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FHN dynamics

$$\begin{aligned}\dot{x}_i &= \frac{1}{\epsilon}(x_i - x_i^3/3 - y_i) + \sum_{j \neq i} g_{ij} A_{ij} h(x_i, x_j) + \eta_i \\ \dot{y}_i &= x_i + \alpha\end{aligned}$$

Function to find peaks

- https://docs.scipy.org/doc/scipy/reference/generated/scipy.signal.find_peaks.html

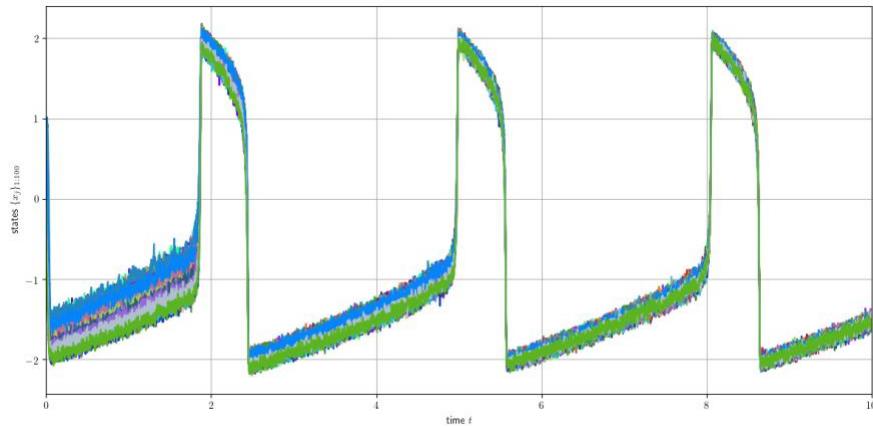
Diffusive FHN model

- Playing around with FHN model

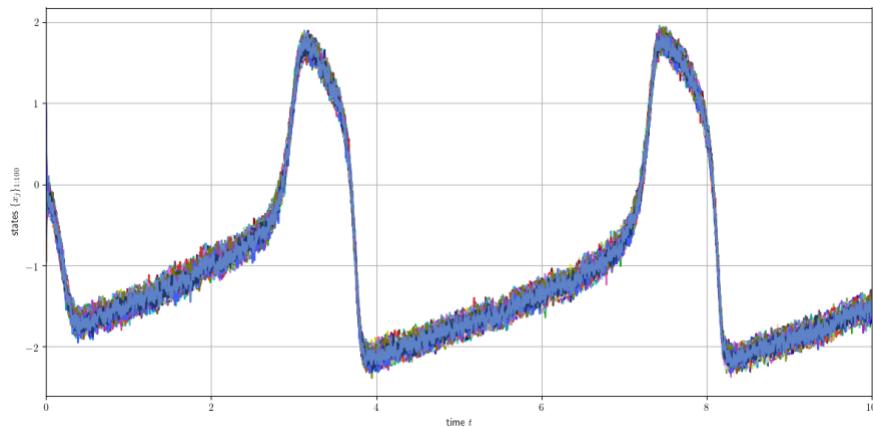
- Parameters

- (a) Random directed weighted network
- (b) Size = 100, connection probability = 0.2, weights $\sim N(10,2)$, $\sigma_i = 1$
- (c) ($\epsilon = 0.01, 0.1, 1, \alpha = 0.95, 1, 1.5$)
- (d) Initial conditions uniform $[-1,1]$

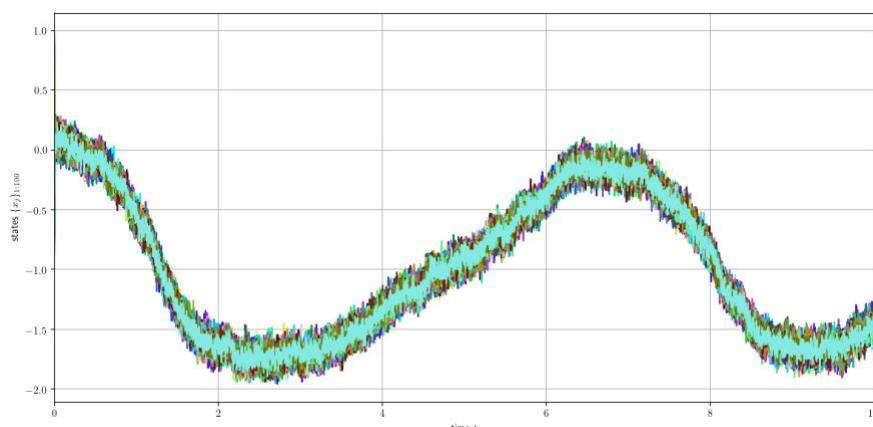
- $\epsilon = 0.01, \alpha = 0.95$



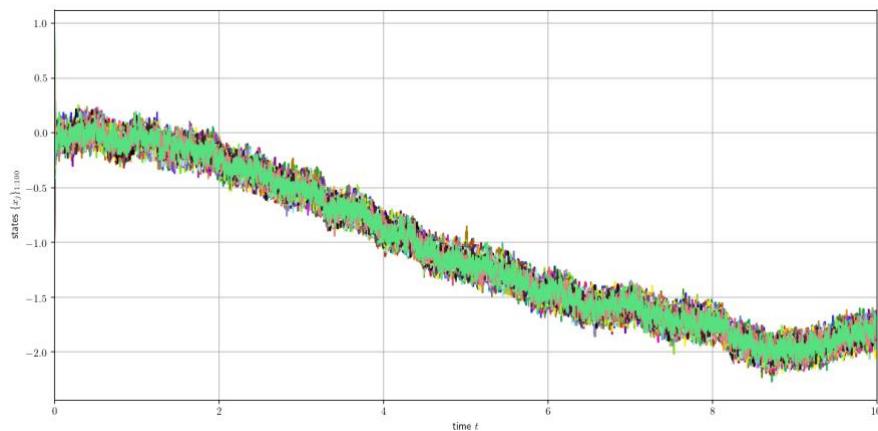
- $\epsilon = 0.1, \alpha = 0.95$



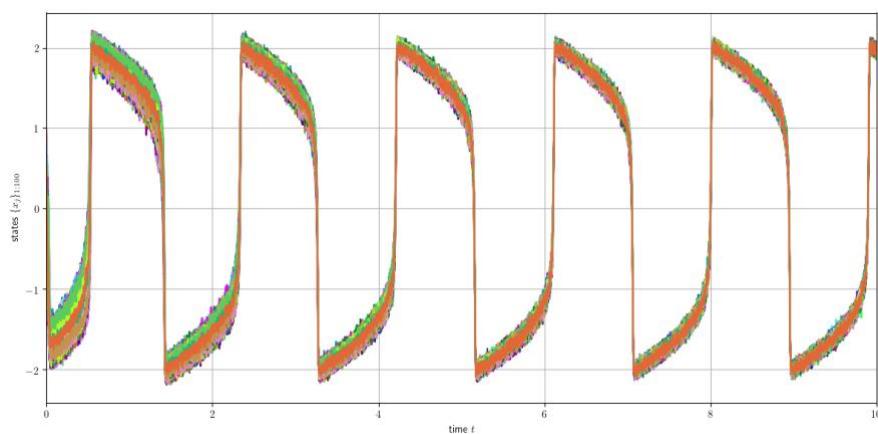
- $\epsilon = 1, \alpha = 0.95$



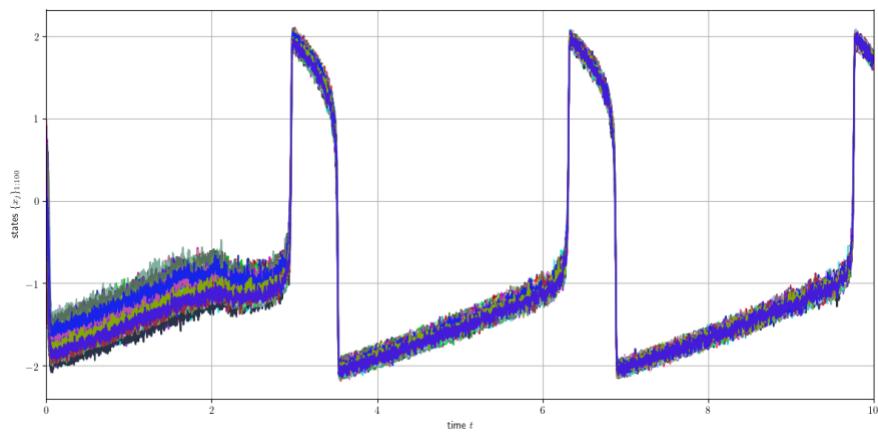
- $\epsilon = 10, \alpha = 0.95$



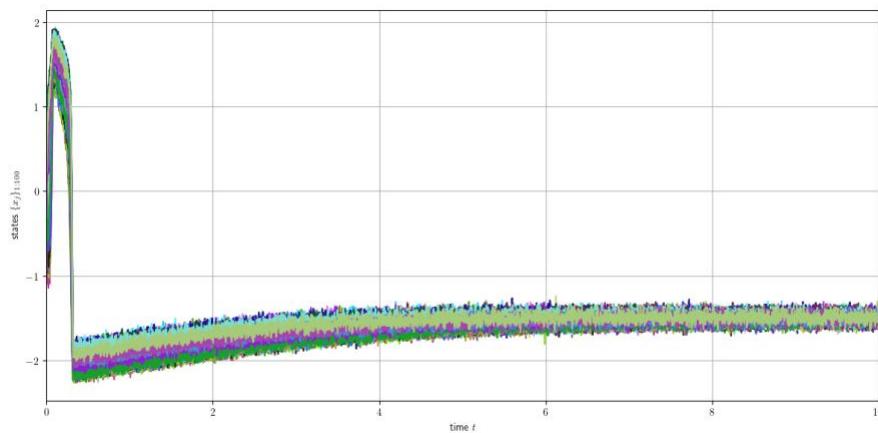
- $\epsilon = 0.01, \alpha = 0$



- $\epsilon = 0.01, \alpha = 1$

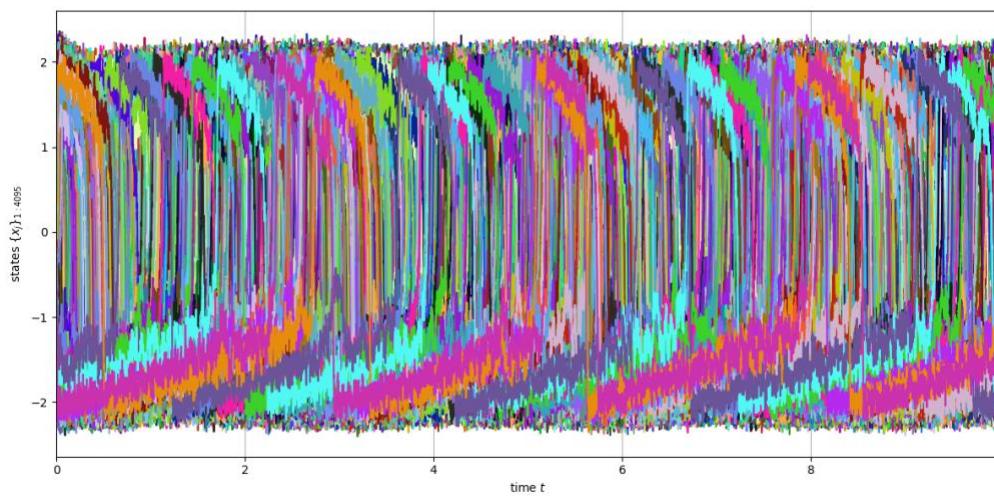


- $\epsilon = 0.01, \alpha = 1.5$

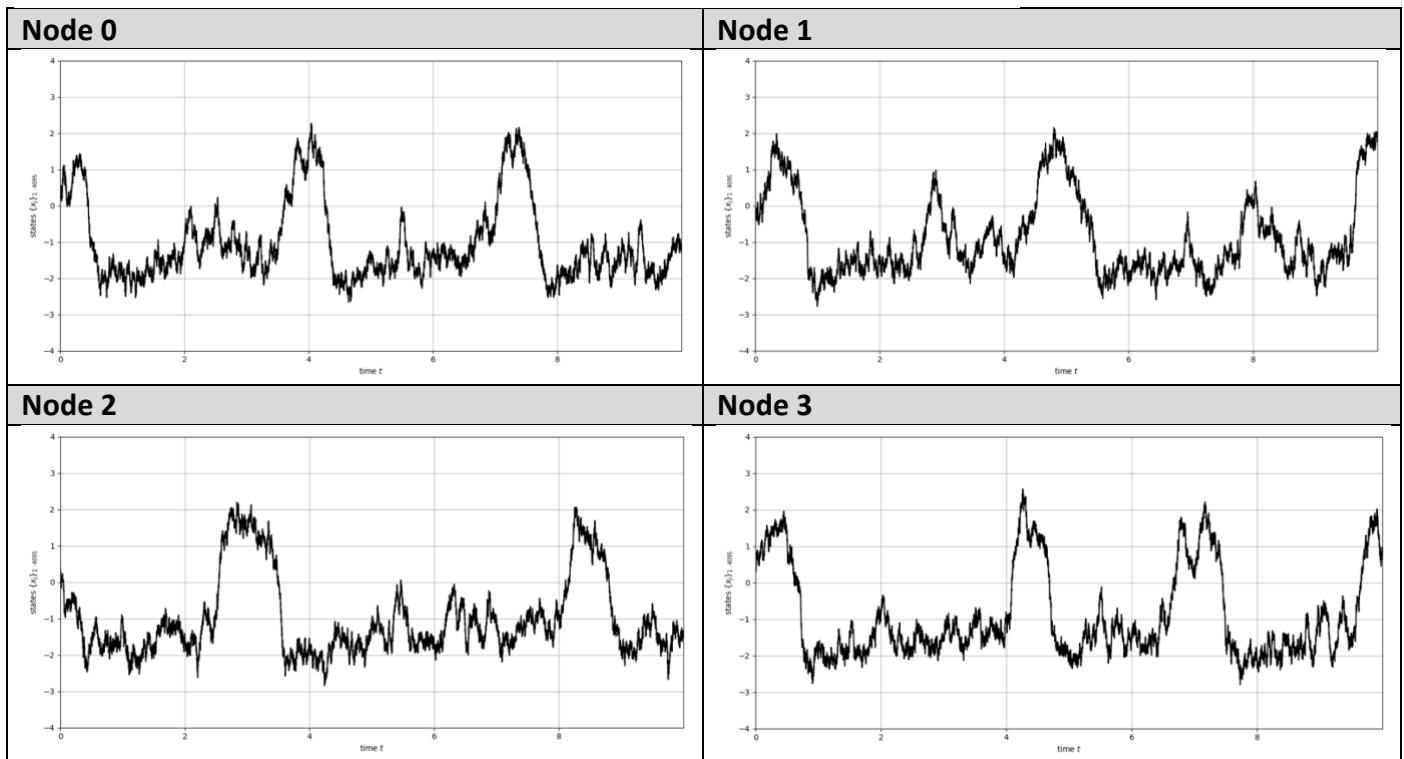
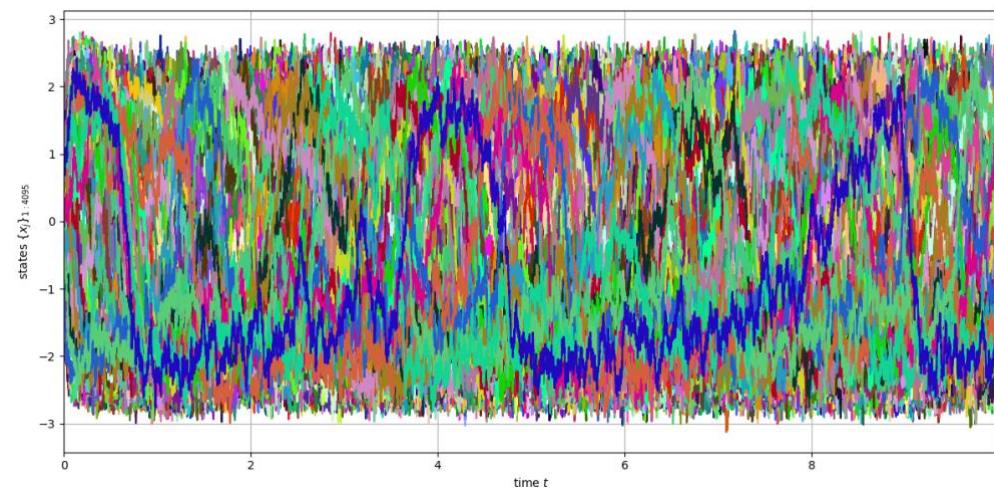


FHN model (DIV25_PREmethod, diffusive coupling function)

- Using g_{ij} from “DIV25_PREmethod” (no multiplier)
- $\epsilon = 0.01, \alpha = 0.95, \sigma_i = 2$

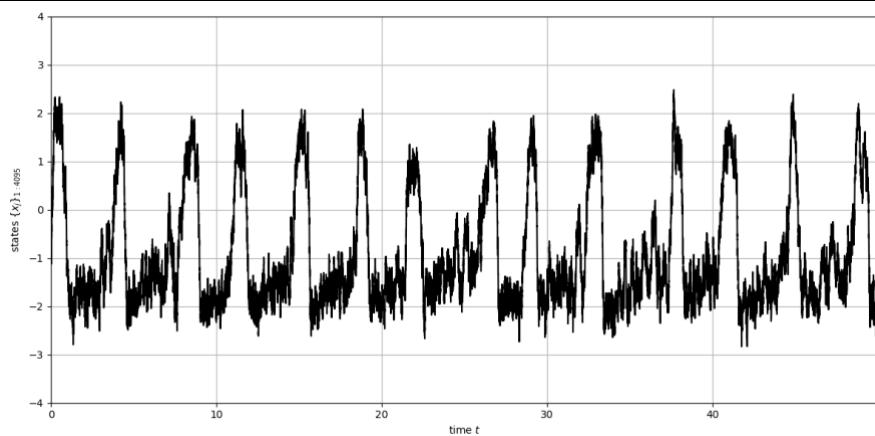


- $\epsilon = 0.1, \alpha = 0.95, \sigma_i = 2$ (run for $t = 0$ to 10)

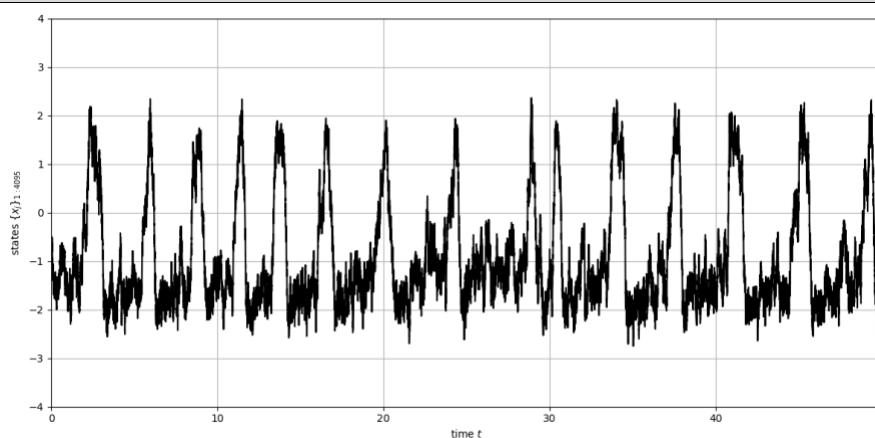


- $\epsilon = 0.1, \alpha = 0.95, \sigma_i = 2$ (run for $t = 0$ to 50)

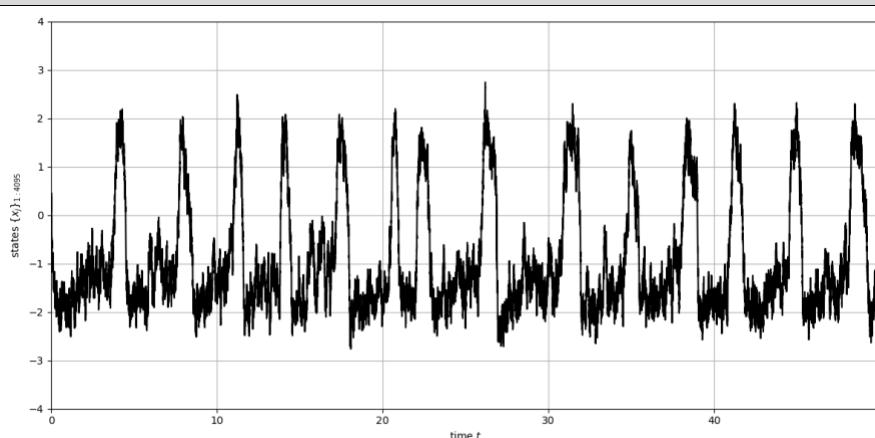
Node 0



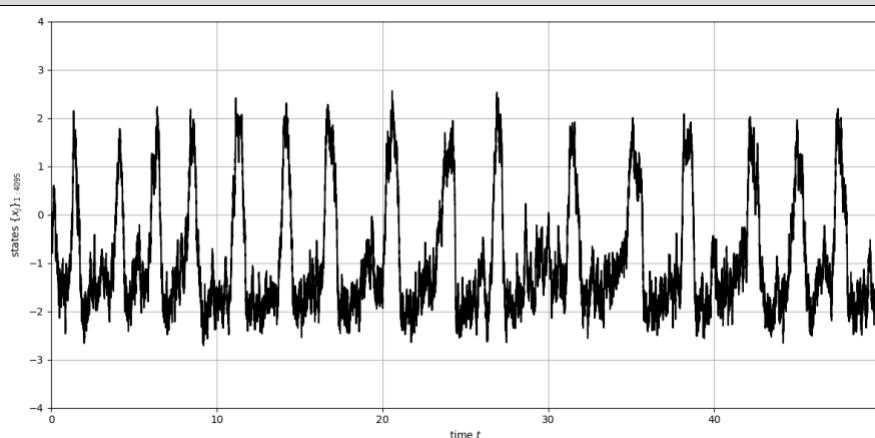
Node 1



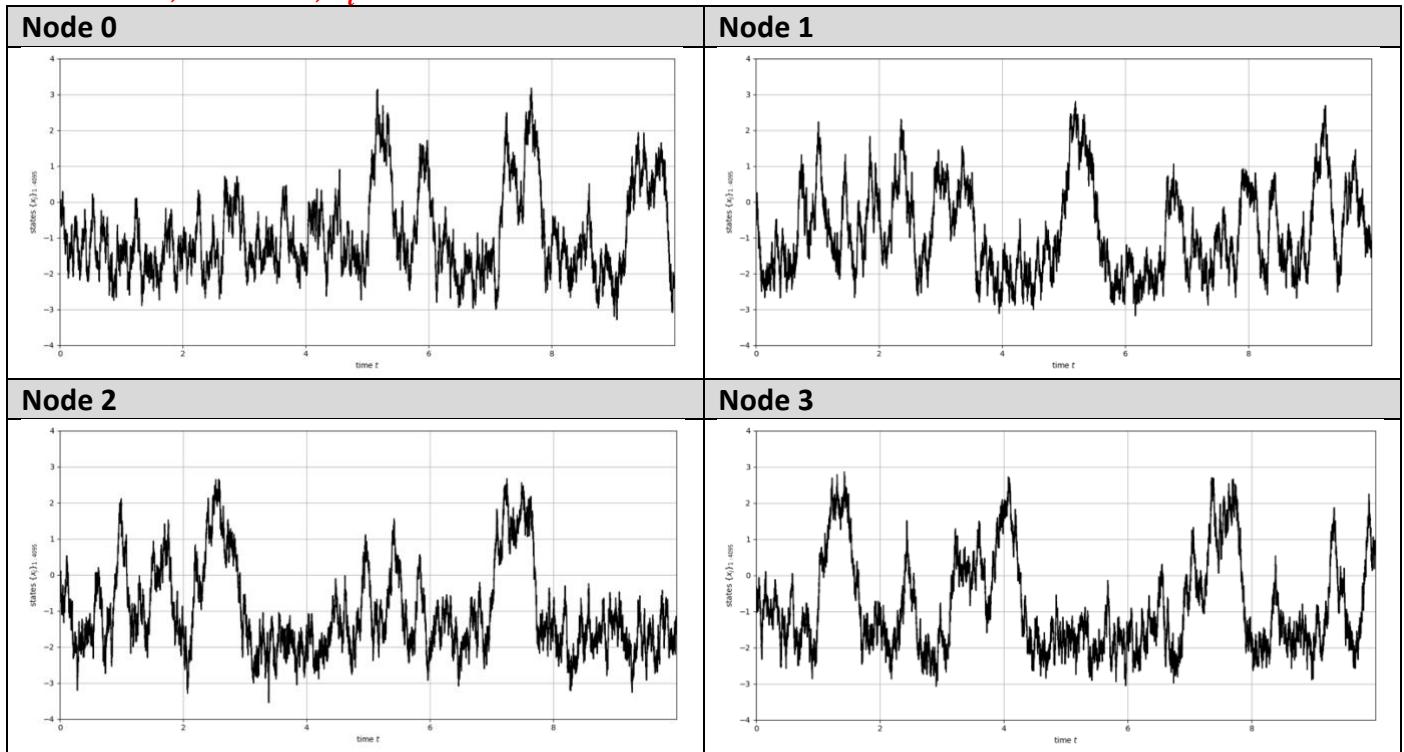
Node 2



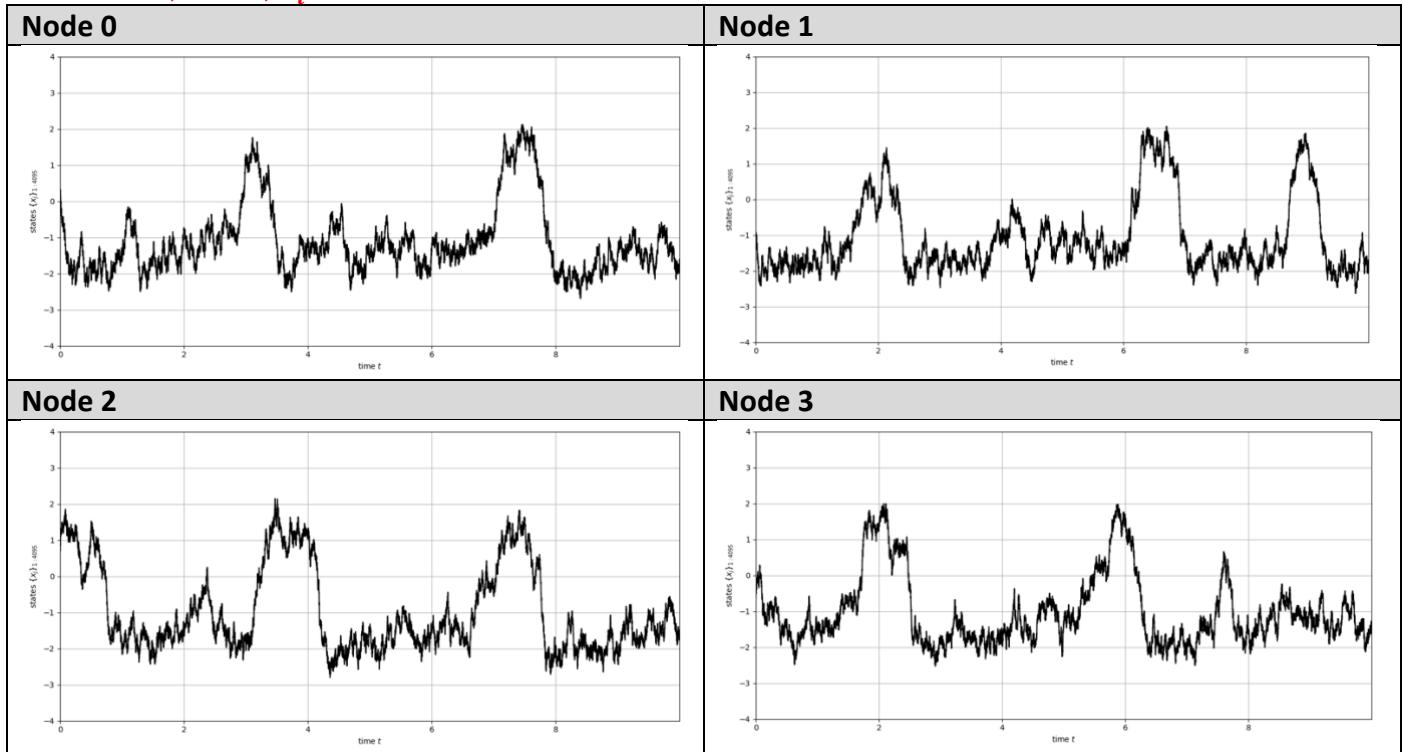
Node 3



- $\epsilon = 0.1, \alpha = 0.95, \sigma_i = 4$

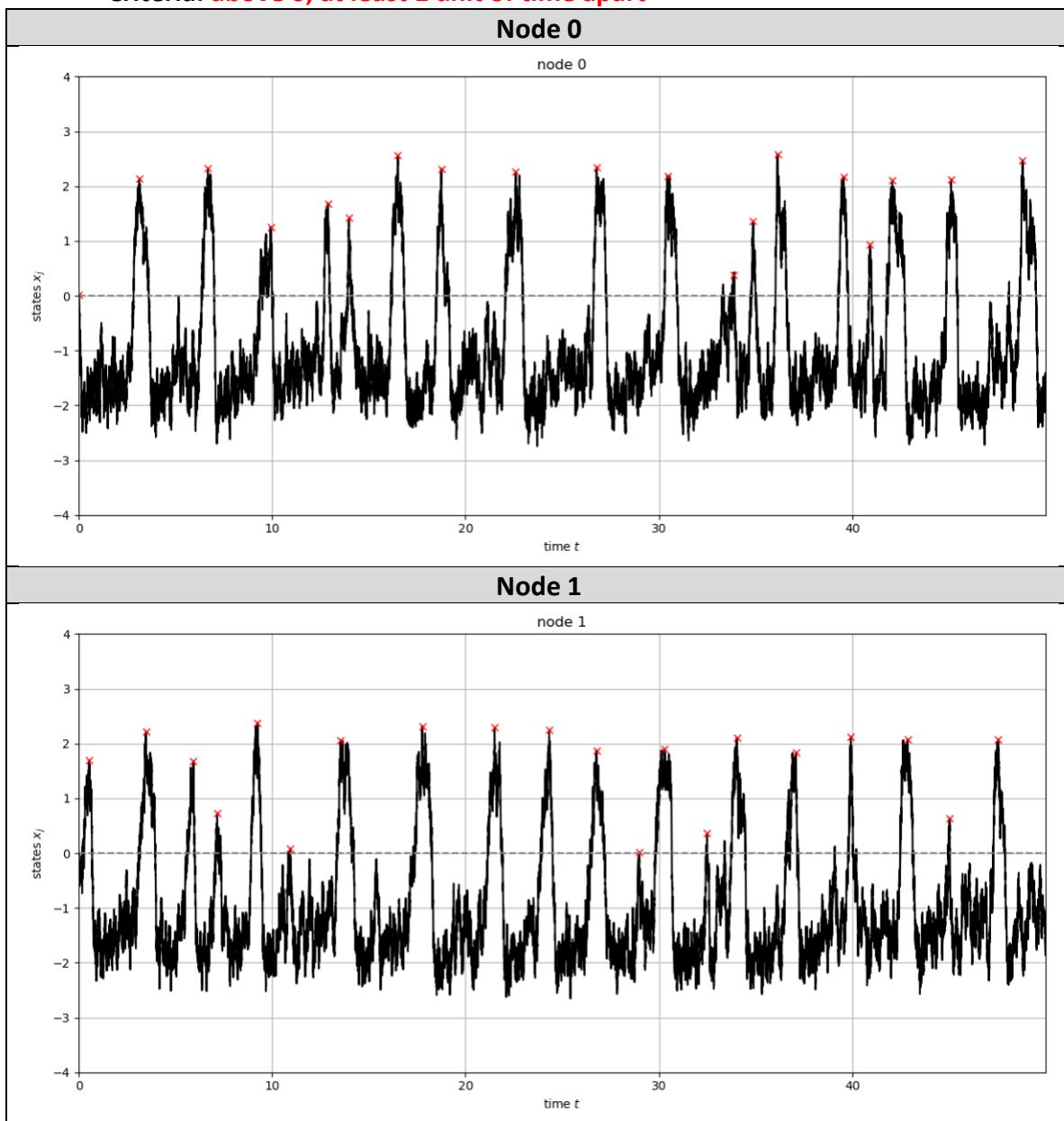


- $\epsilon = 0.1, \alpha = 1, \sigma_i = 2$



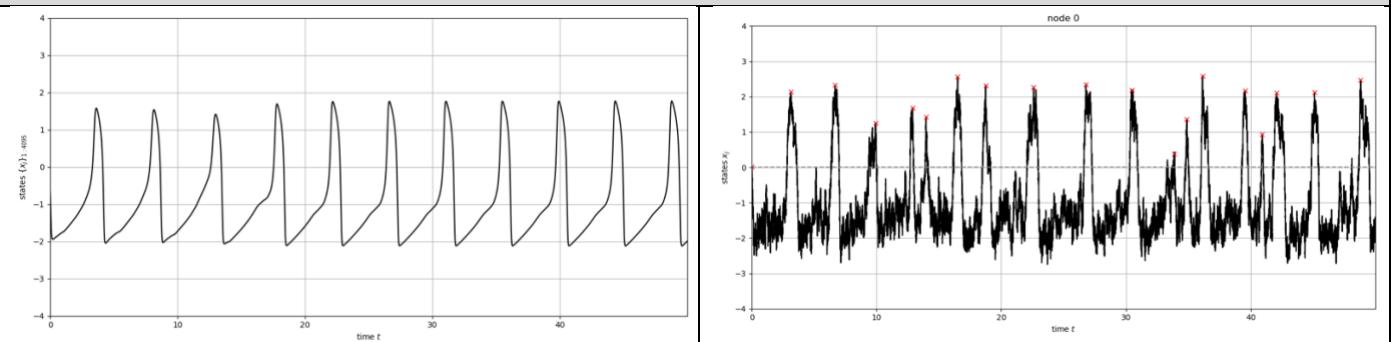
Spike analysis (peak distribution, peak vs strength plot)

- Using g_{ij} from “DIV25_PREmethod” (no multiplier)
- Parameters $\epsilon = 0.1, \alpha = 0.95, \sigma_i = 2$
- Examples of peaks (node 0 & 1)
 - Criteria: above 0; at least 1 unit of time apart

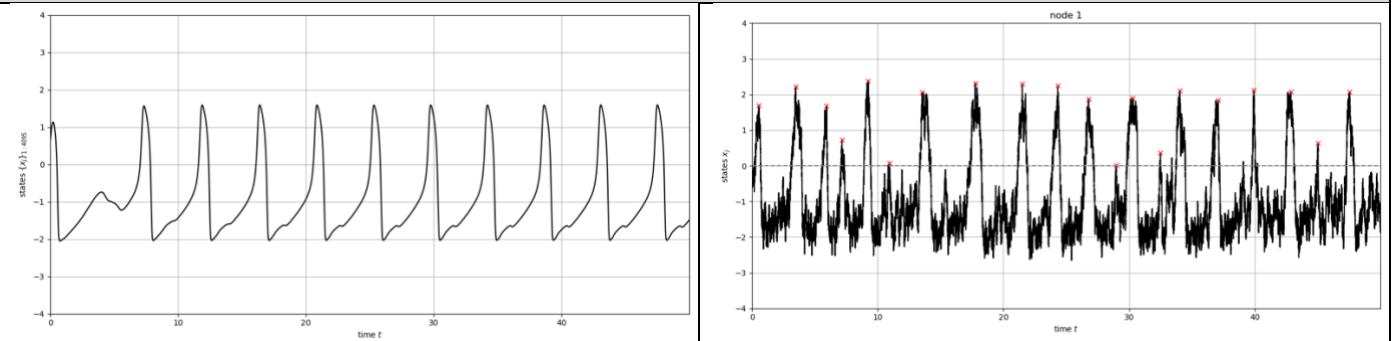


- Noise-free time series as comparison

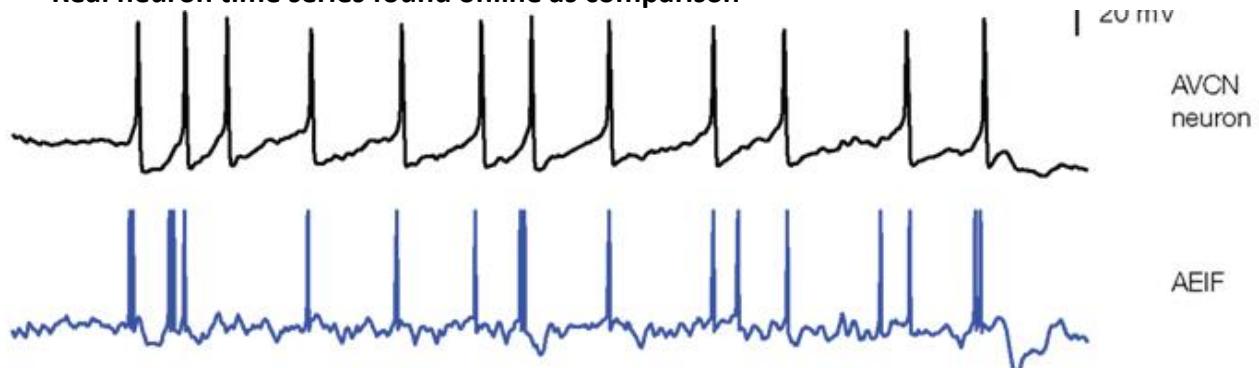
Node 0



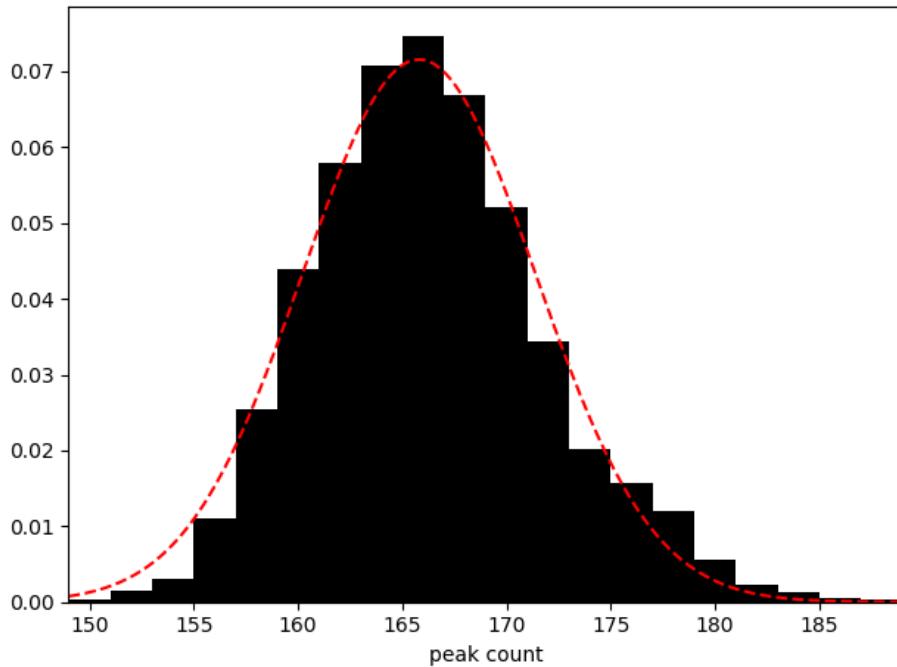
Node 1



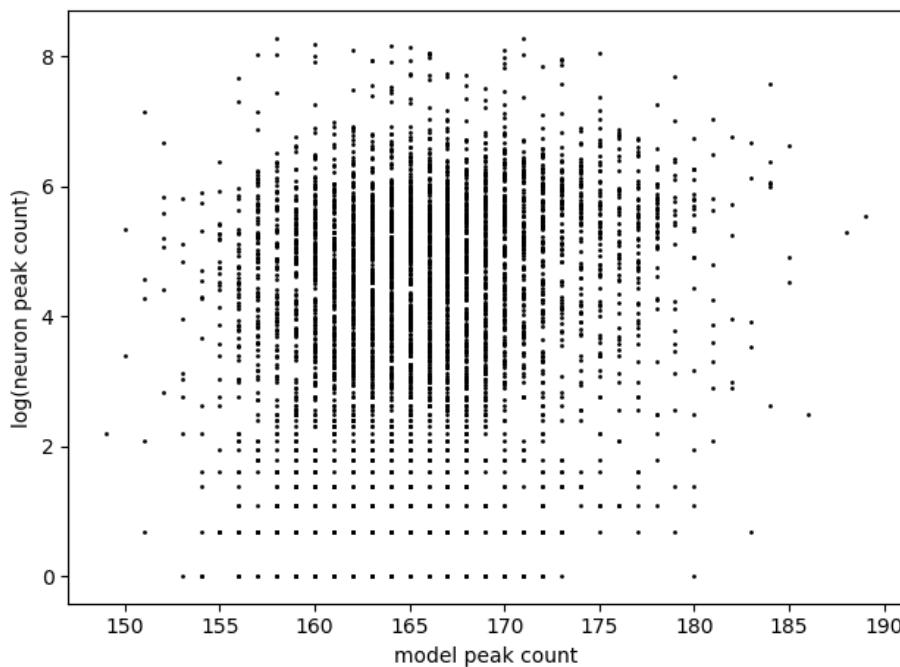
- Real neuron time series found online as comparison



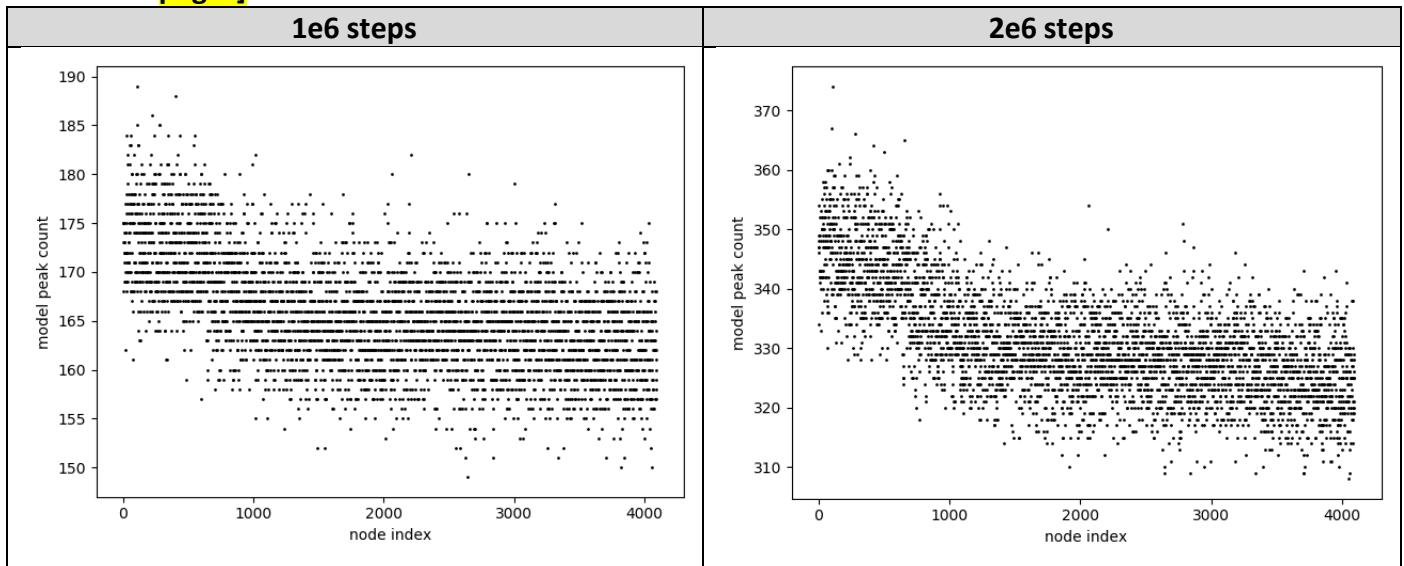
- Distribution of number of peaks (slightly heavy-tailed) [over 1e6 steps]
 - Median = 165, Min = 149, Max = 189 (Max-Median > Median-Min)
 - Mean = 165.7985 > Median
 - Skewness = 0.4002 (right-skewed)
 - (Excess) Kurtosis = 0.2204 (heavier tails than Gaussian)
 - Dotted line: Gaussian with same mean and s.d.



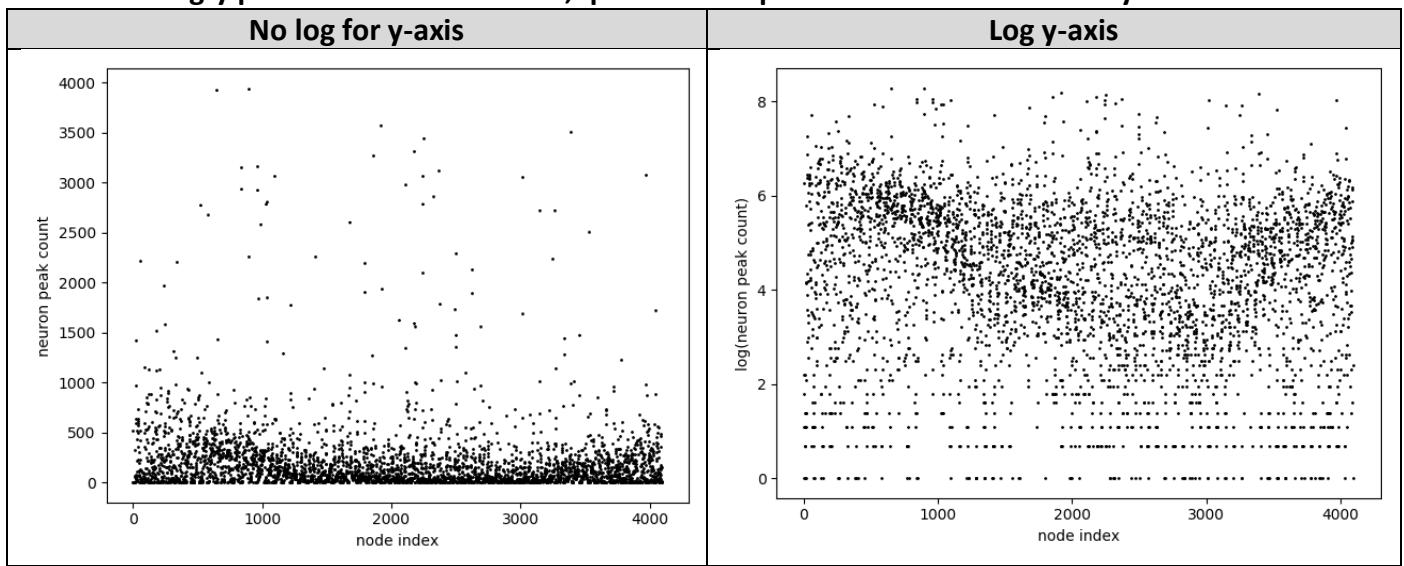
- Model peaks vs experimentally measured peaks (not similar)
 - Log scale for y-axis



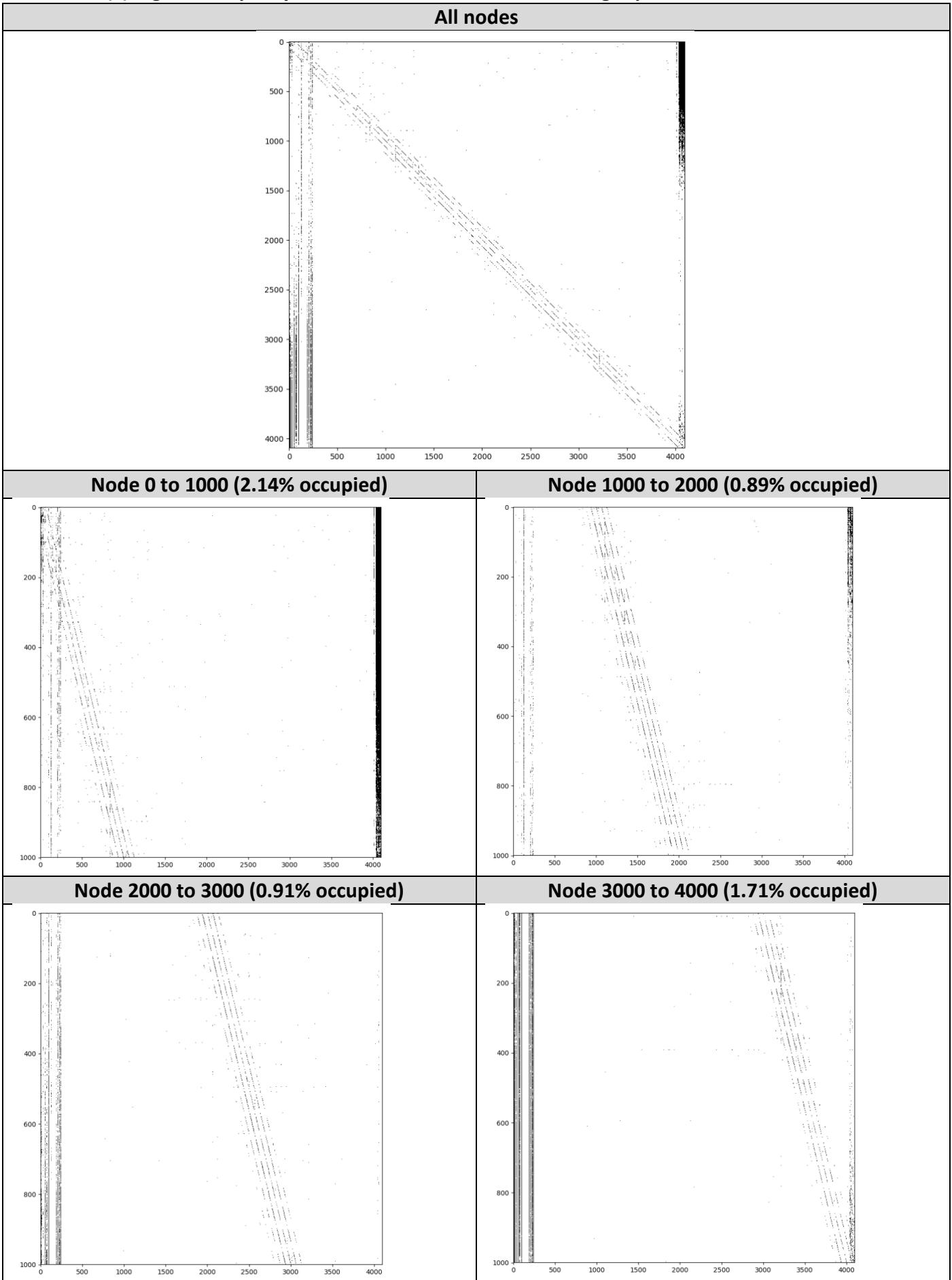
- **Model peaks vs node index**
 - Larger peak counts for first 1000 nodes
 - Fairly uniform for other nodes
 - Does network connectivity (including in-degrees and in-strengths) cause this? [next two pages]



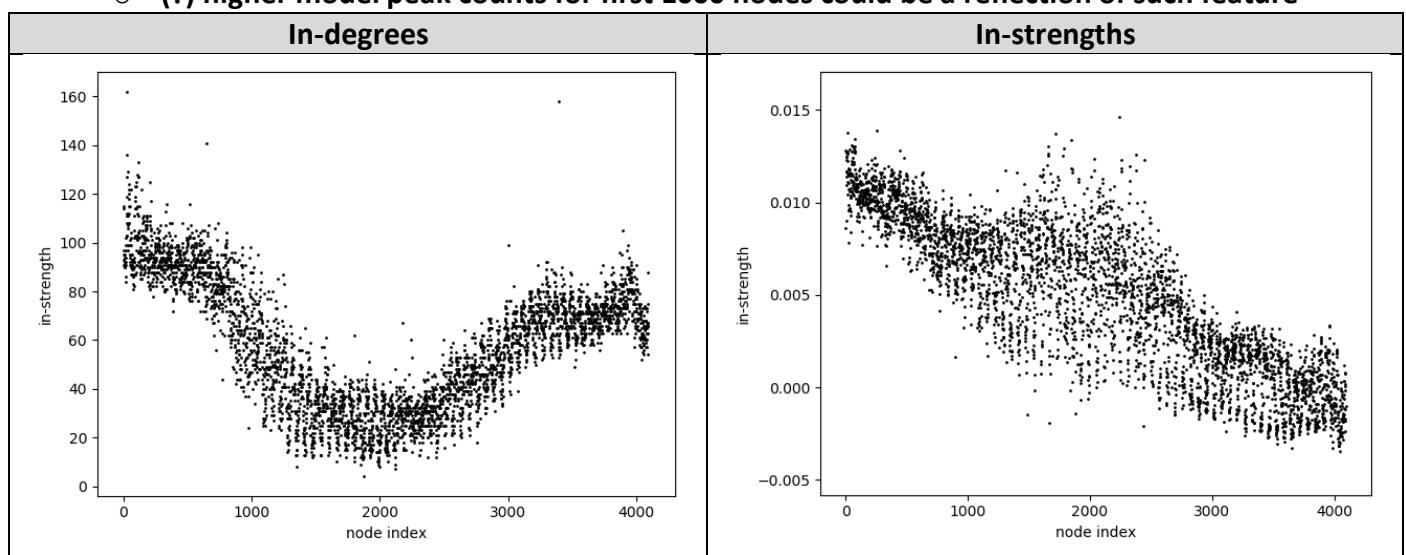
- **Real peaks vs node index**
 - First 1000 nodes generally also have larger peak counts
 - Log-y plot: for first 1000 nodes, quite a lot of points concentrate around y=6



- Analysis of connectivity in adjacency matrix
 - (?) higher occupancy for first 1000 nodes causes the larger peak counts



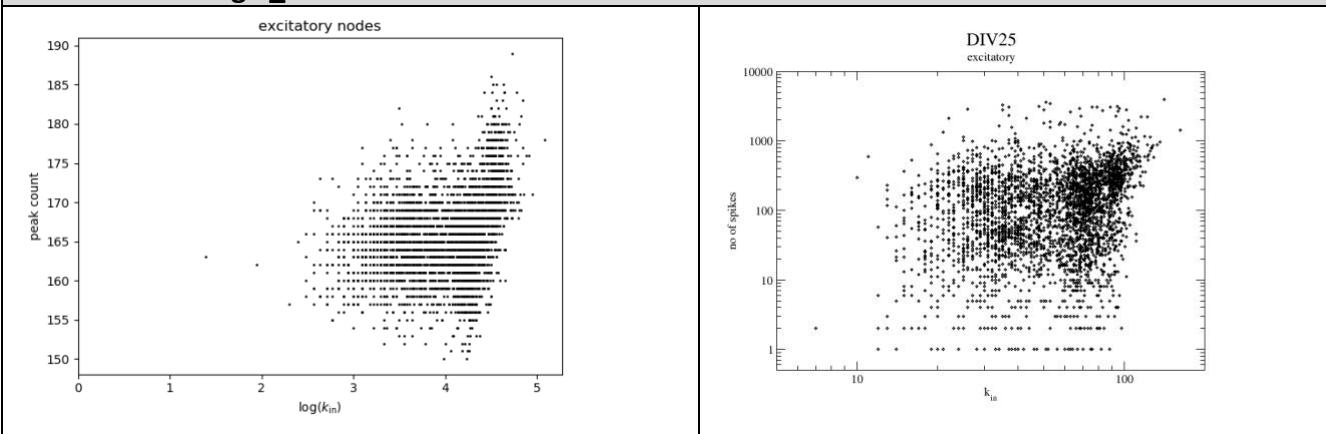
- Analysis of in-degrees and in-strengths (vs node index)
 - First 1000 nodes have relatively more in-degrees and stronger in-strengths than the other nodes
 - (?) higher model peak counts for first 1000 nodes could be a reflection of such feature



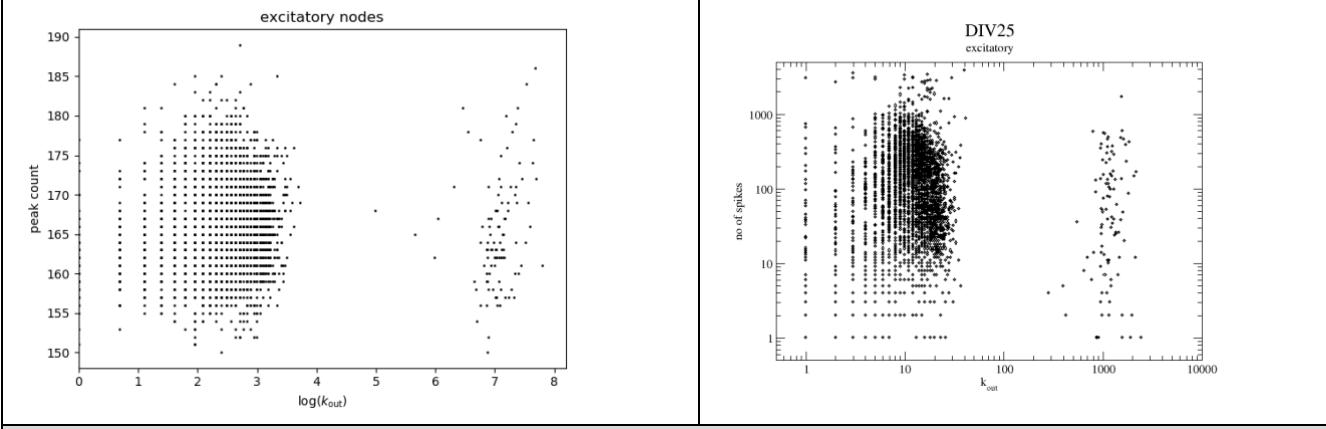
Peak count vs network features

- **Excitatory nodes**

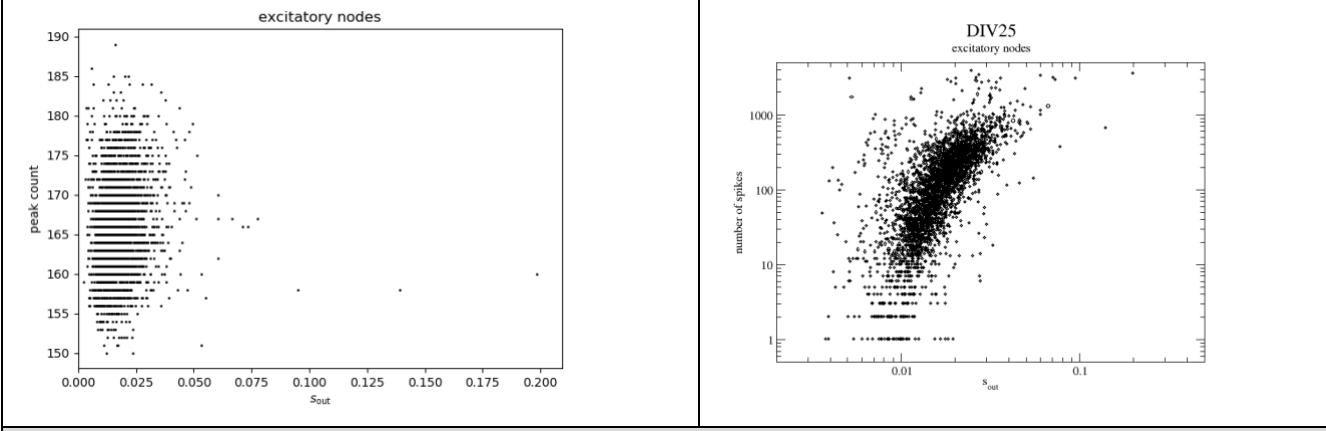
Peak count vs log k_in



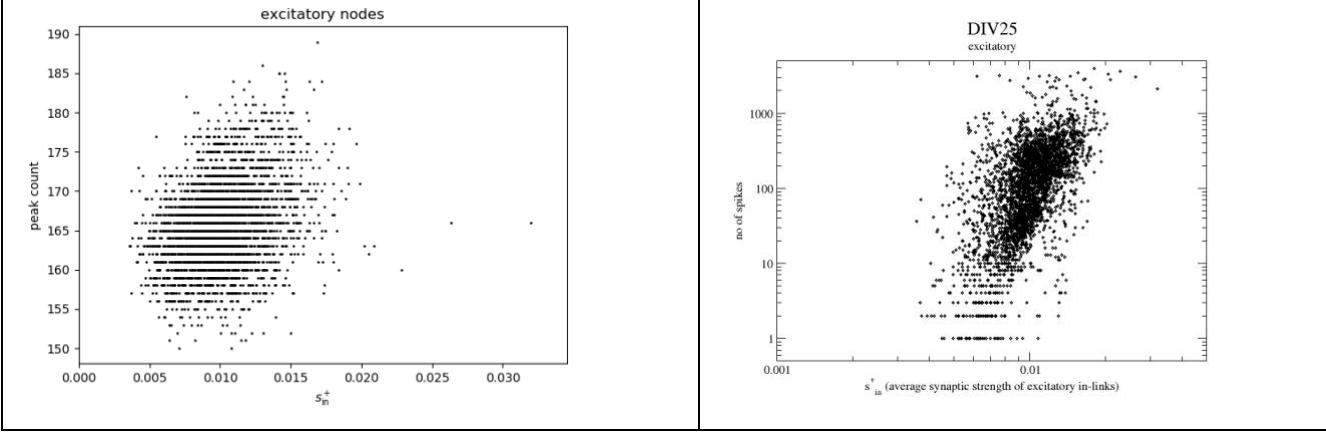
Peak count vs log k_out



Peak count vs s_out

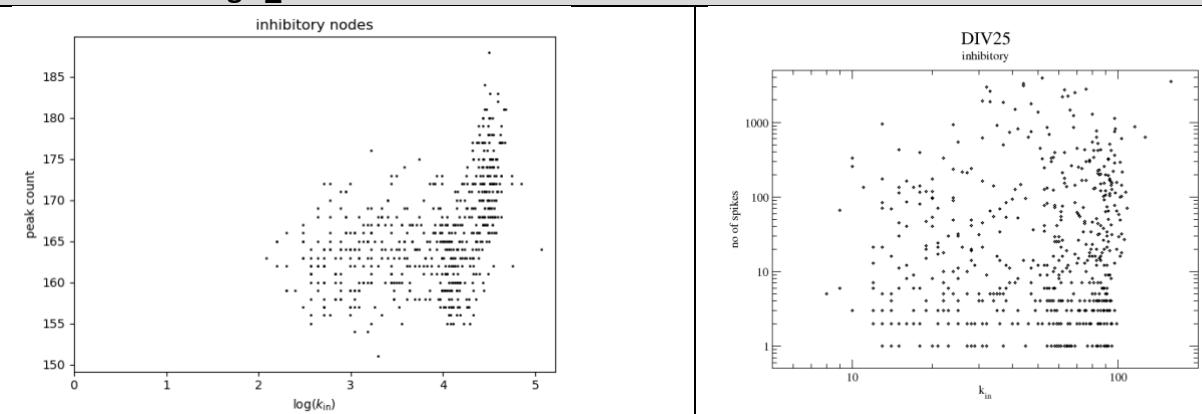


Peak count vs s^+_in

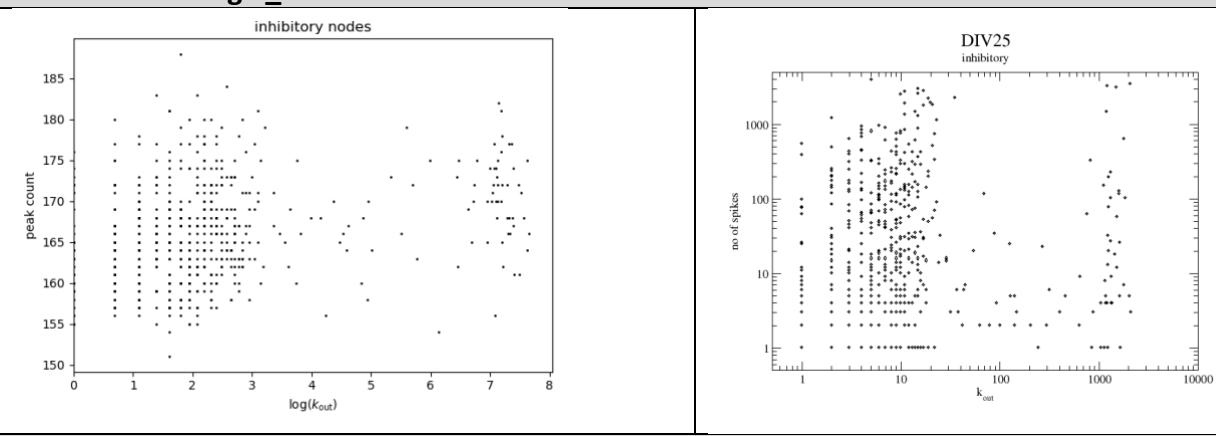


- Inhibitory nodes

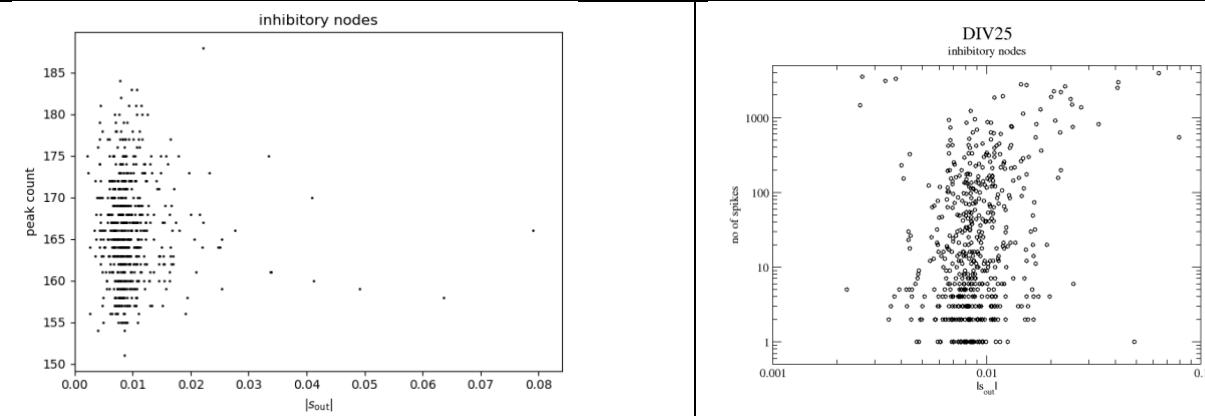
Peak count vs log k_in



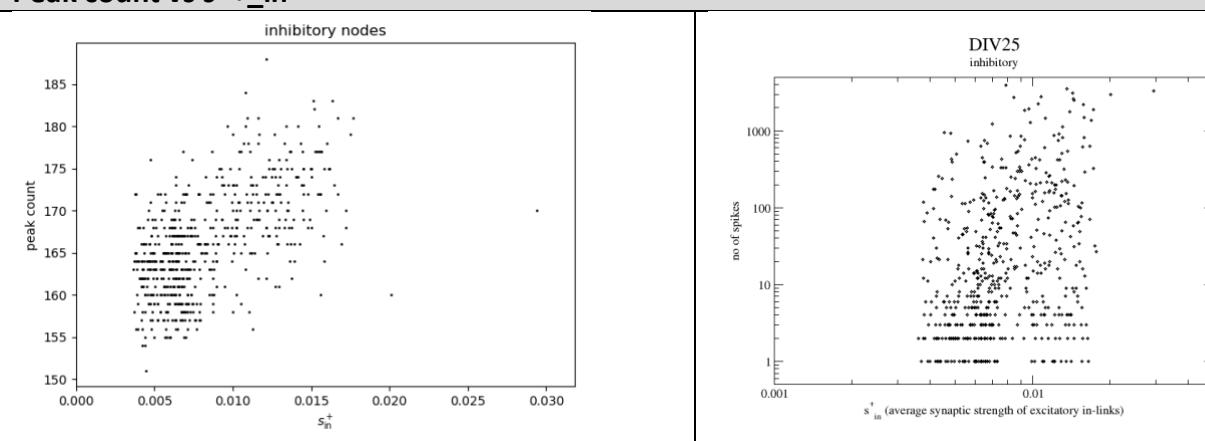
Peak count vs log k_out



Peak count vs |s_out|

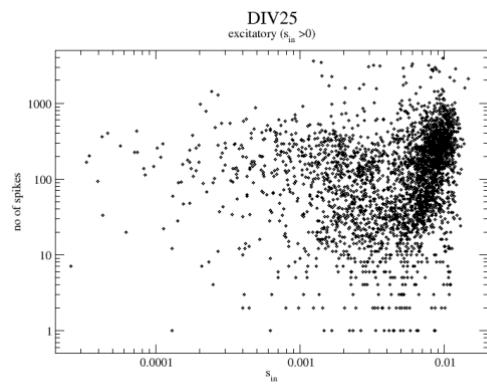
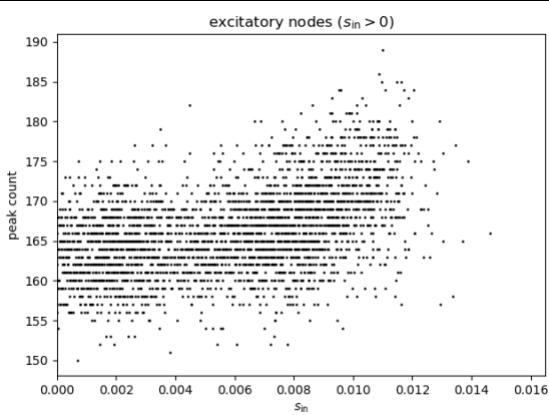


Peak count vs s^+_in

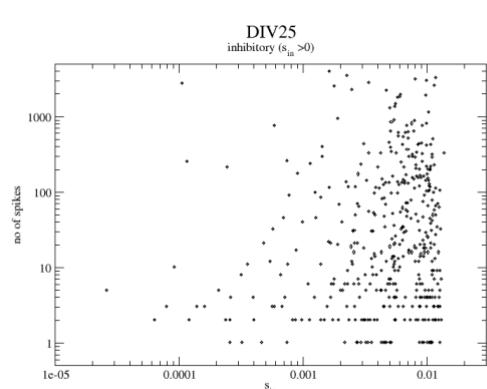
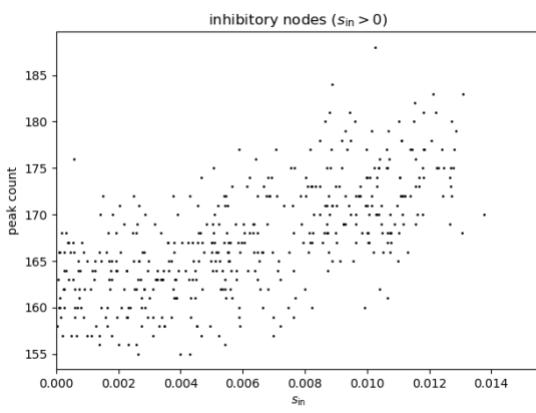


- **Mixed (x-axis: s_{in} or $|s_{in}|$)**

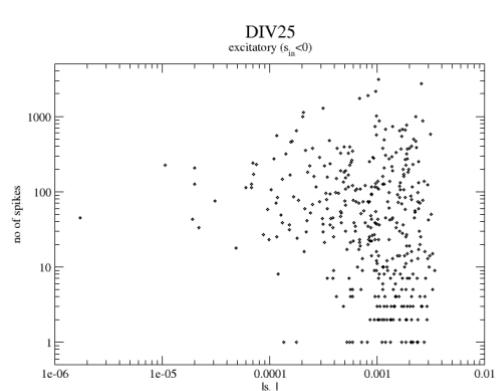
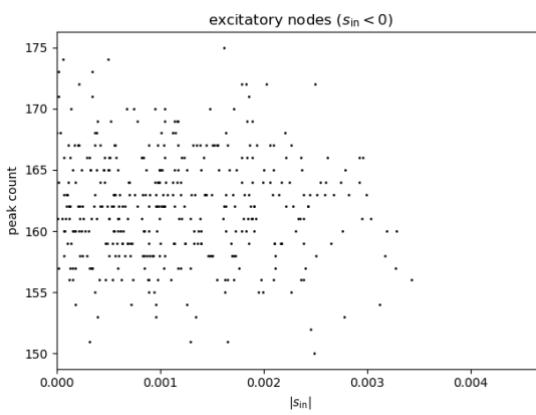
$s_{in} > 0 \& s_{out} > 0$



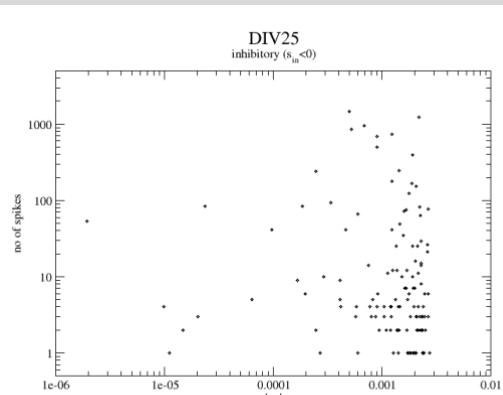
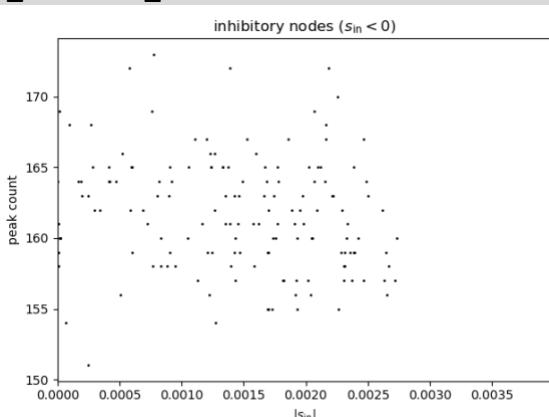
$s_{in} > 0 \& s_{out} < 0$



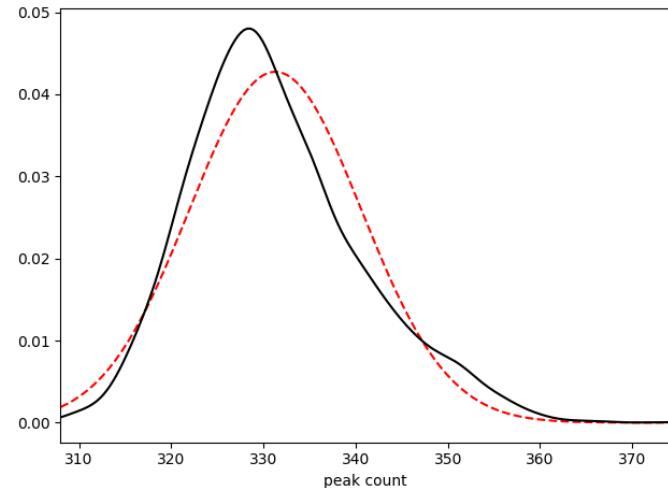
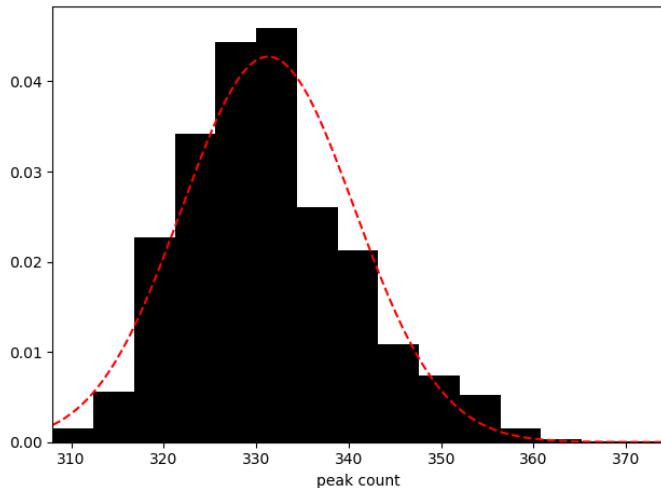
$s_{in} < 0 \& s_{out} > 0$



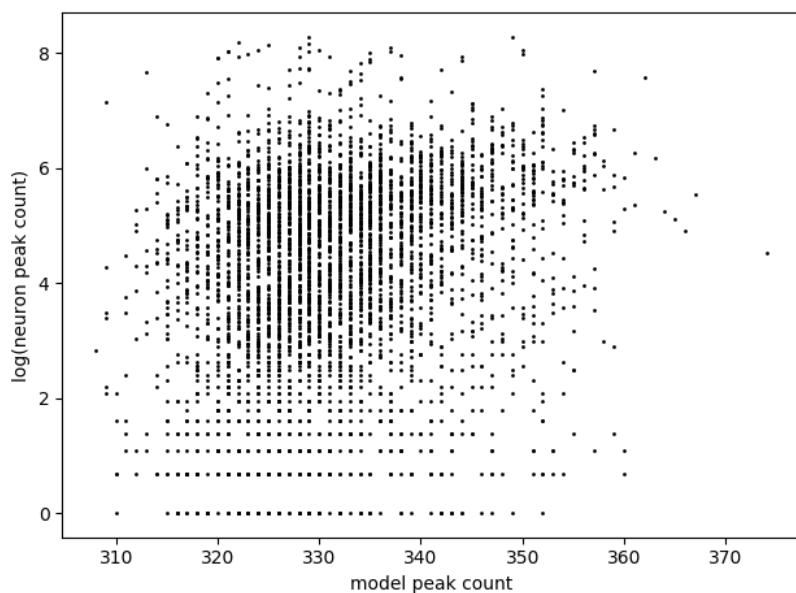
$s_{in} < 0 \& s_{out} < 0$



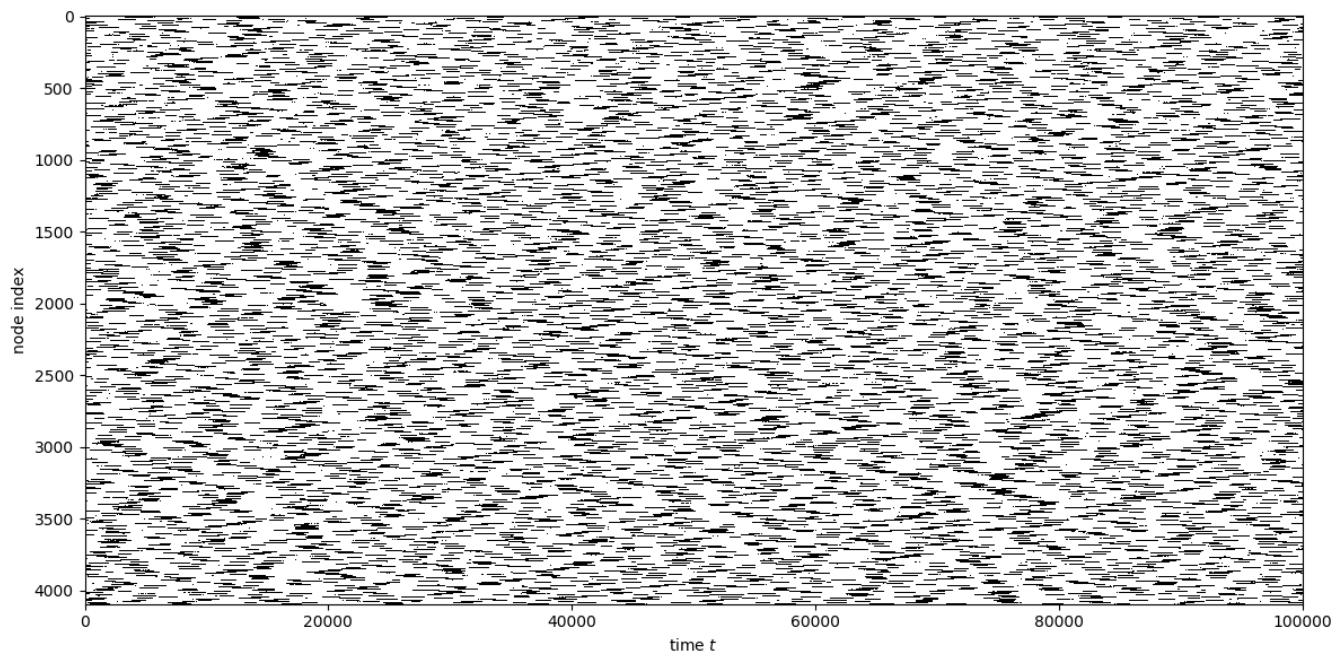
- Distribution of number of peaks (slightly heavy-tailed) [over 2e6 steps]
 - Median = 330, Min = 308, Max = 374 (Max-Median > Median-Min)
 - Mean = 331.2886 > Median
 - Skewness = 0.5945 (right-skewed)
 - (Excess) Kurtosis = 0.2684 (heavier tails than Gaussian)
 - Dotted line: Gaussian with same mean and s.d.



- Model peaks vs experimentally measured peaks (not similar)
 - Log scale for y-axis

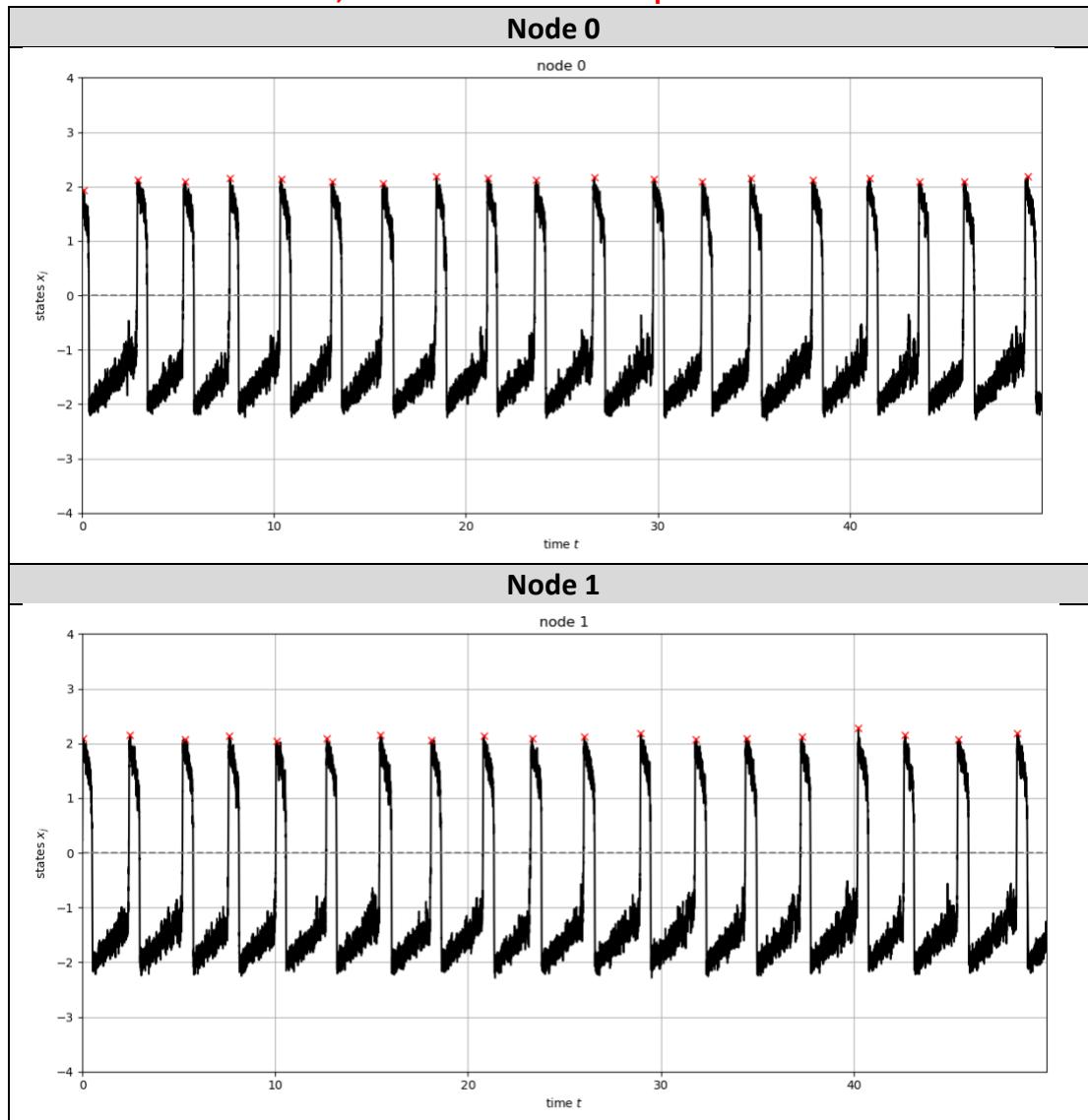


- **Raster plot [1e5 steps]**



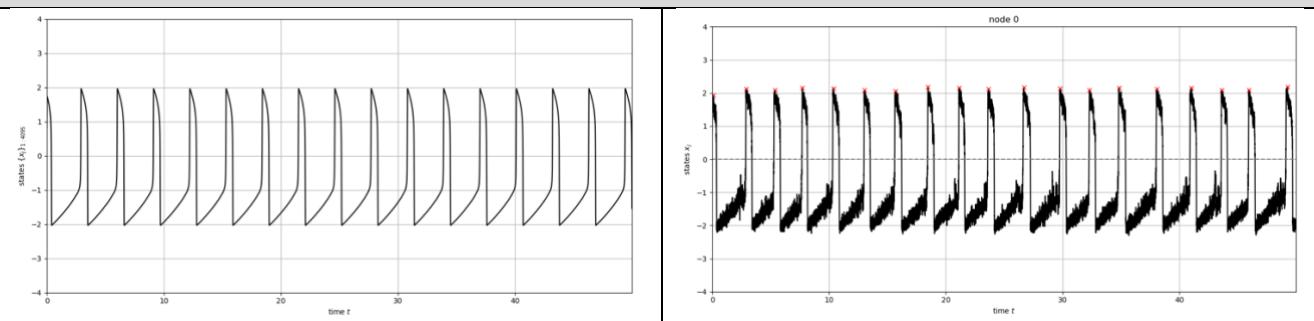
More spike analysis (other parameters)

- Using g_{ij} from “DIV25_PREmethod” (no multiplier)
- Parameters $\epsilon = 0.01, \alpha = 0.95, \sigma_i = 2$
- Examples of peaks (node 0 & 1)
 - Criteria: above 0; at least 1 unit of time apart

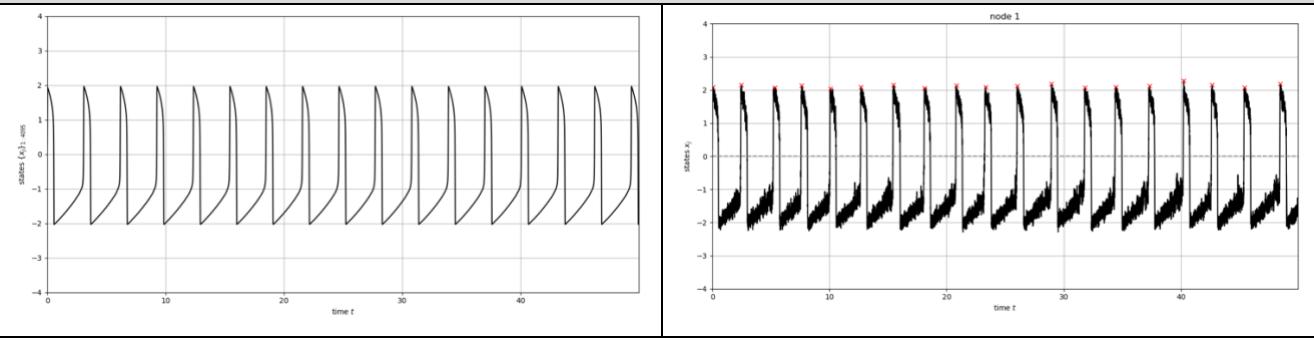


- Noise-free time series as comparison (near periodic)

Node 0

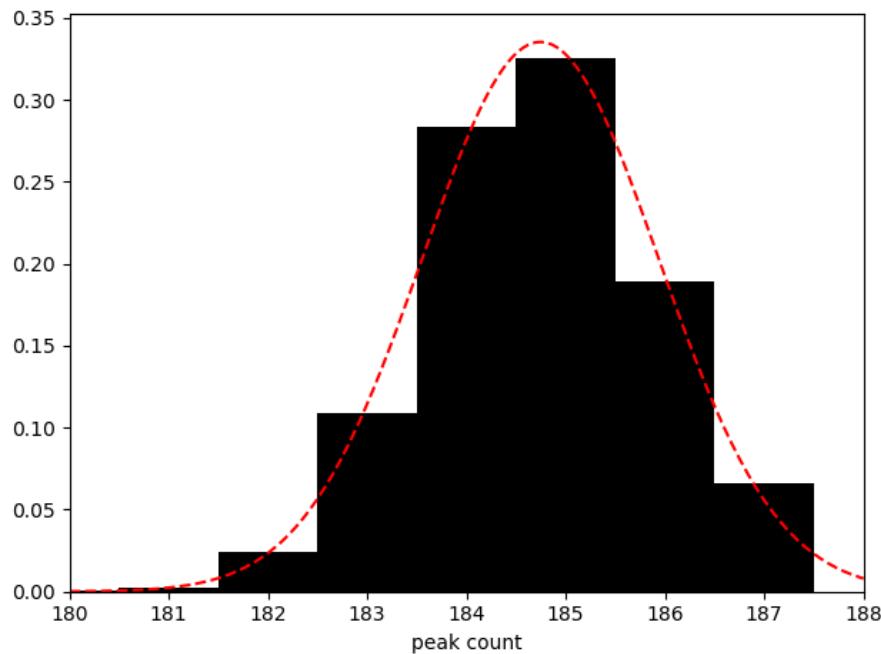


Node 1

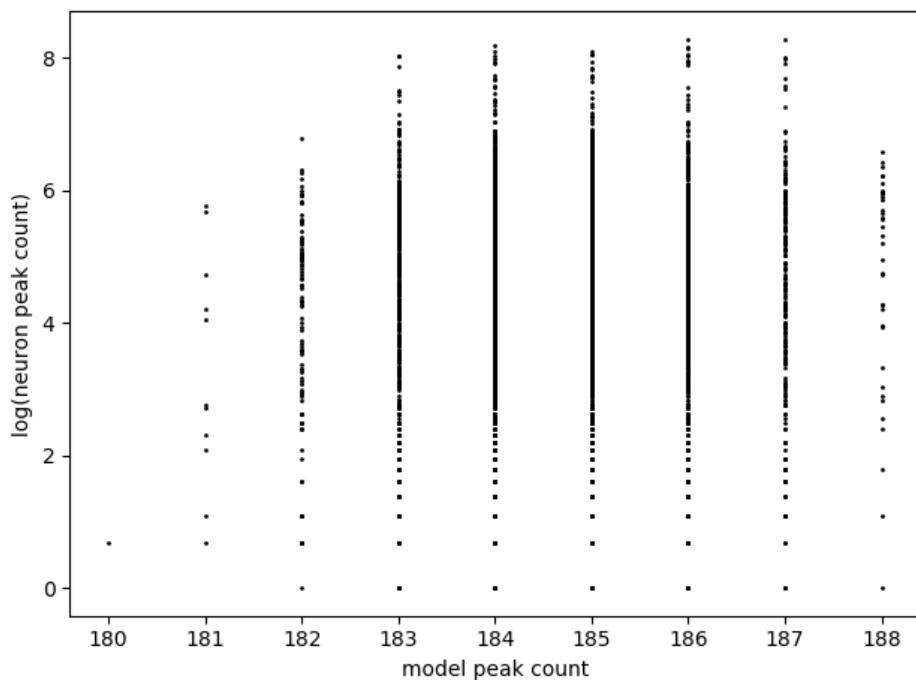


- Distribution of number of peaks (near normal) [over 1e6 steps]

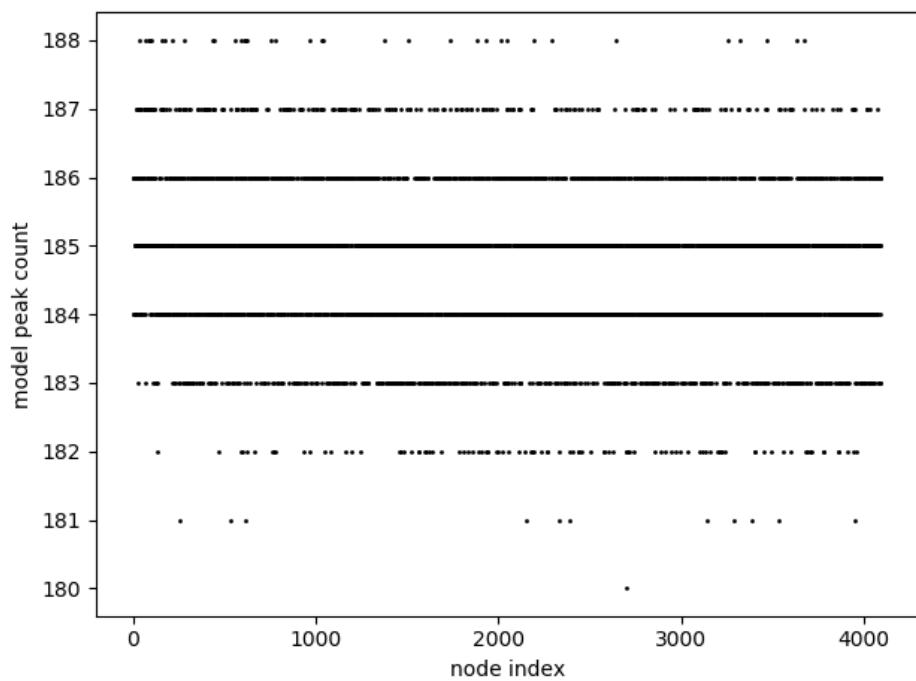
- Median = 185, Min = 180, Max = 188
- Mean = 184.7443 < Median
- Skewness = -0.0100
- (Excess) Kurtosis = 0.0148
- Dotted line: Gaussian with same mean and s.d.



- Model peaks vs experimentally measured peaks (not similar)
 - Log scale for y-axis

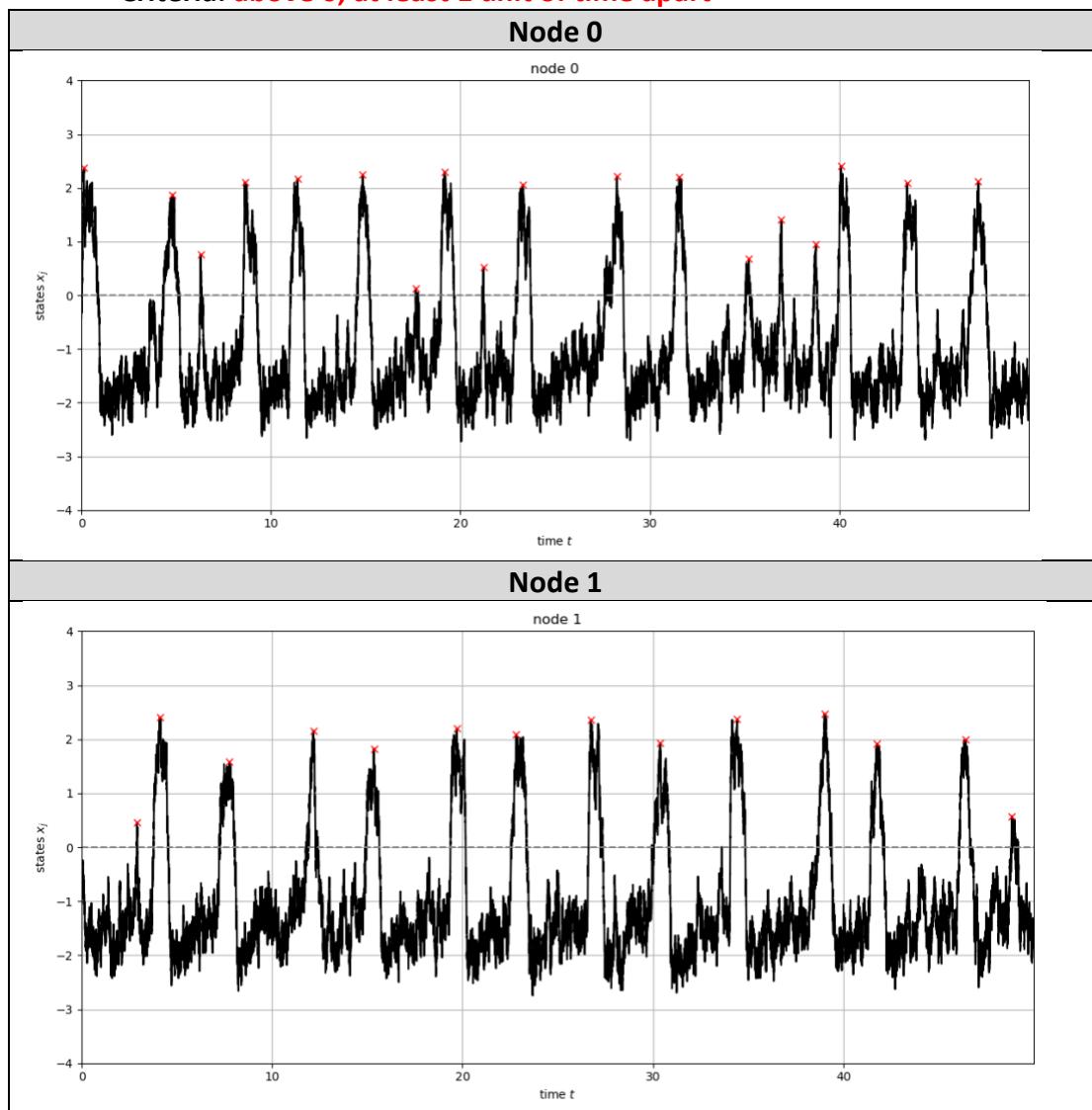


- Model peaks vs node index



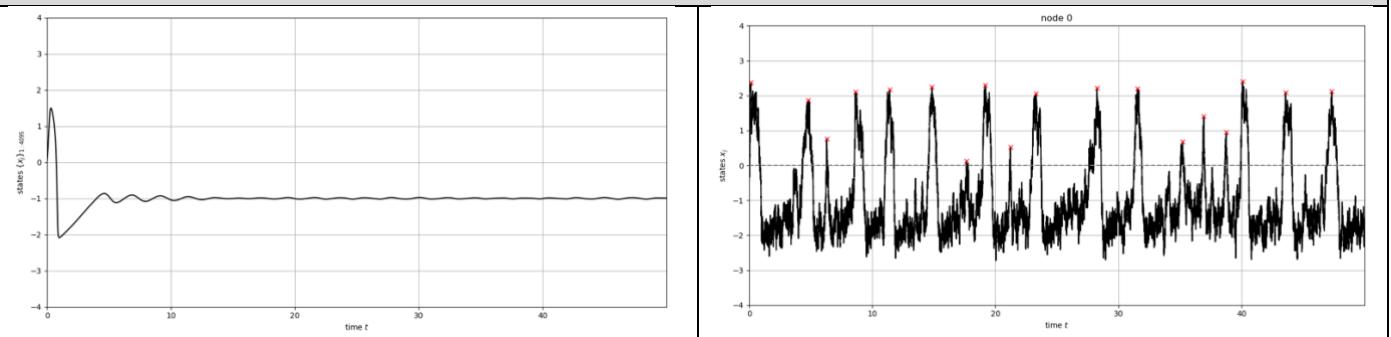
More spike analysis (other parameters)

- Using g_{ij} from “DIV25_PREmethod” (no multiplier)
- Parameters $\epsilon = 0.1$, $\alpha = 1$, $\sigma_i = 2$
- Examples of peaks (node 0 & 1)
 - Criteria: above 0; at least 1 unit of time apart

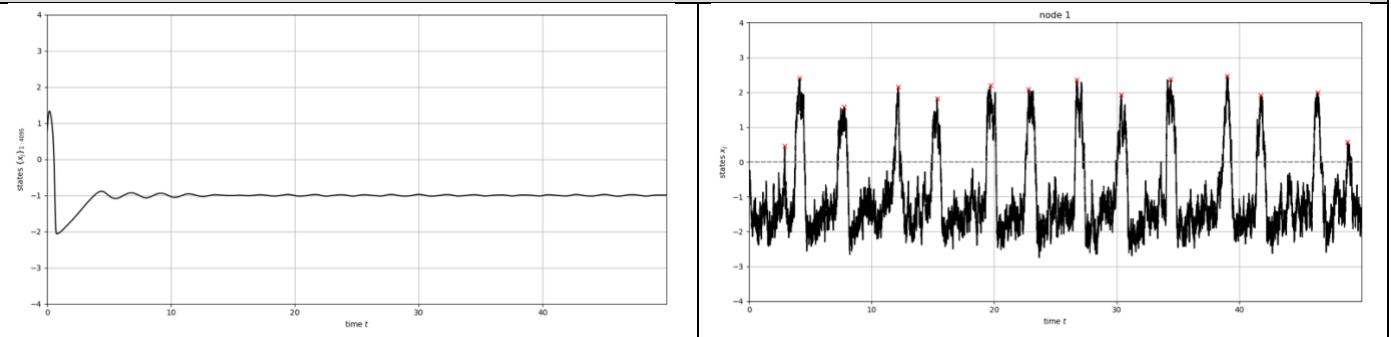


- Noise-free time series as comparison

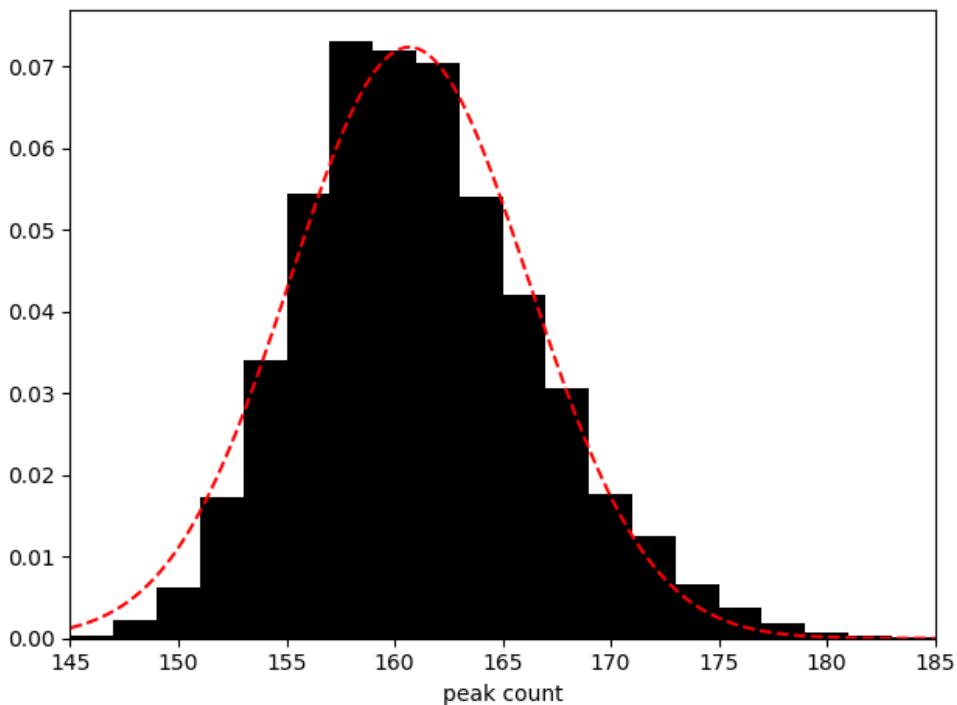
Node 0



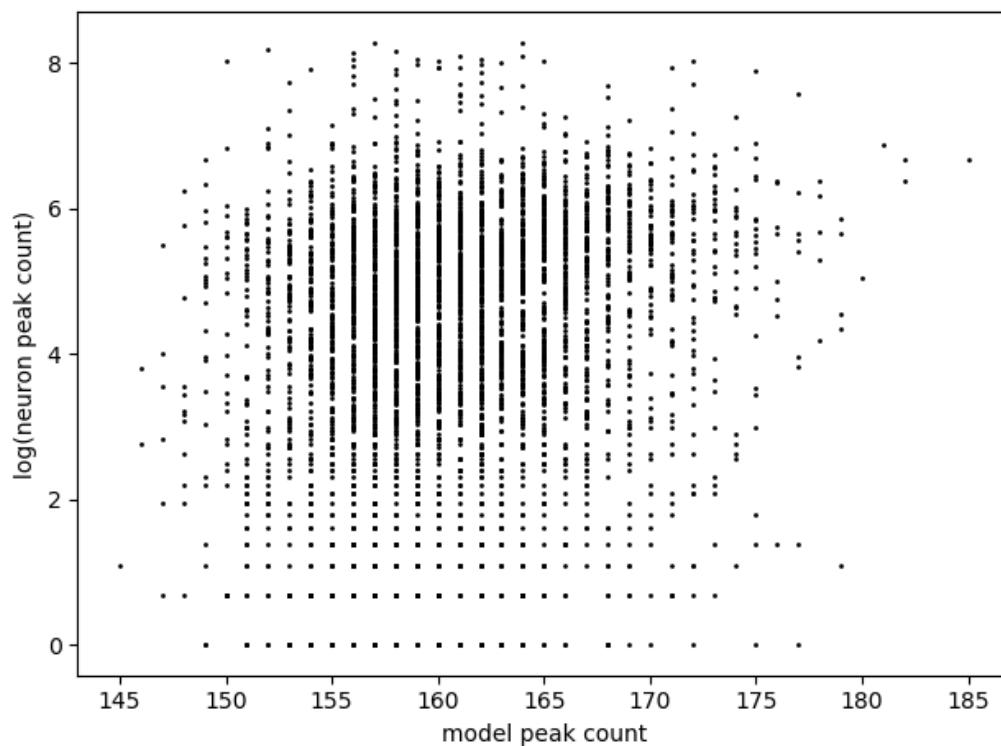
Node 1



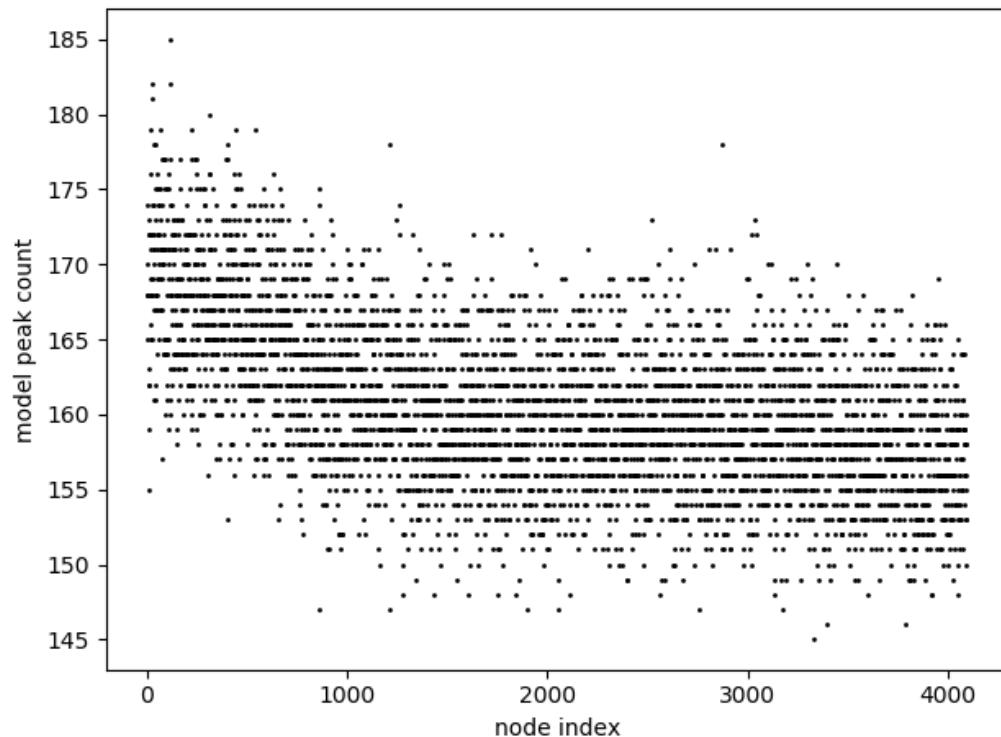
- Distribution of number of peaks (slightly heavy-tailed) [over 1e6 steps]
 - Median = 160, Min = 145, Max = 185 (Max-Median > Median-Min)
 - Mean = 160.6794 > Median
 - Skewness = 0.4302 (right-skewed)
 - (Excess) Kurtosis = 0.2224 (heavier tails than Gaussian)
 - Dotted line: Gaussian with same mean and s.d.



- Model peaks vs experimentally measured peaks (not similar)
 - Log scale for y-axis

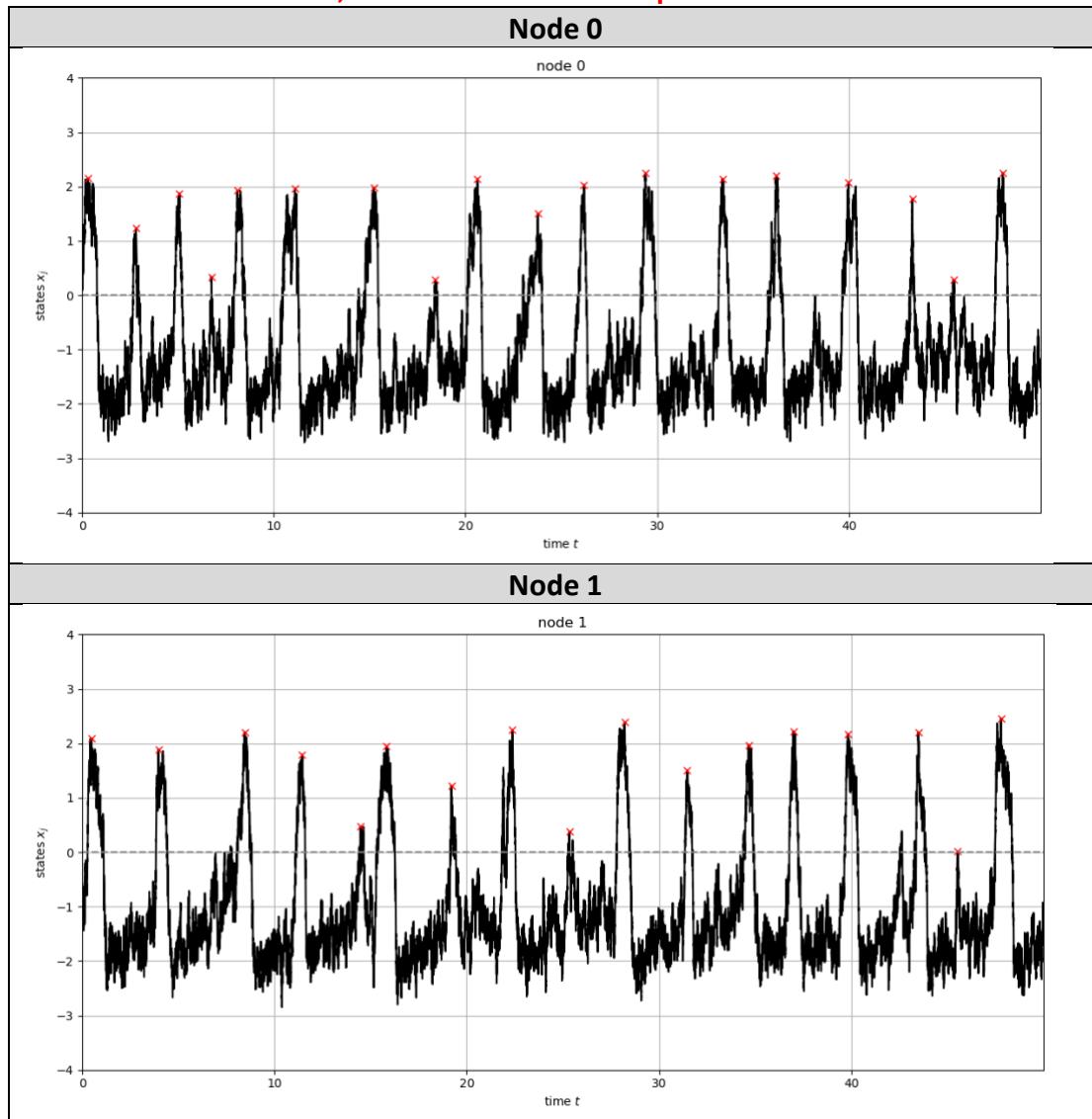


- Model peaks vs node index



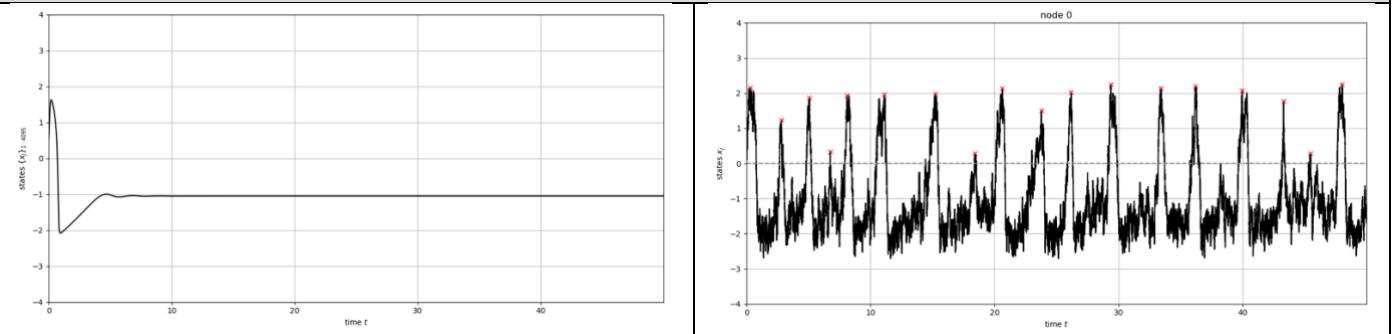
More spike analysis (other parameters)

- Using g_{ij} from “DIV25_PREmethod” (no multiplier)
- Parameters $\epsilon = 0.1$, $\alpha = 1.05$, $\sigma_i = 2$
- Examples of peaks (node 0 & 1)
 - Criteria: above 0; at least 1 unit of time apart

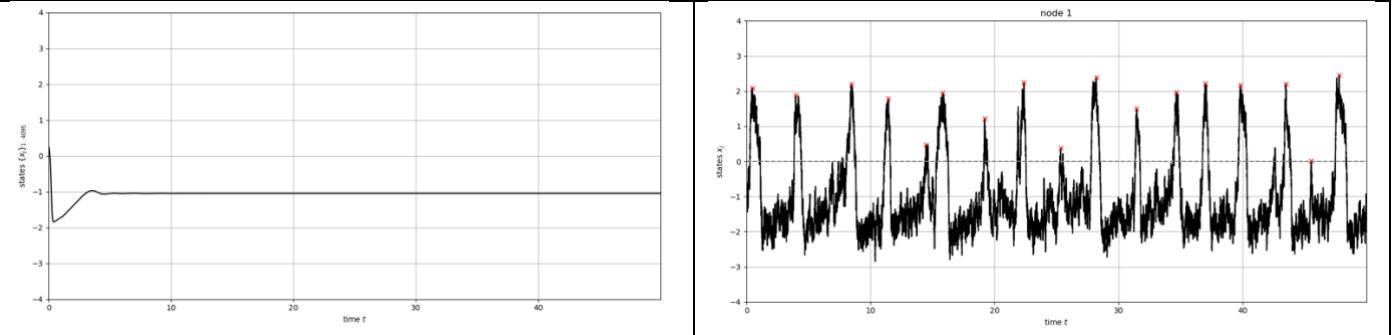


- Noise-free time series as comparison

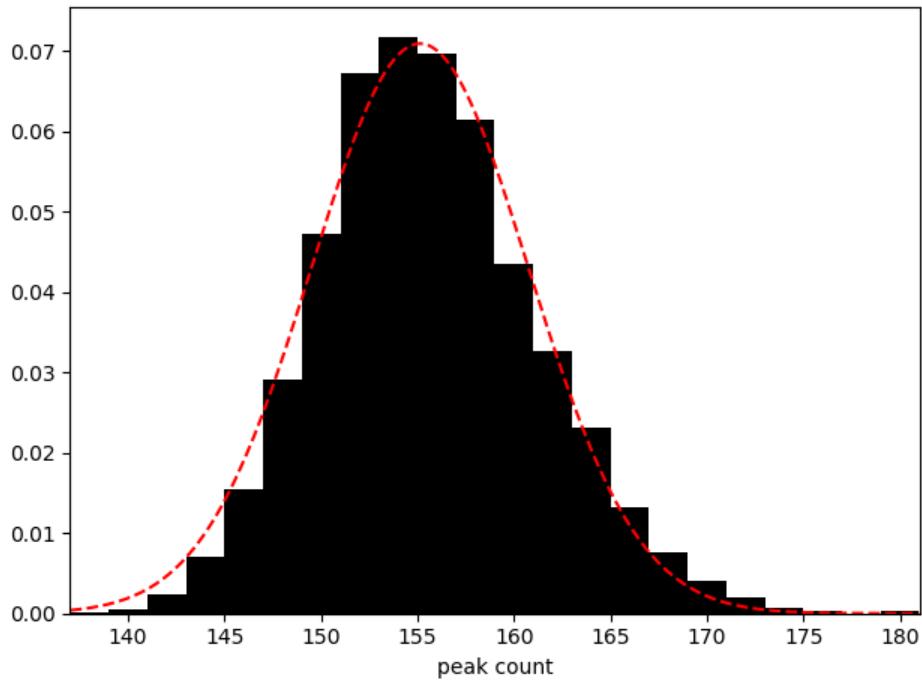
Node 0



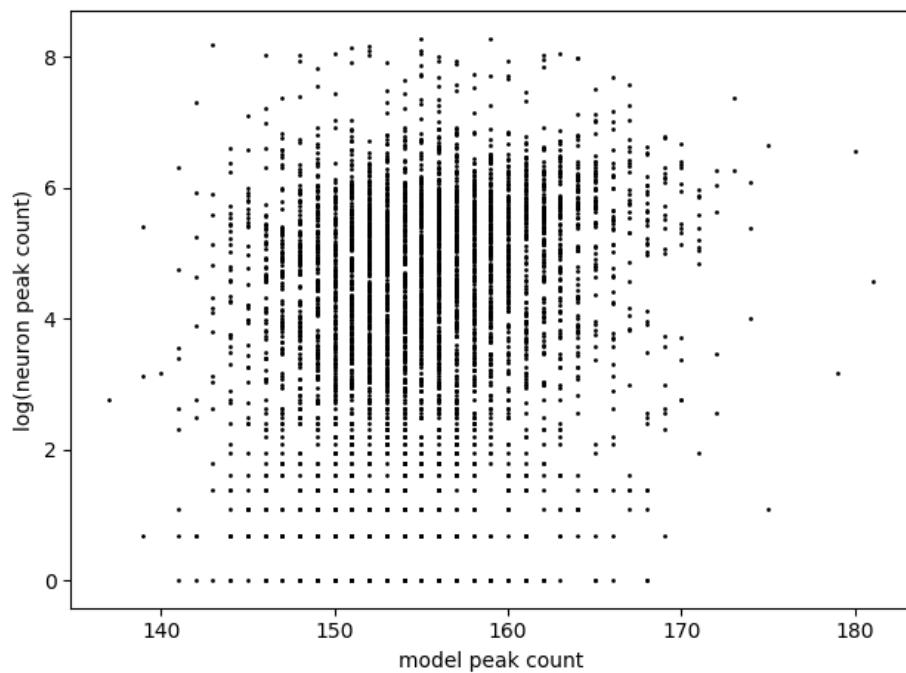
Node 1



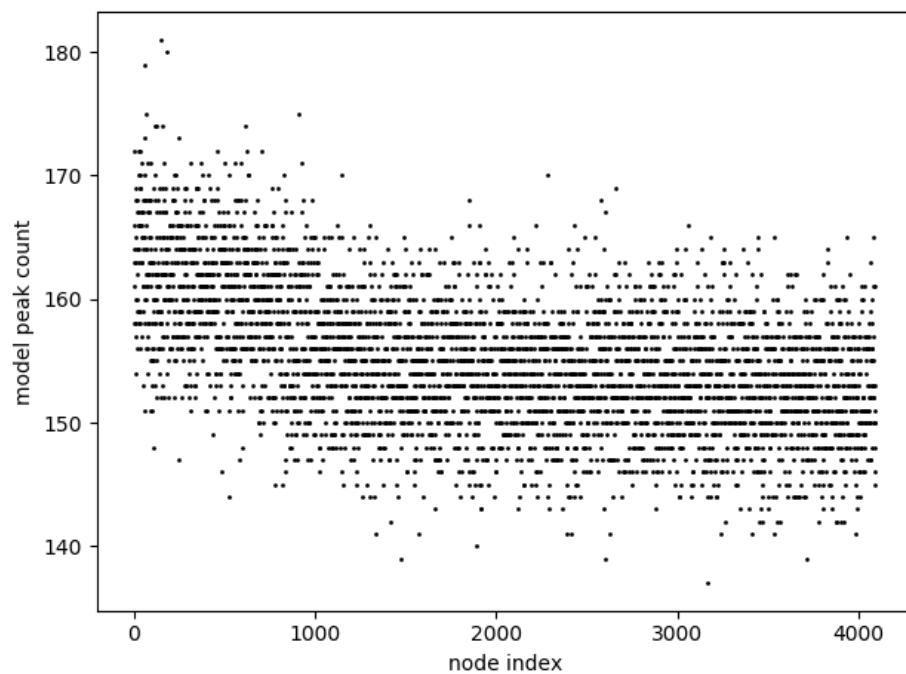
- Distribution of number of peaks (slightly heavy-tailed) [over 1e6 steps]
 - Median = 155, Min = 137, Max = 181 (Max-Median > Median-Min)
 - Mean = 155.1231 > Median
 - Skewness = 0.3537 (right-skewed)
 - (Excess) Kurtosis = 0.2562 (heavier tails than Gaussian)
 - Dotted line: Gaussian with same mean and s.d.



- Model peaks vs experimentally measured peaks (not similar)
 - Log scale for y-axis



- Model peaks vs node index

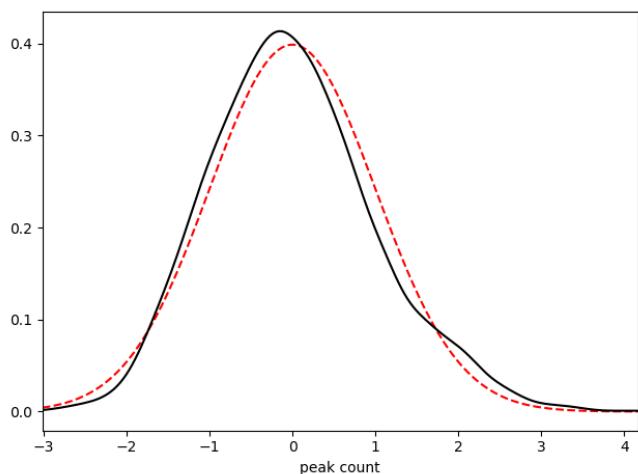


Standardized peak distributions

x-axis: leftmost = min, rightmost = max

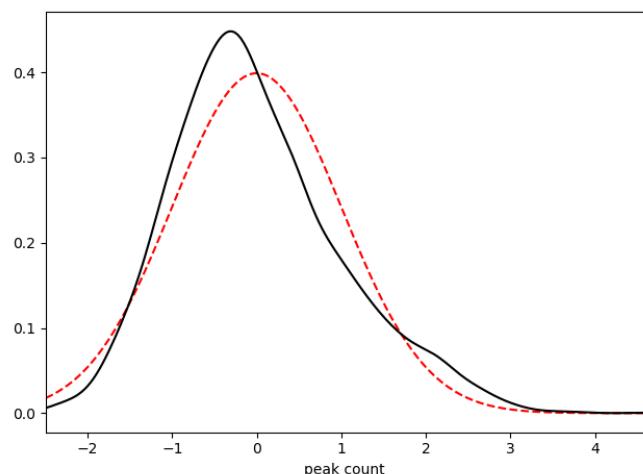
1e6 steps

Case 1: $\epsilon = 0.1, \alpha = 0.95, \sigma_i = 2$



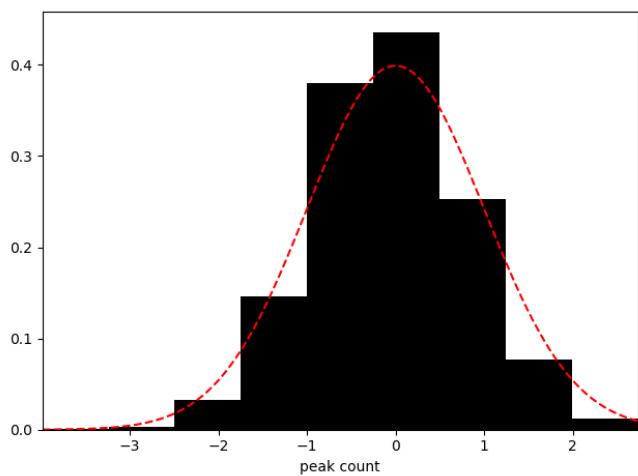
- Median = 165, Min = 149, Max = 189
- Mean = 165.7985
- Skewness = 0.4002 (right-skewed)
- (Excess) Kurtosis = 0.2204 (heavier tails than Gaussian)

2e6 steps

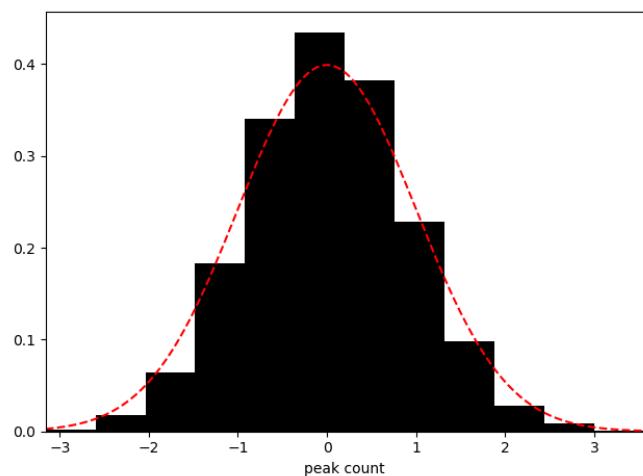


- Median = 330, Min = 308, Max = 374
- Mean = 331.2886
- Skewness = 0.5945 (right-skewed)
- (Excess) Kurtosis = 0.2684 (heavier tails than Gaussian)
- *This is the most right-skewed among the other sets of params*
- *Max = 4 s.d. from mean*

Case 2: $\epsilon = 0.01, \alpha = 0.95, \sigma_i = 2$

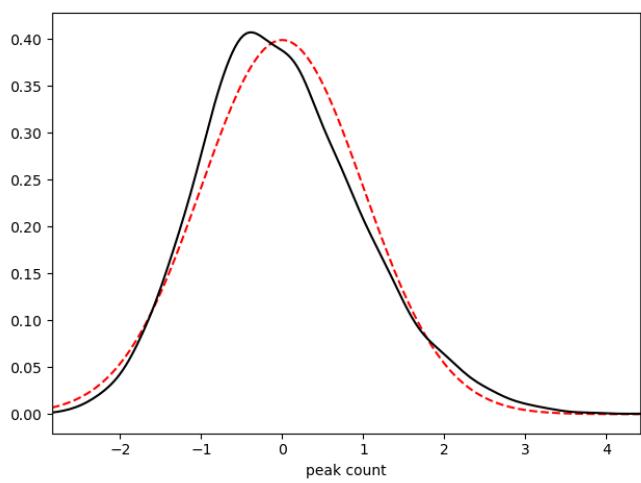


- Median = 185, Min = 180, Max = 188
- Mean = 184.7443
- Skewness = -0.0100
- (Excess) Kurtosis = 0.0148
- Can't properly fit a curve as there are too few data points so I still use histogram

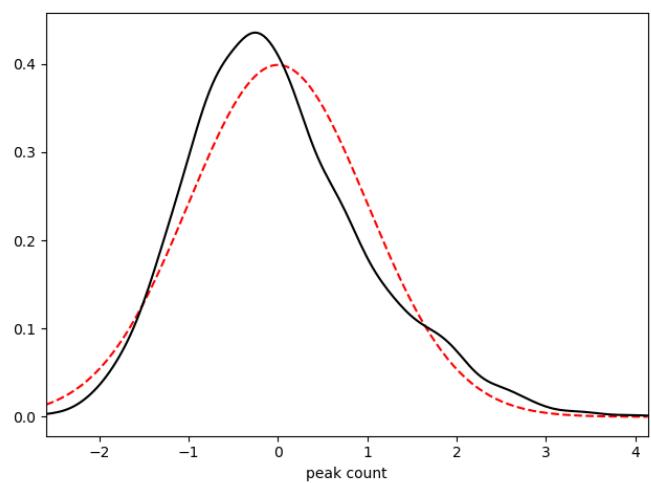


- Median = 369, Min = 364, Max = 375
- Mean = 369.1739
- Skewness = 0.0519
- (Excess) Kurtosis = 0.0033
- Can't properly fit a curve as there are too few data points so I still use histogram

Case 3: $\epsilon = 0.1, \alpha = 1, \sigma_i = 2$

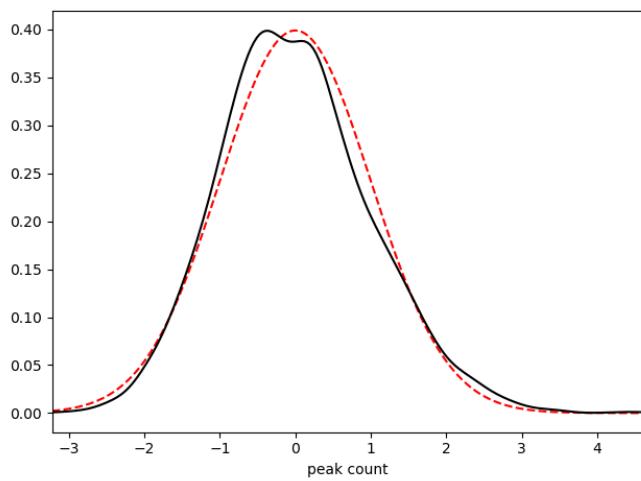


- Median = 160, Min = 145, Max = 185
- Mean = 160.6794
- Skewness = 0.4302 (right-skewed)
- (Excess) Kurtosis = 0.2224 (heavier tails than Gaussian)

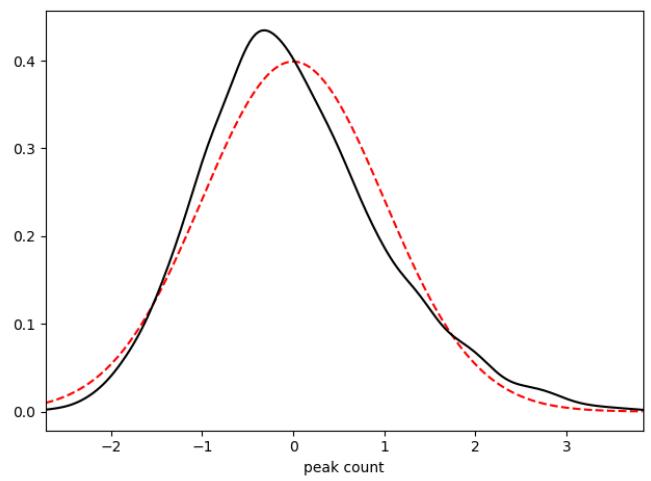


- Median = 320, Min = 297, Max = 359
- Mean = 320.9031
- Skewness = 0.5982 (right-skewed)
- (Excess) Kurtosis = 0.3511 (heavier tails than Gaussian)

Case 4: $\epsilon = 0.1, \alpha = 1.05, \sigma_i = 2$

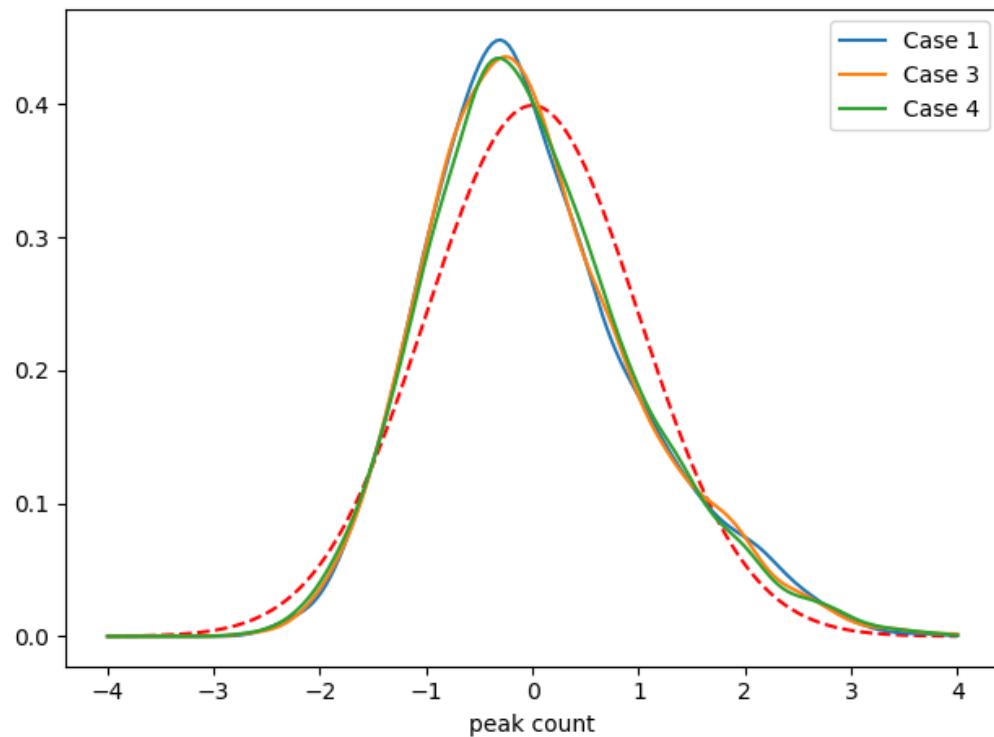


- Median = 155, Min = 137, Max = 181
- Mean = 155.1231
- Skewness = 0.3537 (right-skewed)
- (Excess) Kurtosis = 0.2562 (heavier tails than Gaussian)



- Median = 309, Min = 285, Max = 345
- Mean = 309.8799
- Skewness = 0.5384 (right-skewed)
- (Excess) Kurtosis = 0.3200 (heavier tails than Gaussian)

Case 1,3,4 combined

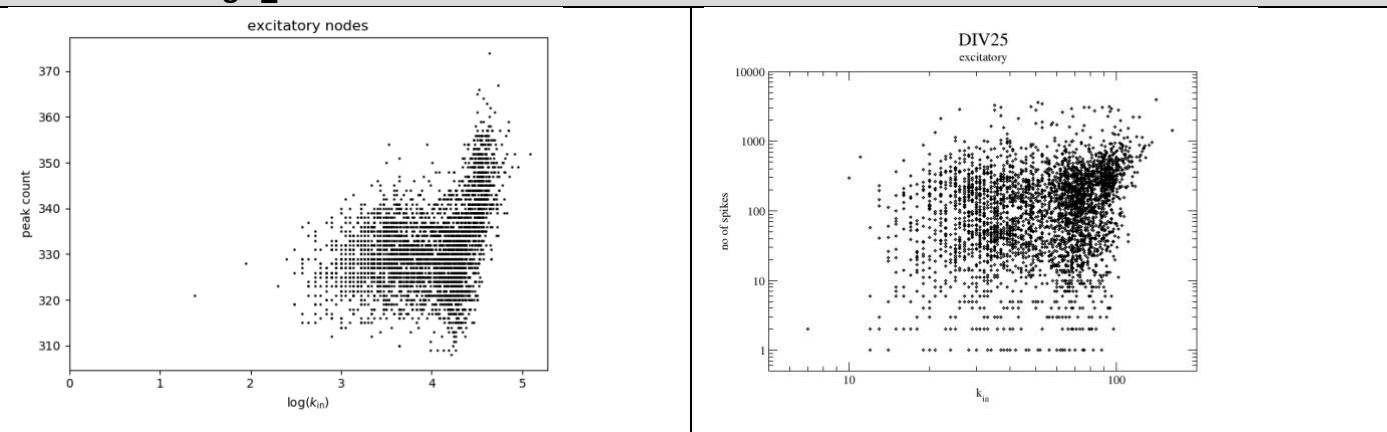


Peak count vs network features (2e6 steps)

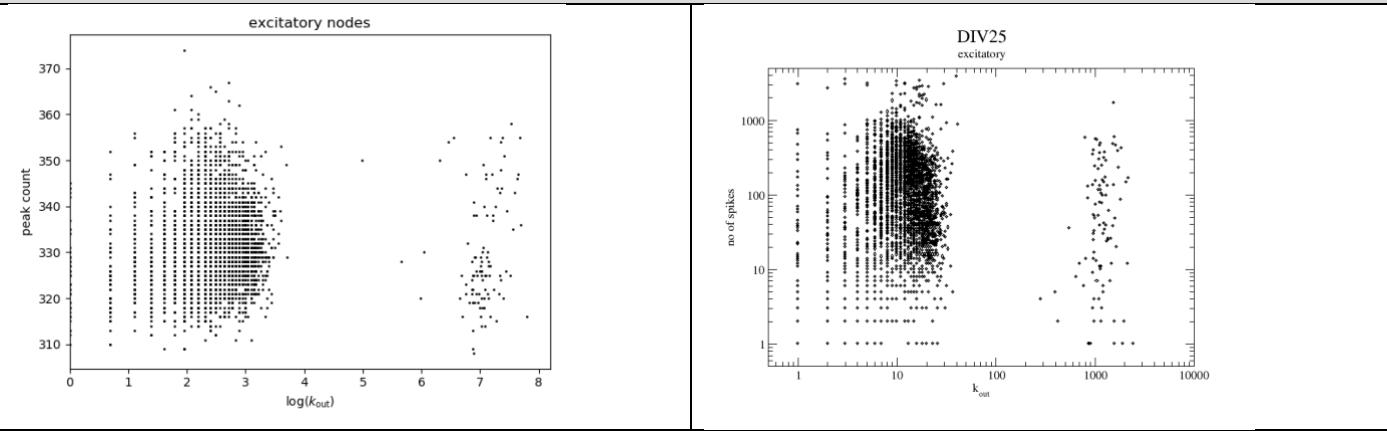
Case 1: $\epsilon = 0.1, \alpha = 0.95, \sigma_i = 2$

- **Excitatory nodes**

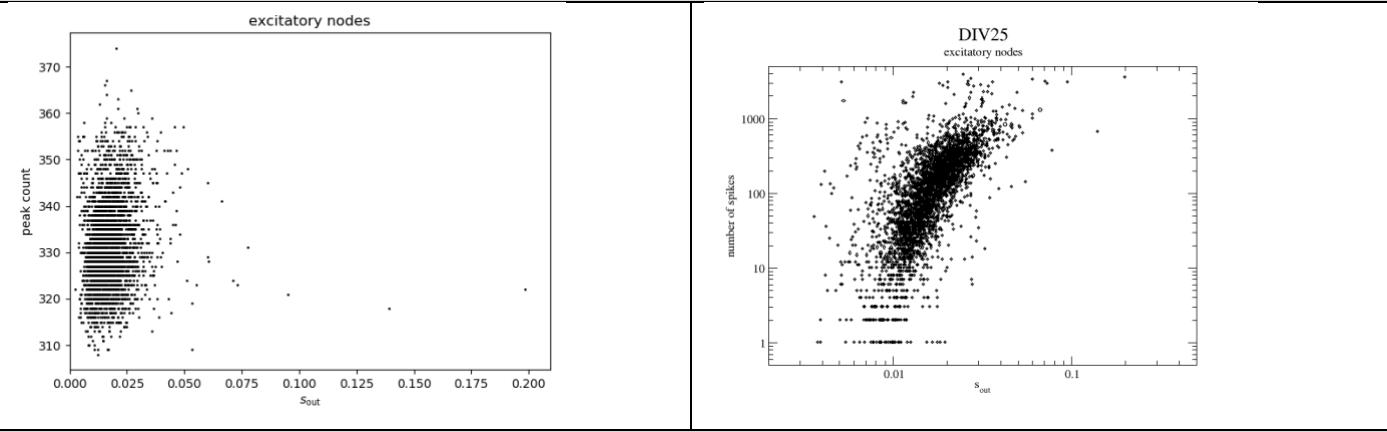
Peak count vs log k_in



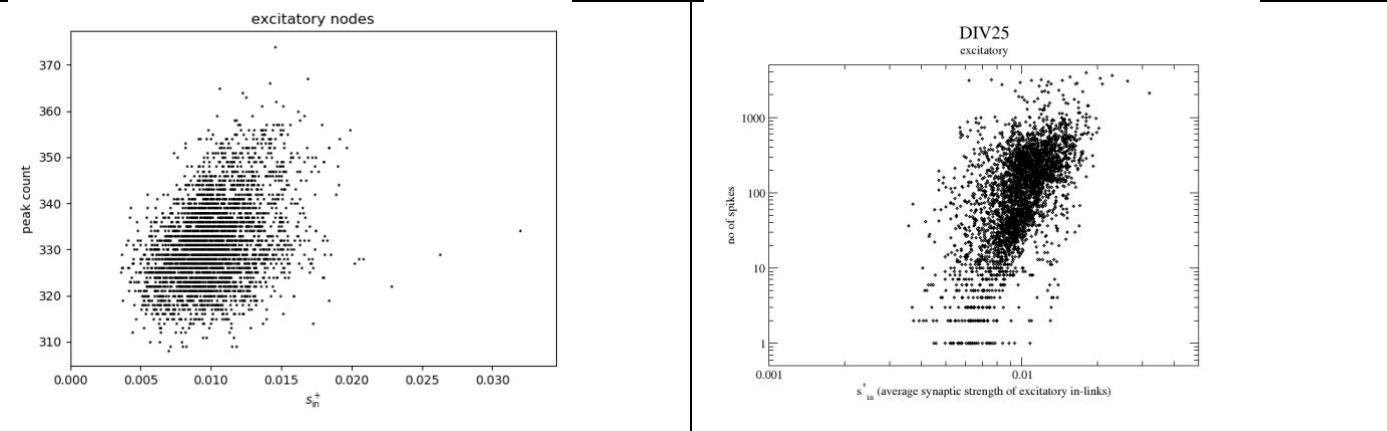
Peak count vs log k_out



Peak count vs s_out

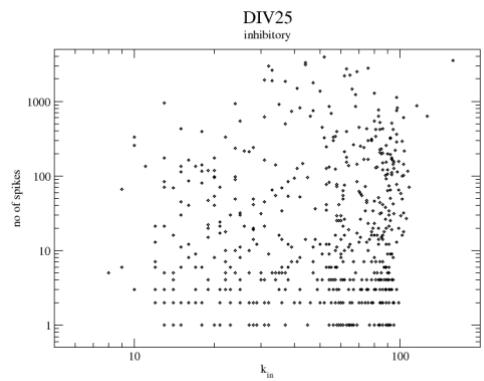
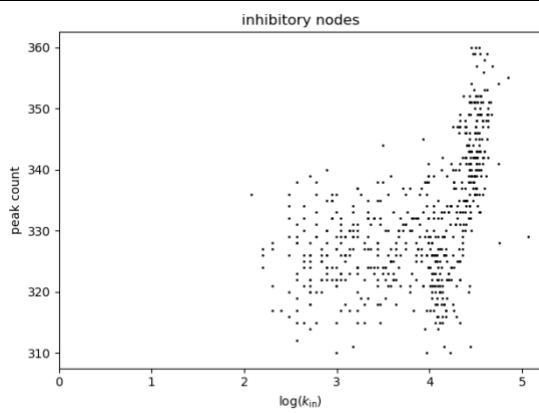


Peak count vs s⁺_in

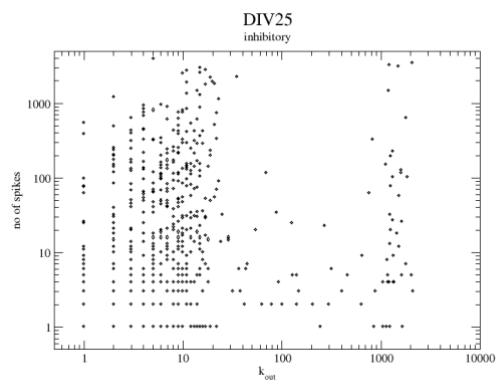
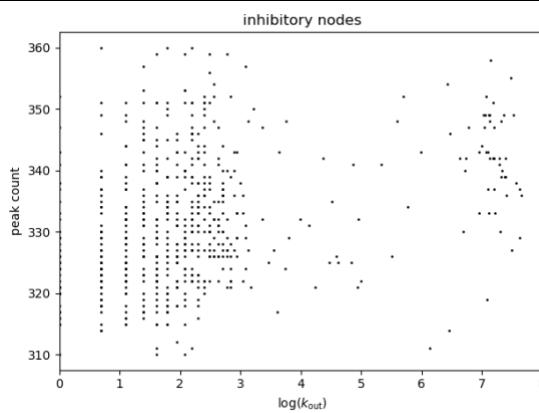


- Inhibitory nodes

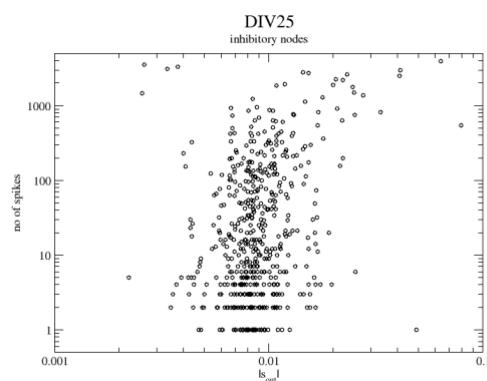
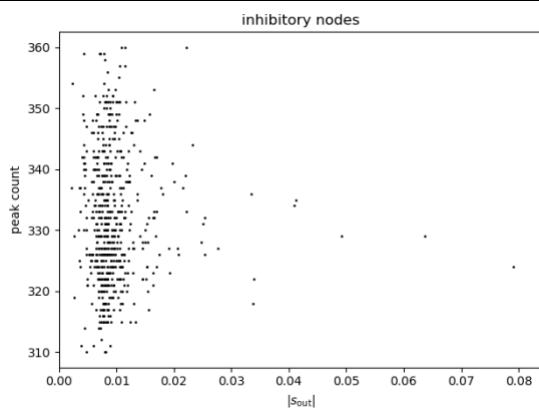
Peak count vs log k_in



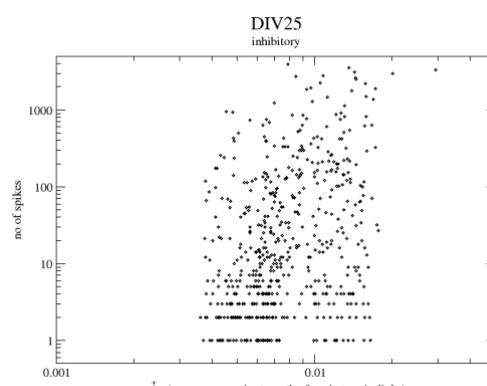
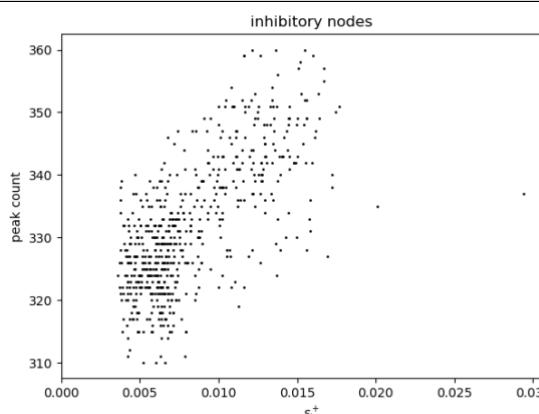
Peak count vs log k_out



Peak count vs |s_out|

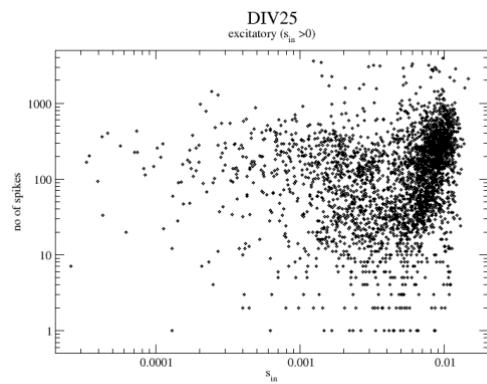
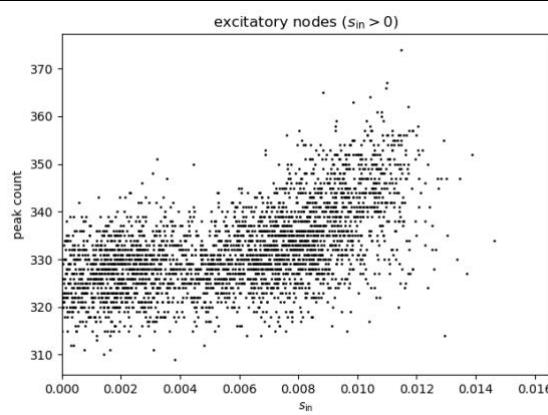


Peak count vs s^+_in

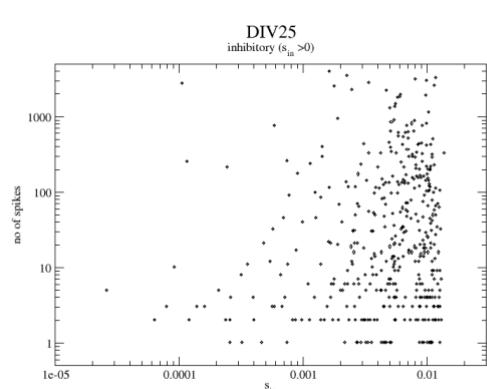
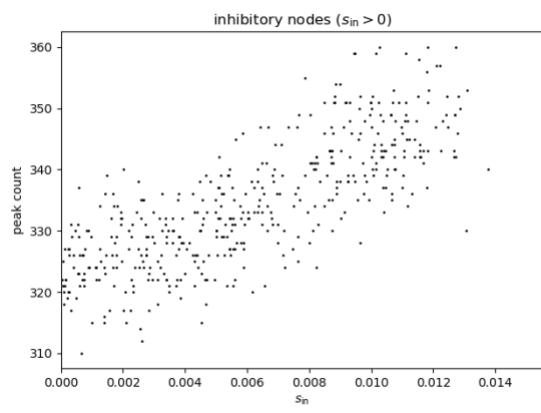


- Mixed (x-axis: s_{in} or $|s_{in}|$)**

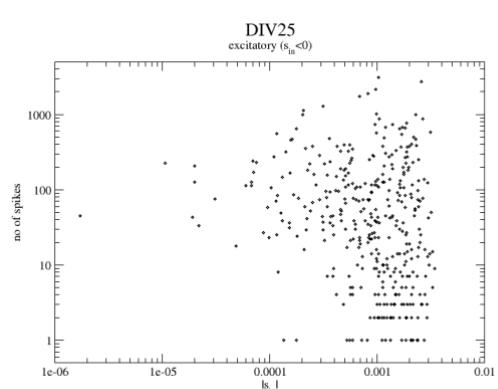
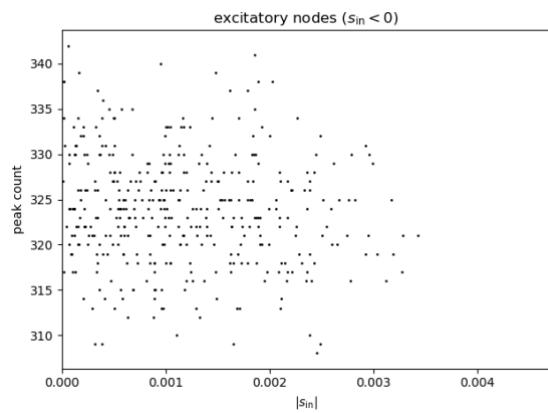
$s_{in} > 0 \& s_{out} > 0$



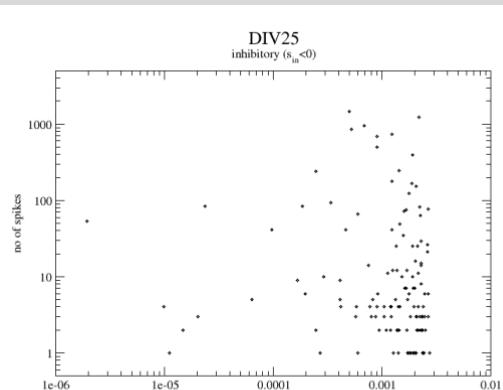
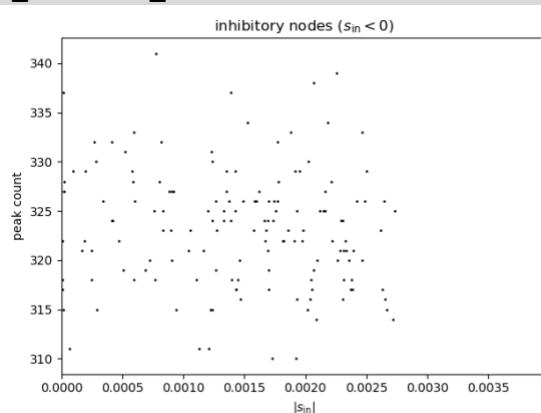
$s_{in} > 0 \& s_{out} < 0$



$s_{in} < 0 \& s_{out} > 0$



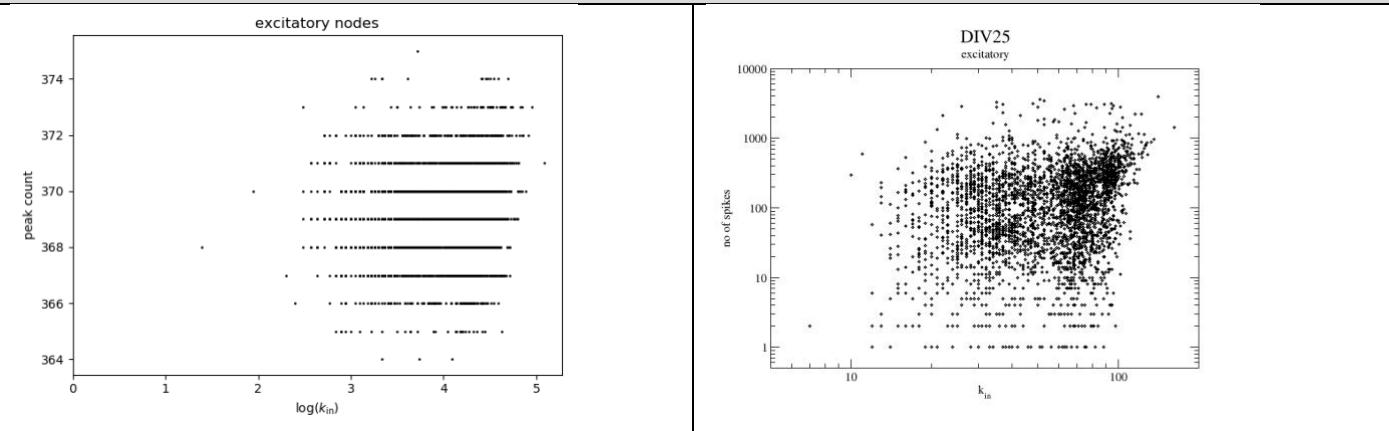
$s_{in} < 0 \& s_{out} < 0$



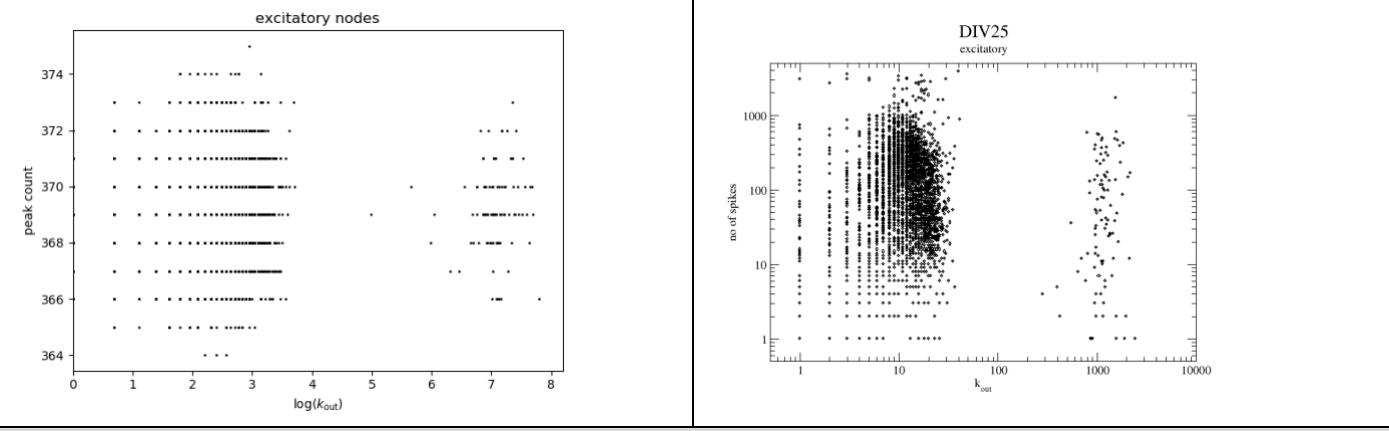
Case 1: $\epsilon = 0.01, \alpha = 0.95, \sigma_i = 2$

- **Excitatory nodes**

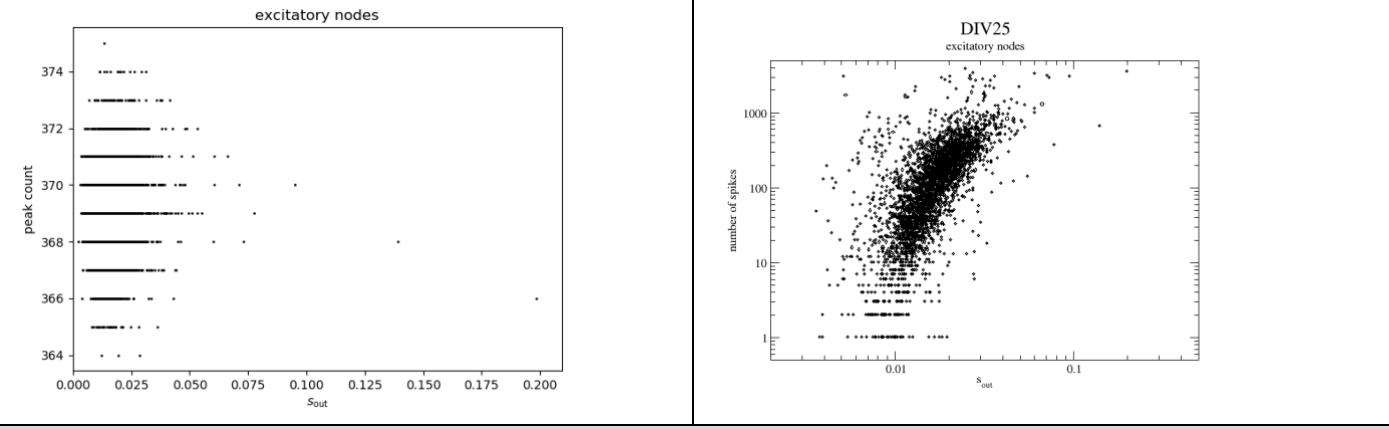
Peak count vs log k_in



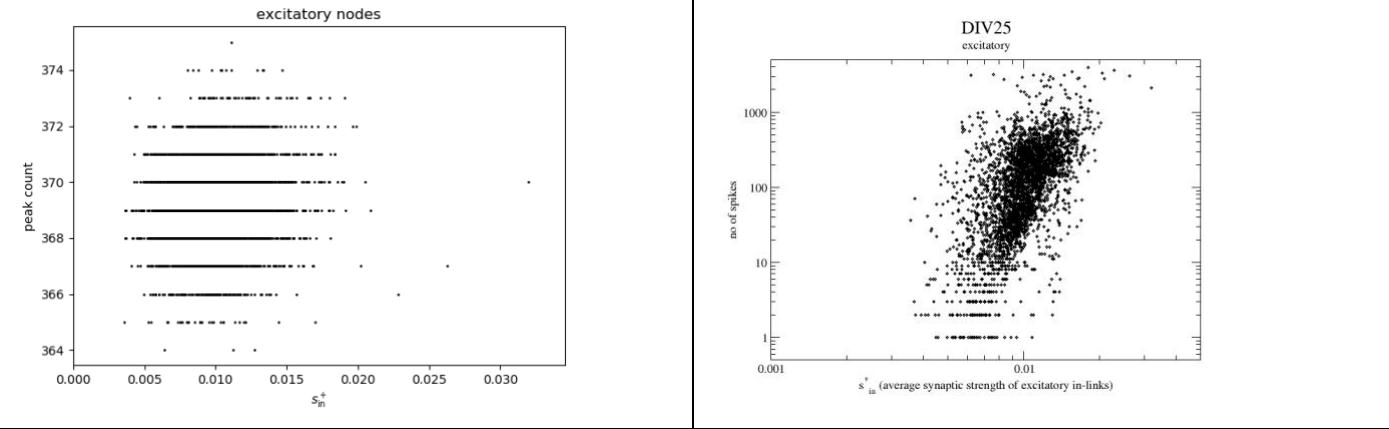
Peak count vs log k_out



Peak count vs s_out

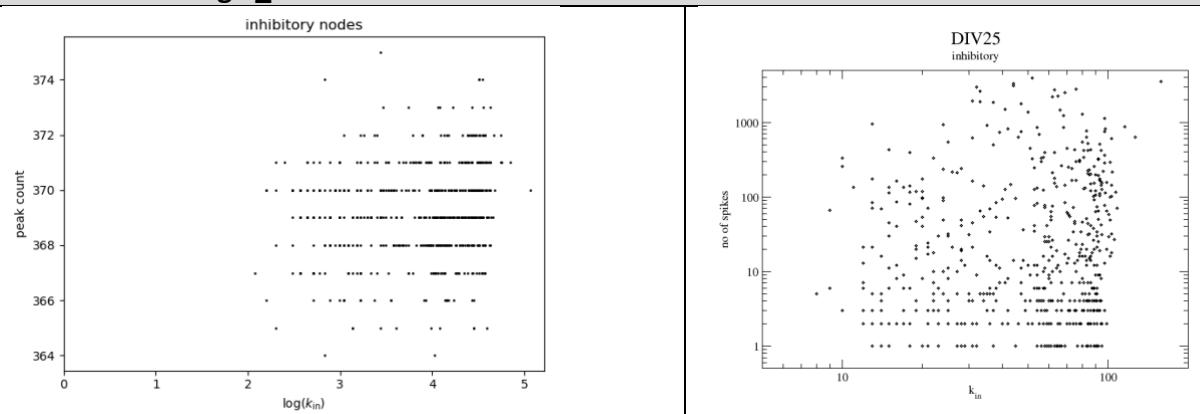


Peak count vs s^+_in

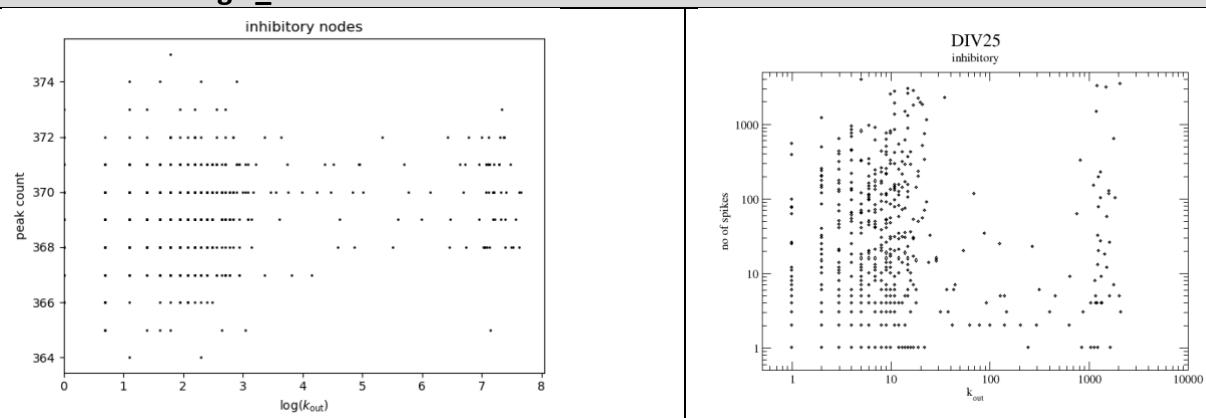


- Inhibitory nodes

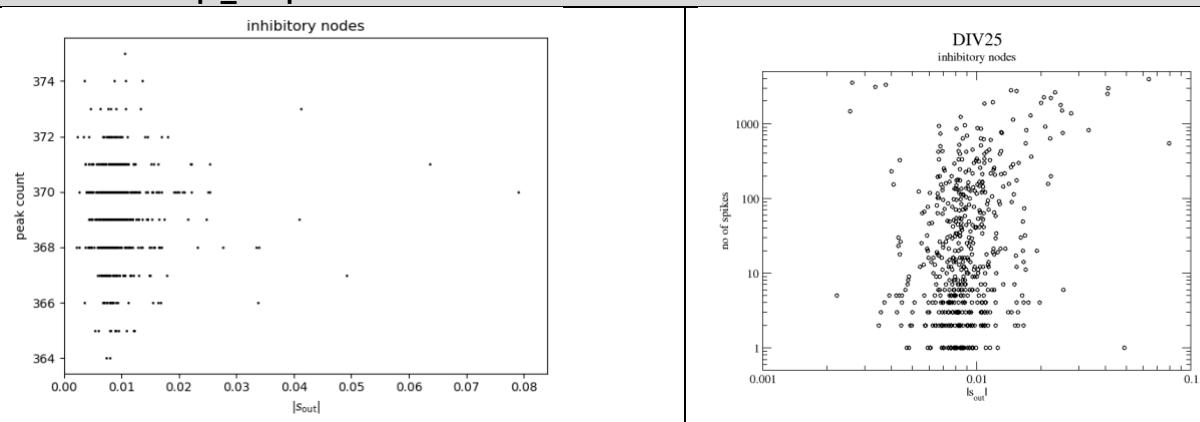
Peak count vs log k_in



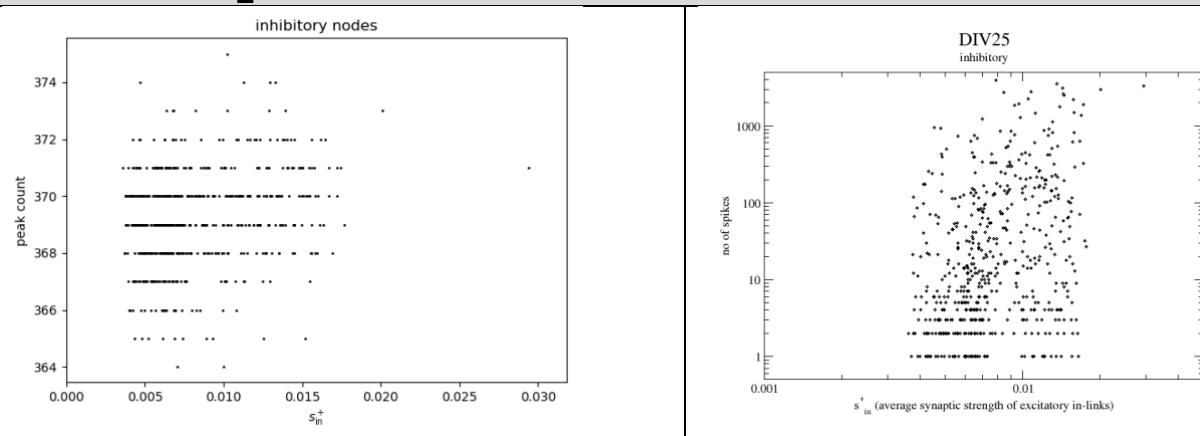
Peak count vs log k_out



Peak count vs |s_out|

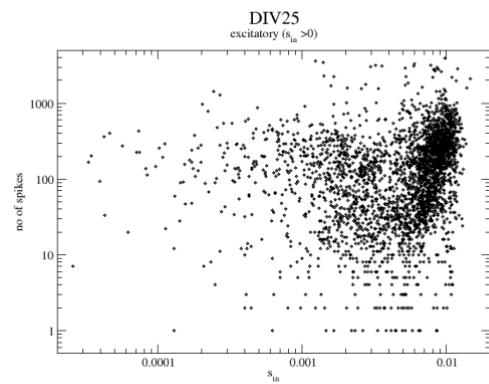
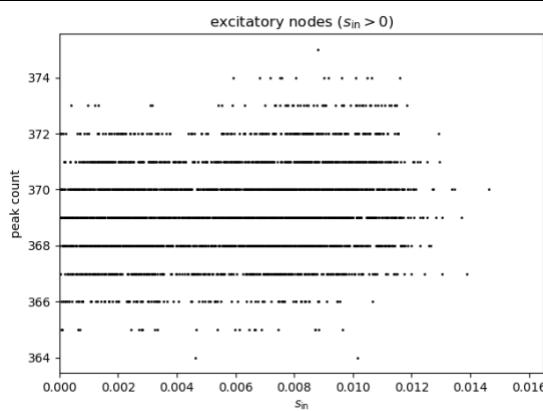


Peak count vs s^+_in

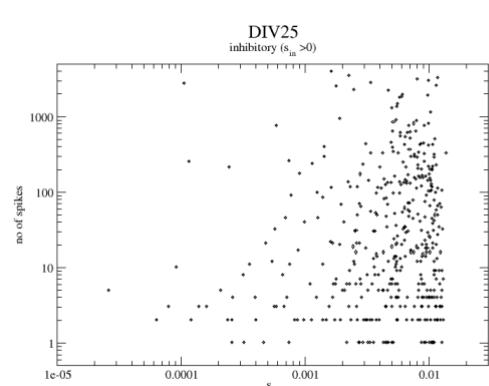
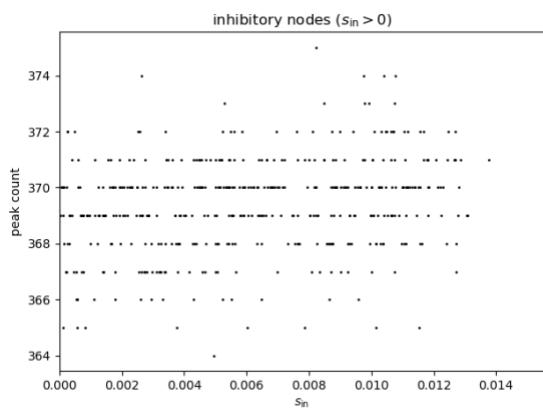


- **Mixed (x-axis: s_{in} or $|s_{in}|$)**

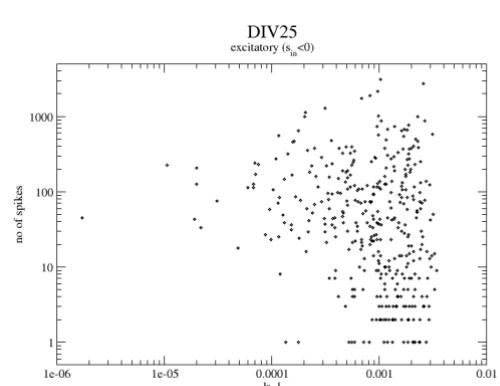
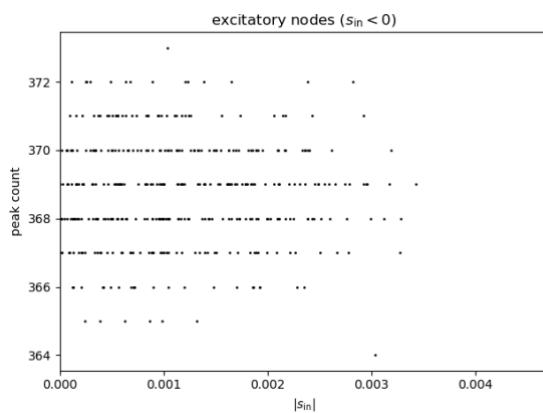
$s_{in} > 0 \& s_{out} > 0$



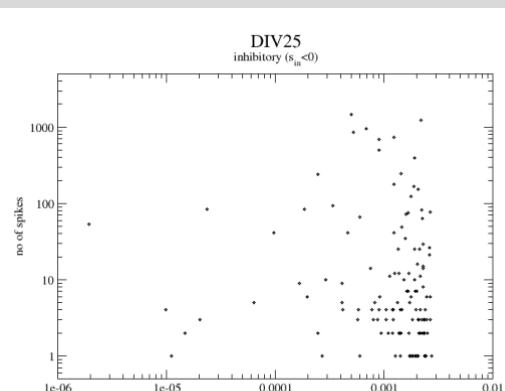
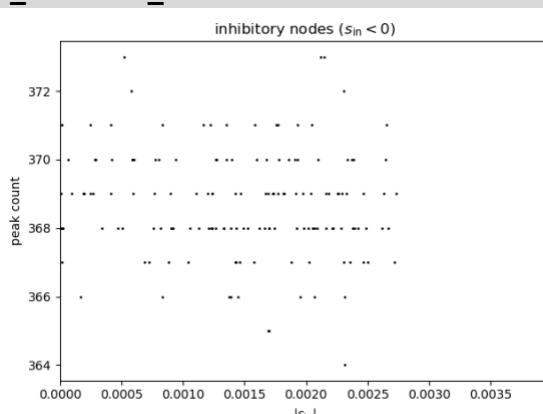
$s_{in} > 0 \& s_{out} < 0$



$s_{in} < 0 \& s_{out} > 0$



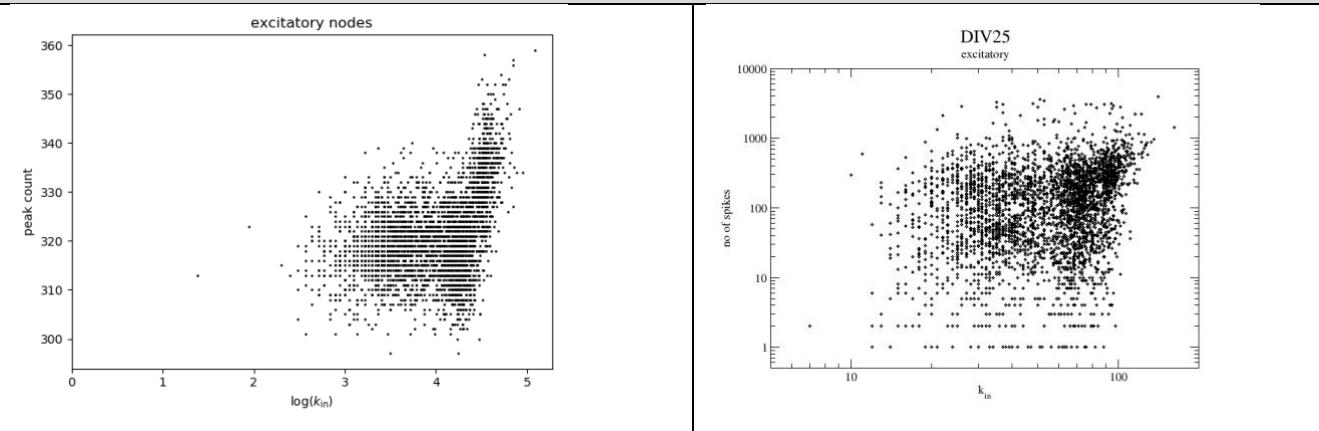
$s_{in} < 0 \& s_{out} < 0$



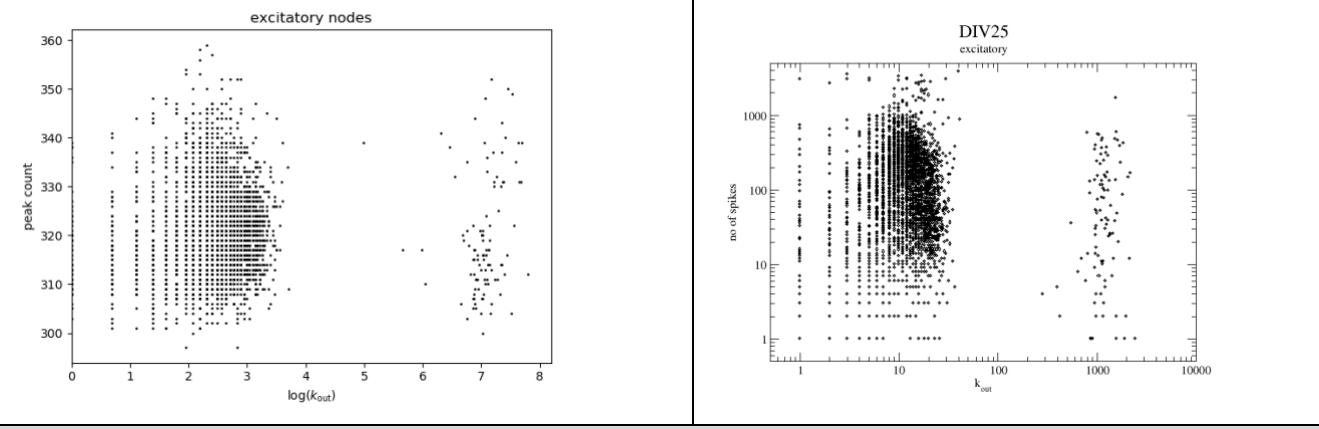
Case 1: $\epsilon = 0.1, \alpha = 1, \sigma_i = 2$

- **Excitatory nodes**

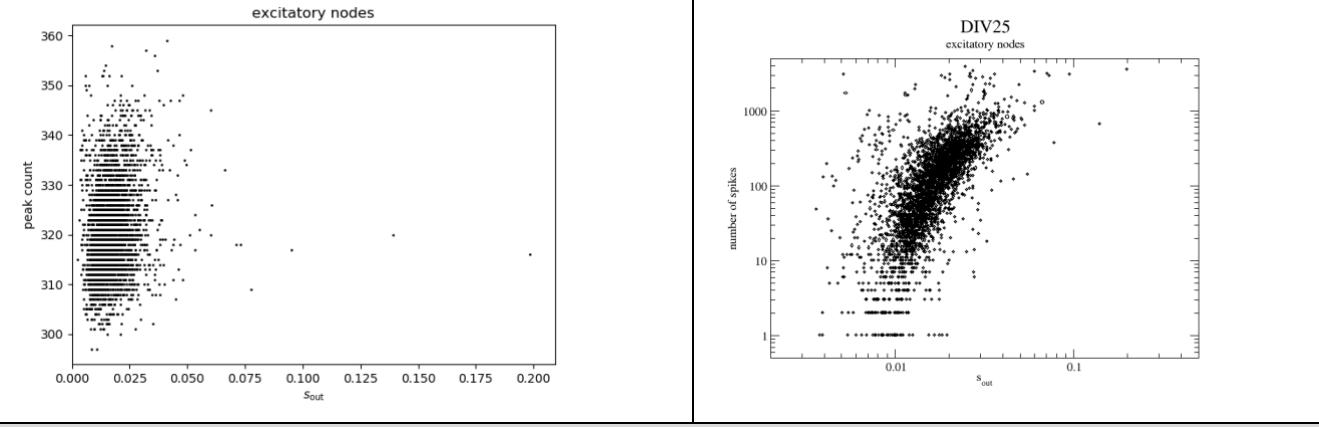
Peak count vs log k_in



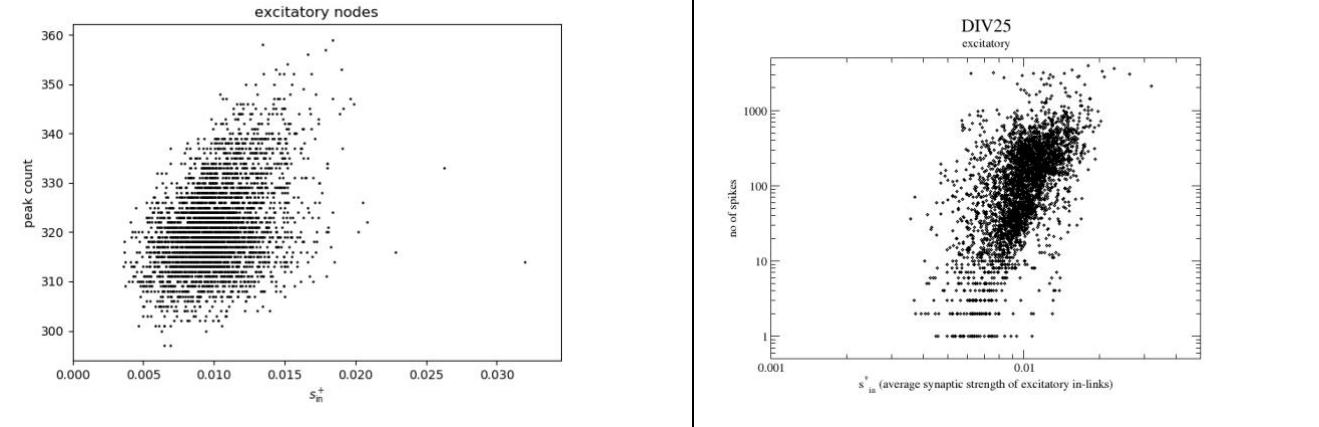
Peak count vs log k_out



Peak count vs s_out

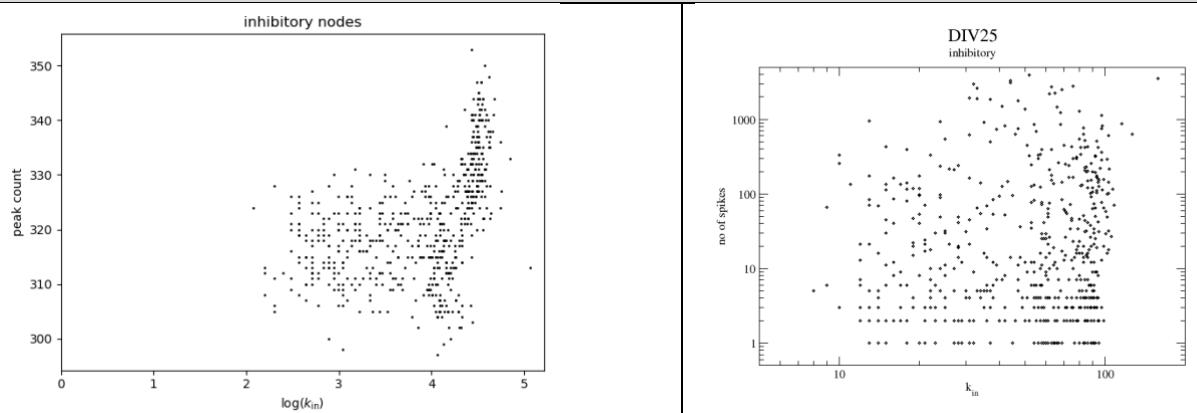


Peak count vs s^+_in

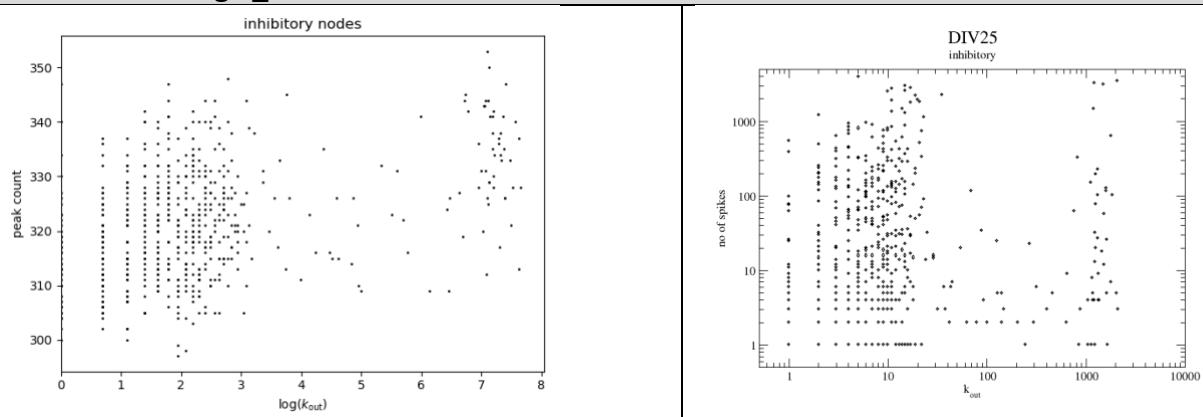


- Inhibitory nodes

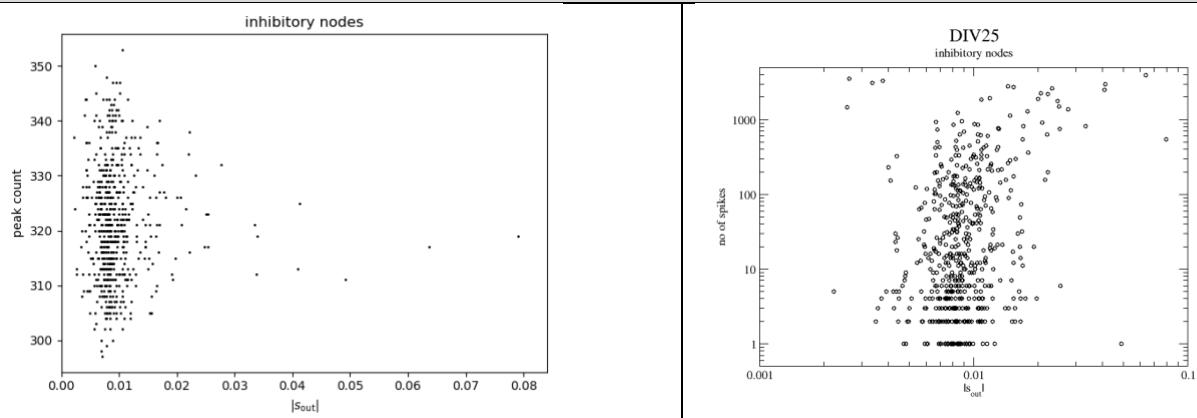
Peak count vs log k_in



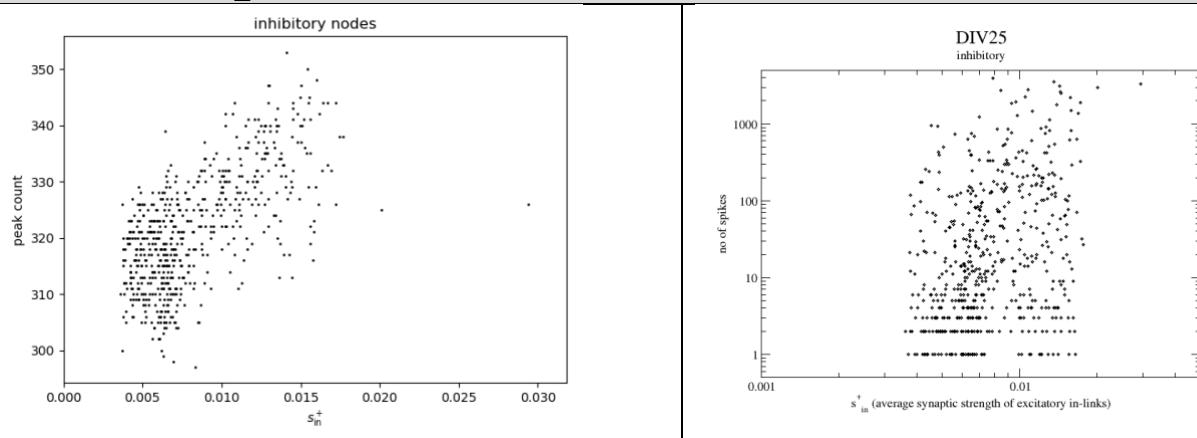
Peak count vs log k_out



Peak count vs |s_out|

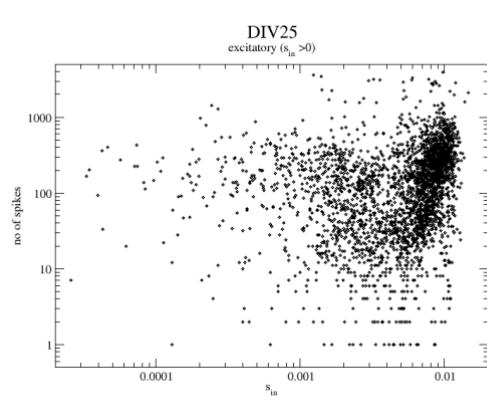
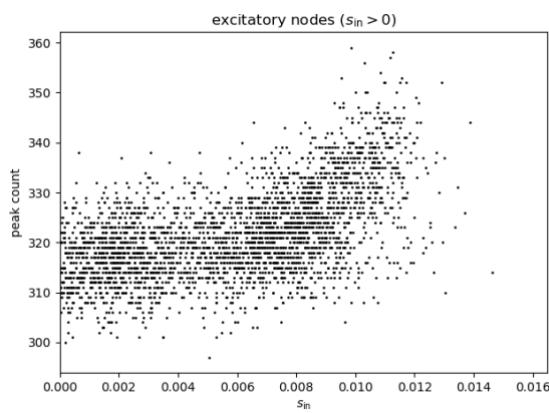


Peak count vs s^+_in

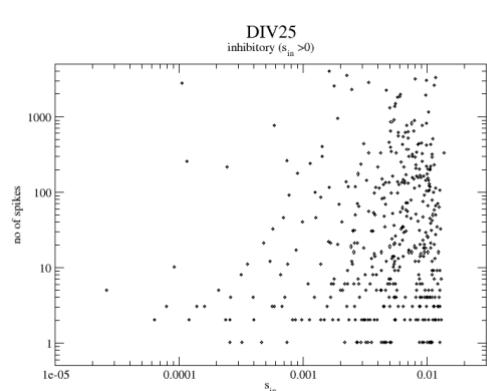
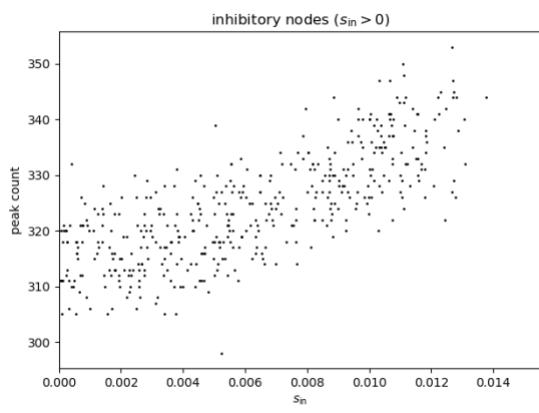


- Mixed (x-axis: s_{in} or $|s_{in}|$)**

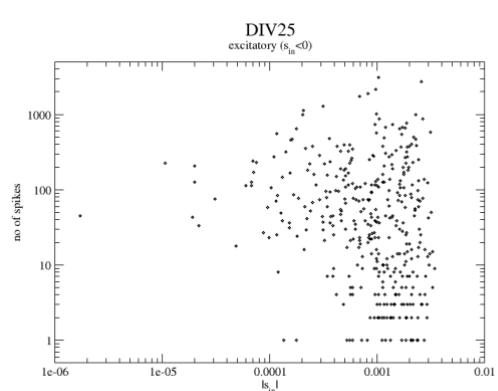
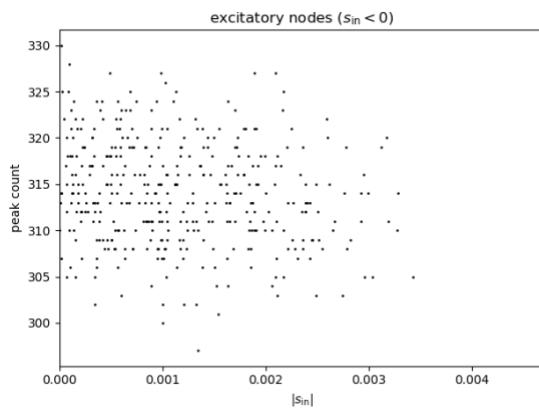
$s_{in} > 0 \& s_{out} > 0$



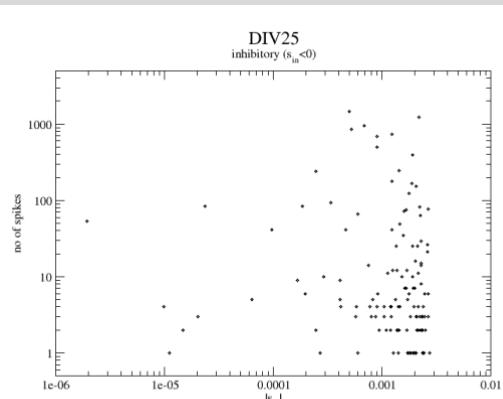
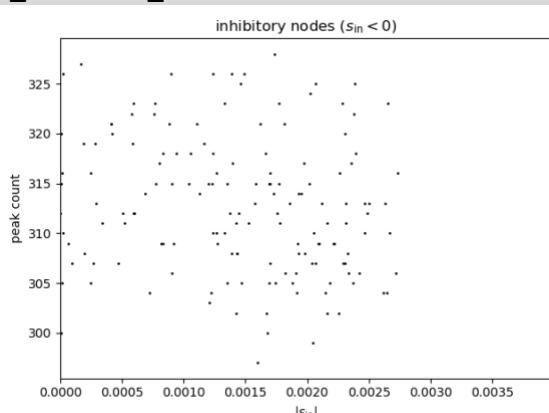
$s_{in} > 0 \& s_{out} < 0$



$s_{in} < 0 \& s_{out} > 0$



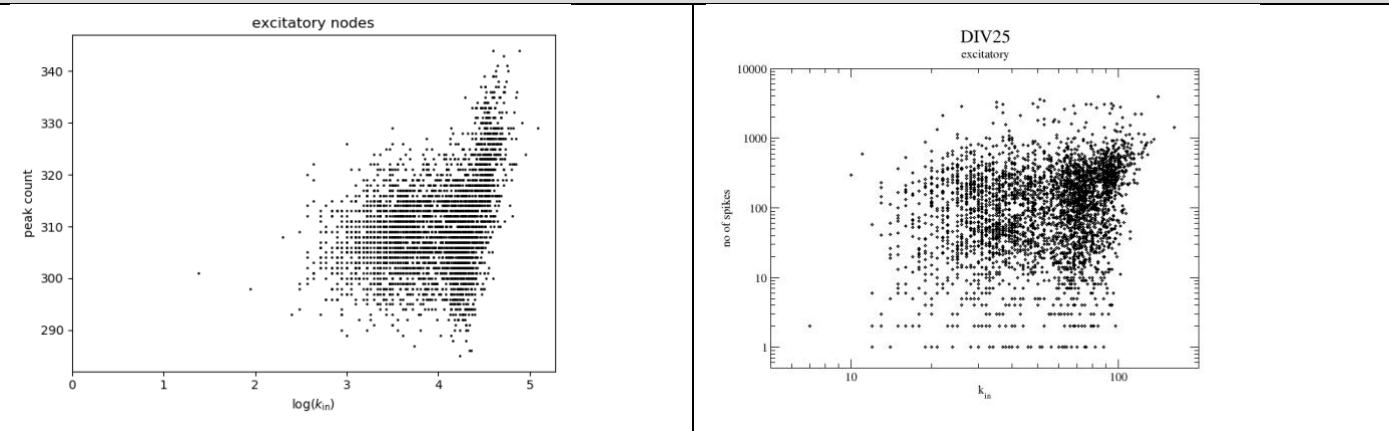
$s_{in} < 0 \& s_{out} < 0$



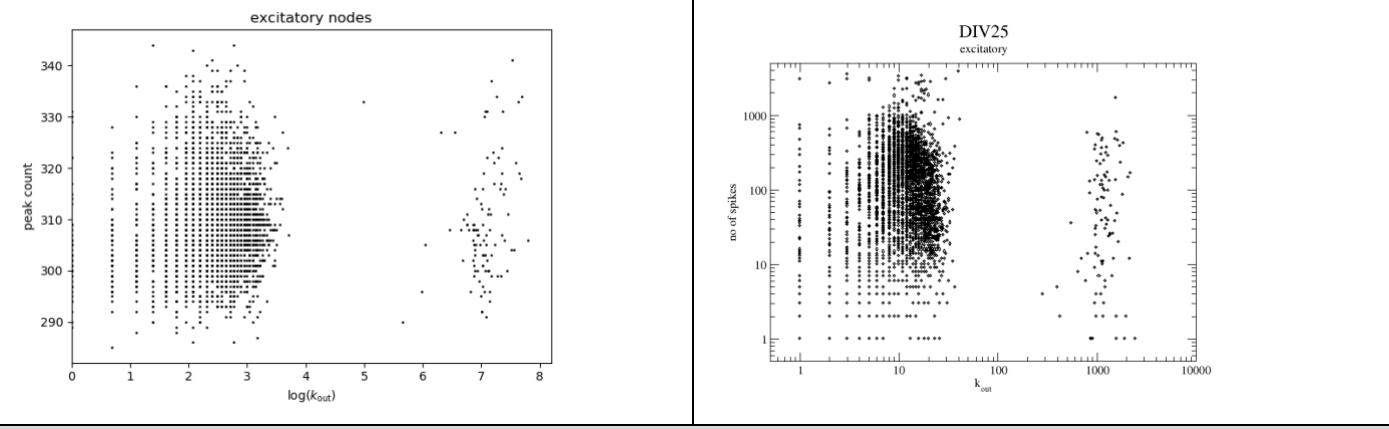
Case 1: $\epsilon = 0.1, \alpha = 1.05, \sigma_i = 2$

- **Excitatory nodes**

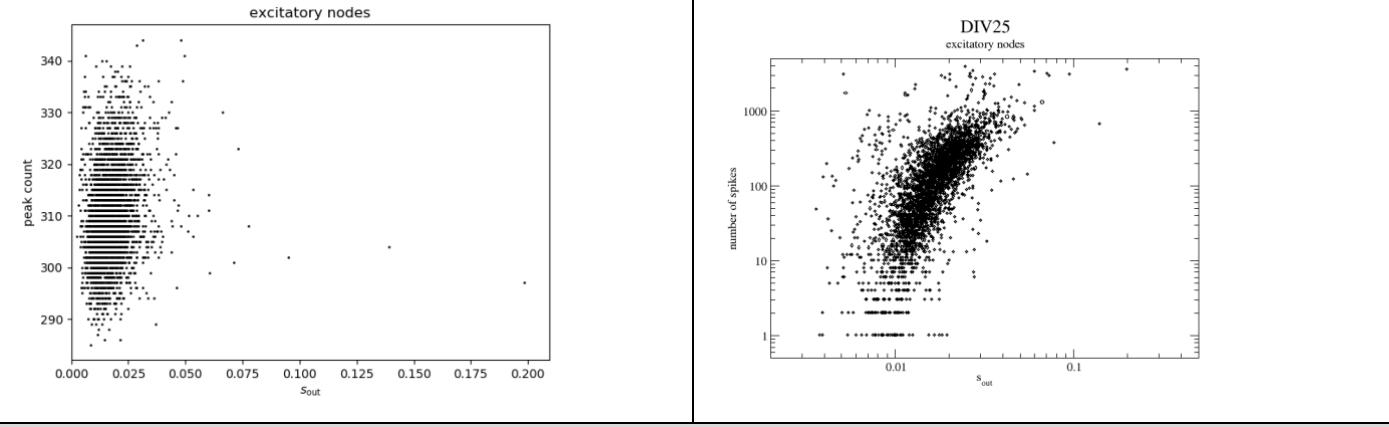
Peak count vs log k_in



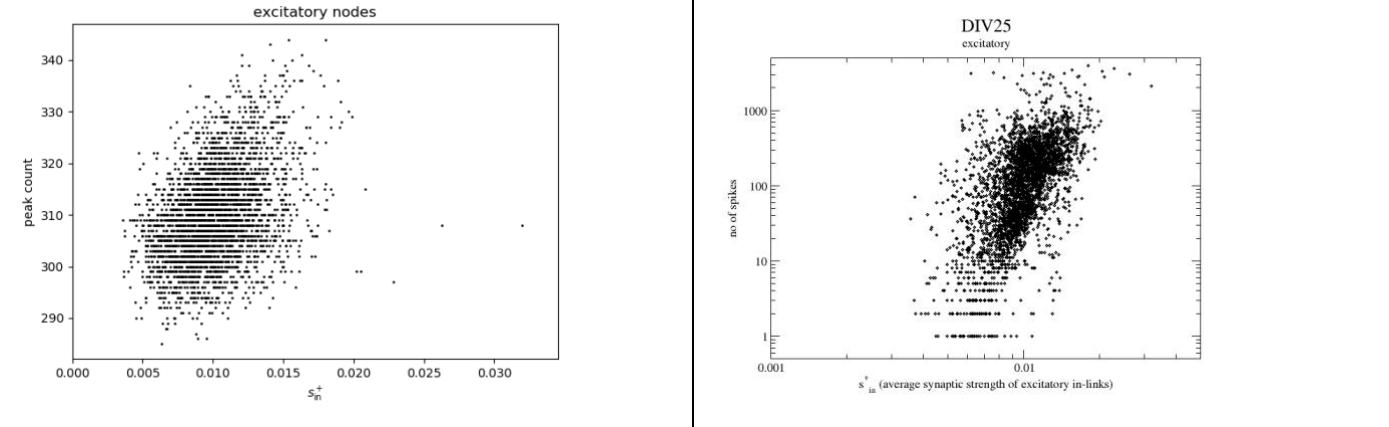
Peak count vs log k_out



Peak count vs s_out

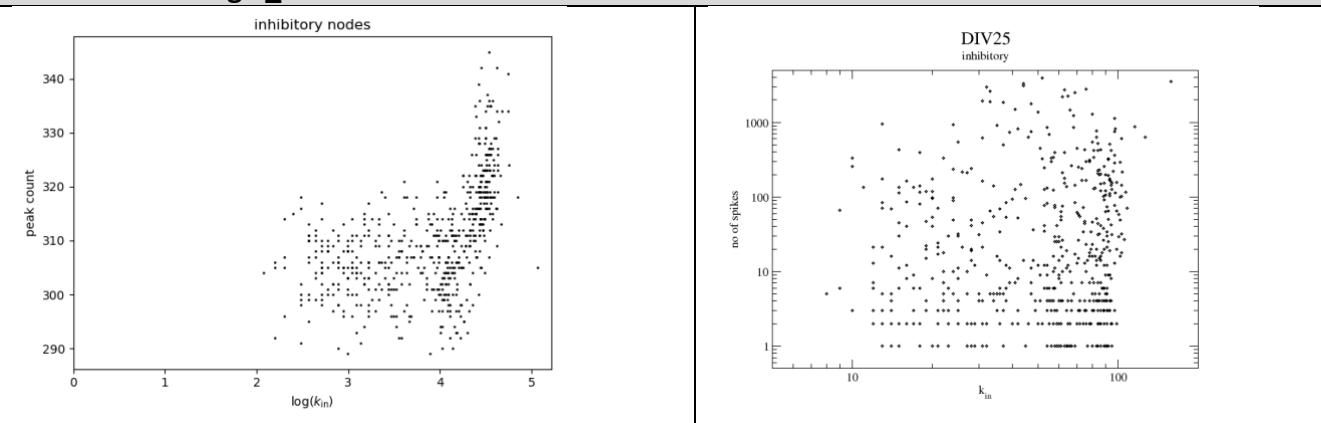


Peak count vs s^+_in

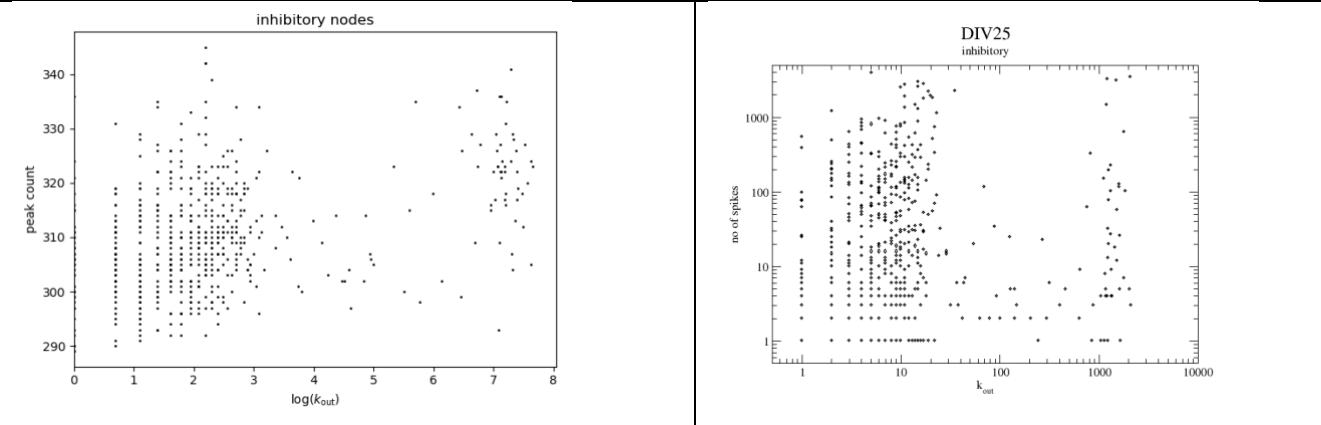


- Inhibitory nodes

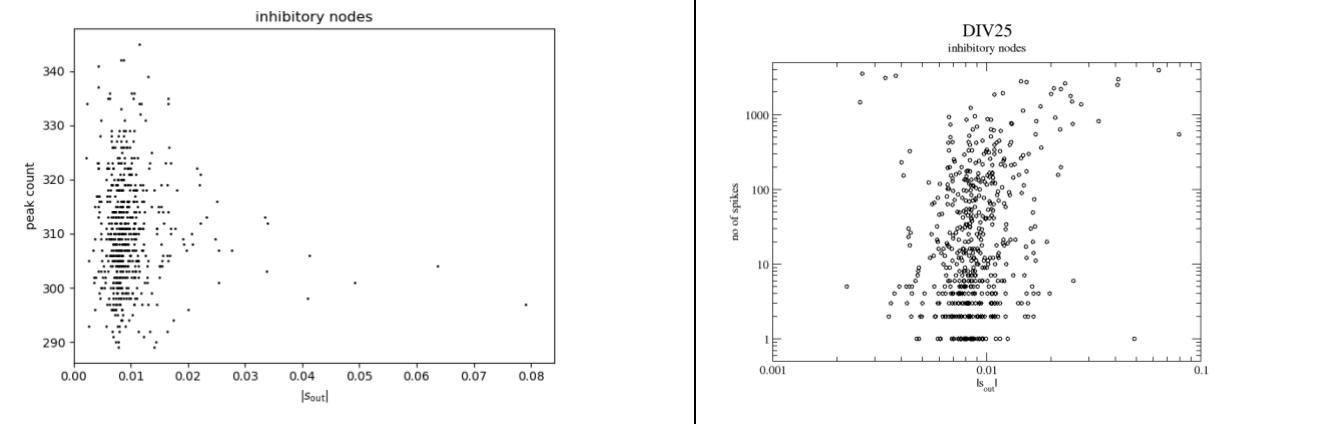
Peak count vs log k_in



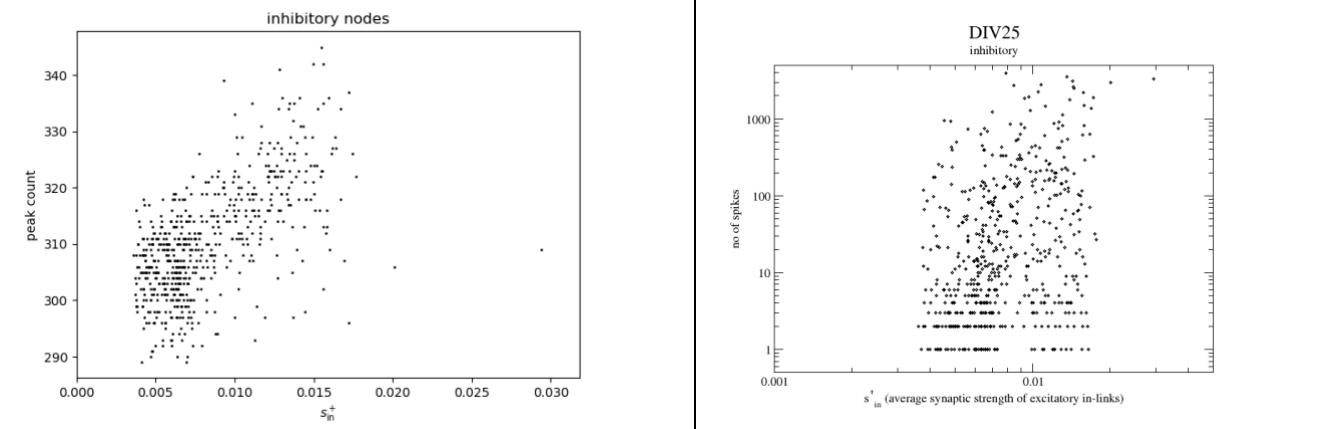
Peak count vs log k_out



Peak count vs |s_out|

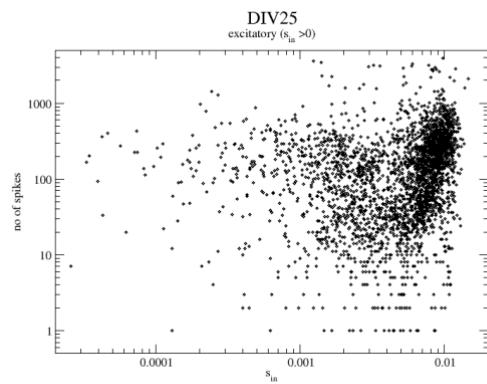
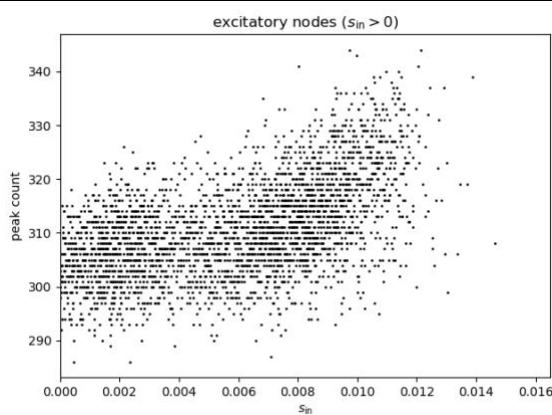


Peak count vs s^+_in

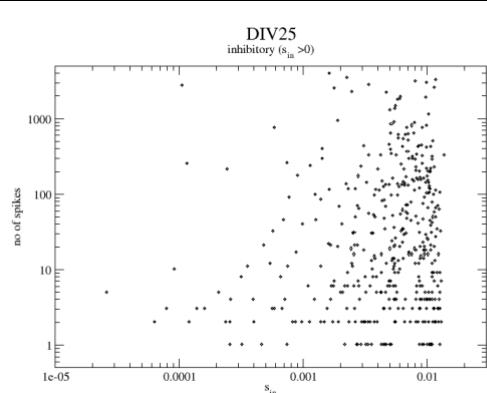
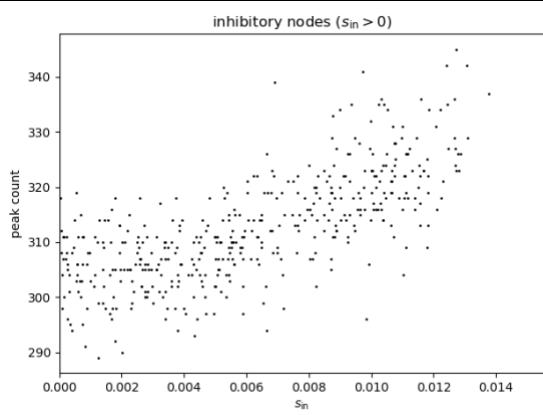


- Mixed (x-axis: s_{in} or $|s_{in}|$)**

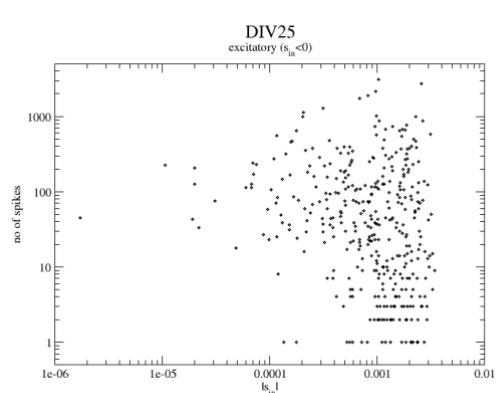
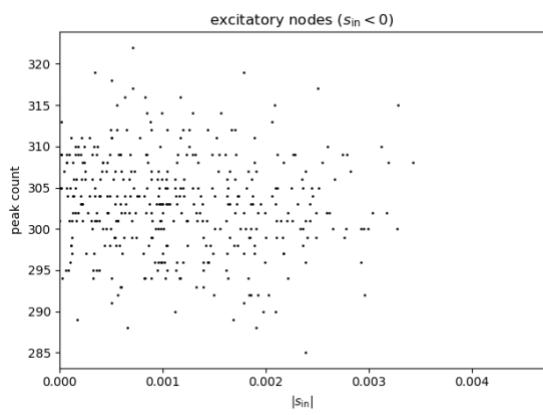
$s_{in} > 0 \& s_{out} > 0$



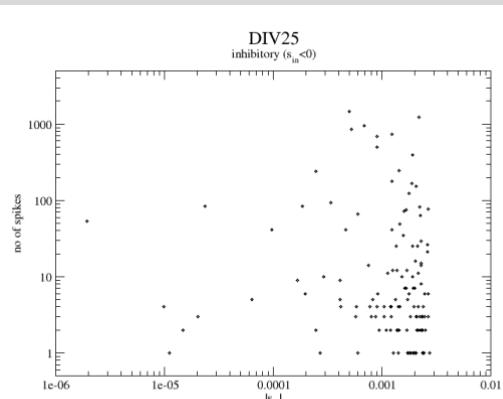
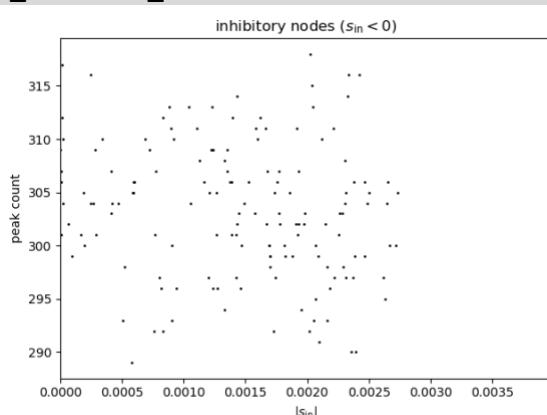
$s_{in} > 0 \& s_{out} < 0$



$s_{in} < 0 \& s_{out} > 0$



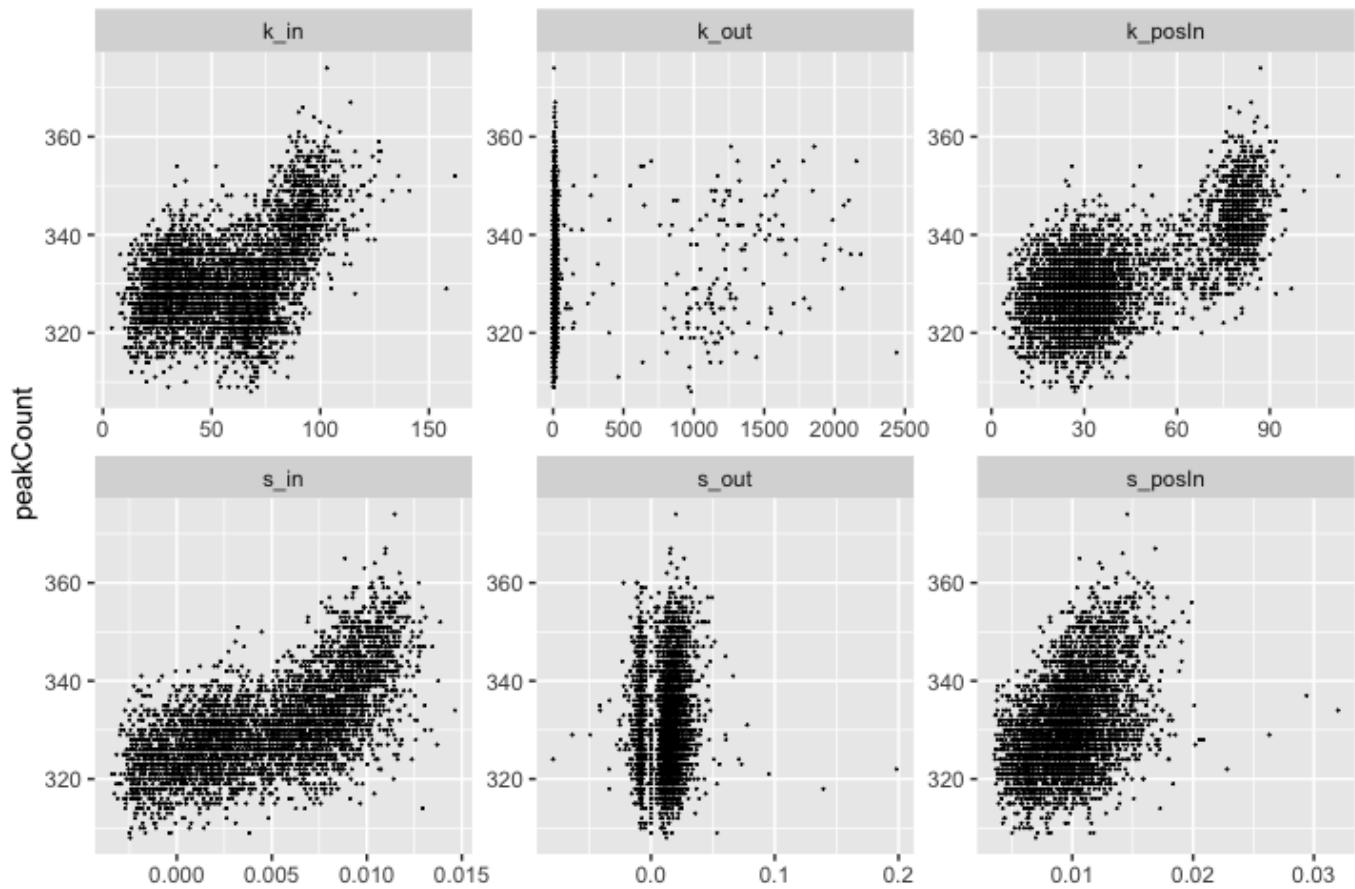
$s_{in} < 0 \& s_{out} < 0$



More analysis on Case 1

- Case 1: $\epsilon = 0.1, \alpha = 0.95, \sigma_i = 2$

- Exploratory analysis



- Regression

- Model 1: peak count vs all explanatory variables (i.e. k_in, k_out, ...)
All coefficients are significant except for k_out's ($|t\text{-statistic}| < 2$)

```
Residuals:
    Min      1Q Median      3Q      Max
-20.111 -4.280 -0.209  4.191  28.313

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 3.187e+02  3.889e-01 819.403 < 2e-16 ***
k_in        -8.728e-02  2.647e-02 -3.297 0.000986 ***
k_out       -3.017e-04  4.199e-04 -0.719 0.472432
k_posIn     2.995e-01  3.398e-02  8.813 < 2e-16 ***
s_in        3.611e+02  1.255e+02  2.877 0.004036 **
s_out       -5.089e+01  9.049e+00 -5.623 2.0e-08 ***
s_posIn     4.670e+02  9.571e+01  4.879 1.1e-06 ***
---
Signif. codes:  0 '****' 0.001 '***' 0.01 '**' 0.05 '*' 0.1 ' ' 1

Residual standard error: 6.286 on 4088 degrees of freedom
Multiple R-squared:  0.5472,   Adjusted R-squared:  0.5465
F-statistic: 823.3 on 6 and 4088 DF,  p-value: < 2.2e-16
```

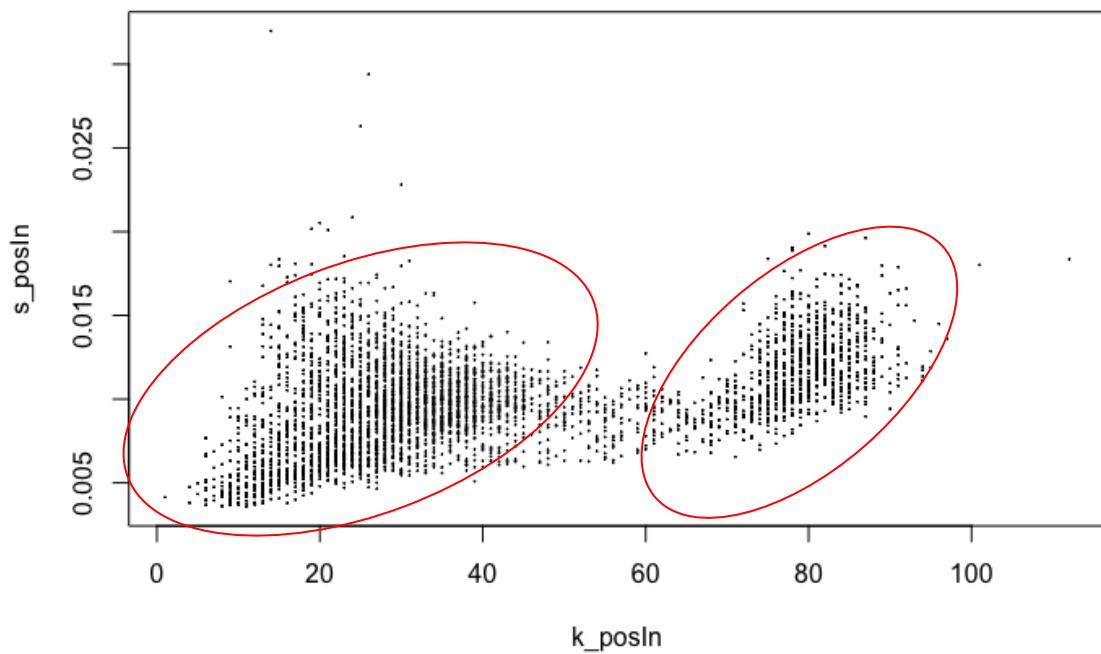
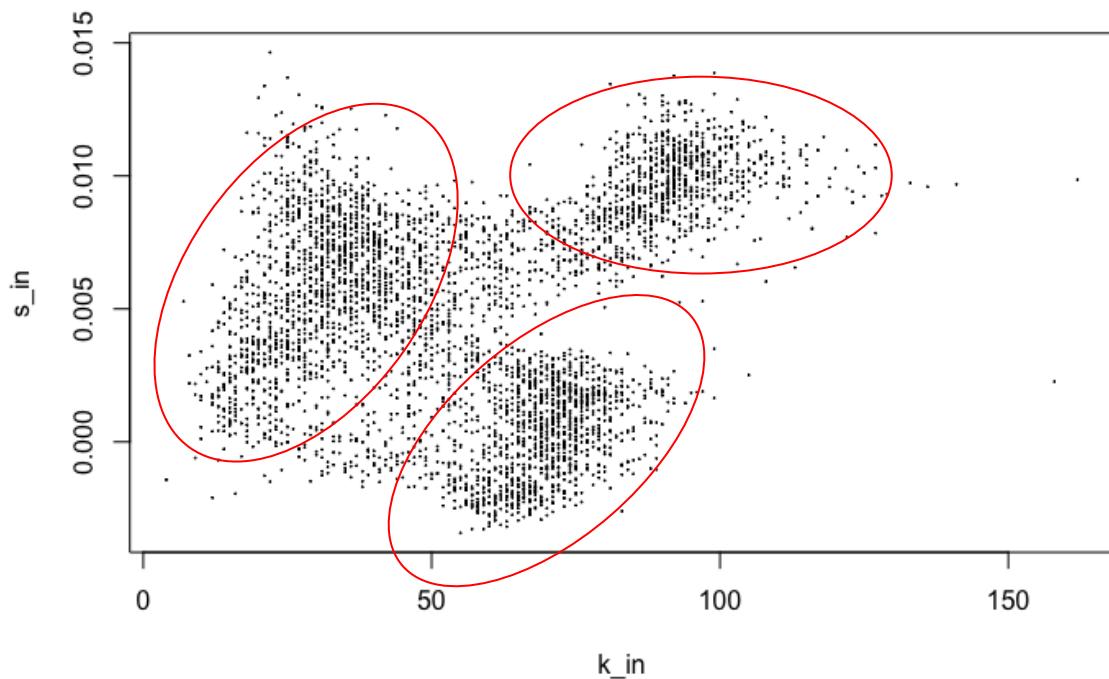
- Model 2: peak count vs all explanatory variables except k_out

```
Residuals:
    Min      1Q Median      3Q      Max
-20.0890 -4.2830 -0.1967  4.2044  28.3507

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 318.74129    0.38589 825.989 < 2e-16 ***
k_in        -0.08758    0.02647 -3.309 0.000944 ***
k_posIn     0.29942    0.03398  8.811 < 2e-16 ***
s_in        363.29479   125.48484  2.895 0.003810 **
s_out       -49.41719    8.81455 -5.606 2.20e-08 ***
s_posIn     460.75478   95.31020  4.834 1.39e-06 ***
---
Signif. codes:  0 '****' 0.001 '***' 0.01 '**' 0.05 '*' 0.1 ' ' 1

Residual standard error: 6.286 on 4089 degrees of freedom
Multiple R-squared:  0.5471,   Adjusted R-squared:  0.5466
F-statistic: 988 on 5 and 4089 DF,  p-value: < 2.2e-16
```

- Possibly classify nodes in degree-strength space?
May use machine learning kNN algorithm

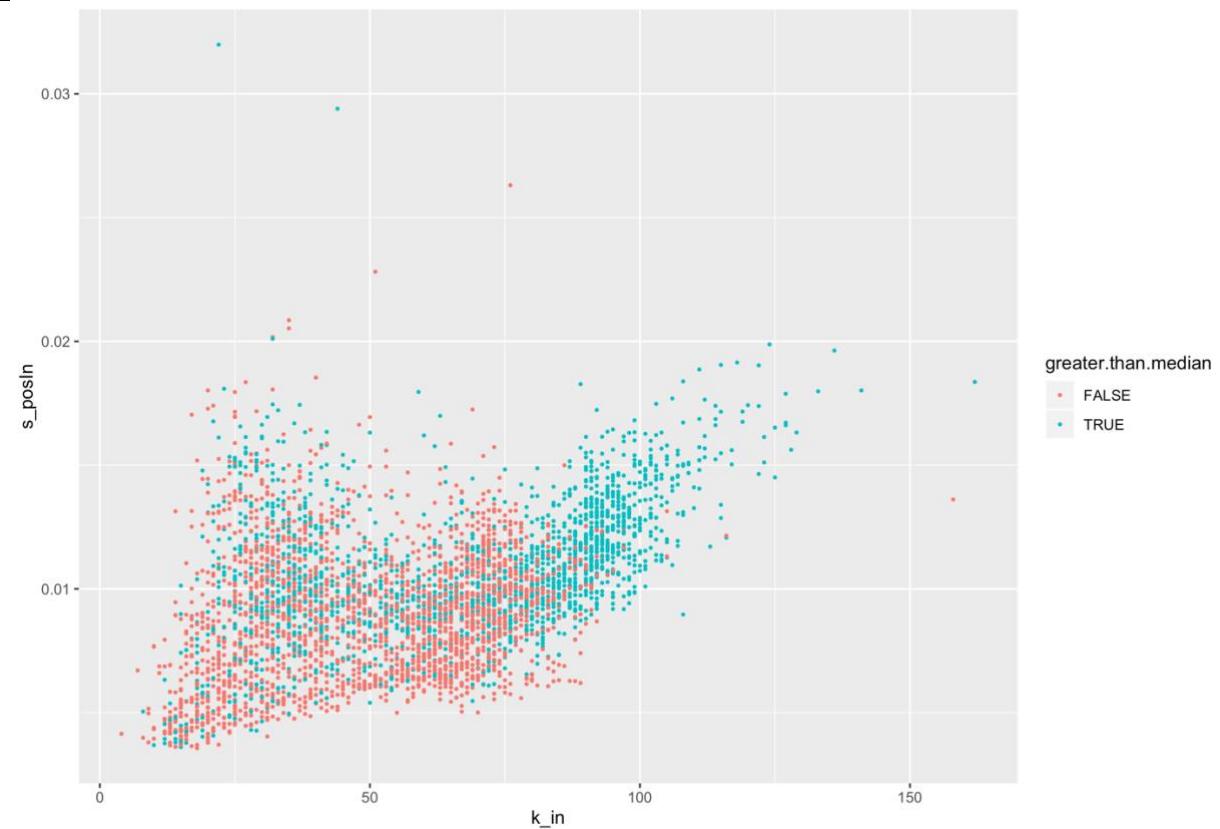


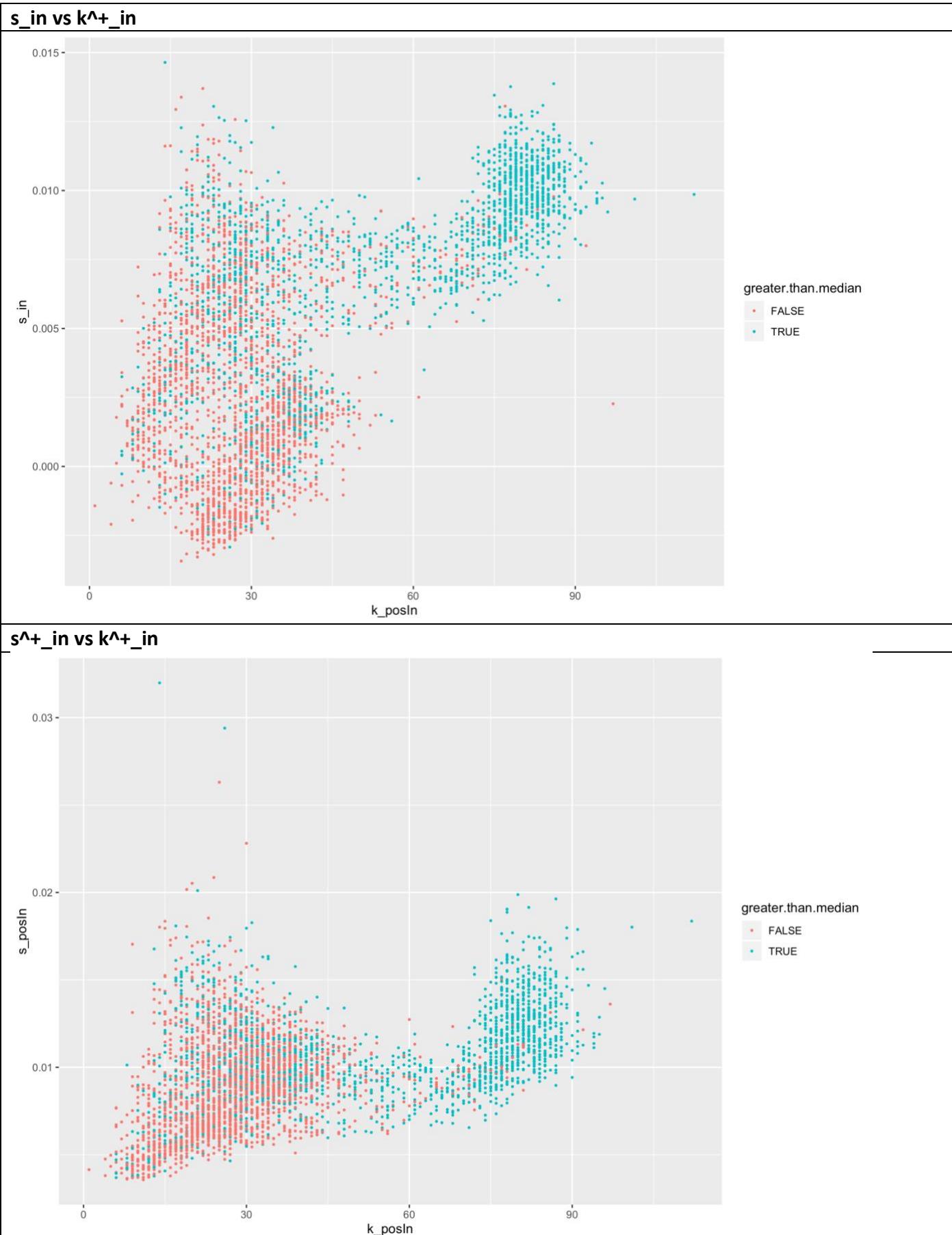
- Peak count in degree-strength space (median as threshold)

s_in vs k_in



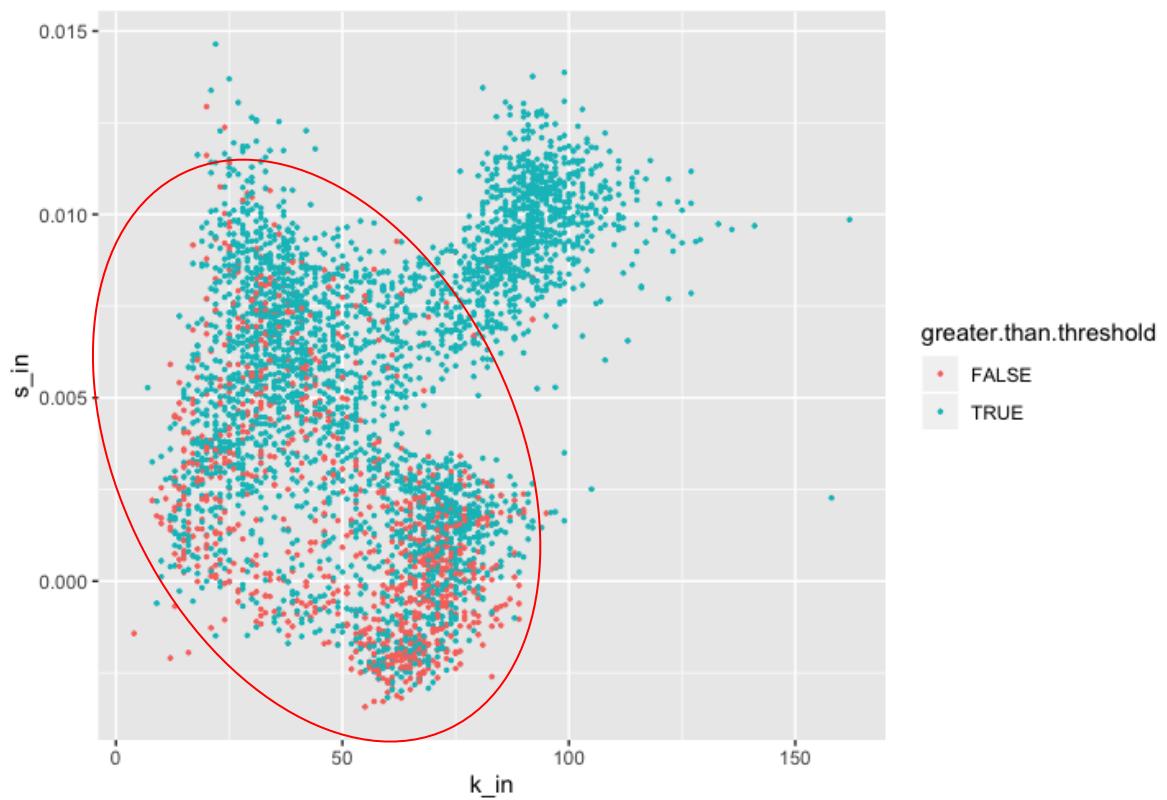
s^{+}_{in} vs k_in



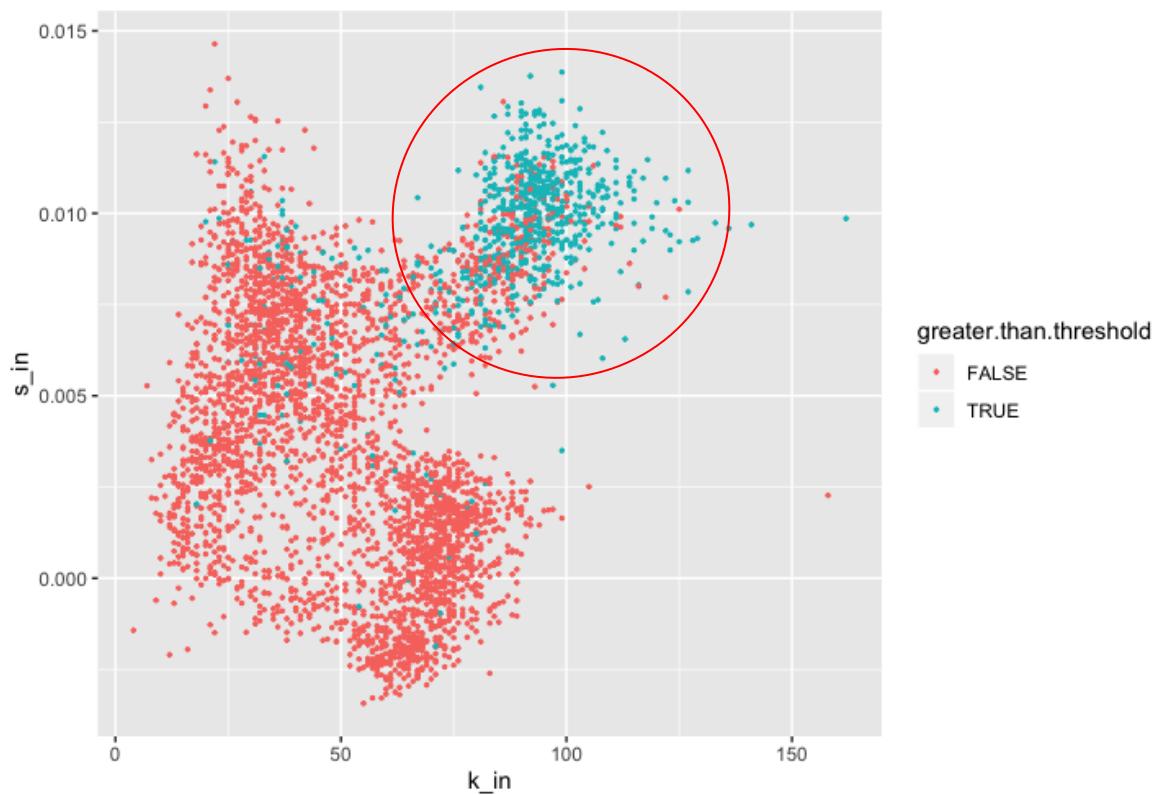


- Peak count in degree-strength space (other thresholds)

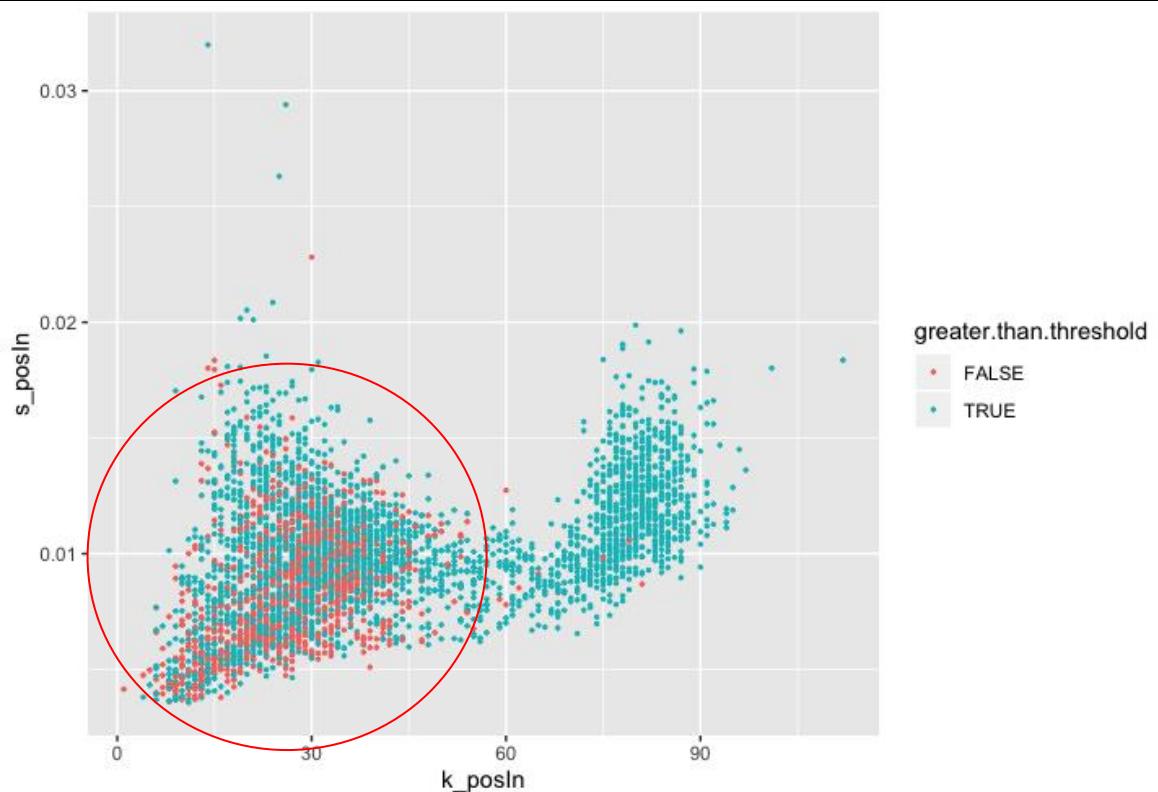
s_in vs k_in (20th percentile threshold)



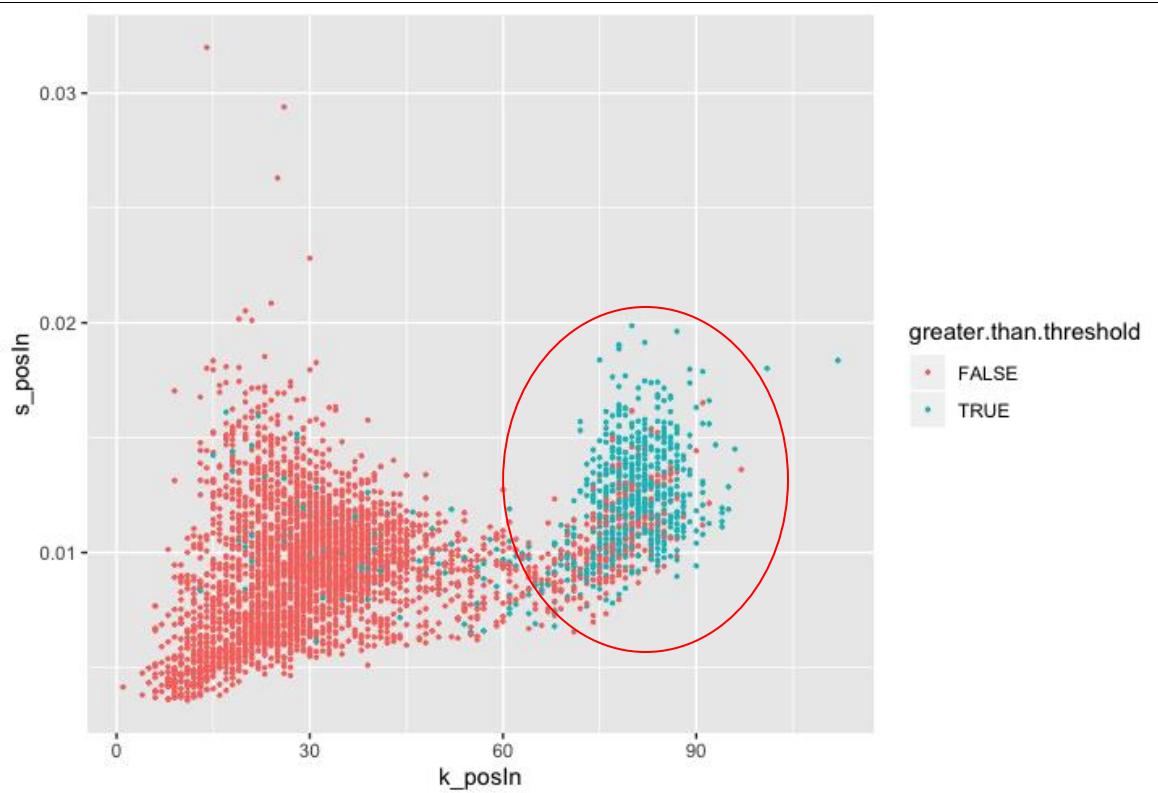
s_in vs k_in (80th percentile threshold)



s⁺_in vs k⁺_in (20th percentile threshold)



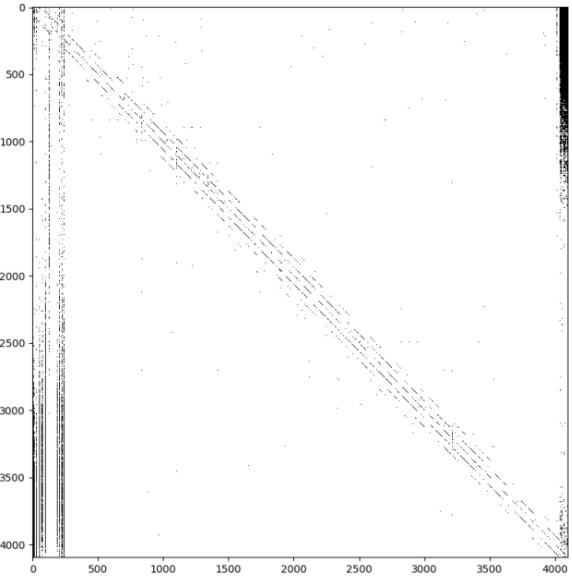
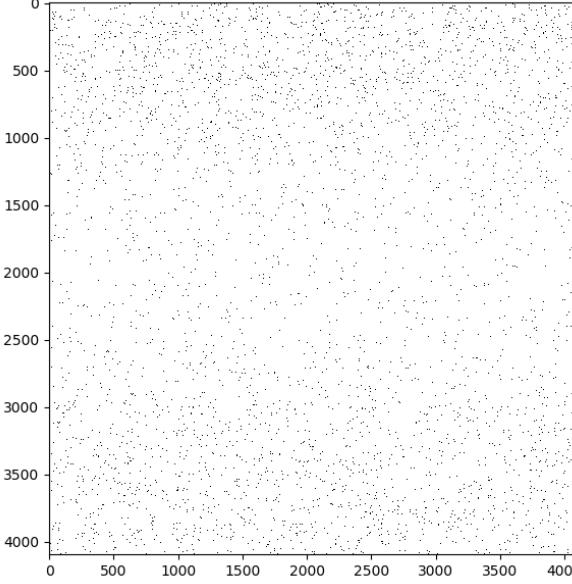
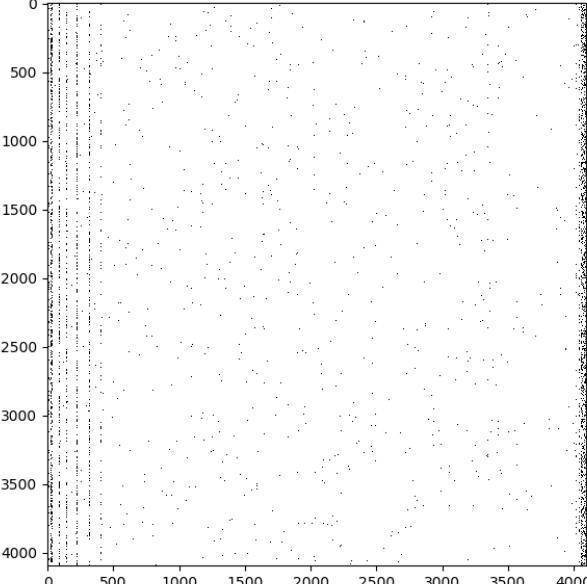
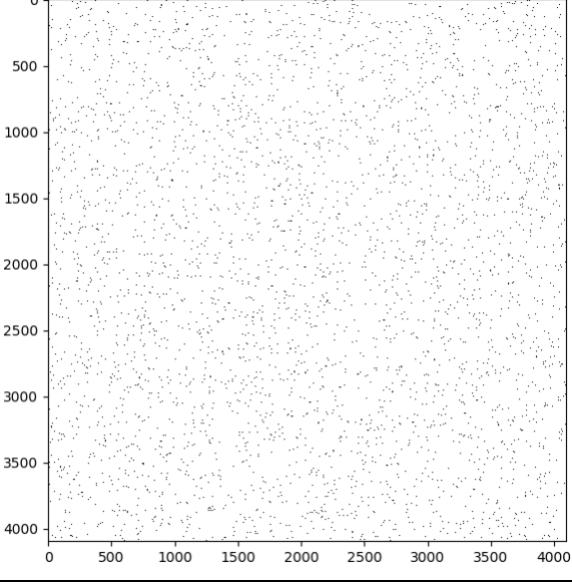
s⁺_in vs k⁺_in (80th percentile threshold)



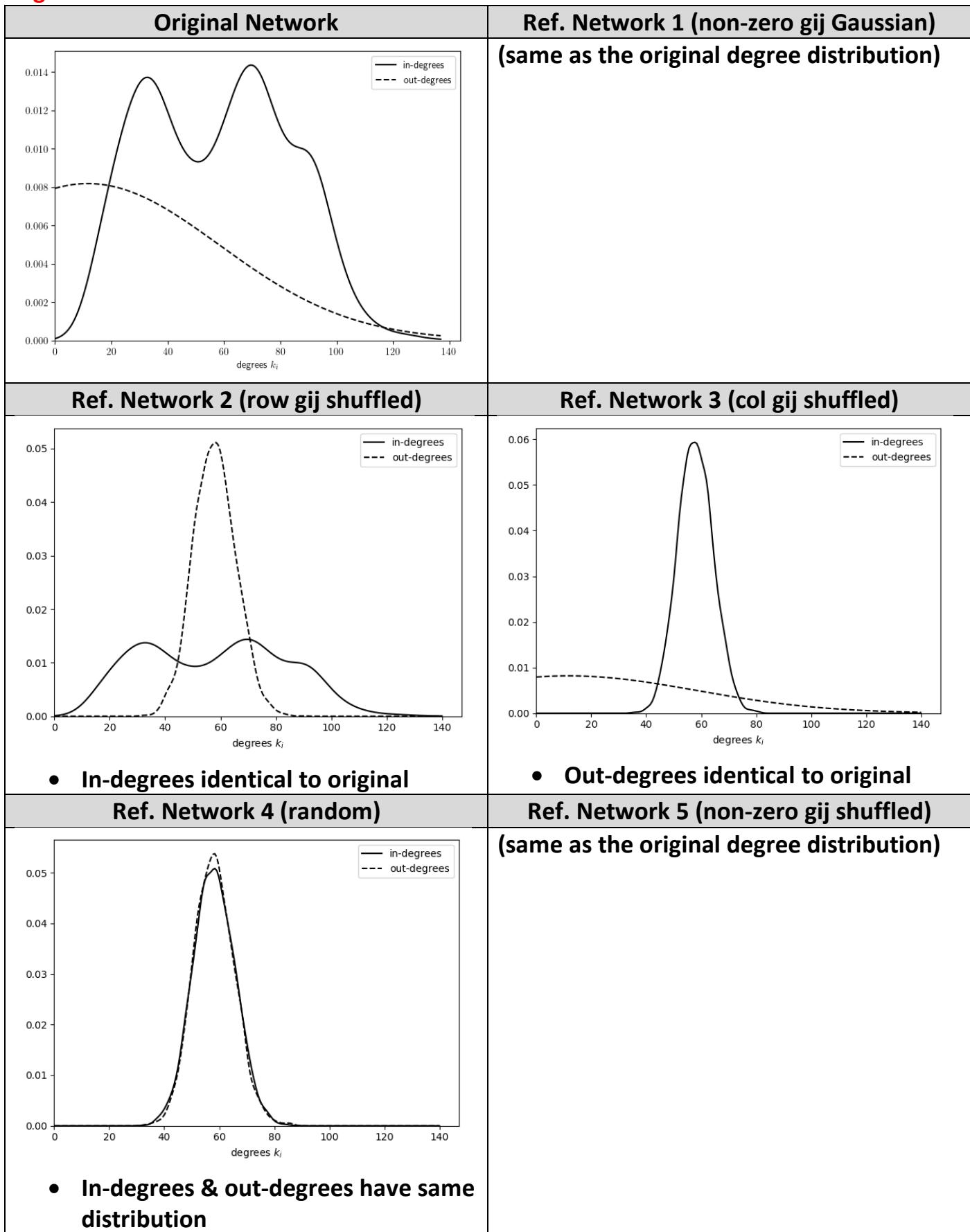
Analysis on reference networks

- (a) reference network 1:** keep A_{ij} but replace nonzero g_{ij} by values taken from a Gaussian distribution of same mean and standard deviation (this time we do not separately consider positive and negative g_{ij} 's). This network has same k_{in} and k_{out} but different s_{in} and s_{out}
- (b) reference network 2:** shuffle g_{ij} for fixed i ; this has same k_{in} and s_{in} but different k_{out} and s_{out}
- (c) reference network 3:** shuffle g_{ij} for fixed j ; this has same k_{out} and s_{out} but different k_{in} and s_{in}
- (d) reference network 4:** random directed network with same connection probability p and g_{ij} from a Gaussian distribution of same mean and standard deviation
- (e) reference network 5:** keep A_{ij} but shuffle non-zero g_{ij}

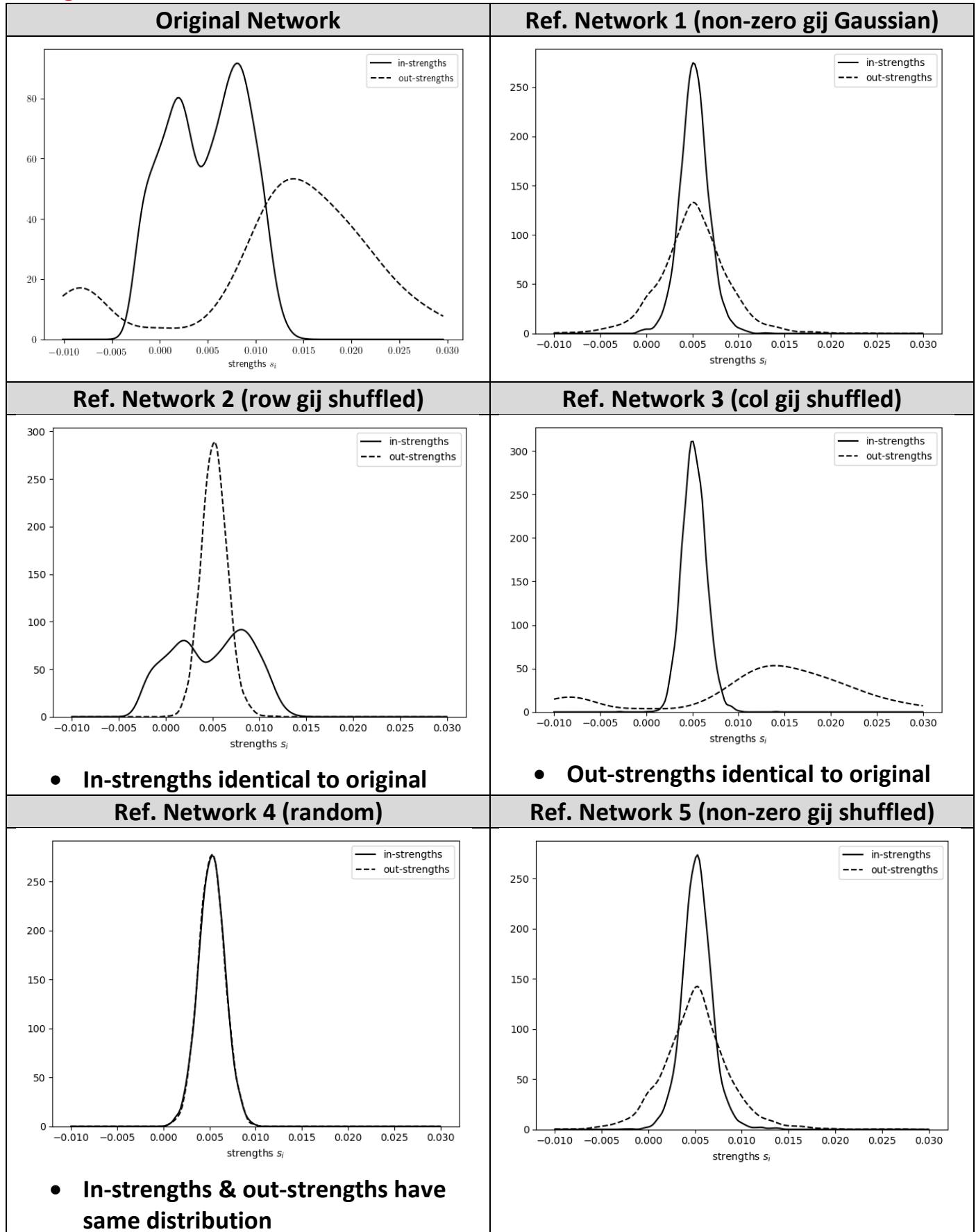
Adjacency matrix

Original Network	Ref. Network 1 (non-zero g_{ij} Gaussian) (same as the original adjacency matrix)
	
Ref. Network 2 (row g_{ij} shuffled)	Ref. Network 3 (col g_{ij} shuffled)
	
Ref. Network 4 (random)	Ref. Network 5 (non-zero g_{ij} shuffled) (same as the original adjacency matrix)
	

Degree distribution

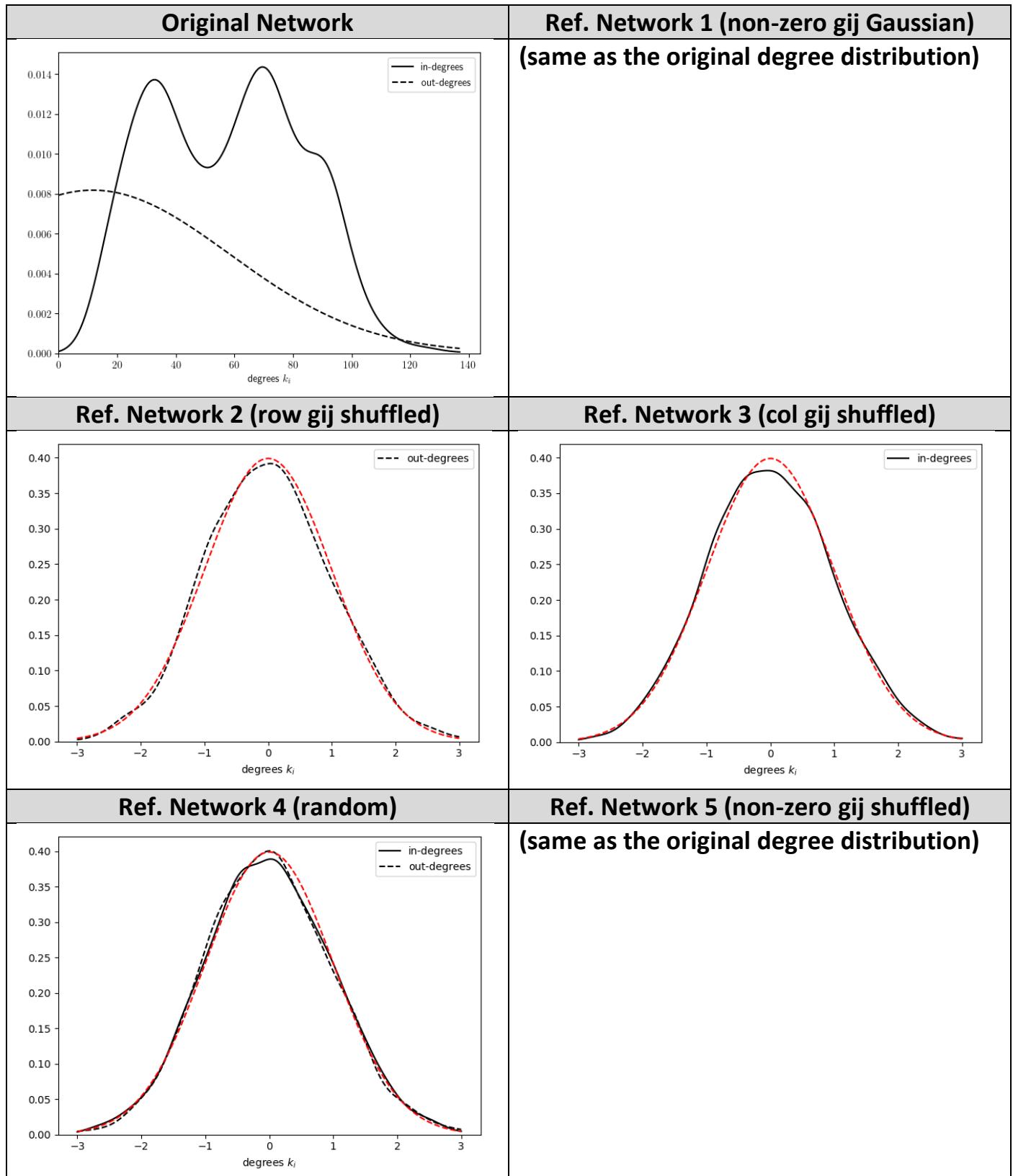


Strength distribution



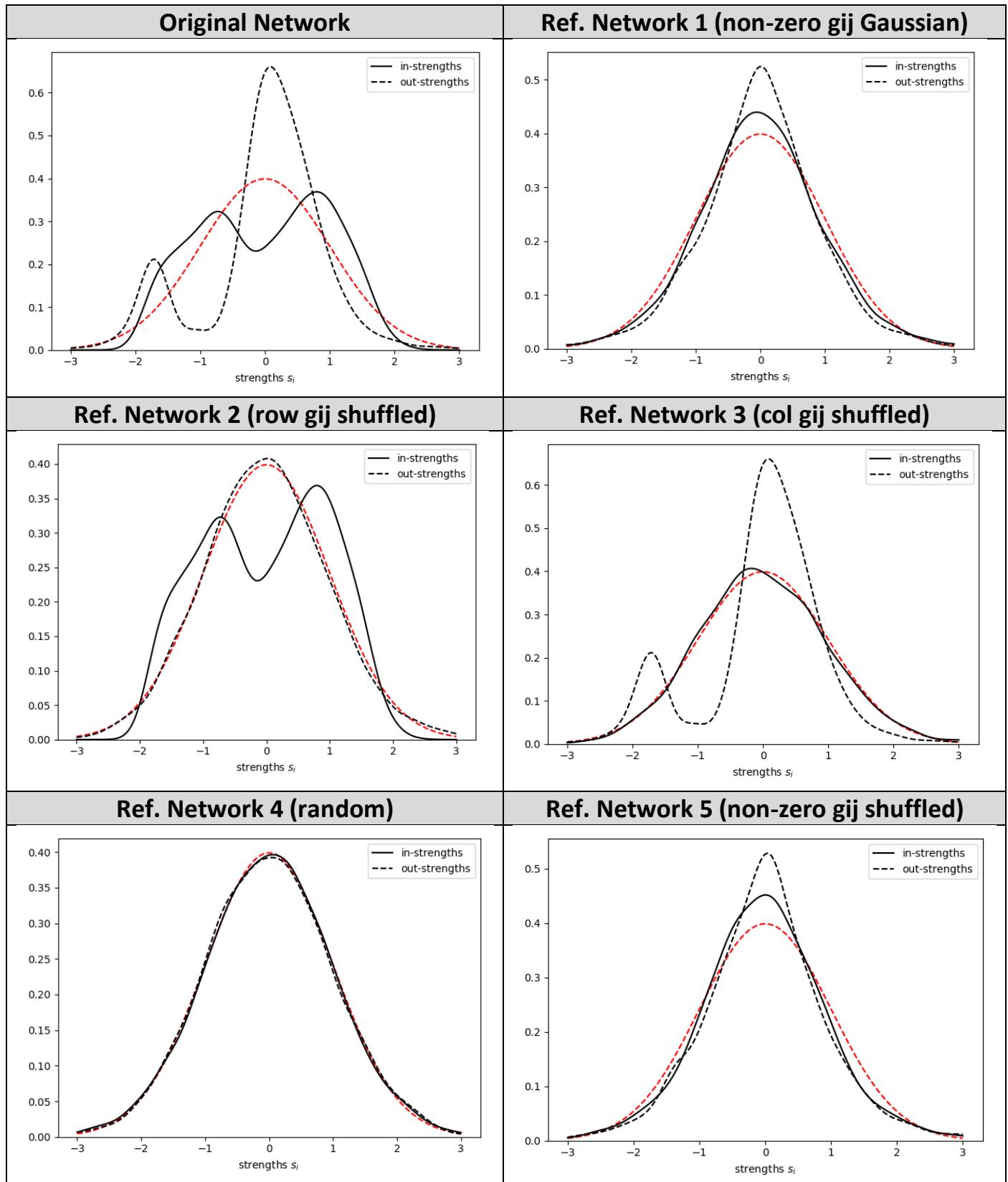
Degree distribution (standardized)

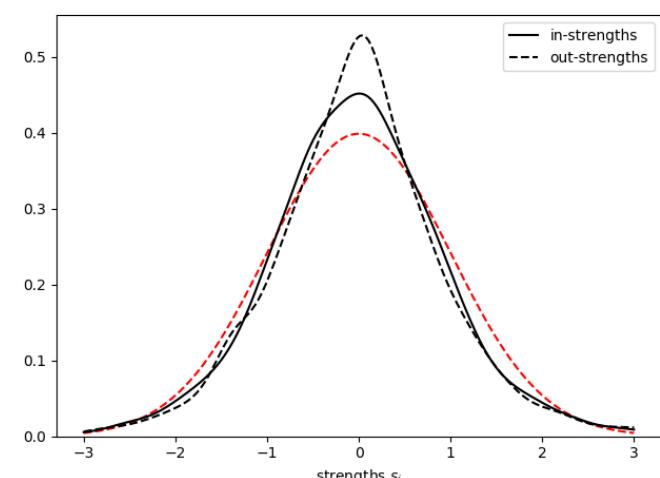
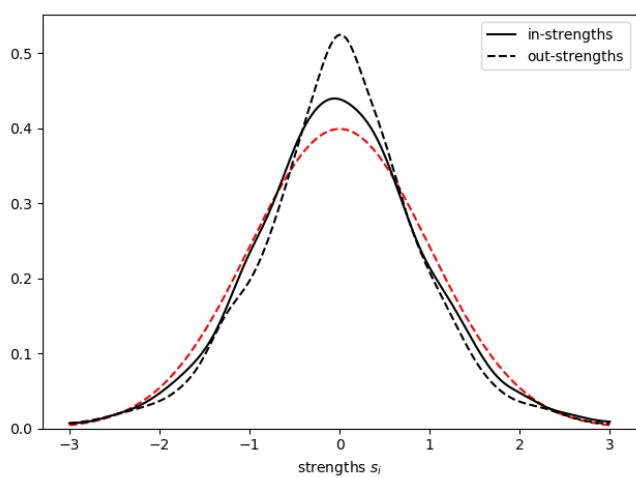
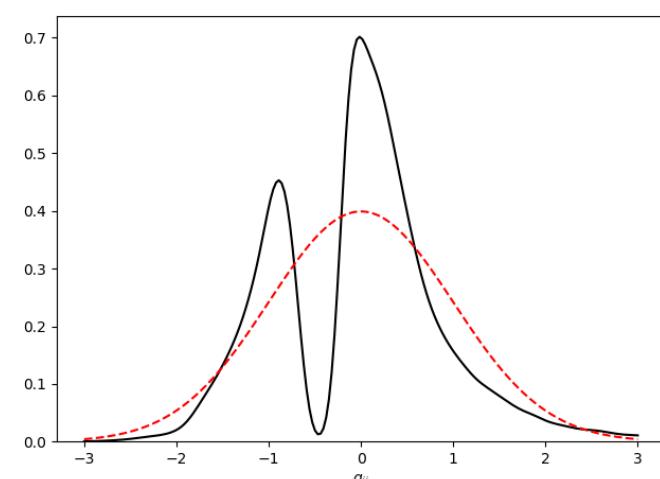
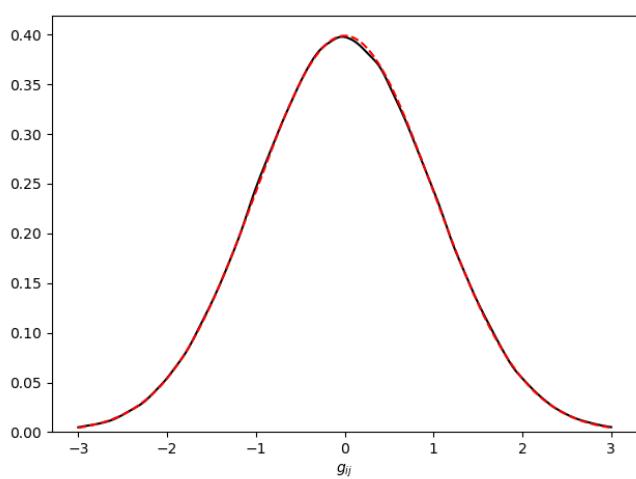
Red dotted = standard Gaussian



Strength distribution (standardized)

Red dotted = standard Gaussian

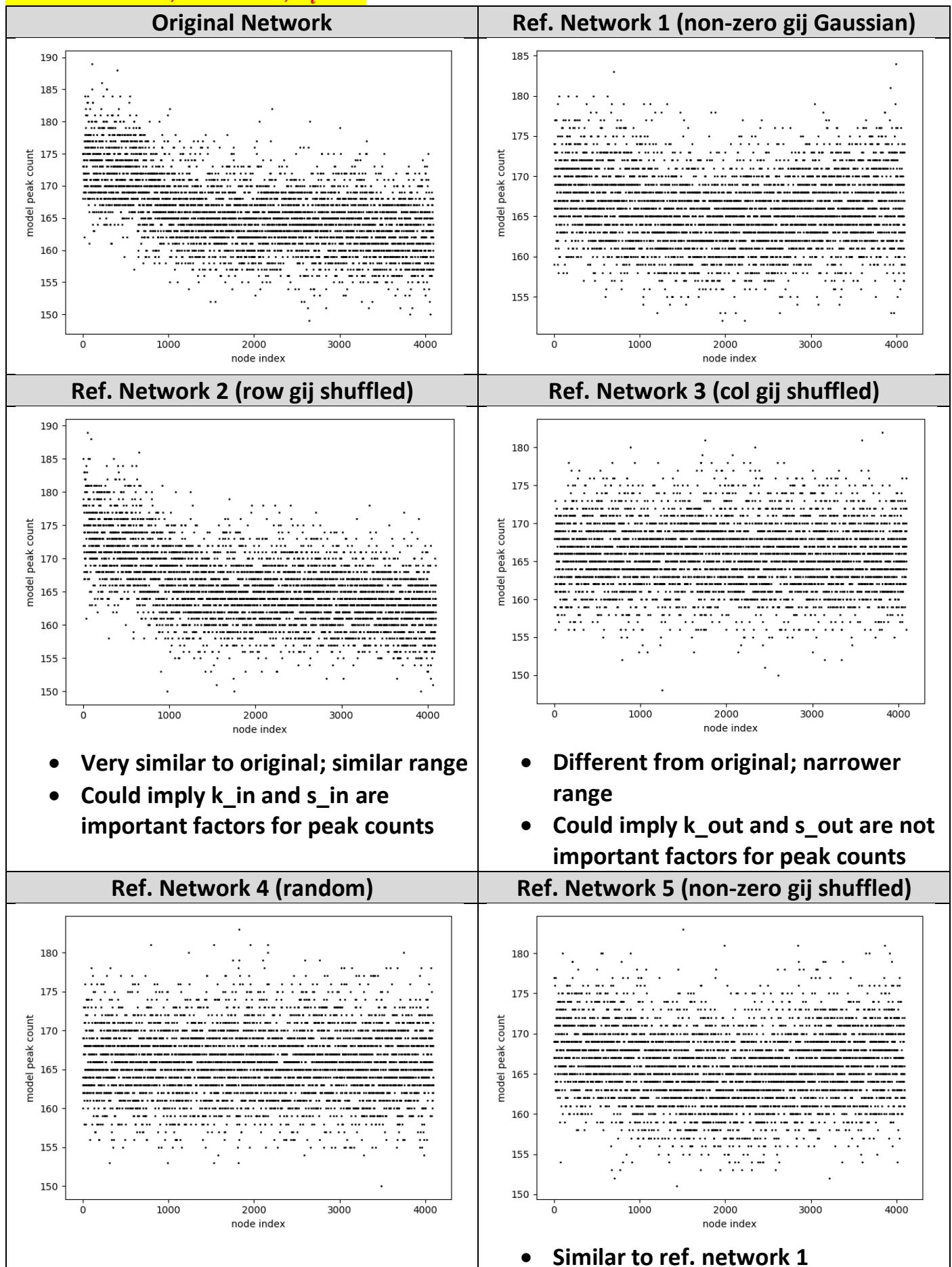


Ref. Network 1 (non-zero g_{ij} Gaussian)**Ref. Network 5 (non-zero g_{ij} shuffled)****Strength distribution (standardized)** **g_{ij} distribution (standardized)**

(same as original network's)

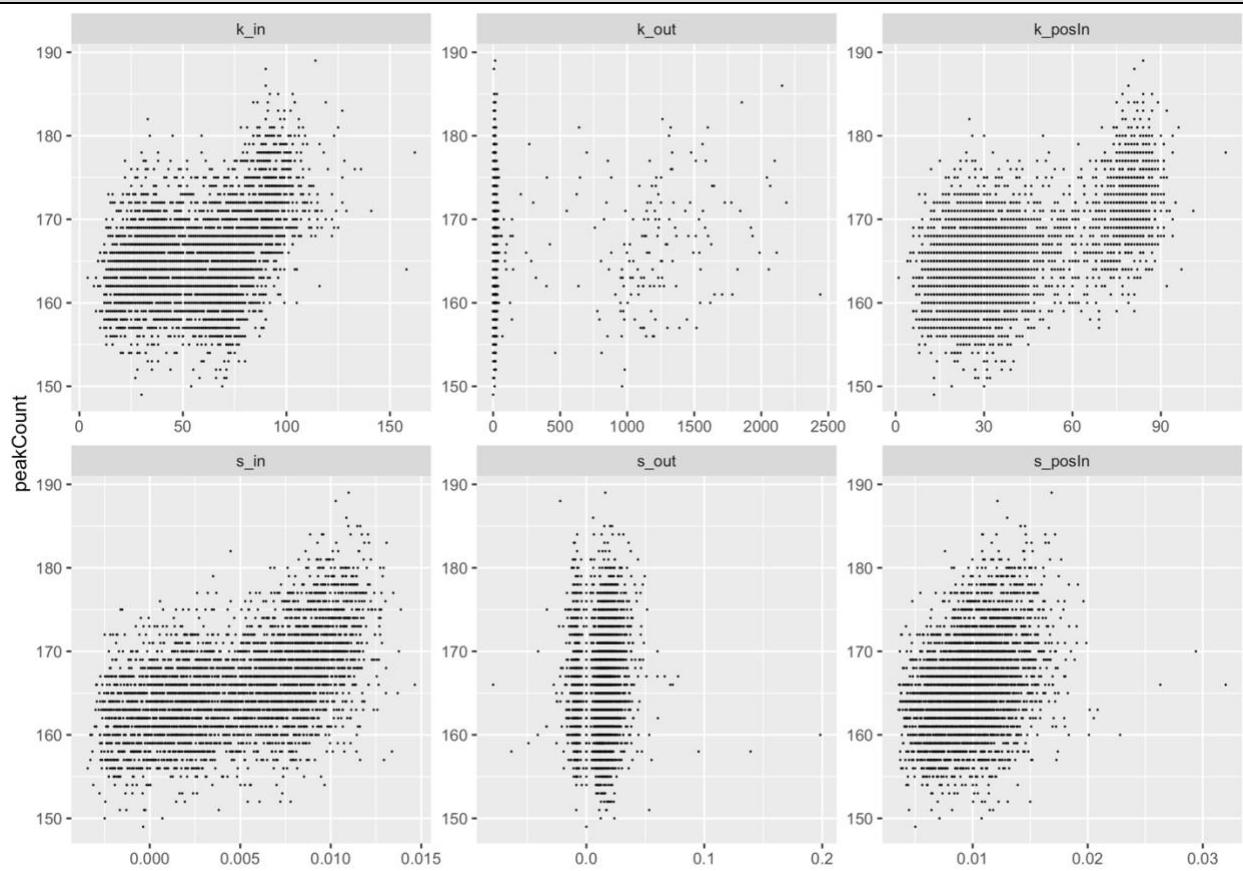
Peak count vs index (1e6 steps)

Case 1: $\epsilon = 0.1, \alpha = 0.95, \sigma_i = 2$

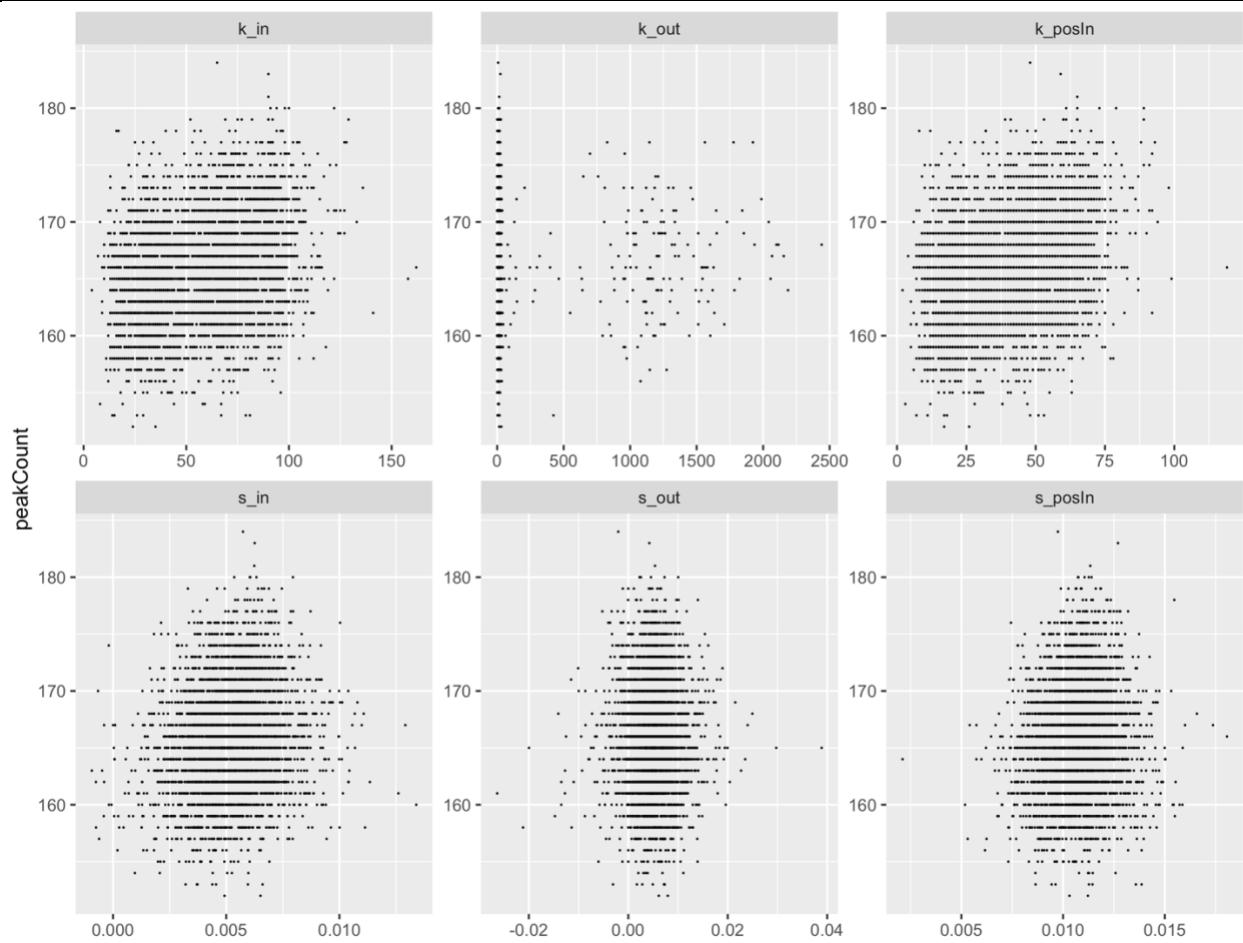


Exploratory analysis (1e6 steps)

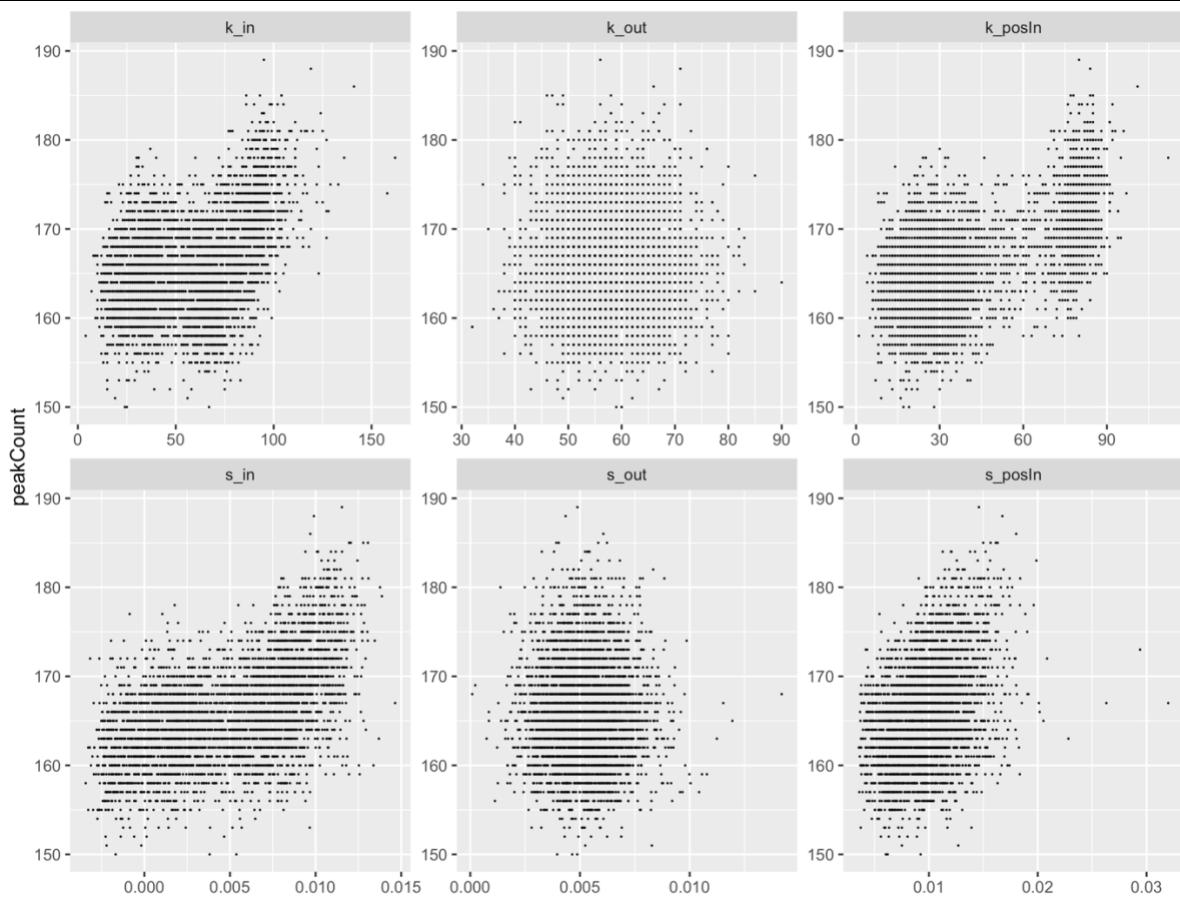
Original Network



Ref. Network 1 (non-zero gjj Gaussian)

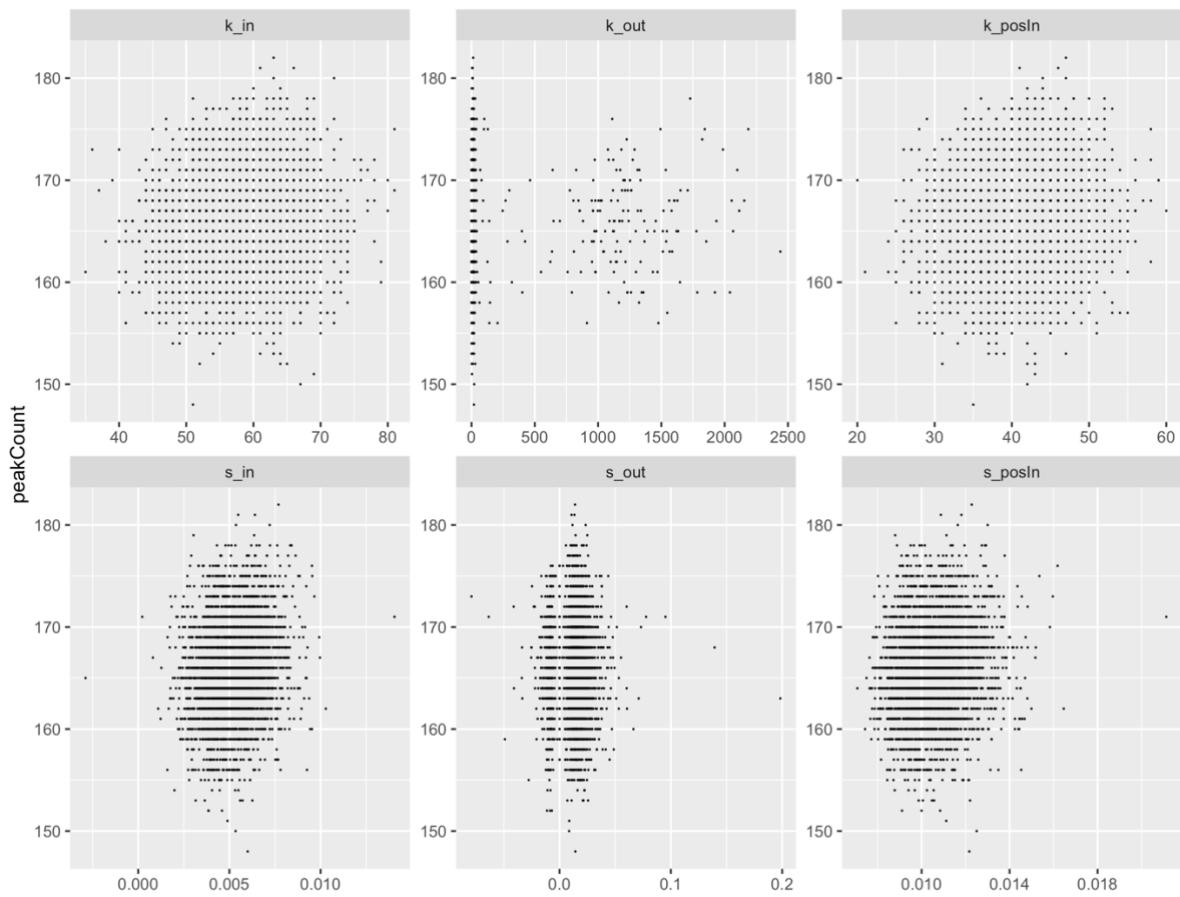


Ref. Network 2 (row gij shuffled)

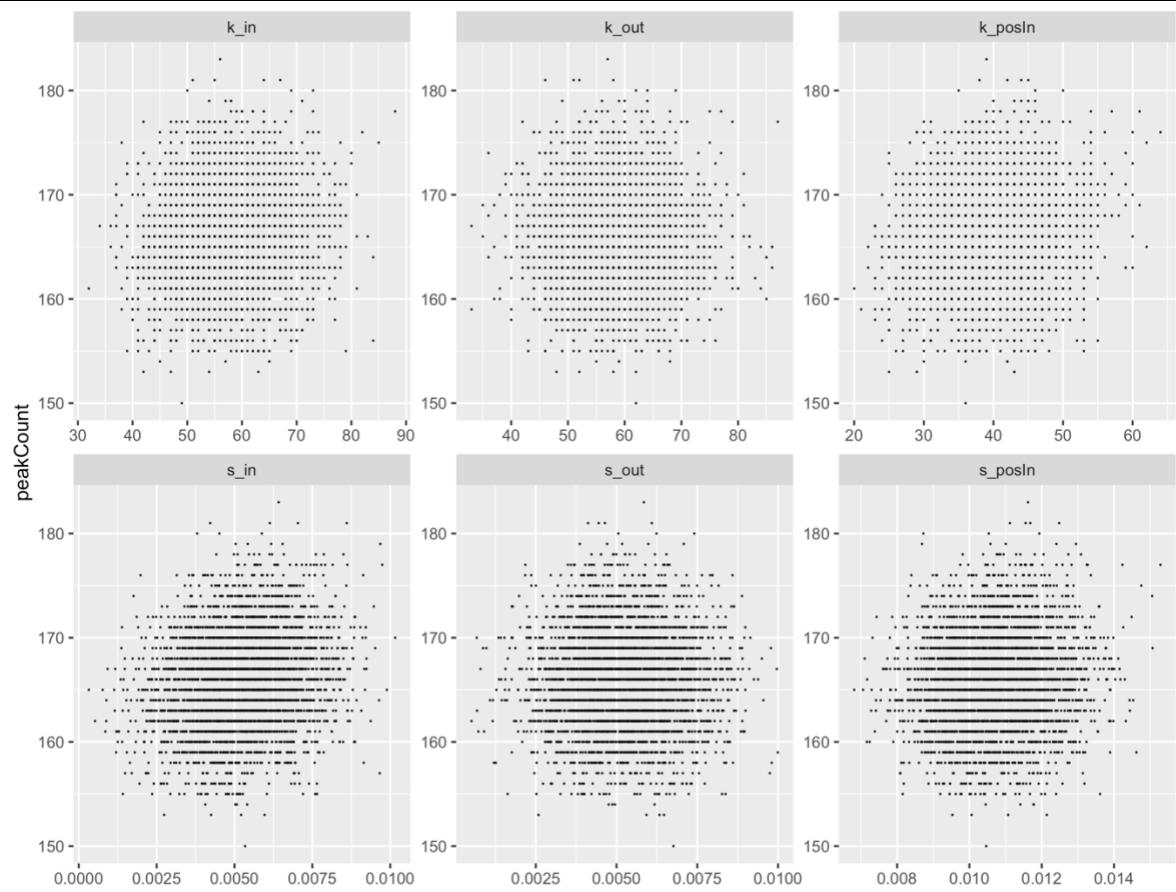


- Very similar to original

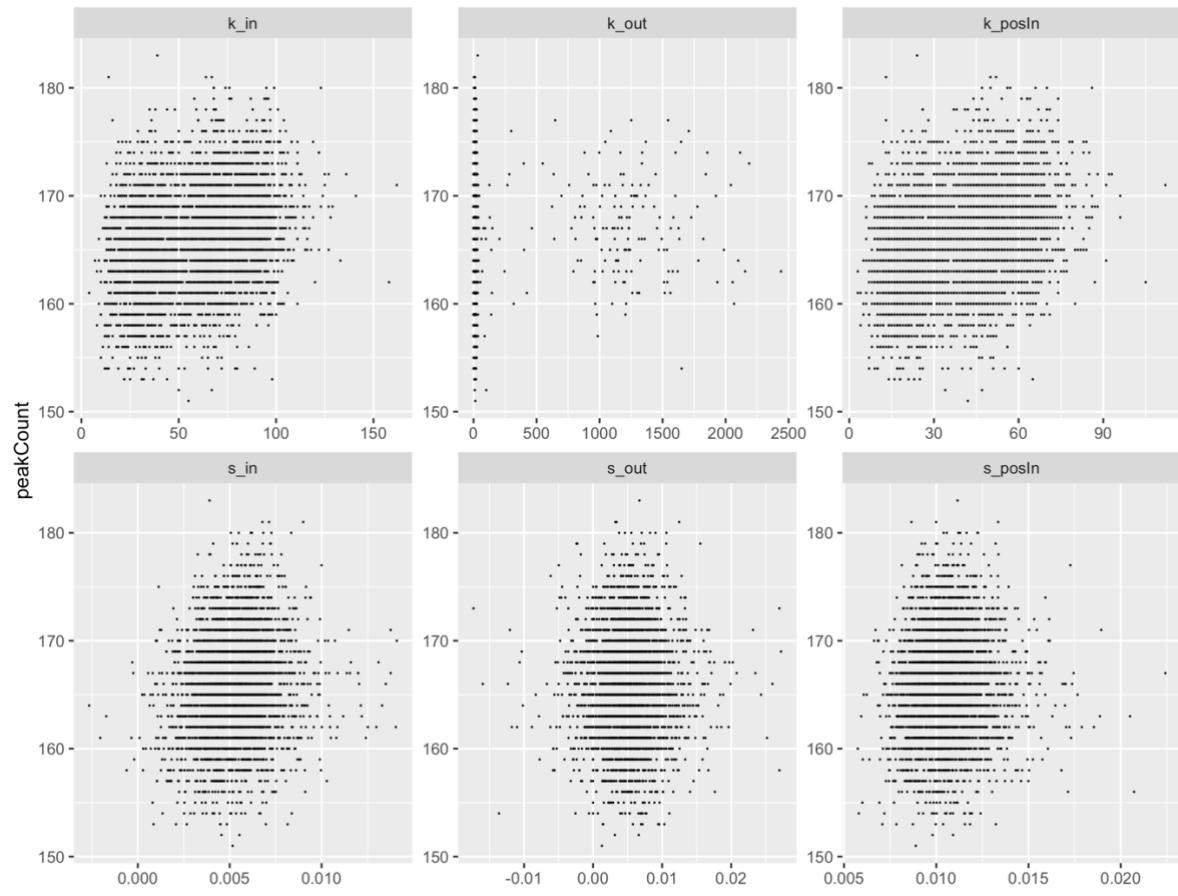
Ref. Network 3 (col gij shuffled)



Ref. Network 4 (random)

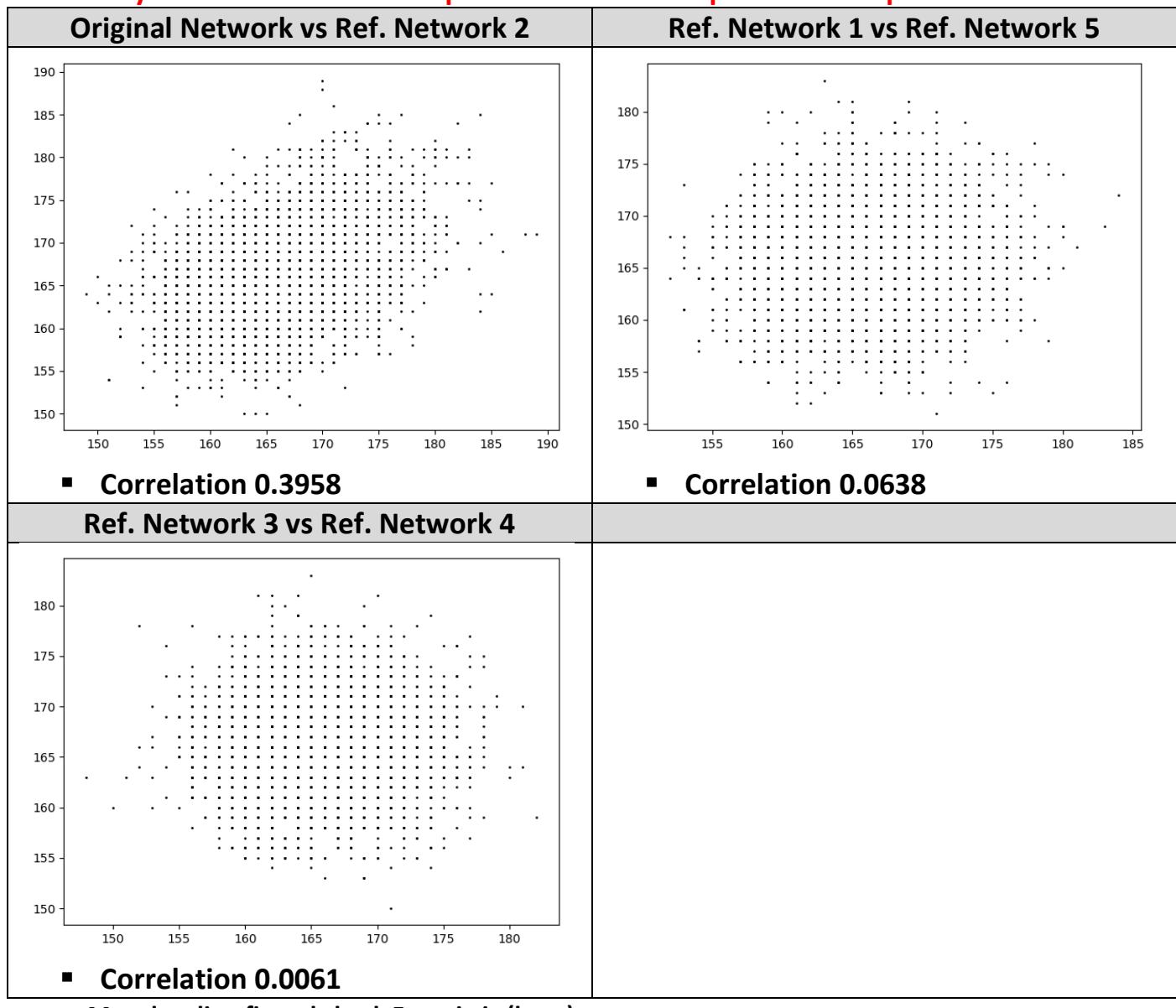


Ref. Network 5 (non-zero gij shuffled)



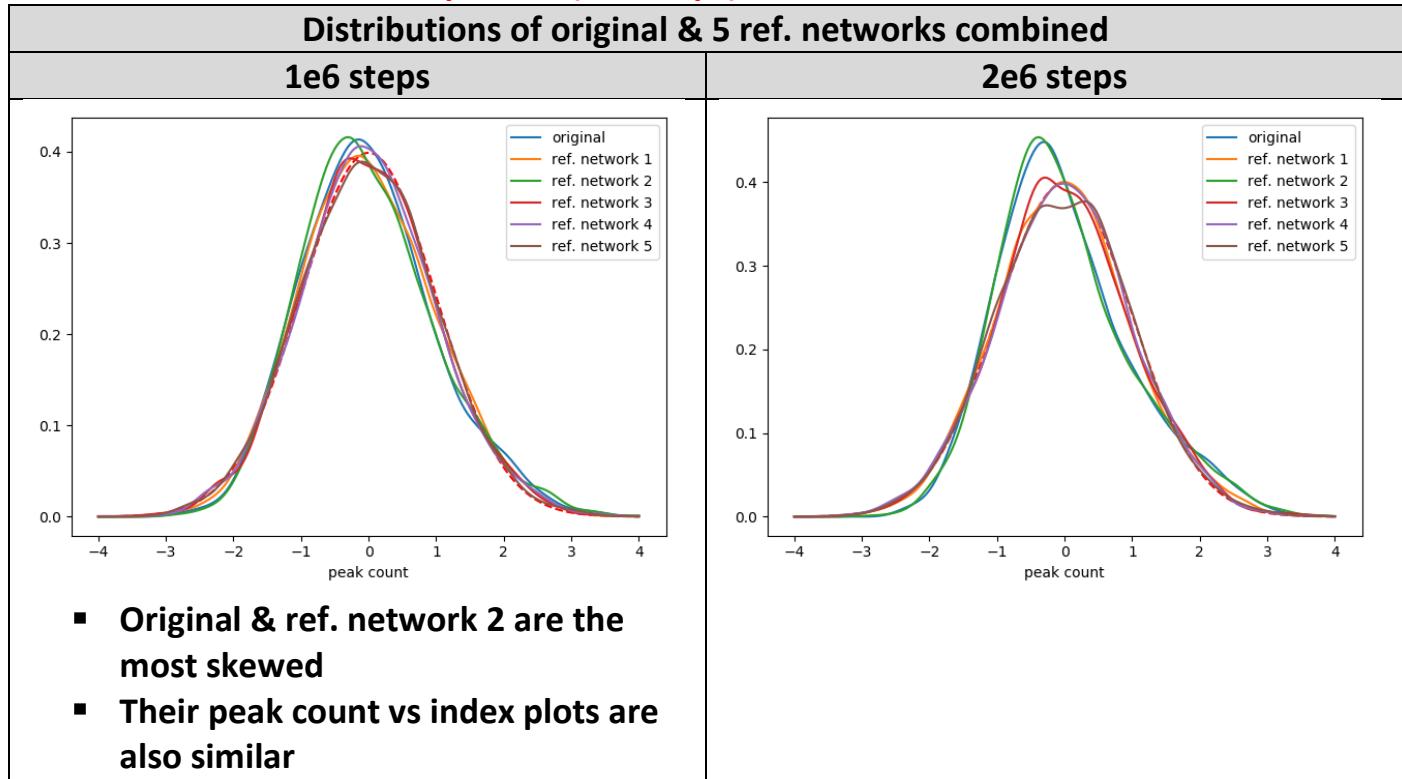
Peak count distribution comparison

- A vs B -> x-axis = A, y-axis = B
- Only networks with similar peak count vs index plots are compared



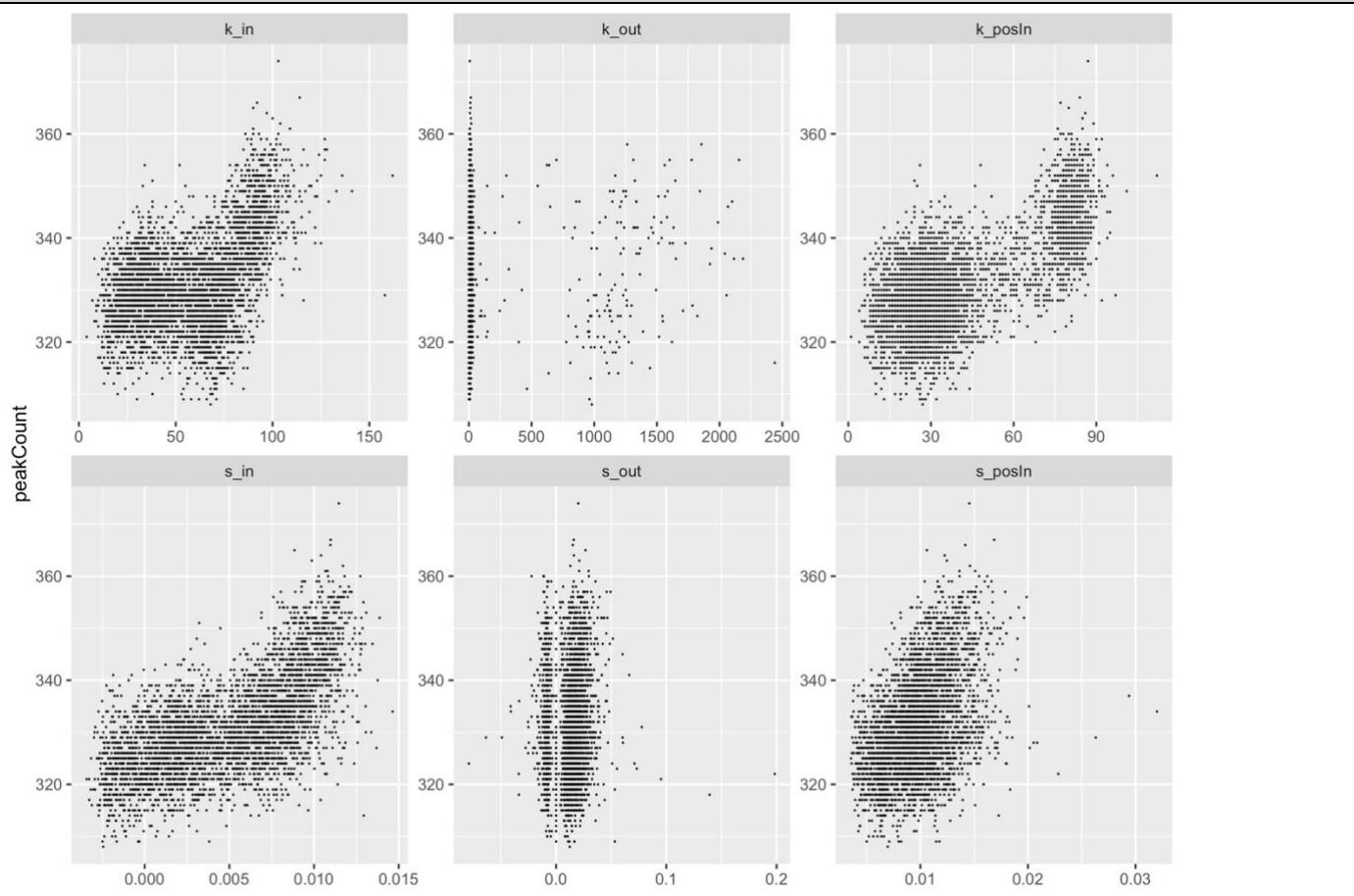
- May do a line fit and check F-statistic (later)

Peak count distribution comparison (2e6 steps)

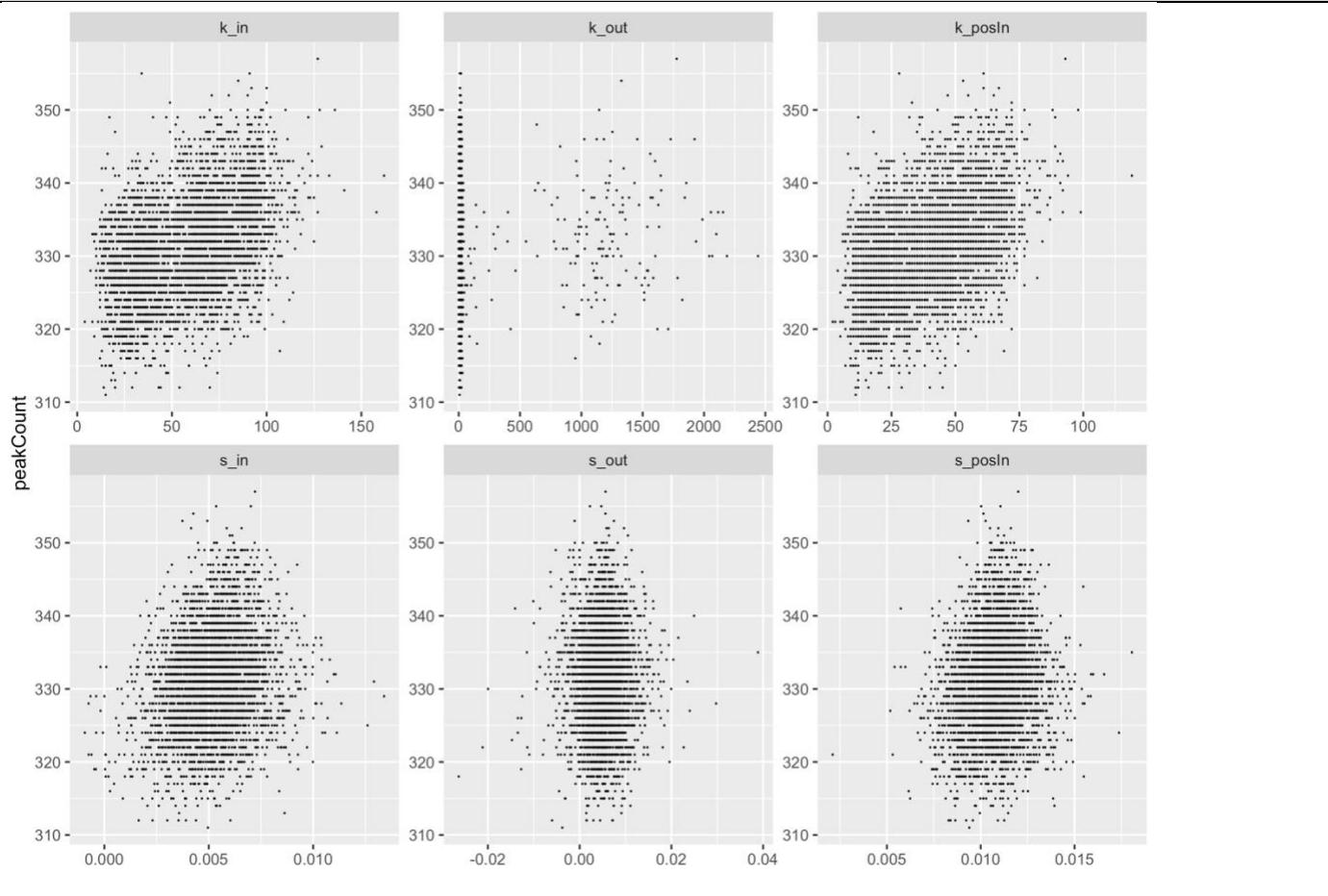


Exploratory analysis (2e6 steps)

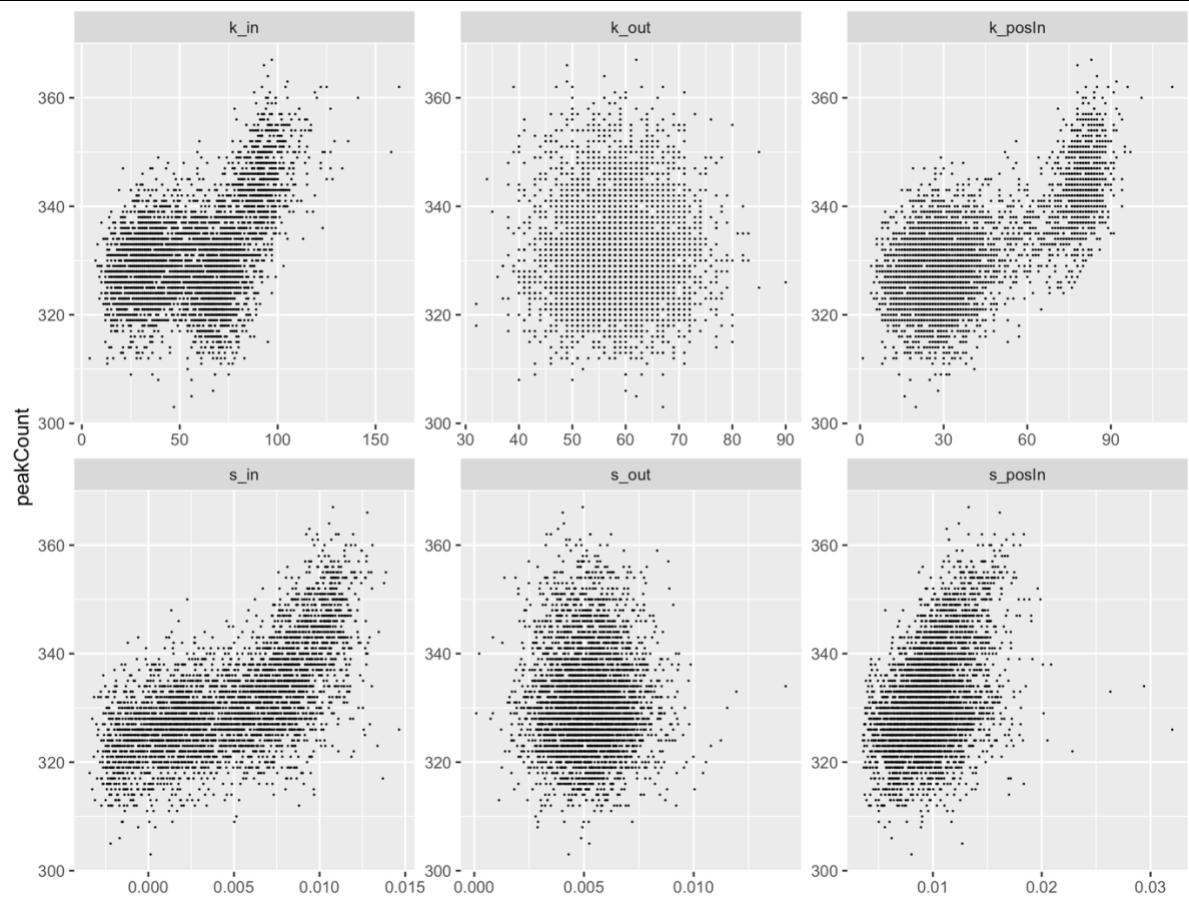
Original Network



Ref. Network 1 (non-zero gjij Gaussian)

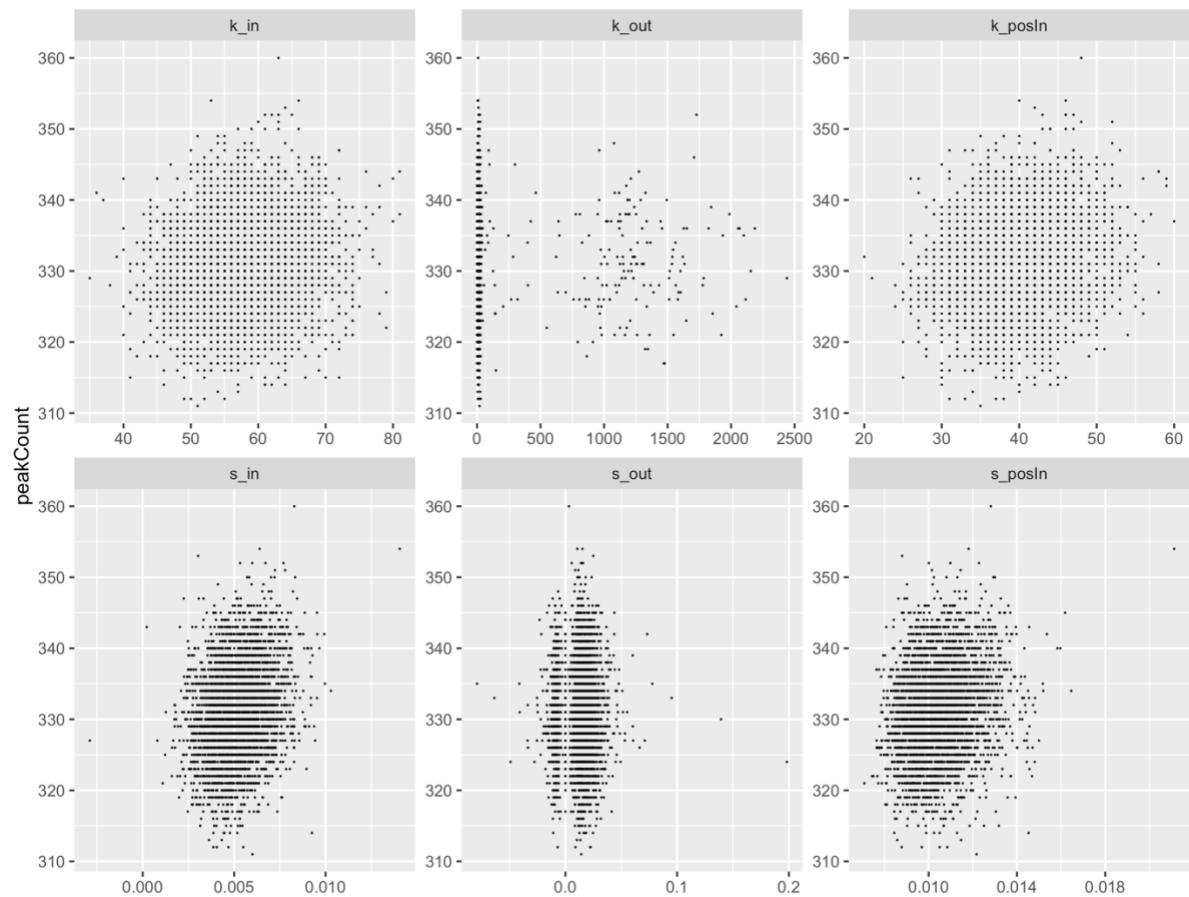


Ref. Network 2 (row gij shuffled)

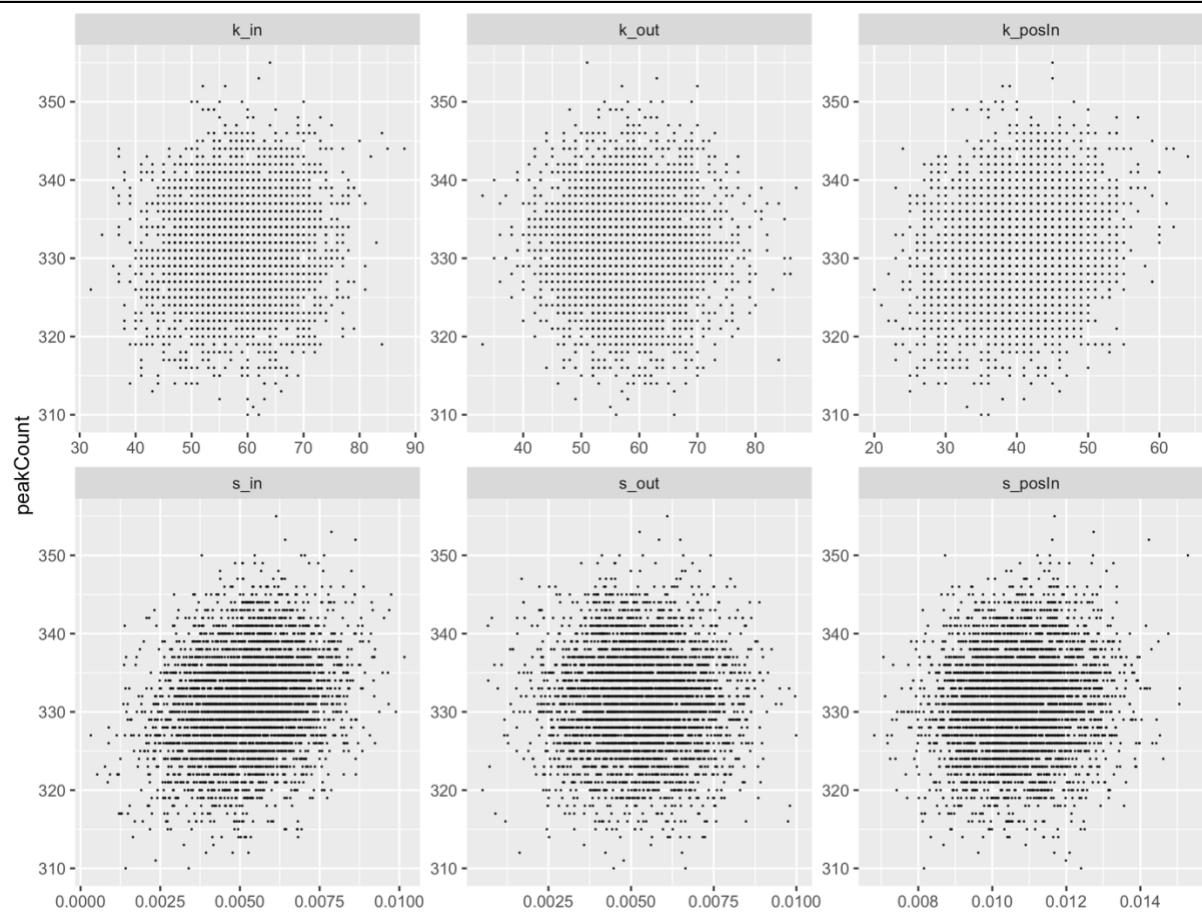


- Very similar to original

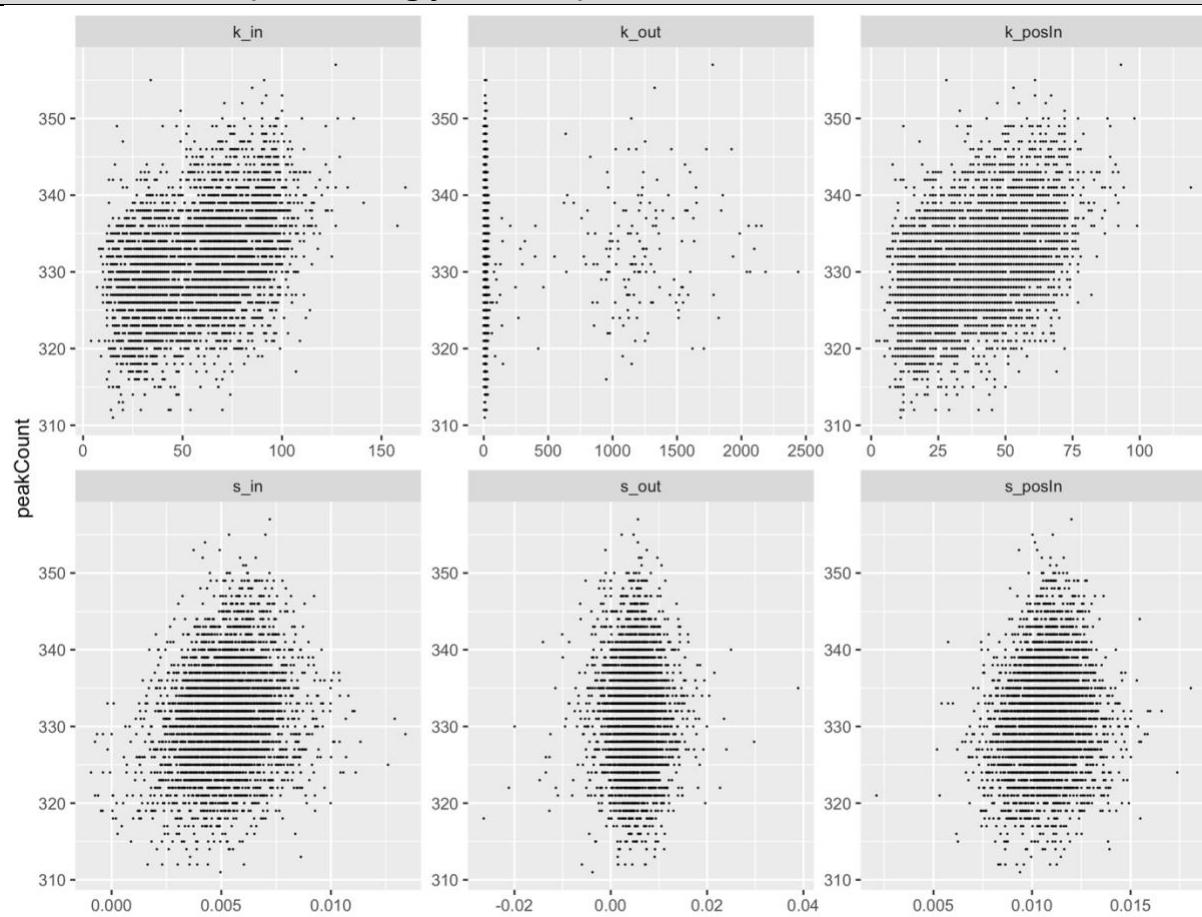
Ref. Network 3 (col gij shuffled)



Ref. Network 4 (random)

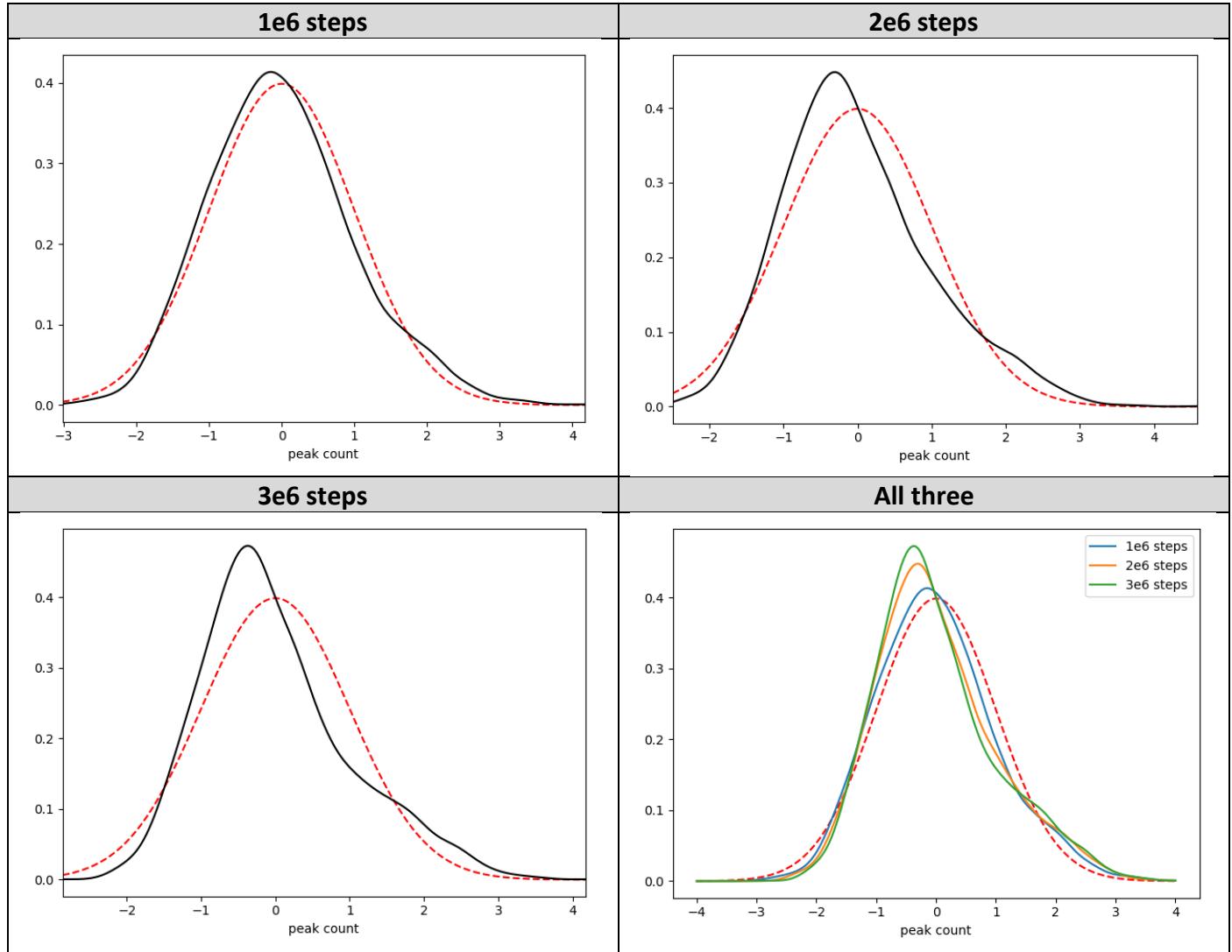


Ref. Network 5 (non-zero gij shuffled)



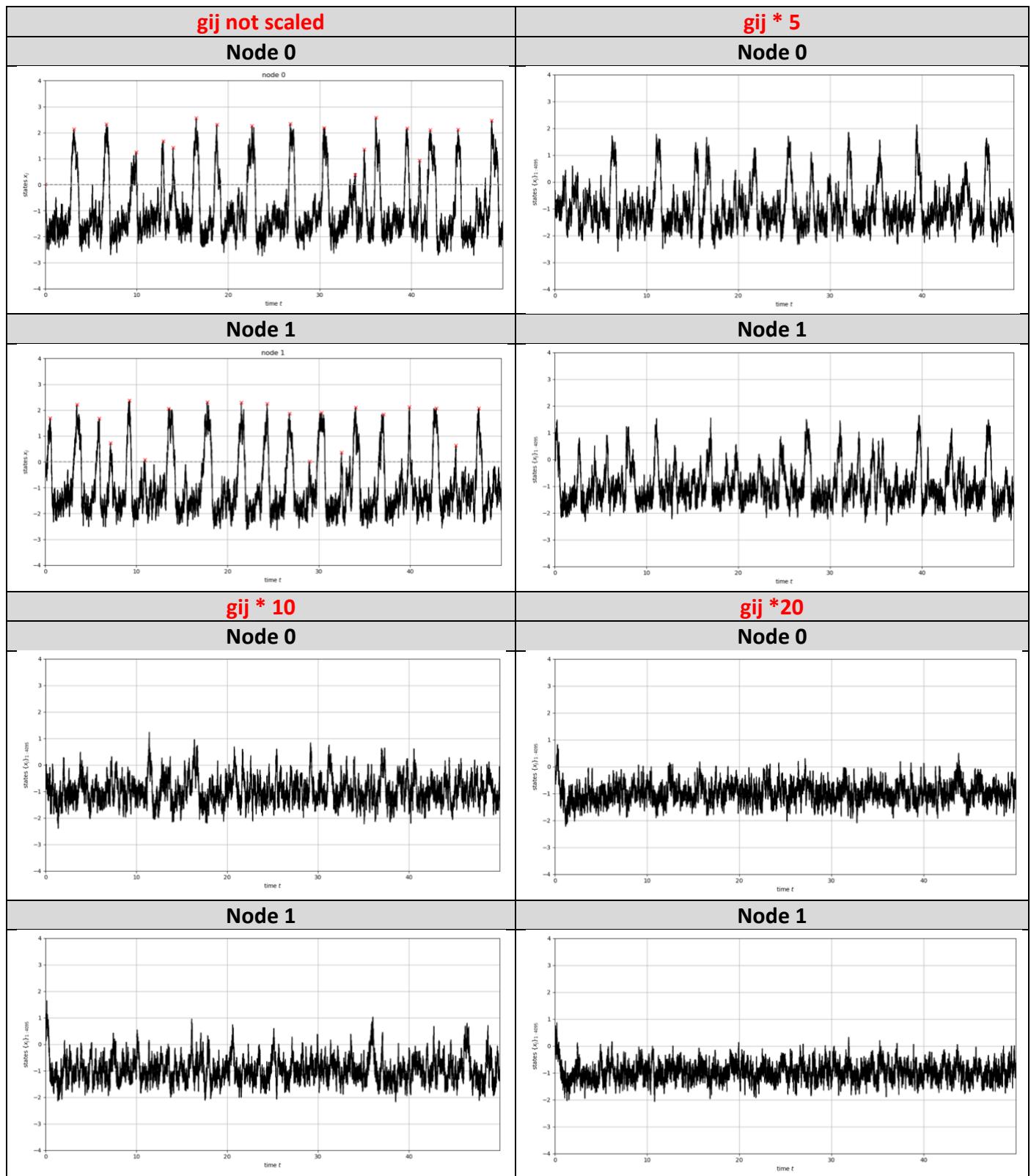
Peak count distribution comparison of Original Network (3e6 steps)

- To check if peak count distribution is stationary

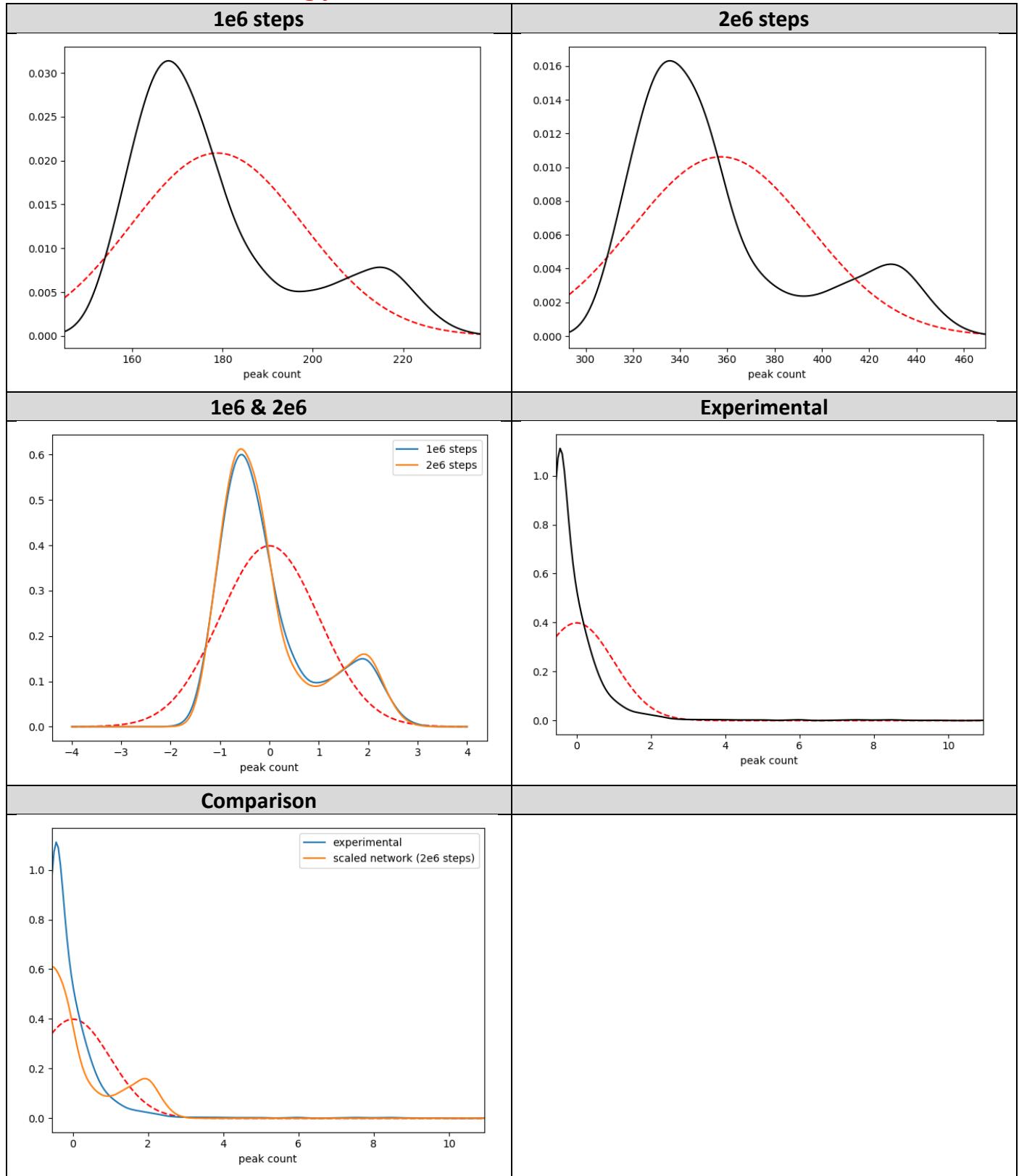


Model with scaled g_{ij}

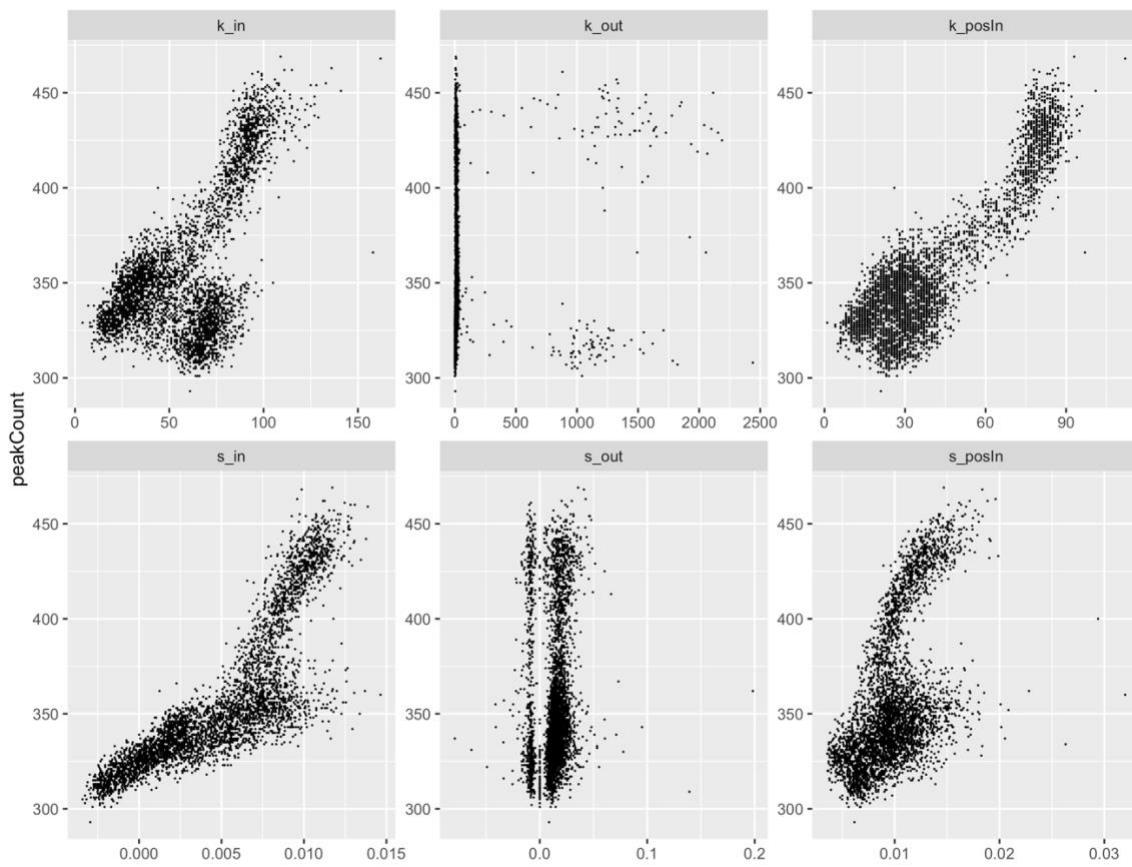
- Parameters $\epsilon = 0.1, \alpha = 0.95, \sigma_i = 2$



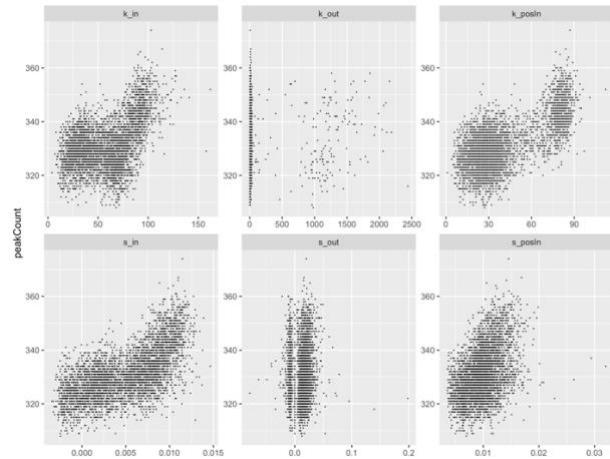
Peak count distribution of $g_{ij} * 5$ network



Exploratory analysis (2e6 steps)



Original network without scaling gjj (P59)



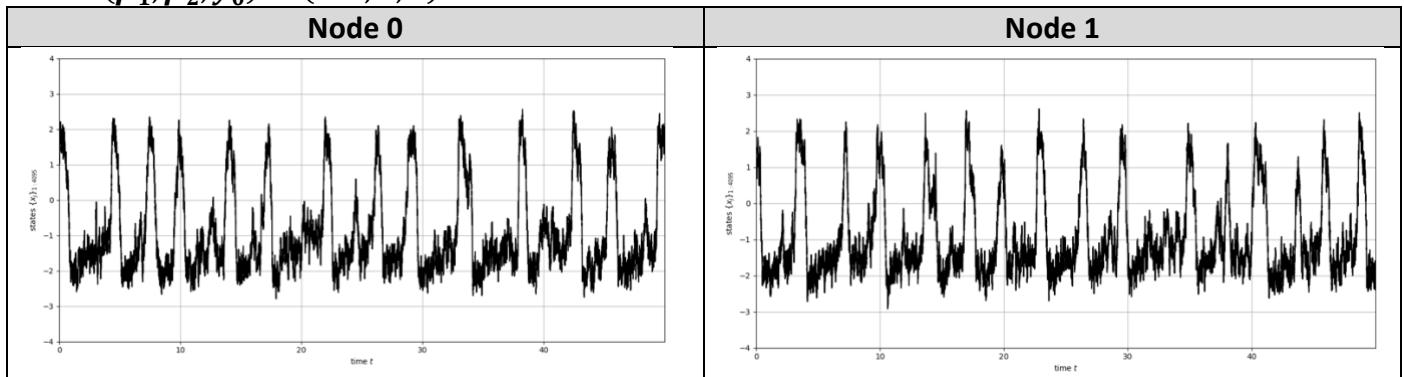
Scaled gjj makes patterns more obvious and more “exaggerated”

Synaptic FHN model

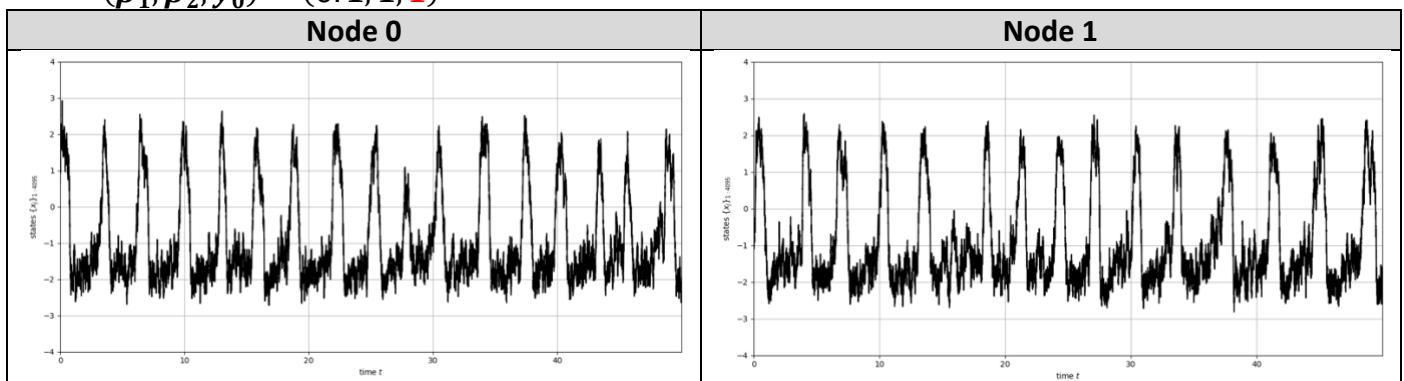
- Play around with different parameters

$$h^{\text{syn}}(x, y) = (1/\beta_1)\{1 + \tanh[\beta_2(y - y_0)]\}.$$

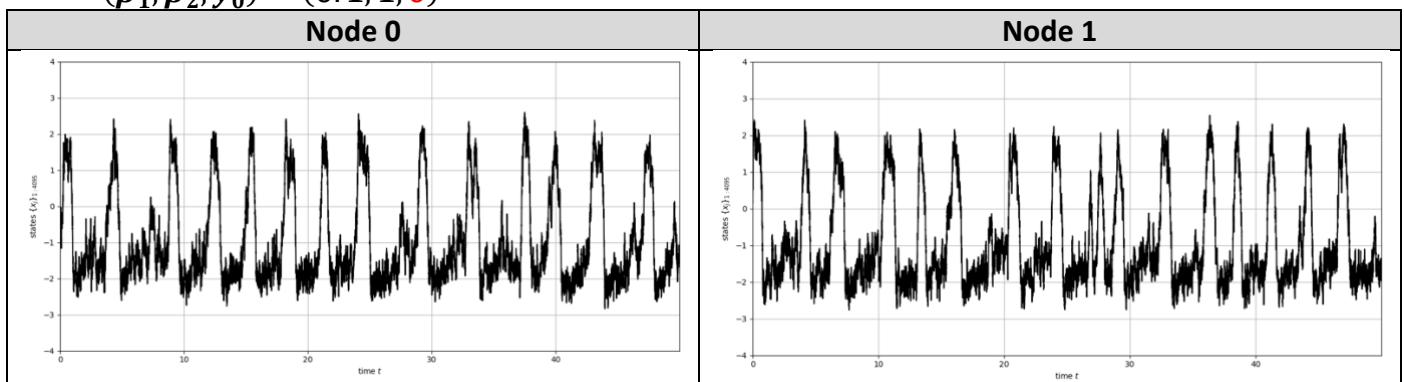
- $(\beta_1, \beta_2, y_0) = (0.5, 1, 1)$



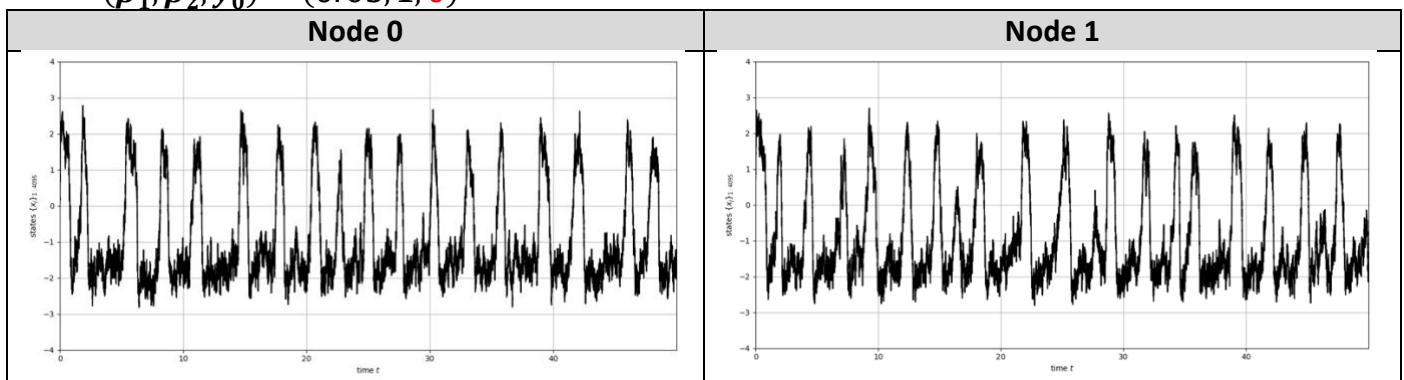
- $(\beta_1, \beta_2, y_0) = (0.1, 1, 1)$



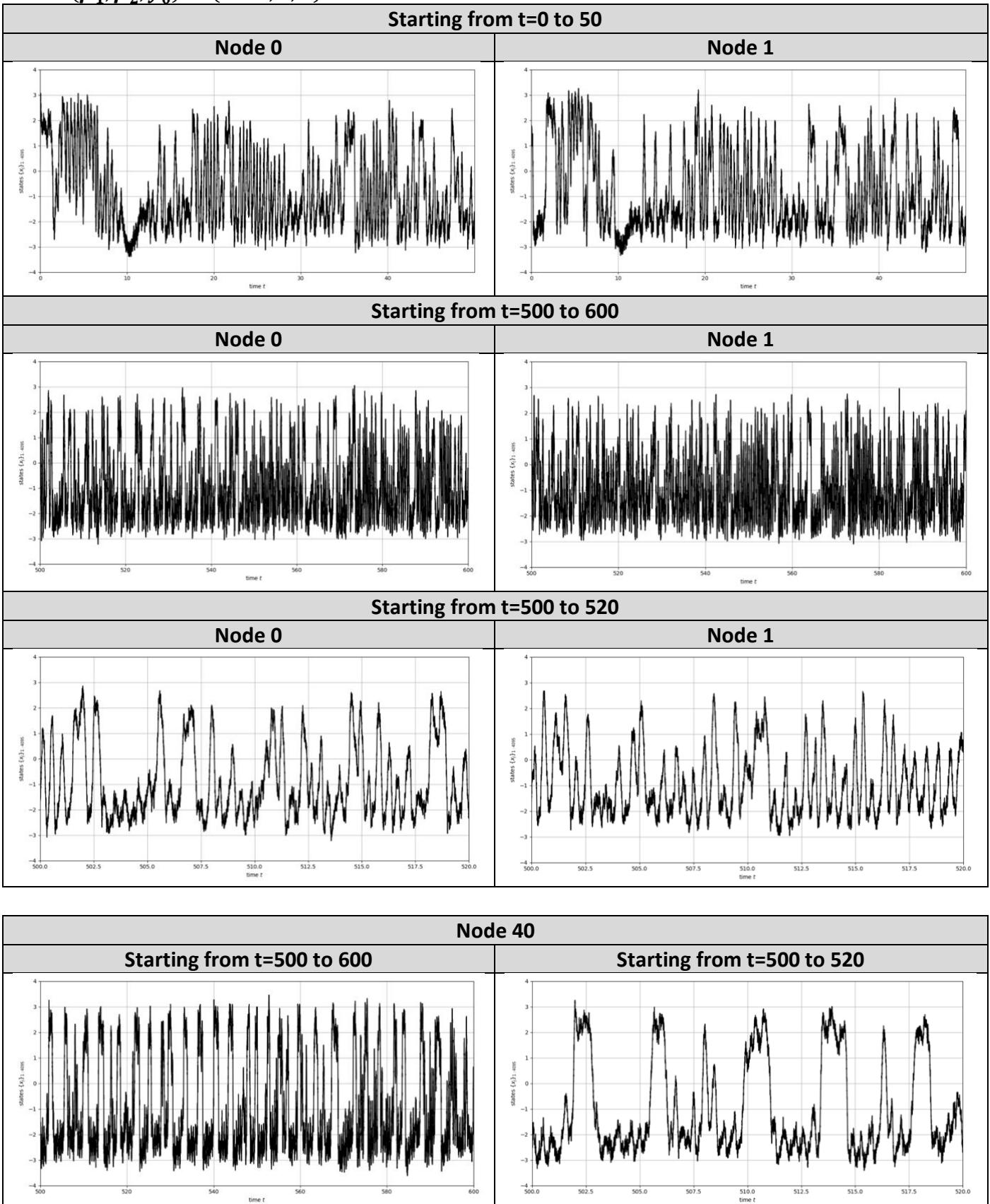
- $(\beta_1, \beta_2, y_0) = (0.1, 1, 0)$



- $(\beta_1, \beta_2, y_0) = (0.05, 1, 0)$

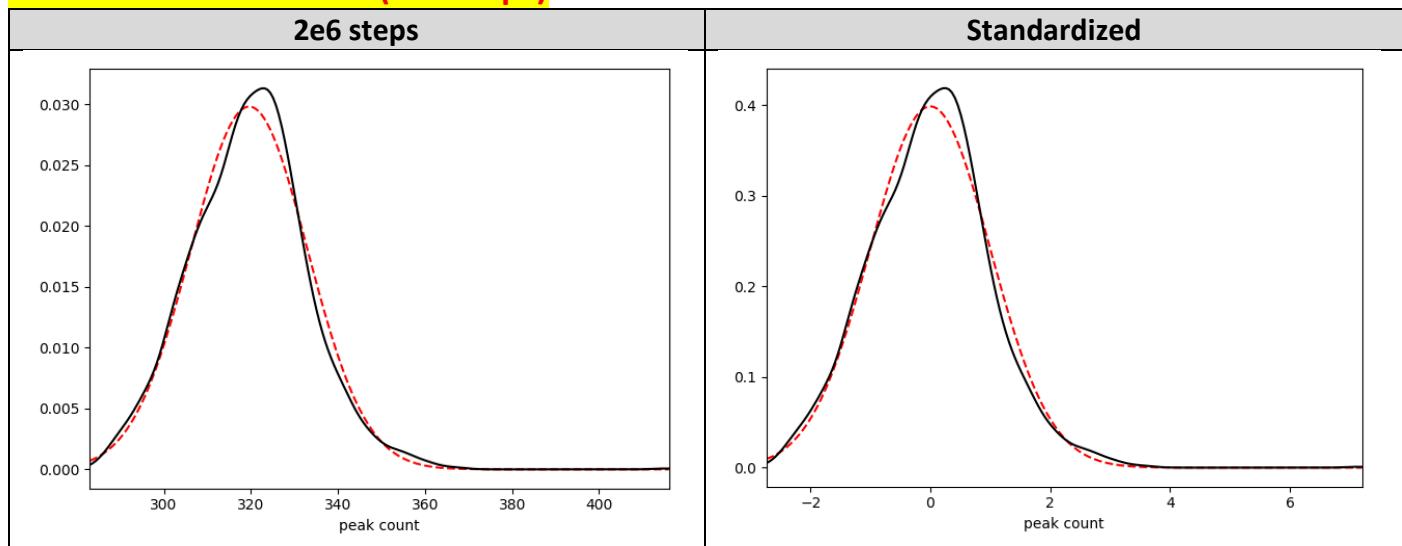


- $(\beta_1, \beta_2, y_0) = (0.01, 1, 0)$

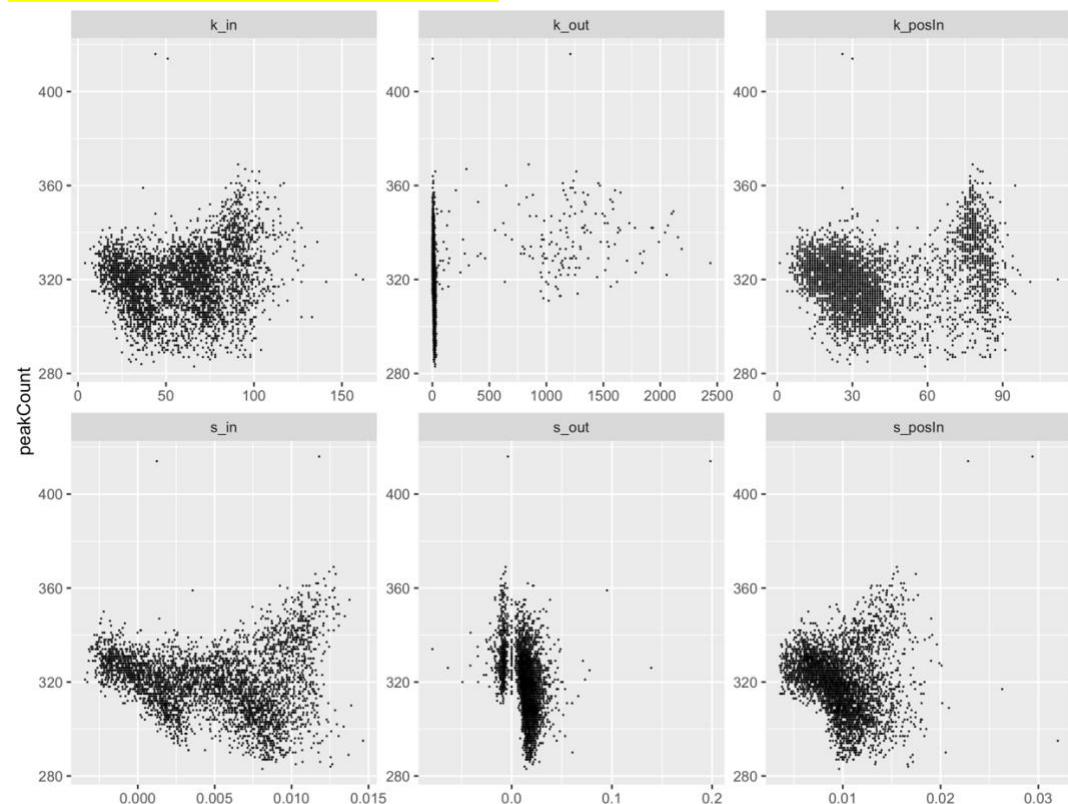


$$(\beta_1, \beta_2, y_0) = (0.05, 1, 0); \epsilon = 0.1, \alpha = 0.95$$

Peak count distribution (2e6 steps)

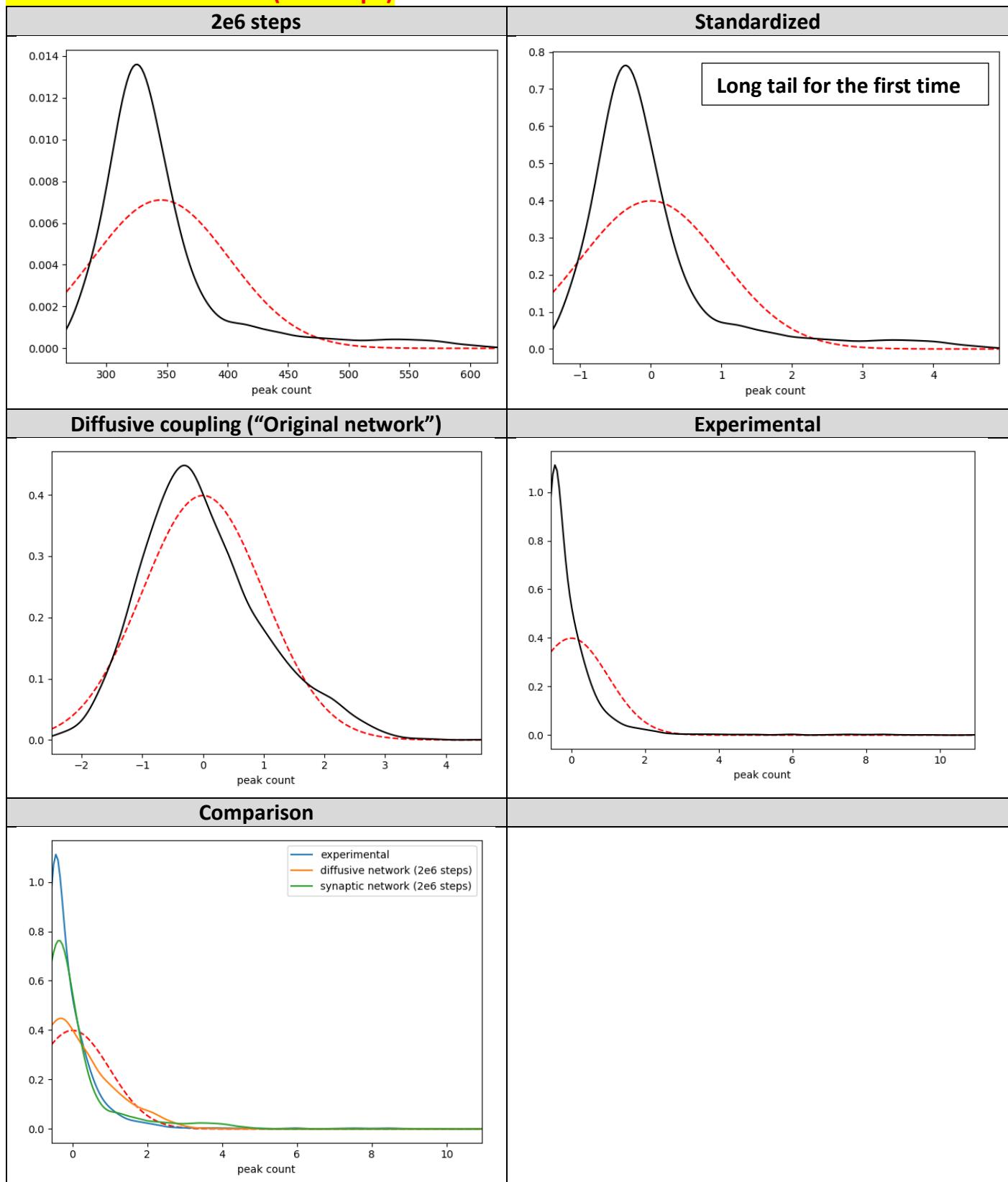


Exploratory analysis (2e6 steps)

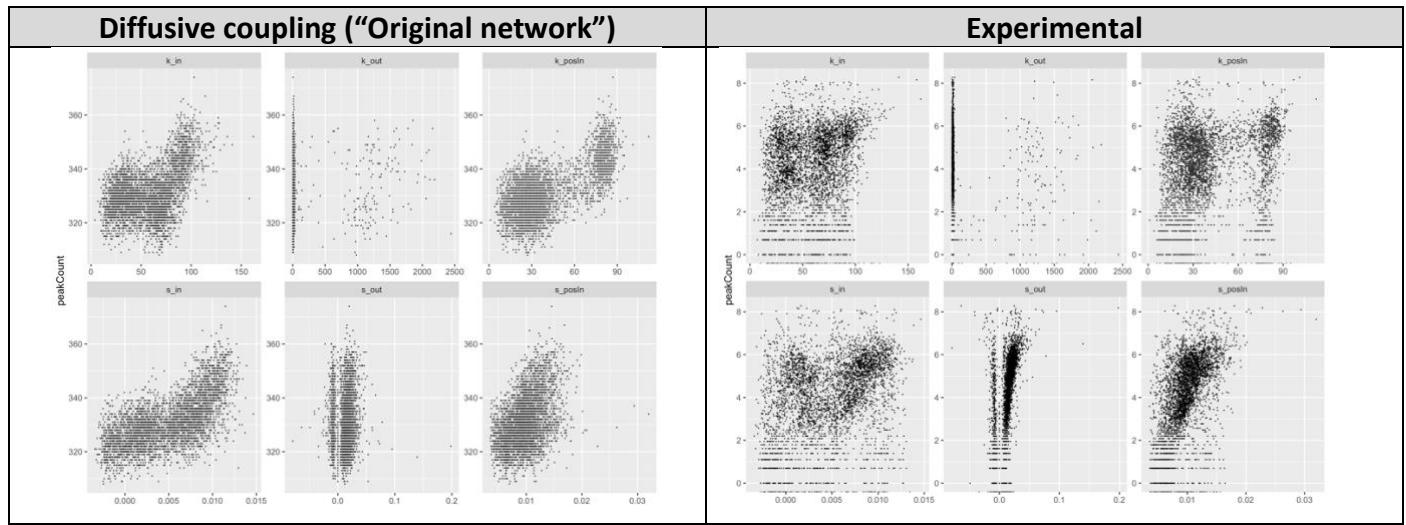
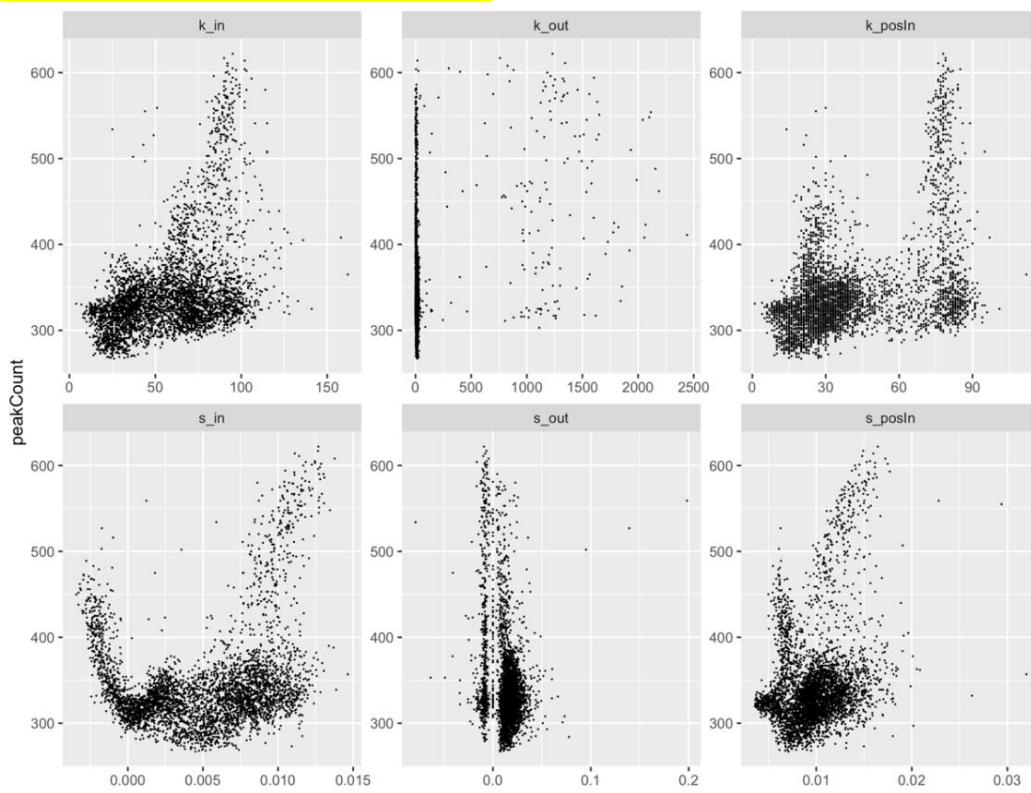


$$(\beta_1, \beta_2, y_0) = (0.01, 1, 0); \epsilon = 0.1, \alpha = 0.95$$

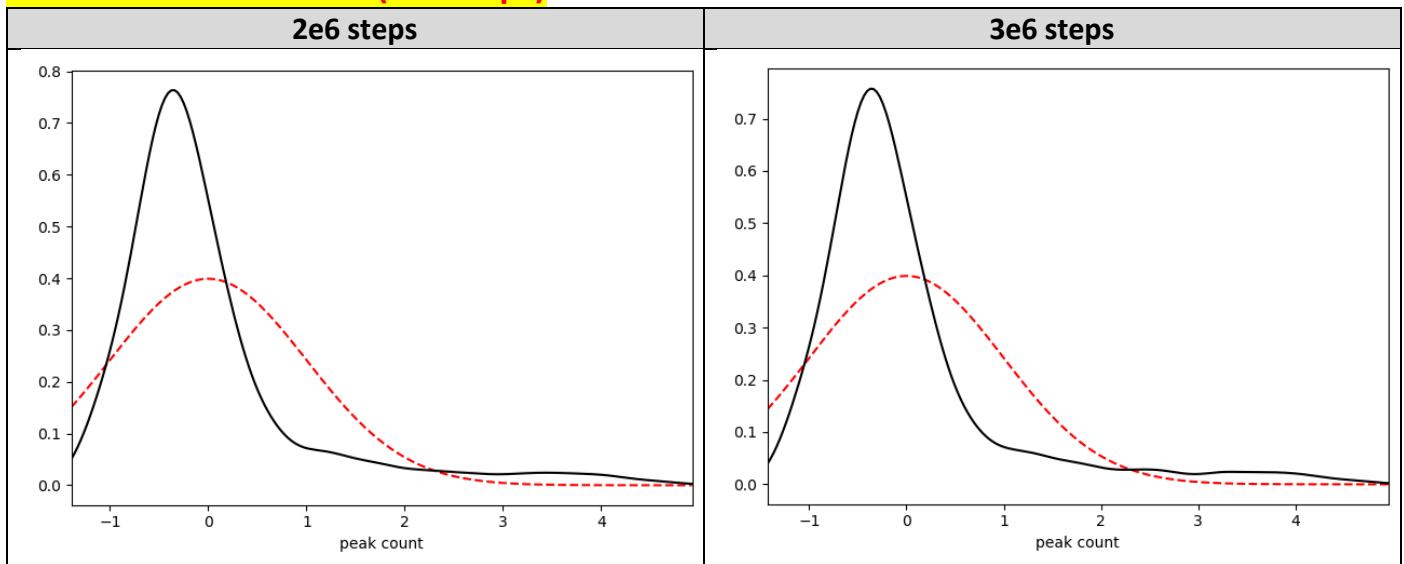
Peak count distribution (2e6 steps)



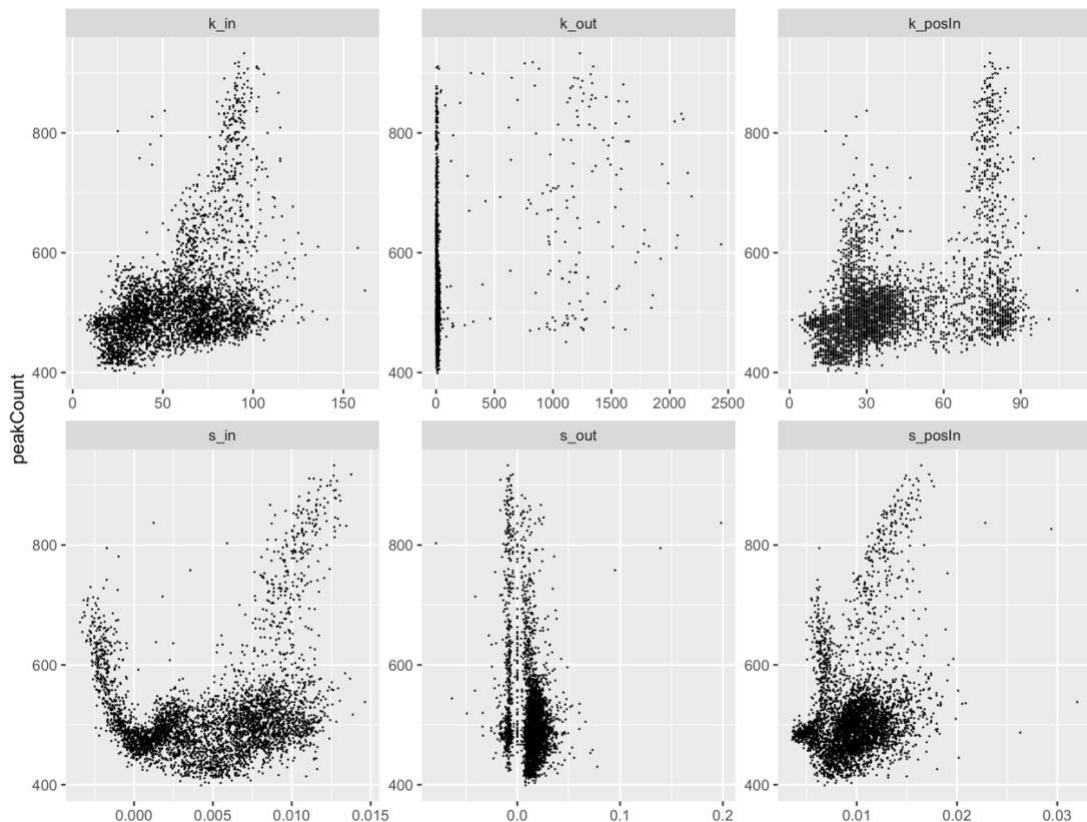
Exploratory analysis (2e6 steps)



Peak count distribution (3e6 steps)



Exploratory analysis (3e6 steps)



Analysis on reference networks

- $(\beta_1, \beta_2, y_0) = (0.01, 1, 0)$; $\epsilon = 0.1, \alpha = 0.95$

(a) reference network 1: keep A_{ij} but replace nonzero g_{ij} by values taken from a Gaussian distribution of same mean and standard deviation (this time we do not separately consider positive and negative g_{ij} 's). This network has same k_{in} and k_{out} but different s_{in} and s_{out}

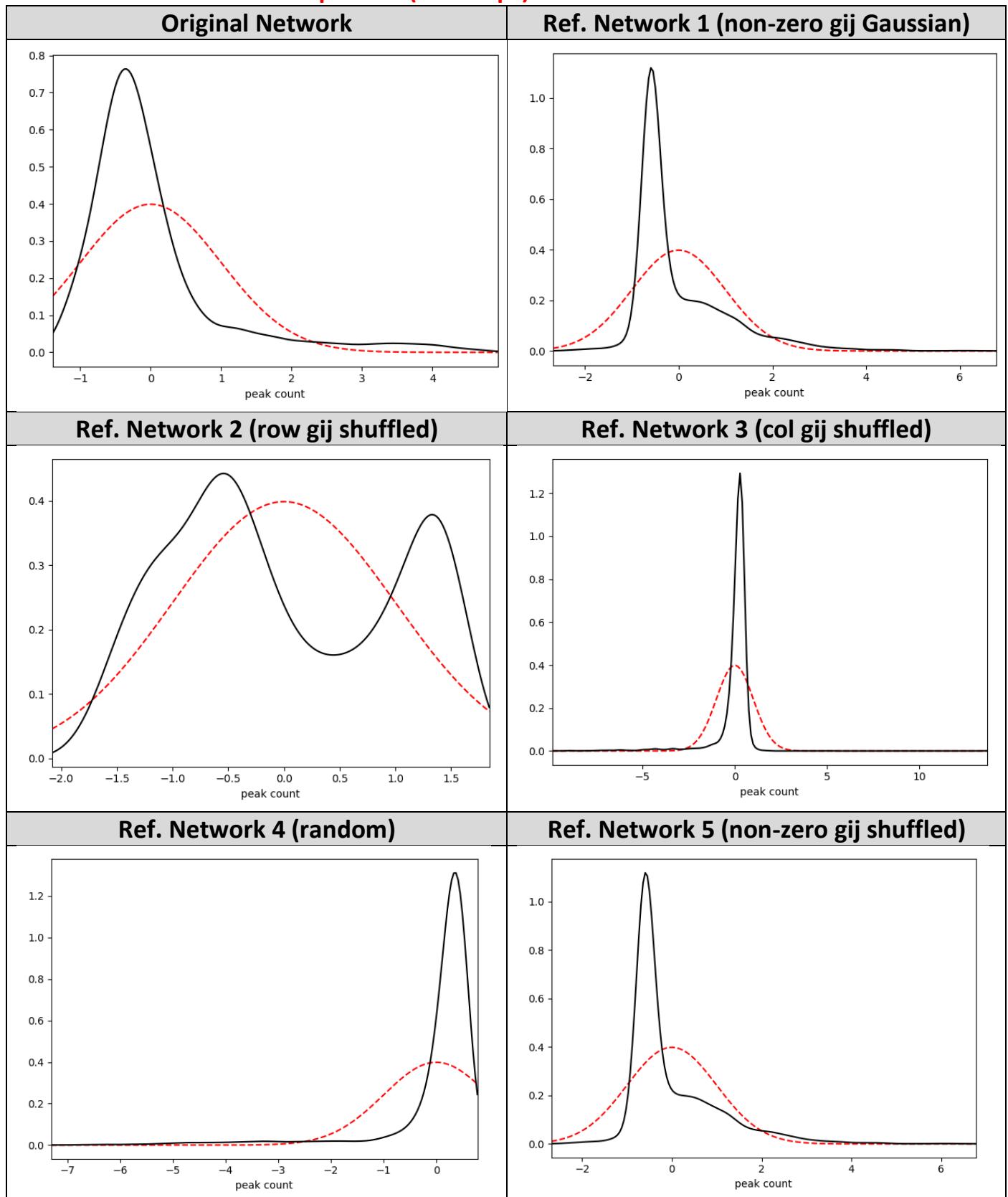
(b) reference network 2: shuffle g_{ij} for fixed i ; this has same k_{in} and s_{in} but different k_{out} and s_{out}

(c) reference network 3: shuffle g_{ij} for fixed j ; this has same k_{out} and s_{out} but different k_{in} and s_{in}

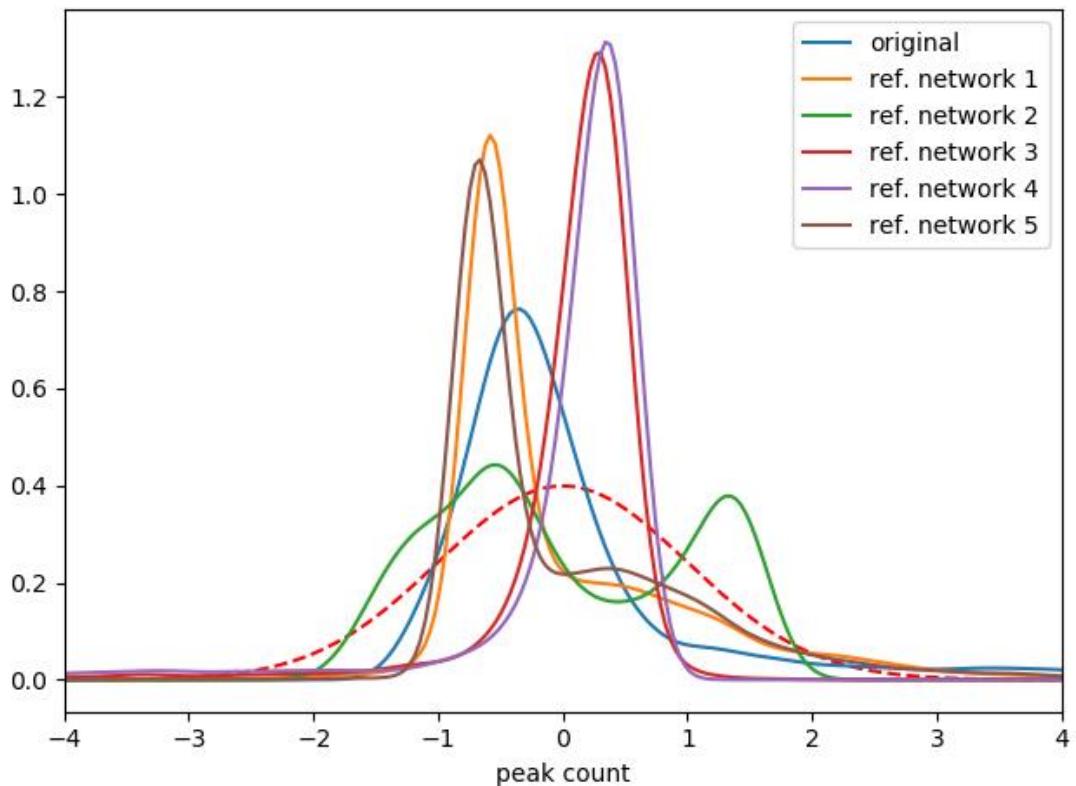
(d) reference network 4: random directed network with same connection probability p and g_{ij} from a Gaussian distribution of same mean and standard deviation

(e) reference network 5: keep A_{ij} but shuffle non-zero g_{ij}

Peak count distribution comparison (2e6 steps)

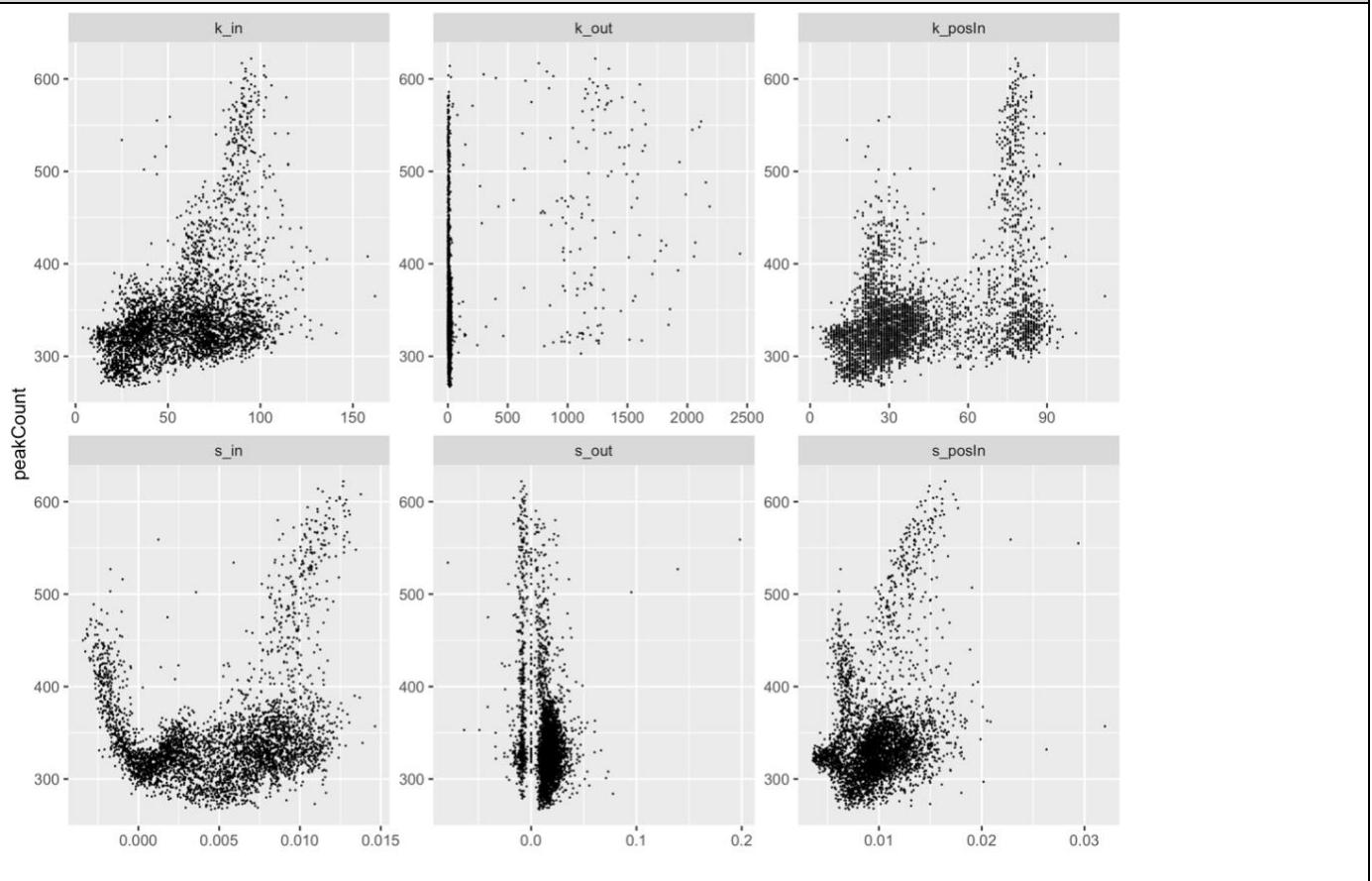


Combined

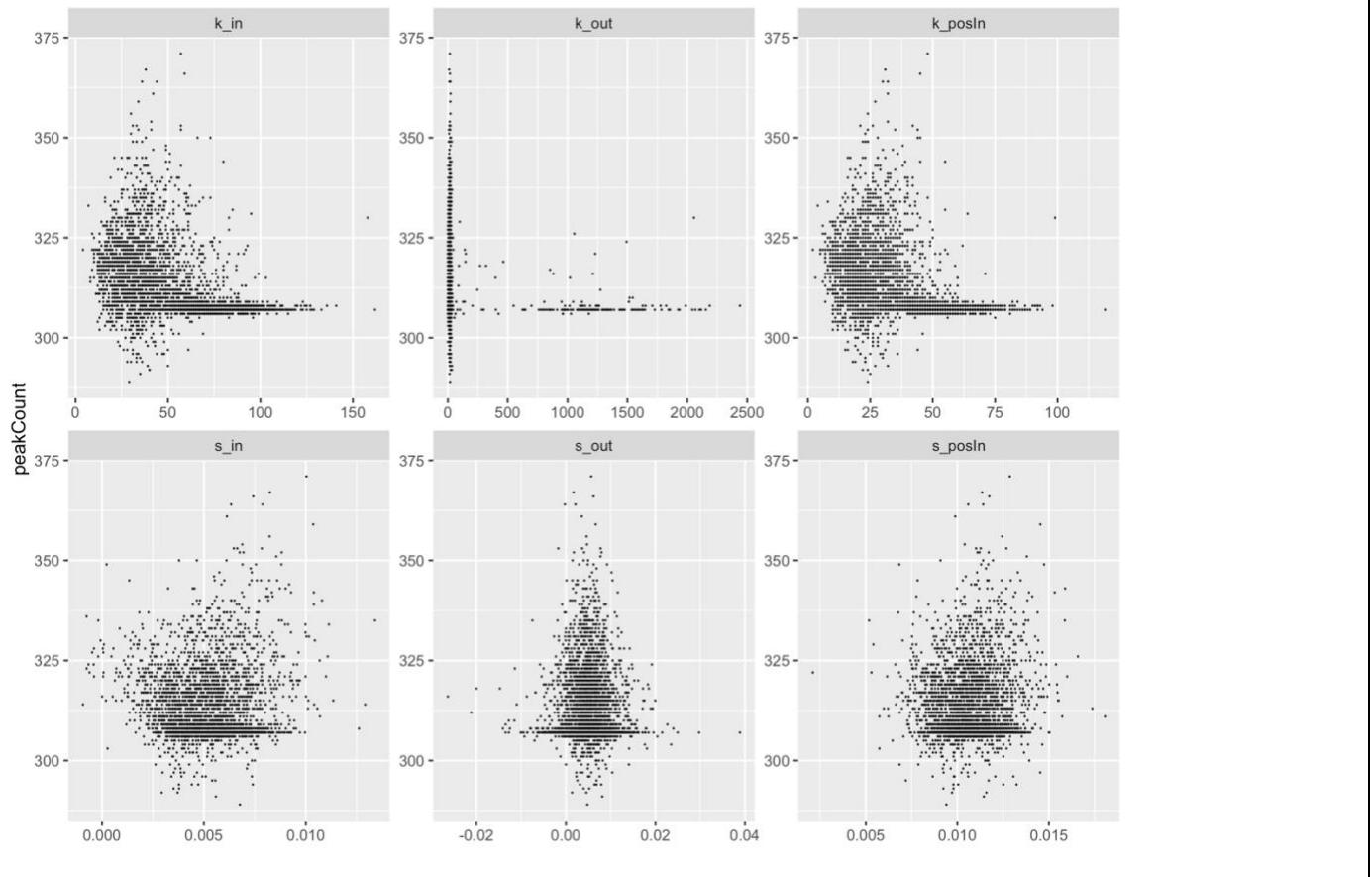


Exploratory analysis (2e6 steps)

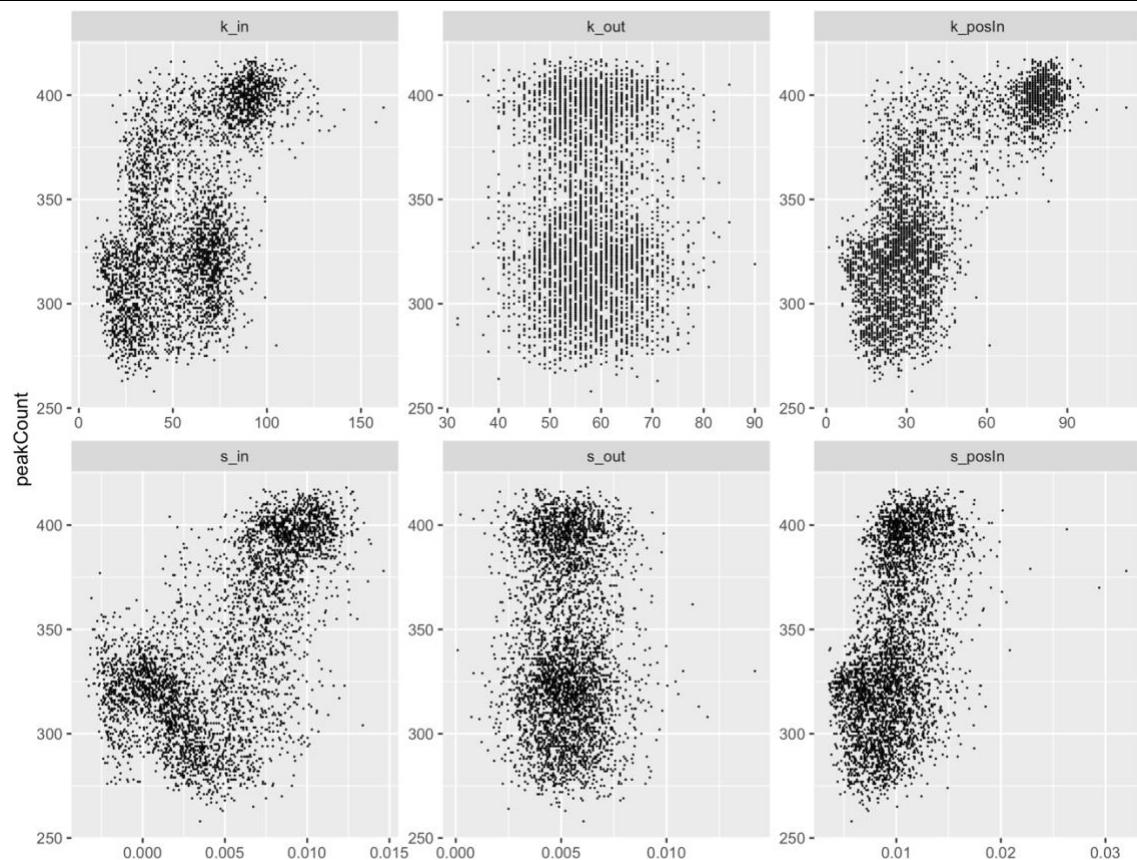
Original Network



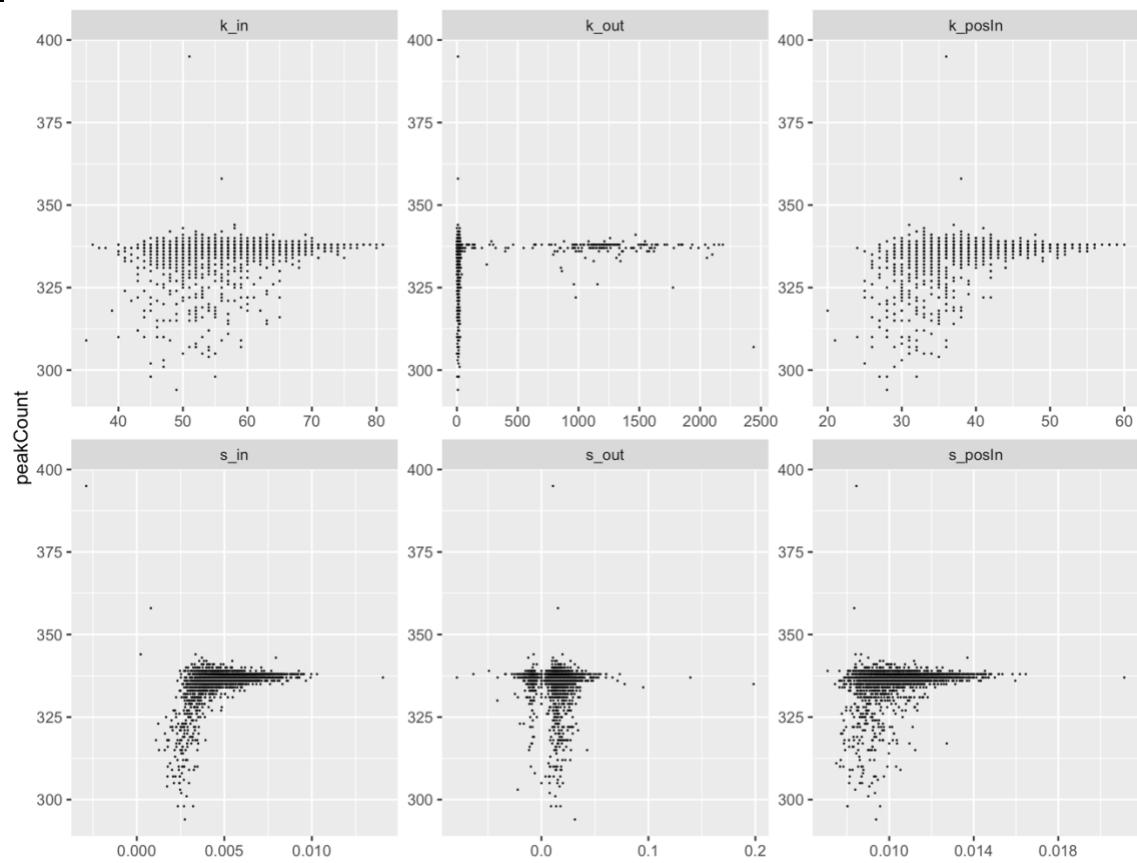
Ref. Network 1 (non-zero gjij Gaussian)



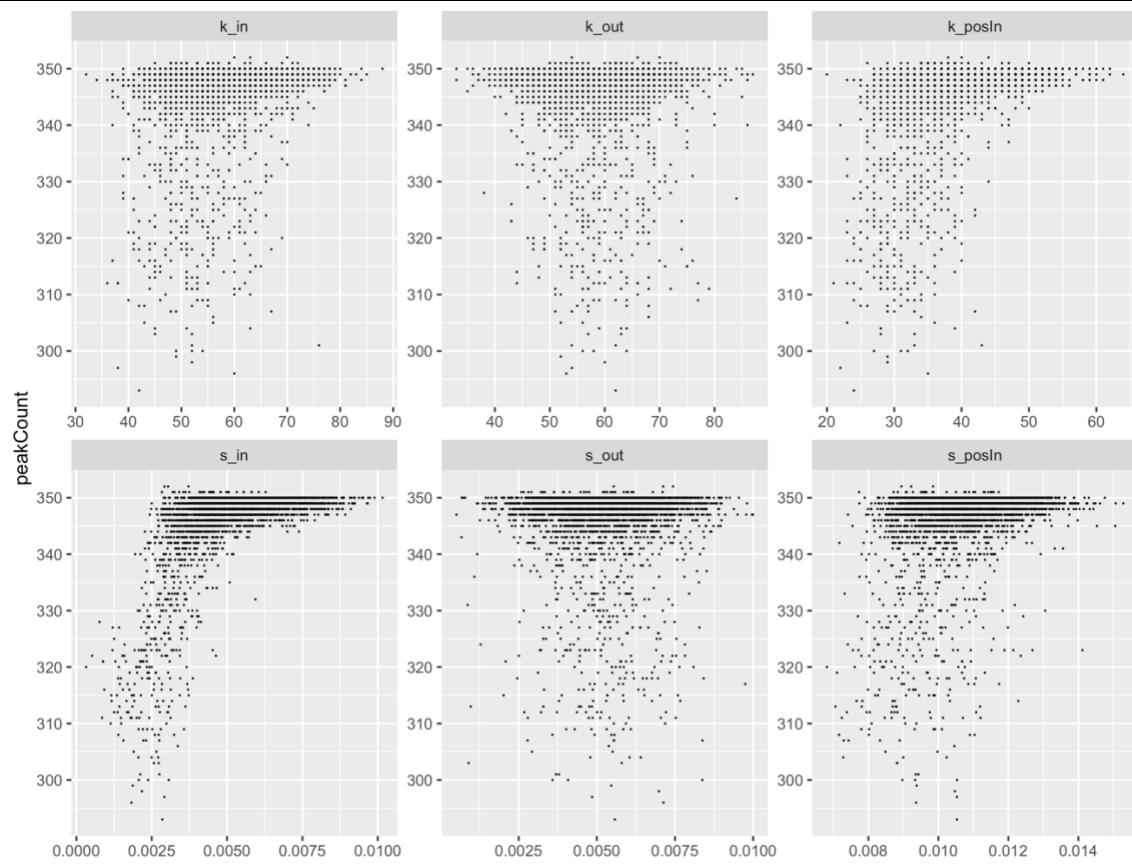
Ref. Network 2 (row gij shuffled)



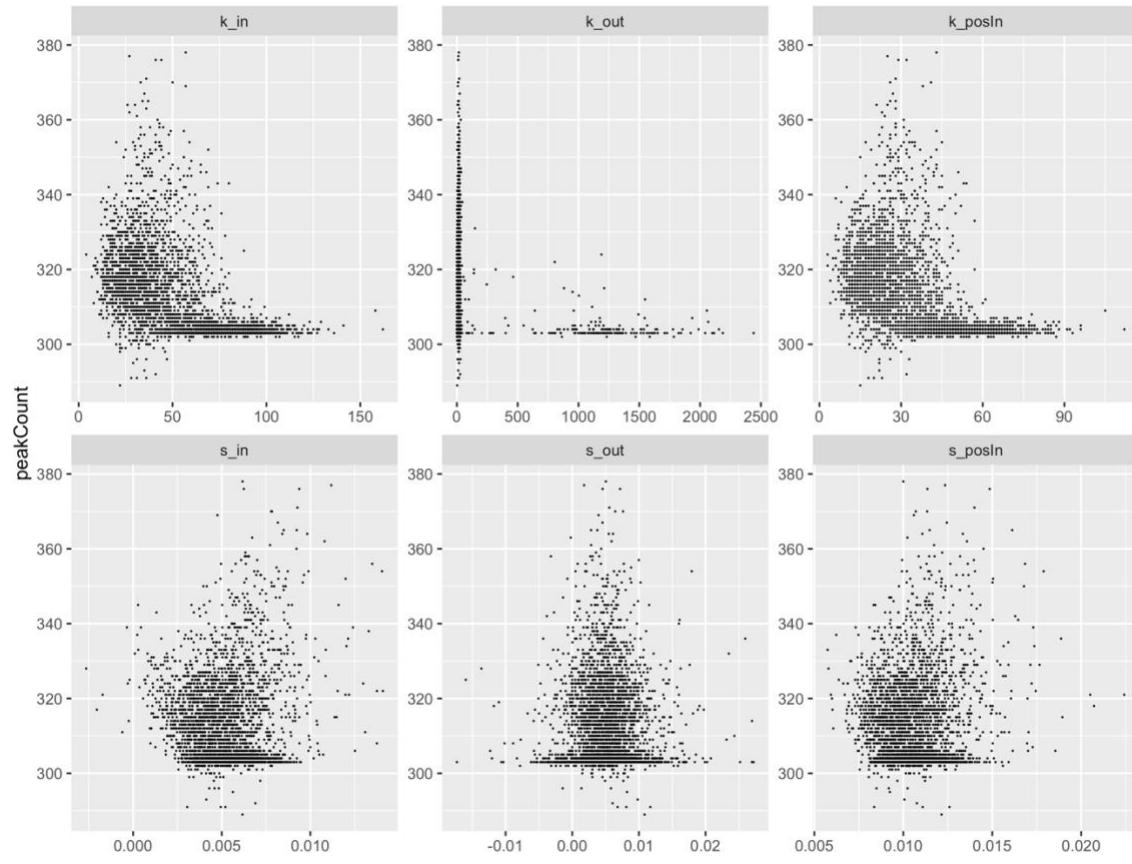
Ref. Network 3 (col gij shuffled)



Ref. Network 4 (random)

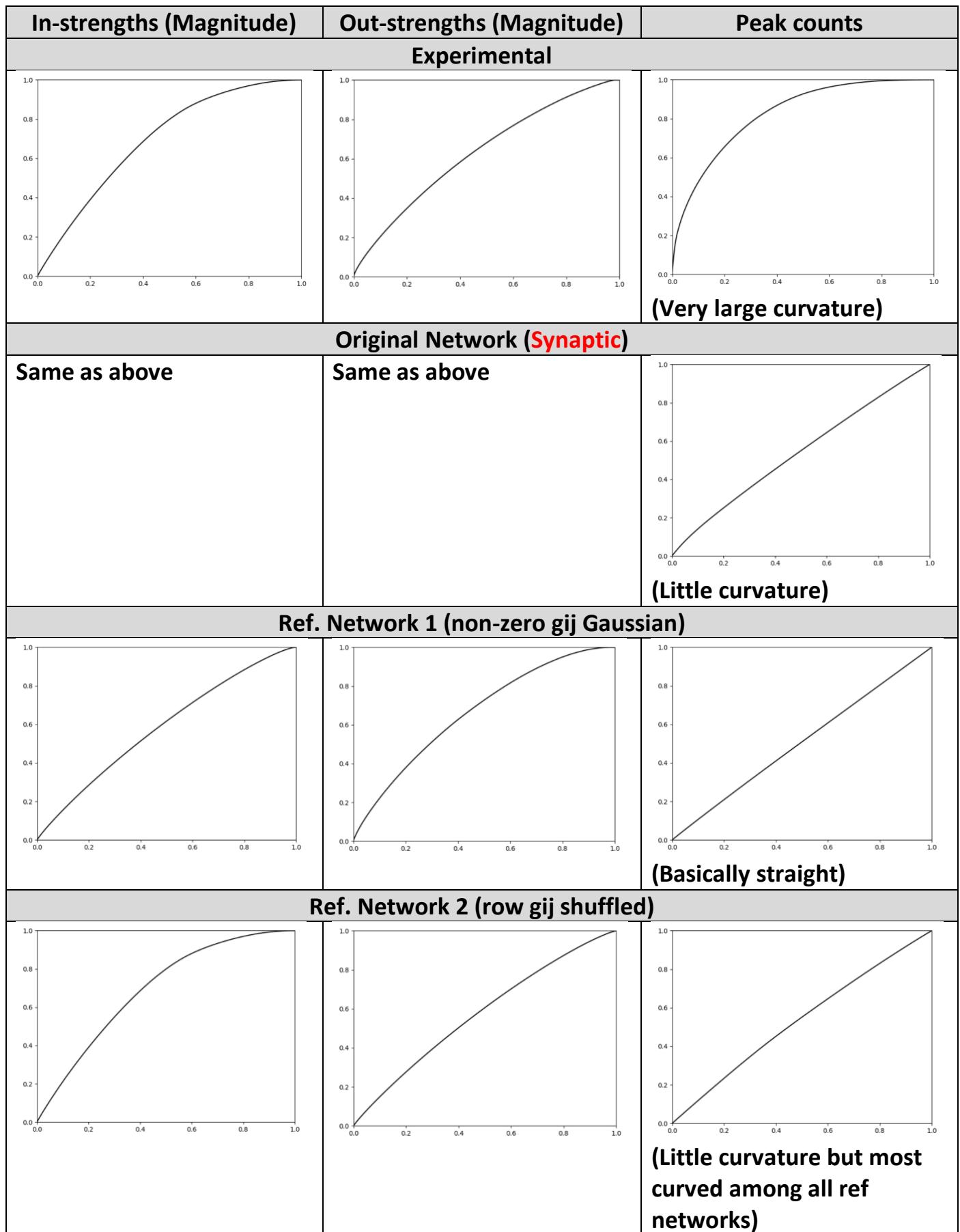


Ref. Network 5 (non-zero gij shuffled)

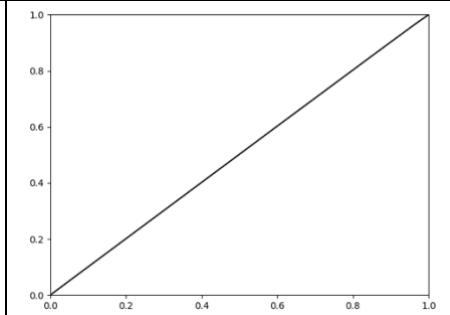
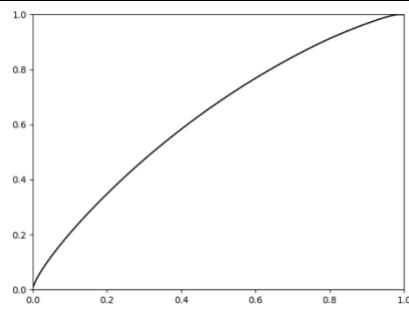
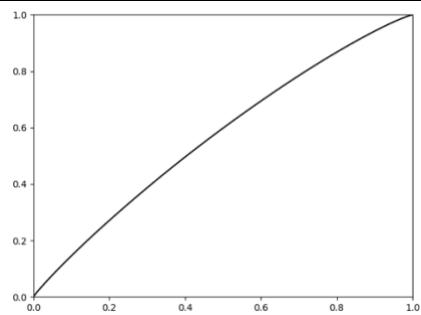


Analysis on peak counts

- Fraction of total accounted for by the top quantiles

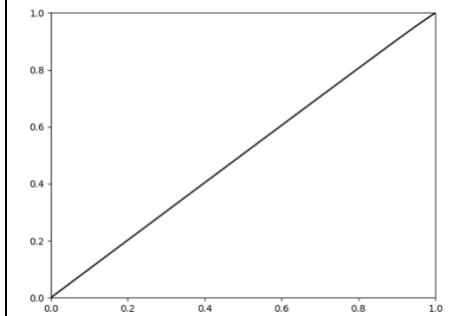
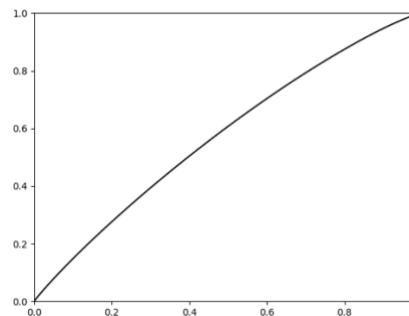
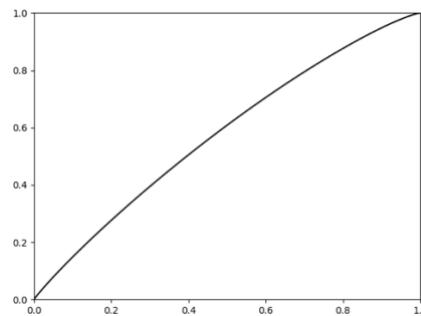


Ref. Network 3 (col gij shuffled)



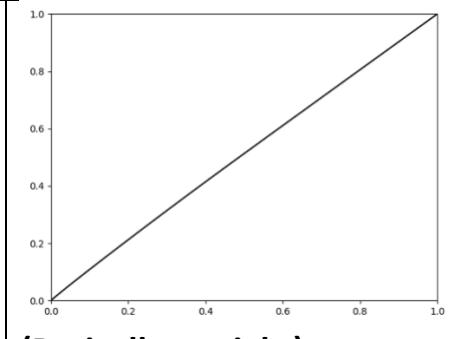
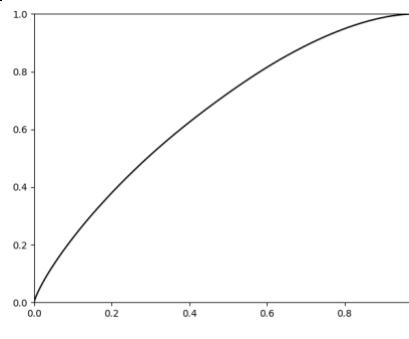
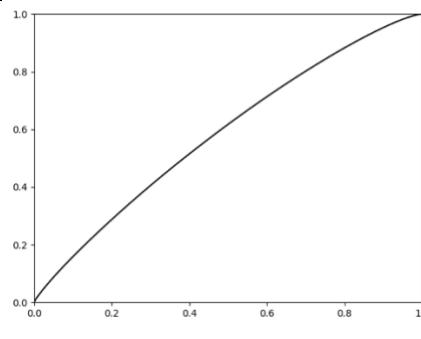
(Basically straight)

Ref. Network 4 (random)

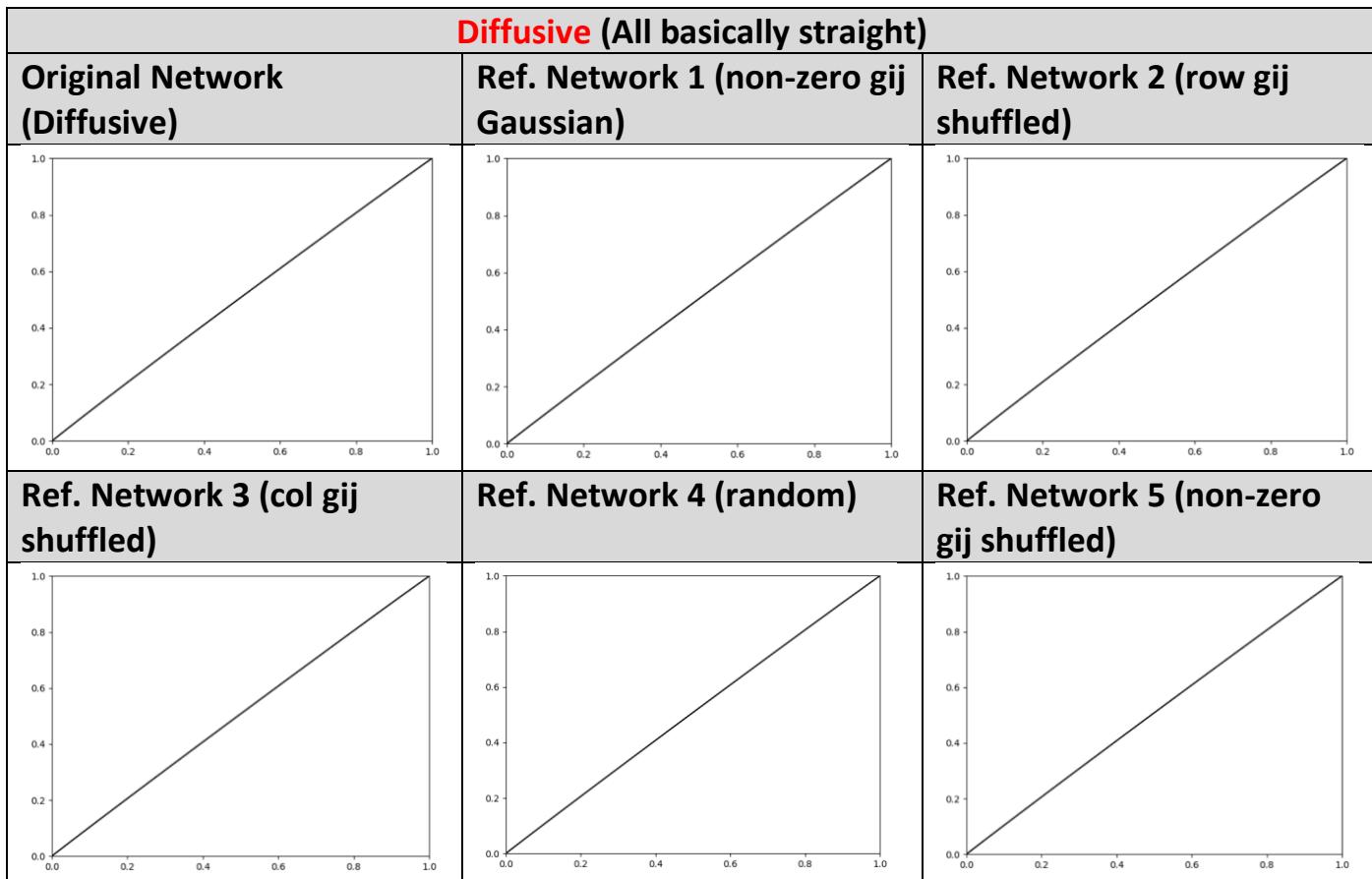


(Basically straight)

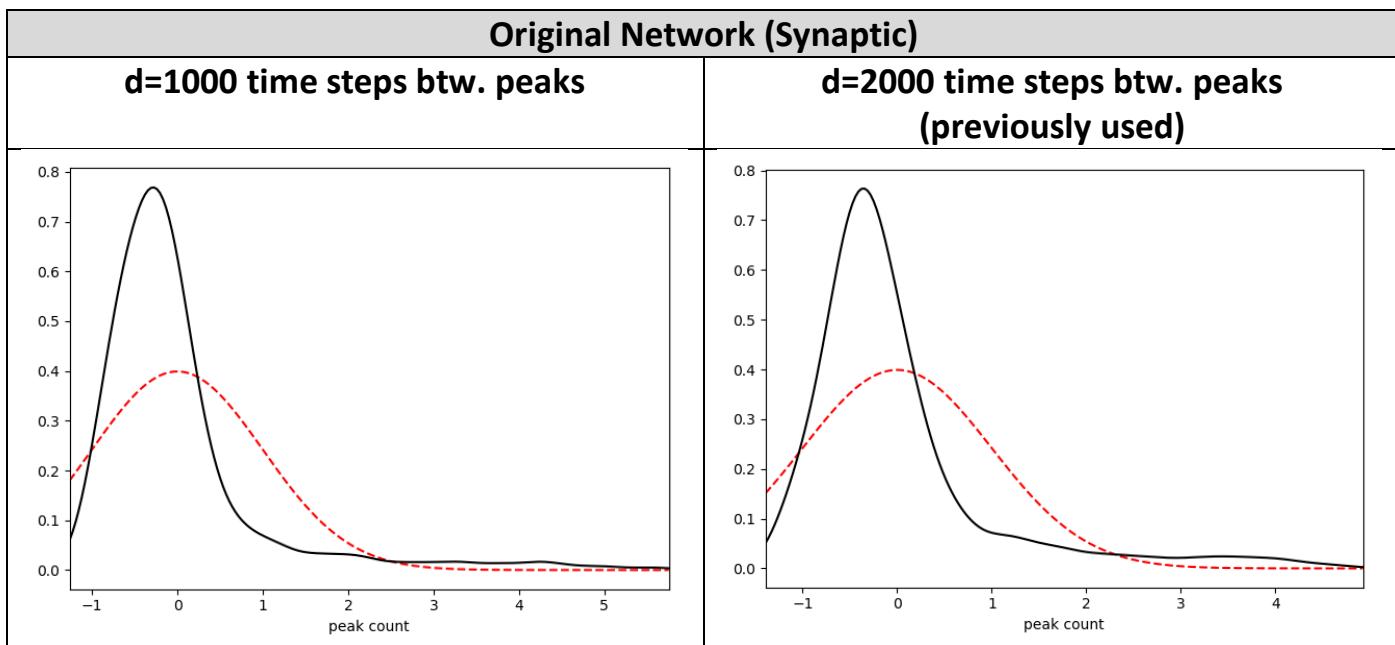
Ref. Network 5 (non-zero gij shuffled)



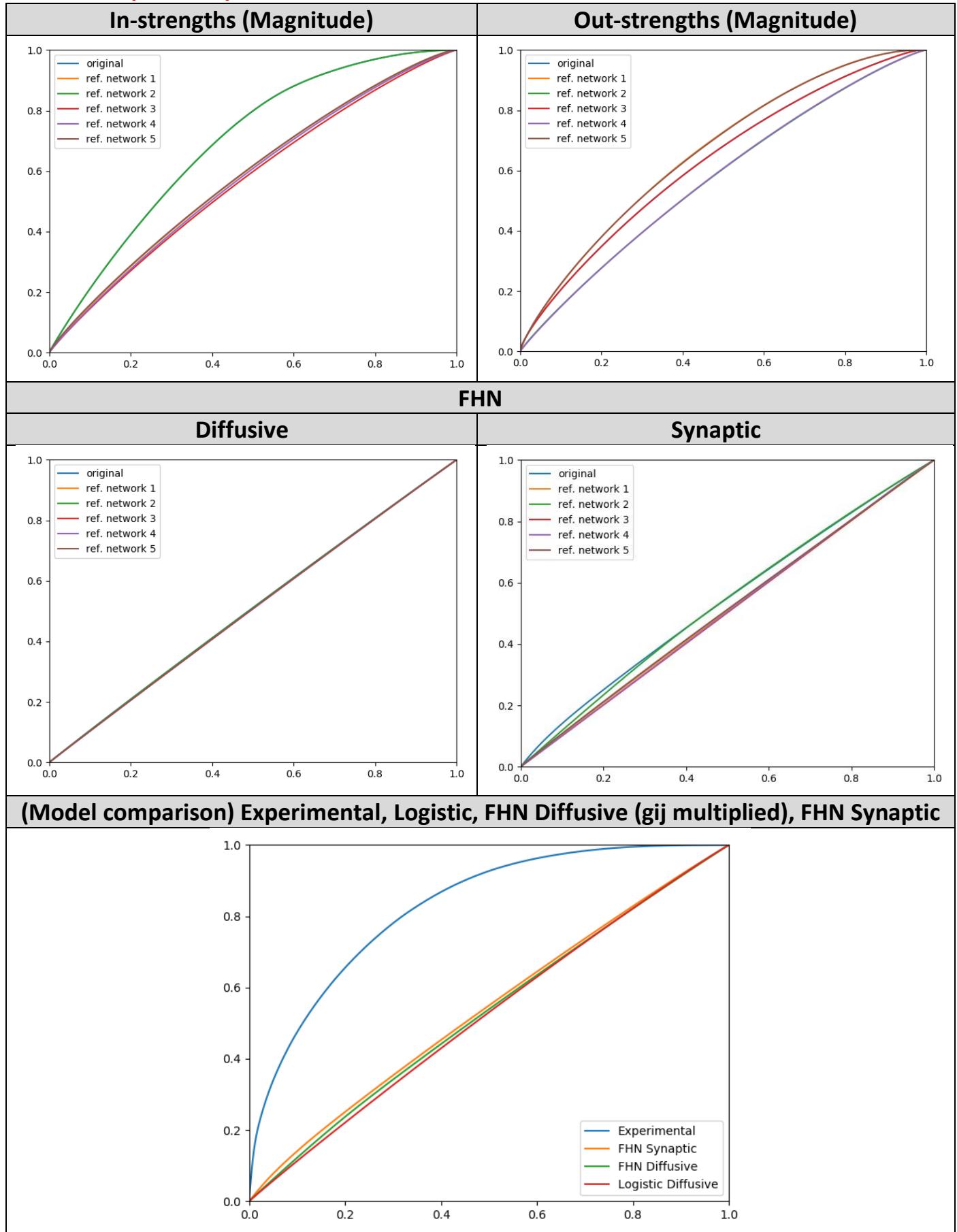
(Basically straight)



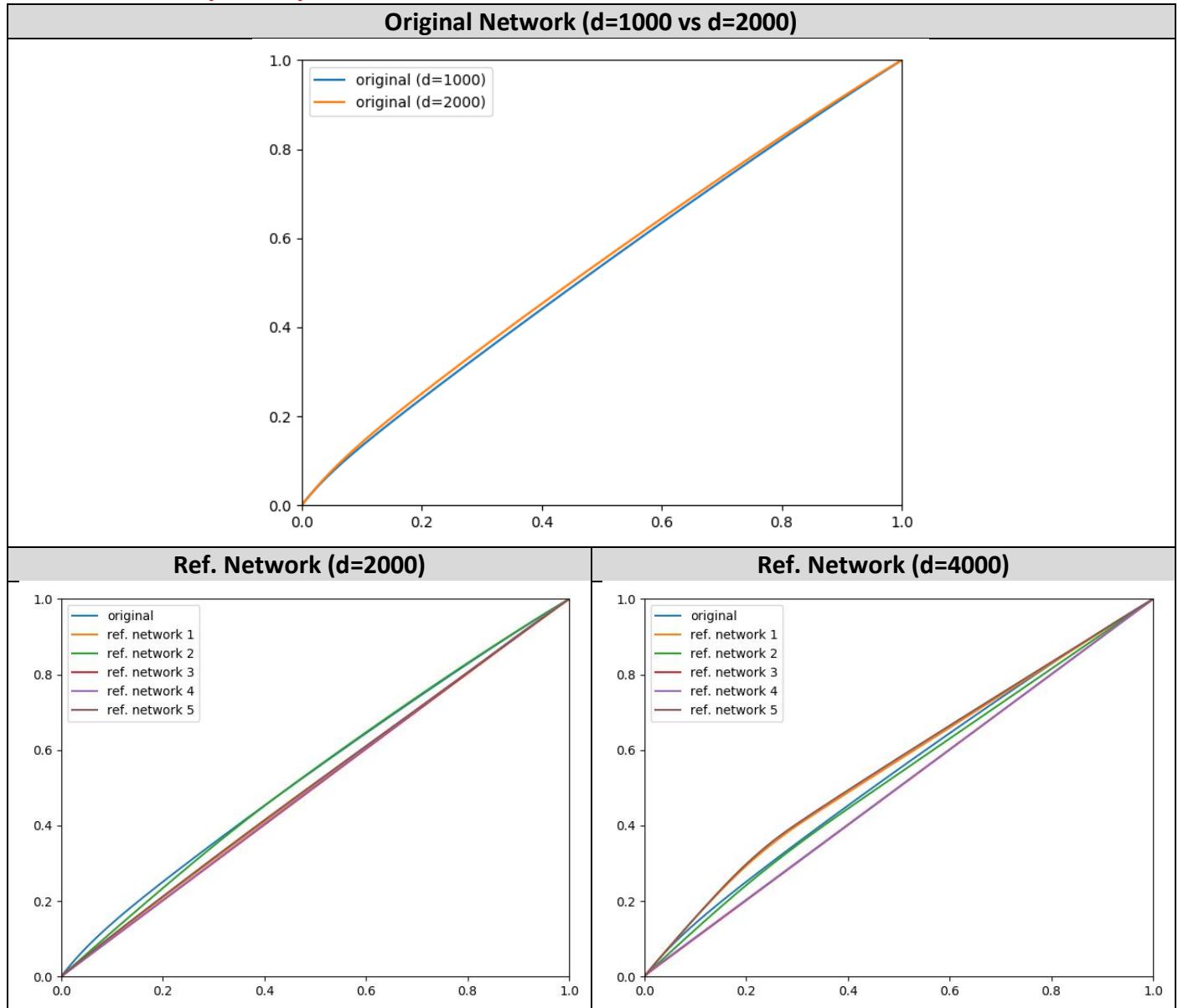
Peak detection method



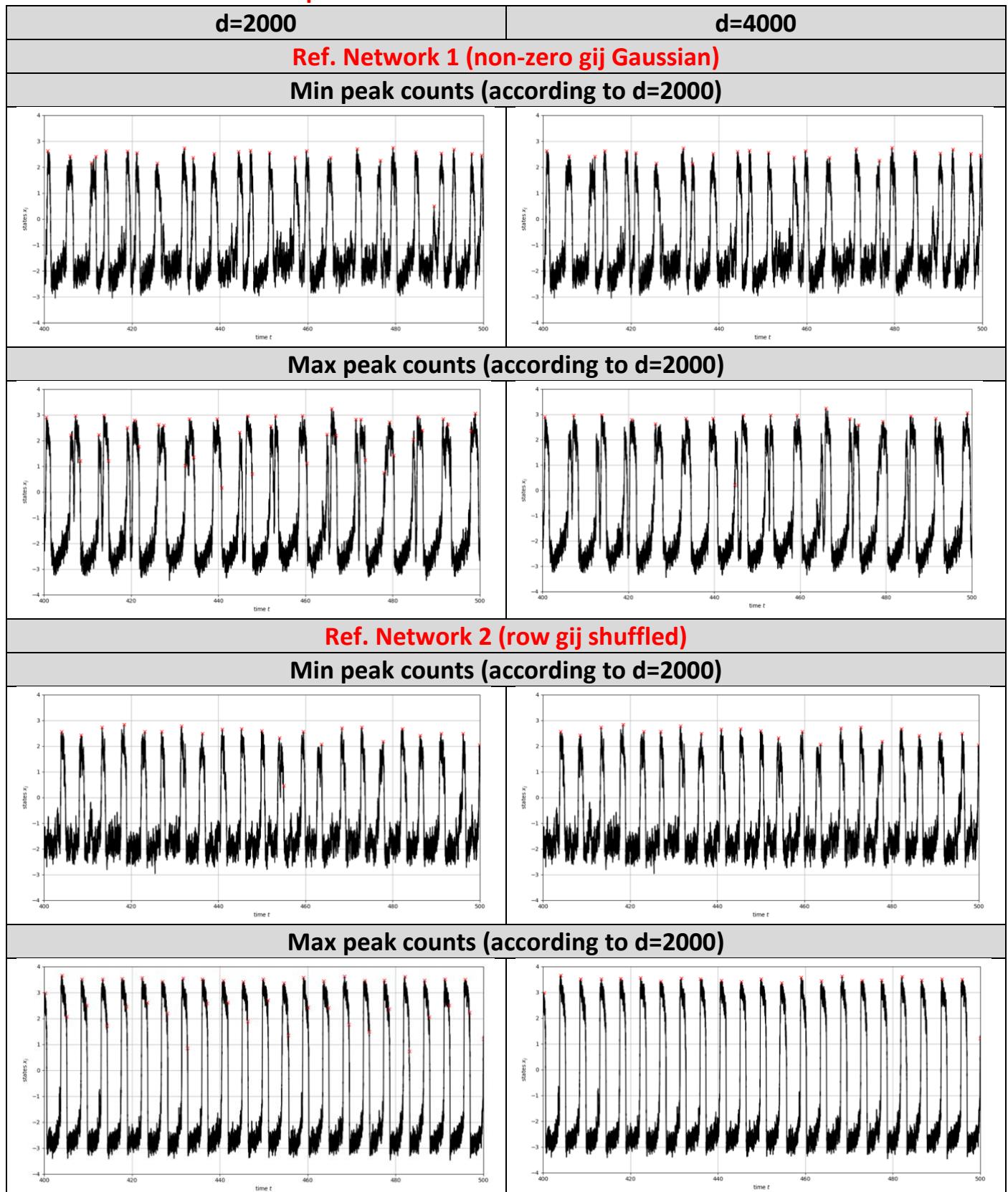
Dominance plot comparison



Effect of min sep d on peak counts

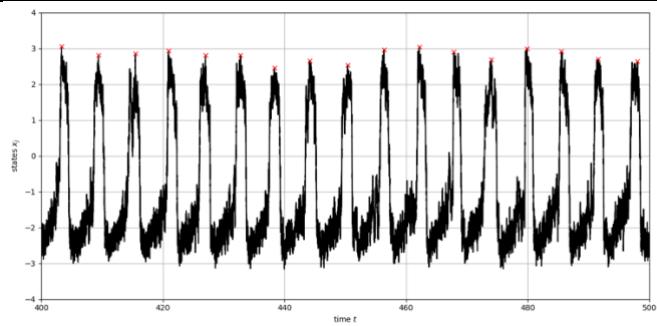
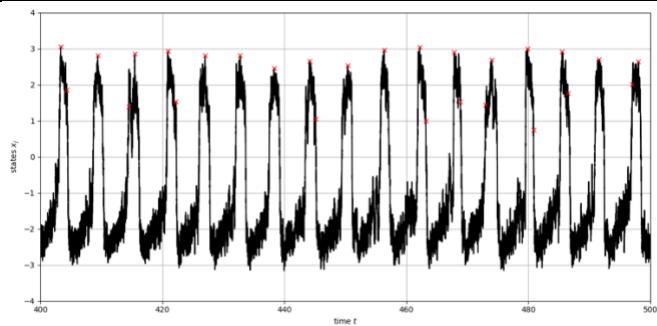


Refence network detected peaks

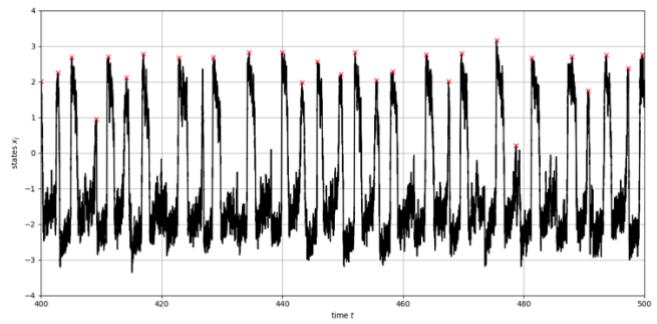
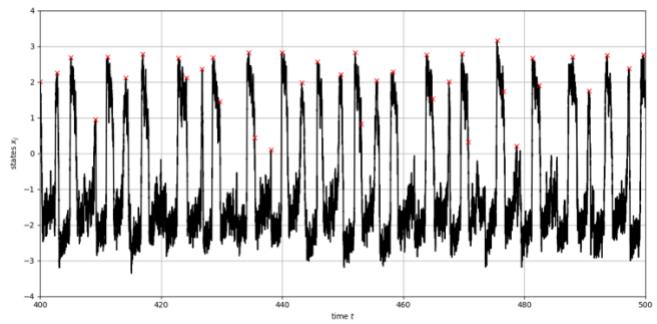


Ref. Network 3 (col g_{ij} shuffled)

Min peak counts (according to $d=2000$)

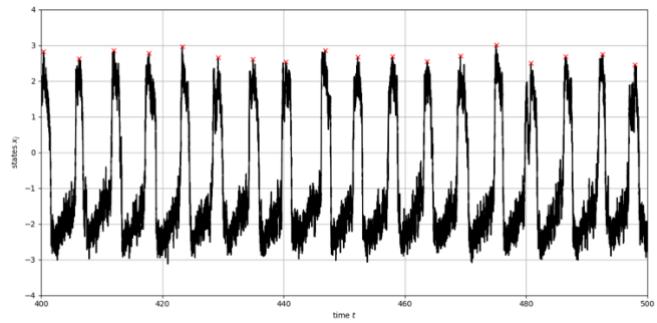
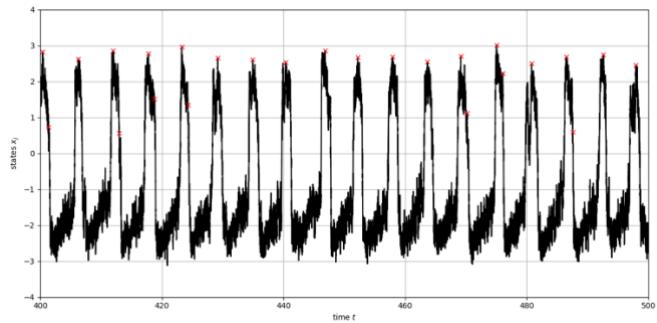


Max peak counts (according to $d=2000$)

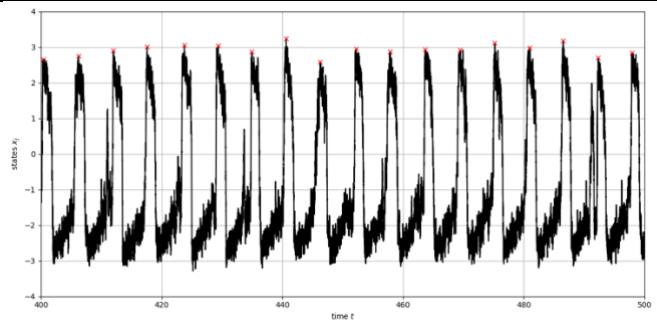
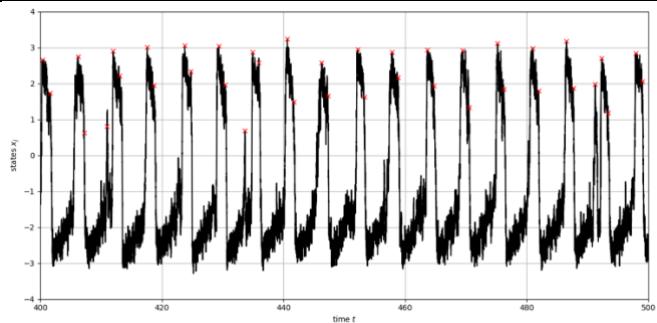


Ref. Network 4 (random)

Min peak counts (according to $d=2000$)

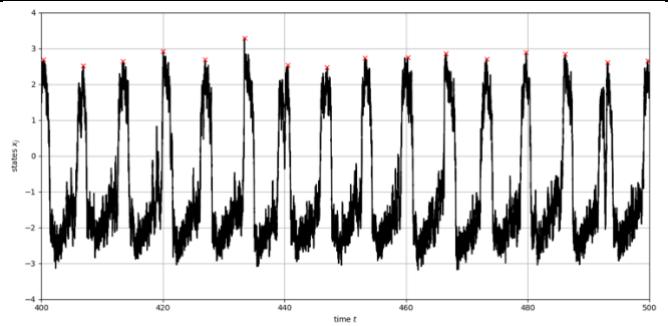
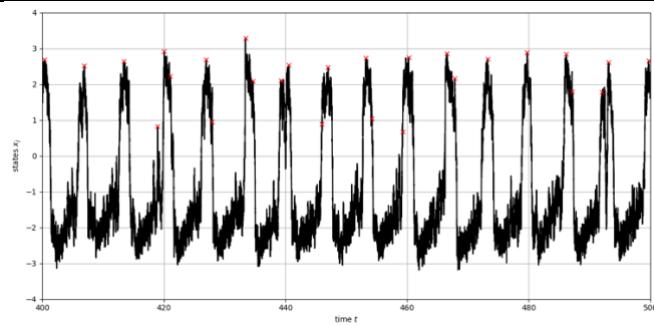


Max peak counts (according to $d=2000$)

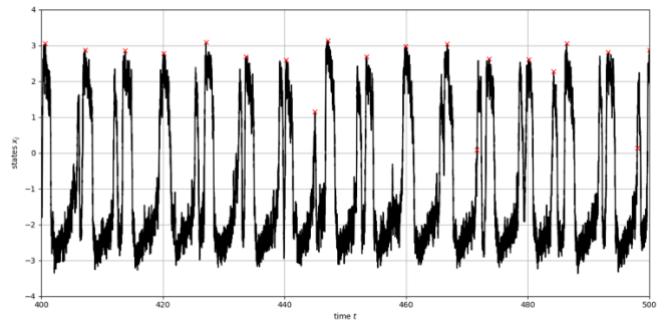
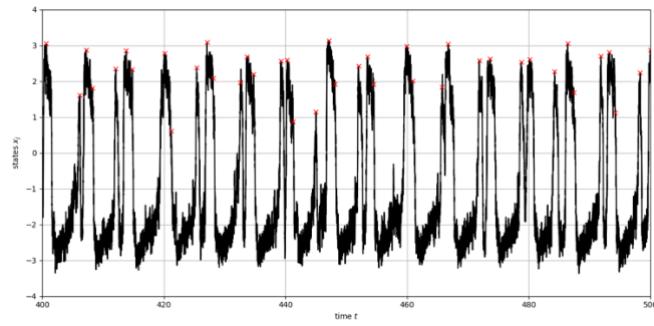


Ref. Network 5 (non-zero g_{ij} shuffled)

Min peak counts (according to $d=2000$)



Max peak counts (according to $d=2000$)

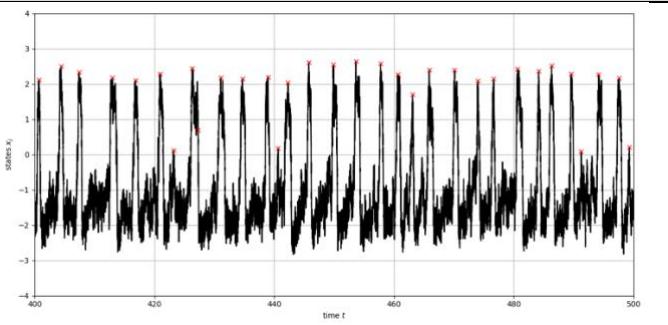
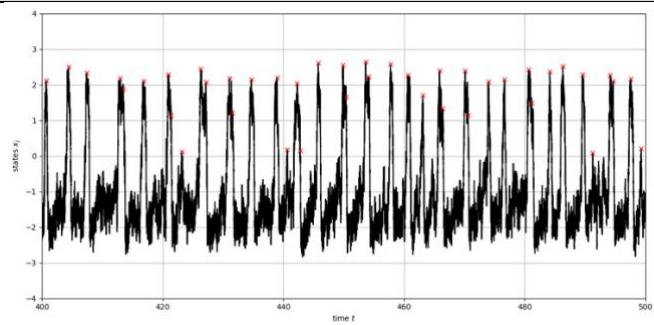


Original network detected peaks

$d=1000$

$d=2000$

Min peak counts (according to $d=1000$)



Max peak counts (according to $d=1000$)

