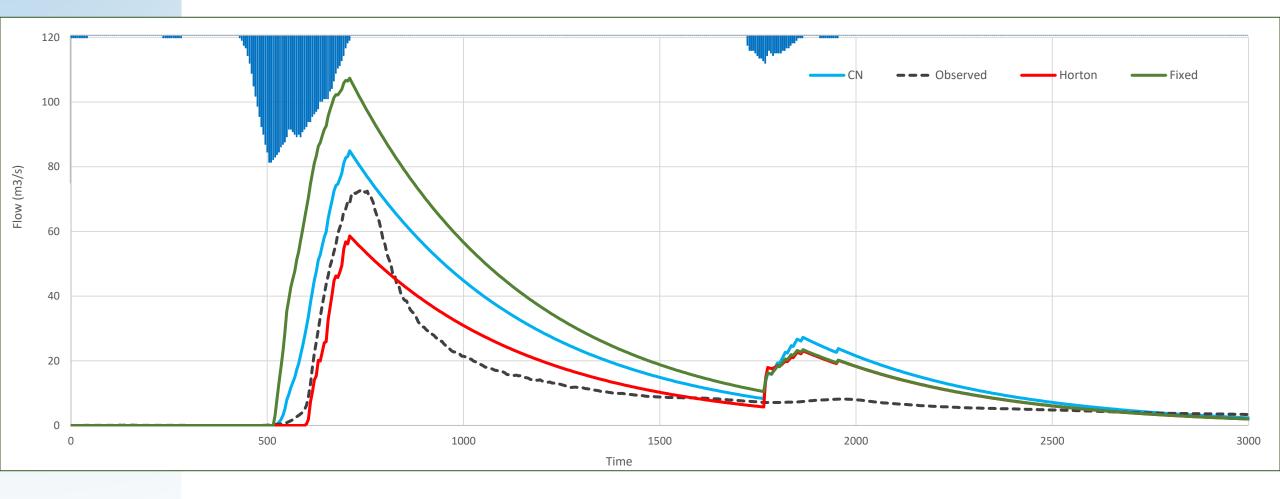


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Selwyn at Whitecliffs – Event 2



## Selwyn at Whitecliffs – Event 3



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### Selwyn at Whitecliffs - Score

Model Type	Mean Score	Range	Result
CN	-0.35	0.39	Unacceptable
Horton	-0.02	1.07	Unacceptable
Fixed	-1.27	1.77	Unacceptable

- < 0 = unacceptable
- 0.0 0.2 = poor calibration
- 0.2 0.4 = moderate calibration
- 0.4 0.6 = good calibration
- > 0.6 = excellent calibration

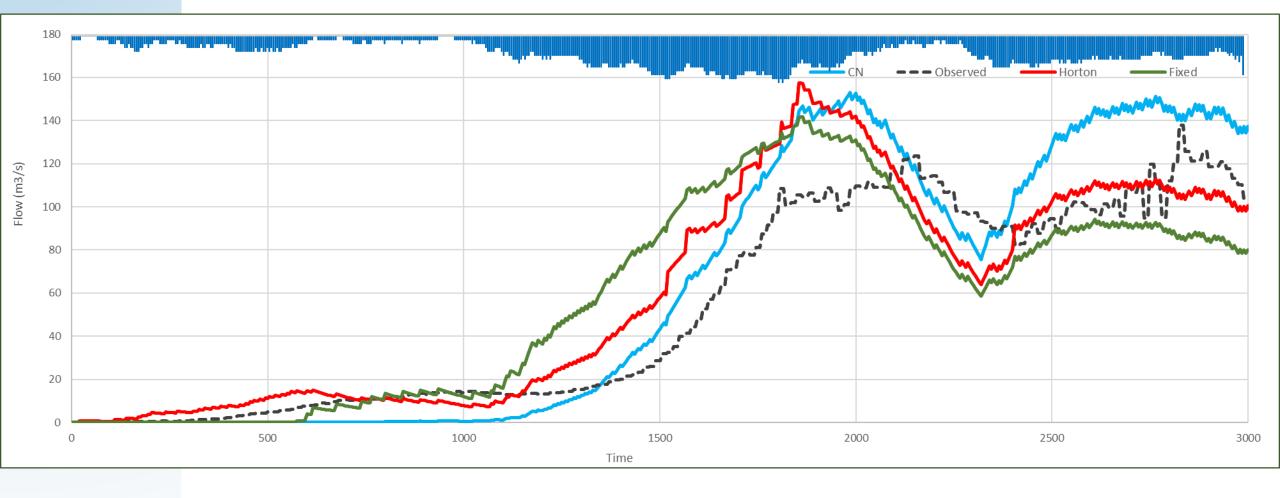
## **North Ashburton River**

- 276 km<sup>2</sup> (approx.)
- 242 minute Tc
- Flow gauged at Cookies Hut
- Rainfall gauge near flow gauge used

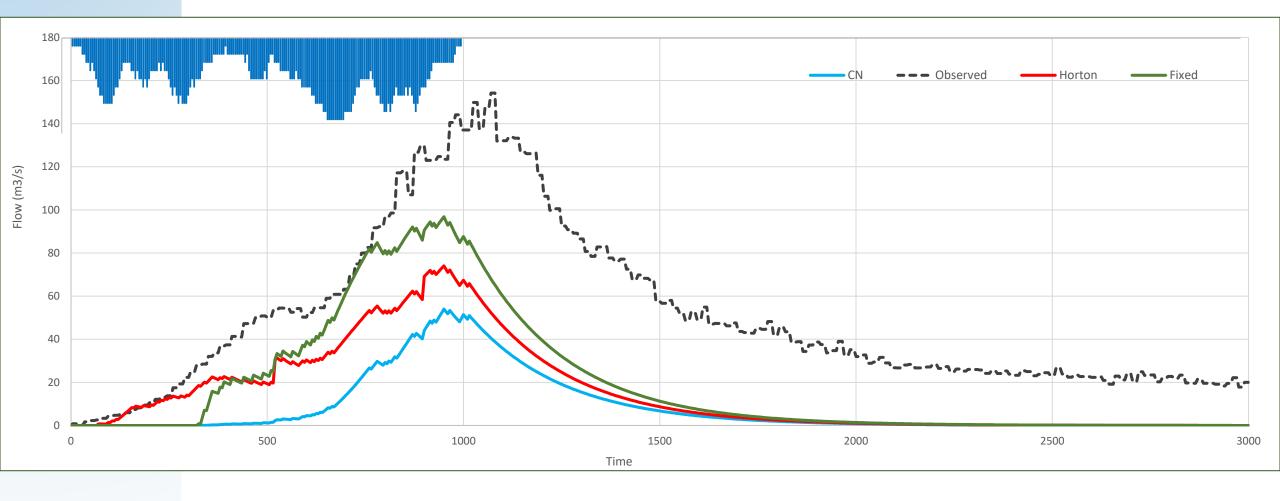
*Too large to be represented with one rain gauge and one catchment in practice* 

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#### **Ashburton River at Cookies Hut - Event 1**

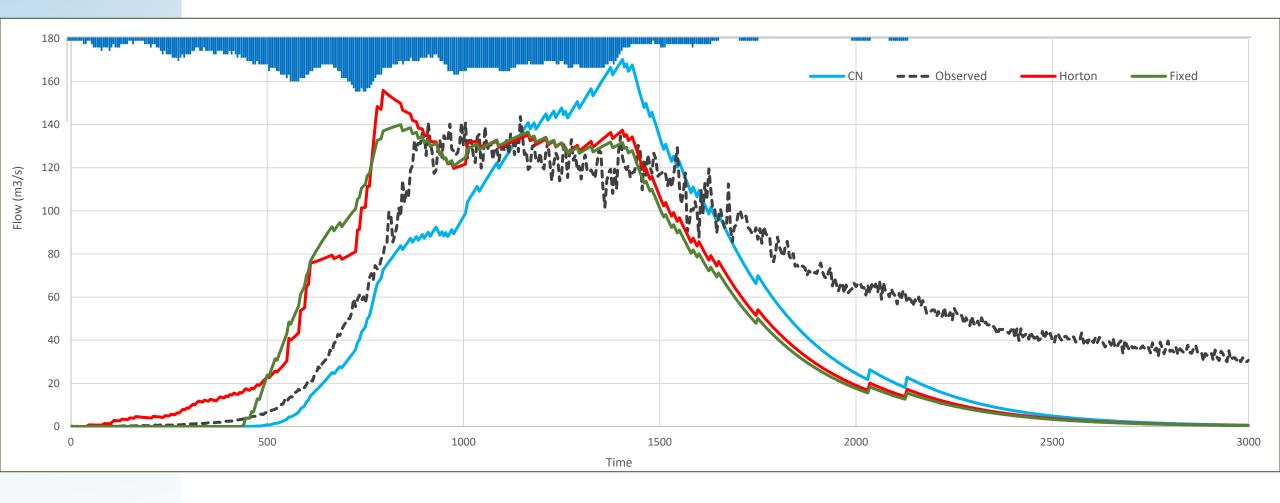


#### **Ashburton River at Cookies Hut - Event 2**



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#### **Ashburton River at Cookies Hut - Event 3**

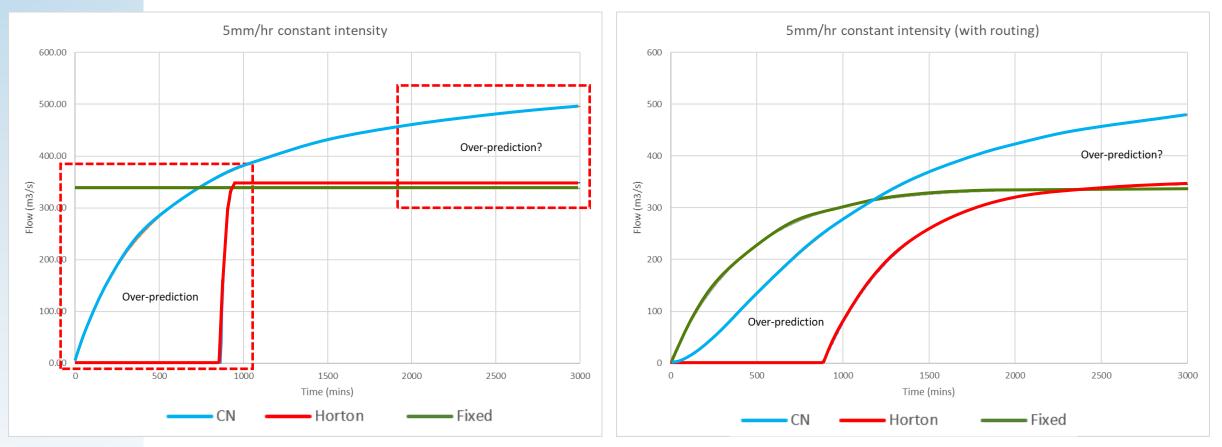


#### **Ashburton River at Cookies Hut - Score**

Model Type	Mean Score	Range	Result
CN	-0.37	2.04	Unacceptable
Horton	-0.19	1.49	Unacceptable
Fixed	-0.18	0.80	Unacceptable

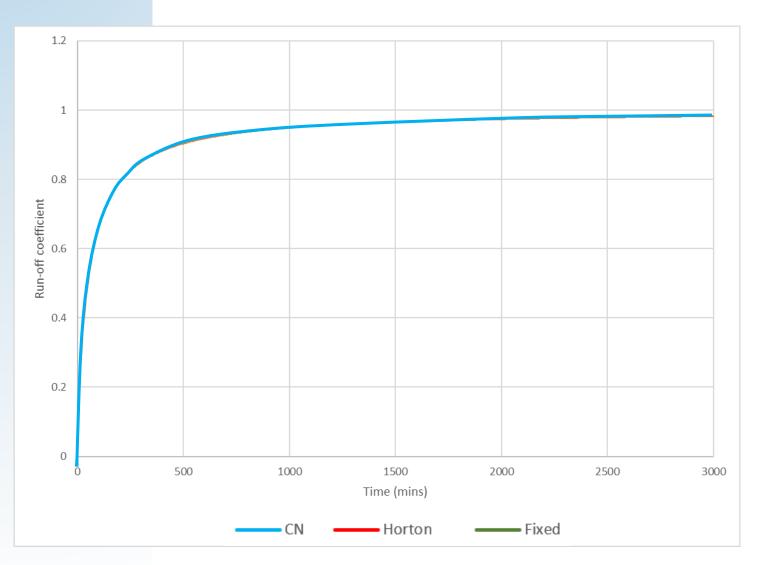
- < 0 = unacceptable
- 0.0 0.2 = poor calibration
- 0.2 0.4 = moderate calibration
- 0.4 0.6 = good calibration
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## **Differences (Calibrated to Target)**





## **CN Model**



- Run-off increases until unity is achieved.
- Is 0% loss appropriate?
- The NZBC applies 10% loss to impervious surfaces (e.g. roofs)
- 42% is the observed upper bound loss (C = 0.58) (Hutt river example, from observed results)

#### Conclusions

- Avoid R<sup>2</sup> to assess calibration `fit'.
- Best to use a combination of RMSE and NSE.
- Using this approach allows for automation in calibration. This was successfully implemented in excel.
- CN performs better than a simple fixed run-off model for two of three catchments tested.
- Horton overall performs the best in terms of fit.
- Horton has the least variance across events for two of three catchments tested.

Thank you

Questions?

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