# Minimum Required Data Length to Reconstruct GC Network

November 10, 2013

#### 1 Task

Compute the minimum required data length when calculating GC. Use IF neural model as an example.

### 2 Analysis

Suppose there are two random variables *x* and *y*.

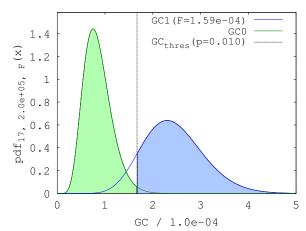
Recall: Distribution of GC obey

$$L \cdot \hat{F}_{x \to y} \stackrel{a}{\sim} \chi'^{2}(m, L \cdot F_{x \to y})$$

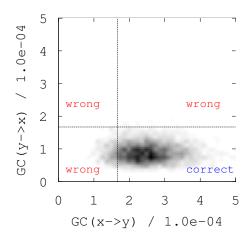
in the large L limit ( $\stackrel{a}{\sim}$ ). Where L is number of data samples,  $\hat{F}_{x\to y}$  is calculation value of true GC  $F_{x\to y}$ , m is fitting order.  $\chi'^2$  is Noncentral chi-squared distribution (see ref. [1] for definition and properties).

For convenience, we denote the probability density function (pdf) of  $\hat{F}$  as  $\rho_{m,L,F}(x)$ .

Now we want to know: for a given L, m and true  $F_{x\to y}$ ,  $F_{y\to x}$ , what's the expected correct rate? Obviously, 25% will be the lowest bound (random guess, without any other knowledge about the two neuron network).



(a) theoretical asymptotic pdf of  $\hat{F}_{x\to y}$  and  $\hat{F}_{y\to x}$  (separately).



(b) density of simultaneous distribution of  $\hat{F}_{x\to y}$  and  $\hat{F}_{y\to x}$ . Obtained from 4000 experiment and counting by divide each axis into 40 uniform bins.

Figure 1: GC pdf under one set of typical parameter:  $\mu = 1.0 \, \text{kHz}$ , f = 0.012, S = 0.01, using  $L = 2.0 \times 10^5 \, (T = 1 \times 10^5 \, \text{ms})$ , m = 17, the true GC is  $F_{x \to y} \approx 1.592 \times 10^{-4}$  and  $F_{y \to x} \approx 0.006 \times 10^{-4}$  (obtained by  $L = 1 \times 10^8$ ). The black line represent the GC thresholding value (GC<sub>thres</sub>) that we used to judge whether there is connection or not. Here GC<sub>thres</sub> satisfis  $P(\hat{F}_{y \to x} < \text{GC}_{\text{thres}}) = 0.01$  and  $F_{y \to x} = 0$  (our null hypothesis), i.e. false positive error rate is 1%.

In order to measure the correctness of GC analysis, here define the correct rate  $p_{\text{correct}}$  as following: in a two-variable GC test, the true GC  $F_{x\to y} > 0$ ,  $F_{y\to x} = 0$ , the GC test use a fixed p-value (false positive error rate, say 0.01) and fixed fitting order m, sample number L to determine a GC threshold  $GC_{\text{thres}}$ . Then compare calculated GC  $\hat{F}$  to  $GC_{\text{thres}}$  to guess whether there is connection or not. We denote the ratio between number of correct guesses (network is x-y) and number of total guess as  $p_{\text{correct}}$ .

If  $\hat{F}_{x\to y}$  and  $\hat{F}_{y\to x}$  are independent, then the expression for  $p_{\text{correct}}$  will be very simple:

$$p_{\text{correct}} = \int_0^{F_{thres}} \rho_{m,L,F0}(F) \, \mathrm{d}F \, \left( 1 - \int_0^{F_{thres}} \rho_{m,L,F1}(F) \, \mathrm{d}F \right), \tag{1}$$

that is the product of areas of green and blue region in Fig.(1a). Otherwise, we have to count the volumn of lower right part of Fig.(1b).

## **2.1** Is $\hat{F}_{x\to y}$ and $\hat{F}_{y\to x}$ are independent?

Geweke said (Ref.[2])  $\hat{F}_{x\to y}$  and  $\hat{F}_{y\to x}$  are asymptotically independent. But how "asymptotically". First, is the asymptotic pdf of  $\rho_{m,L,F}(x)$  accurate?

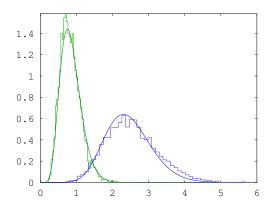
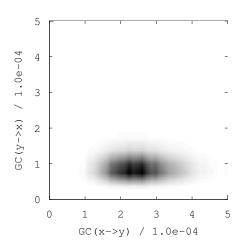


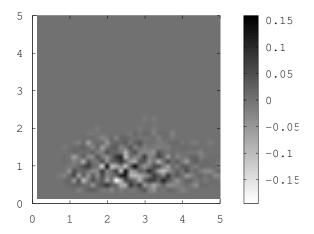
Figure 2: Comparison of statistic data and asymptotic pdf of  $\rho_{m,L,F}(x)$ . Same parameter as Fig.(1).

Second, by the 4000 GC data point mensioned in Fig.(1b), we can calculate the correlation of  $\hat{F}_{x\to y}$  and  $\hat{F}_{y\to x}$ . The result is -0.012, which can be explaind by statistic error  $(1/\sqrt{4000}\approx 0.016)$ .

Further, we compare the joint distribution to the product of marginal distribution.



(a) product of marginal distribution, looks similar to Fig.(1b).



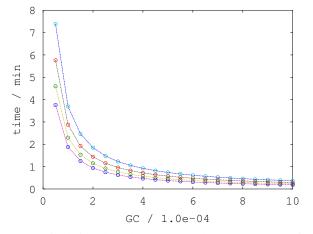
(b) pdf of  $(\hat{F}_{x\to y}, \hat{F}_{y\to x})$  subtract product of marginal distribution, normalized to distribution peak equals one.

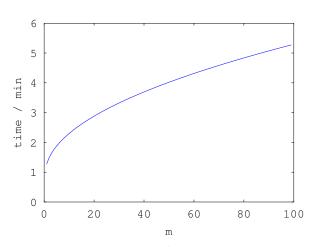
Figure 3

In this parameter (See description of Fig.(1)), the 4000 experiements tell us the correct rate is about 87.1%, while using asymptotic pdf and assume  $\hat{F}_{x\to y}$  and  $\hat{F}_{y\to x}$  are independent, we get 88.9% from Eq.(1) through m, L,  $F_{x\to y}$ ,  $F_{y\to x}=0$  in Fig.(1). Several more experiments are done, they're all matched good enough(< 5%).

#### 2.2 Minimum required data length

From Eq.(1) it's now possible to solve the minimum required data length  $L_{\min}$  or data time length  $T_{\min}$  (instead of doing a lot of numerical experiements), if m and  $F_{x\to y}$  are known.





(a) Required time length v.s. GC value. Four curves from up to down corresponding to  $m = \{5, 10, 20, 40\}$ , circle dot is obtained by solving Eq.(1), dashed line is obtained from Eq.(2)(see below).

(b) Required time length v.s. fitting order. Fix GC value to  $F_{\rm true} = 1.0 \times 10^{-4}$ 

Figure 4: False positive error rate set to 0.01, required correct ratio set to 90%.

In the case of false positive error rate set to 0.01, required correct ratio set to 90%, there is a good approximation of minimum length (relative error of  $T_{min}$  is about 0.1%):

$$T_{\min} \approx \frac{10.00}{F_{\text{true}}} \left( 1.153 + \frac{\sqrt{m - 0.513}}{1.917} \right) \Delta t, \ (m \in \{1, 2, \dots, 100\})$$
 (2)

where  $1/\Delta t$  is sample rate,  $\Delta t = 0.5$  ms in all above cases.

Recall that  $F_{\text{true}}/\Delta t \to \text{const.}$  in the limit  $\Delta t \to 0$ , so Eq.(2) tell us that there is no benefit to use small  $\Delta t$ , because in that case  $m \propto 1/\Delta t$  which make  $T_{\text{min}}$  larger.

Bigger positive error rate will lead to smaller  $T_{\min}$ , but effect is limited. e.g. set positive error rate to 0.05 will decrease  $T_{\min}$  about 12%, while set positive error rate to 0.005 will increase  $T_{\min}$  about 7%.

Now remaining problem is: what is the true GC and corresponding m. We once have done that:

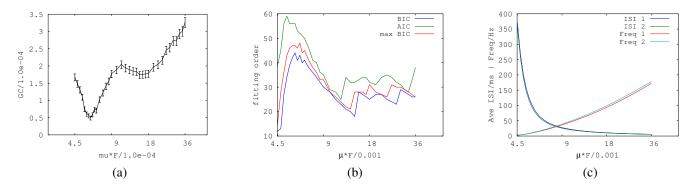


Figure 5: Scan F.  $\mu = 0.01$ , S = 0.01,  $T = 1.0 \times 10^4 \text{ sec}$ ,  $\Delta t = 0.5 \text{ ms}$ 

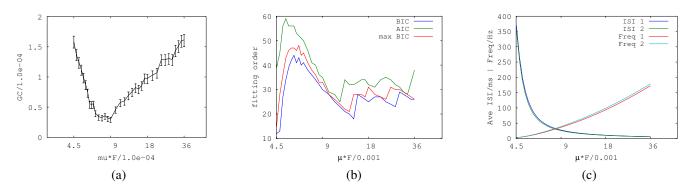


Figure 6: Scan F.  $\mu = 0.02$ , S = 0.01,  $T = 1.0 \times 10^4 \text{ sec}$ ,  $\Delta t = 0.5 \text{ ms}$ 

Importing these data to Eq.(2) we get:

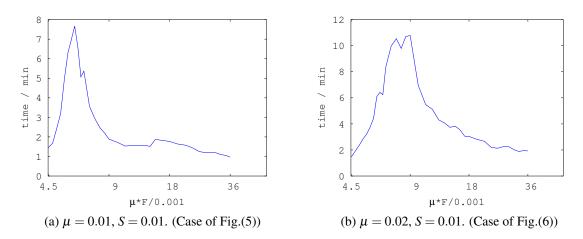


Figure 7: Required time length

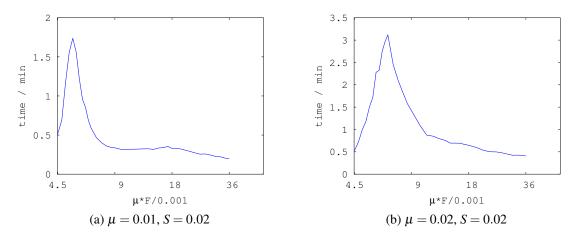


Figure 8: Required time length

As what we expected, twice the S, roughly quarter the required time.

#### Case of using spike train:

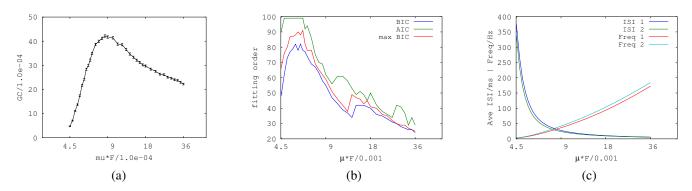


Figure 9: Using spike train. Scan F.  $\mu = 0.01$ , S = 0.02,  $T = 1.0 \times 10^4$  sec,  $\Delta t = 0.5$  ms

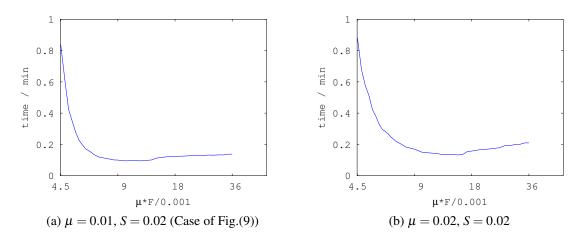


Figure 10: Required time length. (Use spike train)

Note:  $EPSP \approx 107 \, \text{SmV}$  or  $EPSP \approx 7 \, \text{S}$  in the normalized unit.

# 3 Big ISI Case

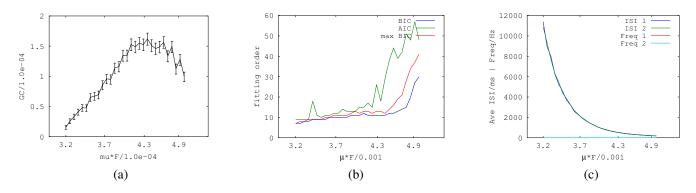


Figure 11: Scan F.  $\mu = 0.01$ , S = 0.01,  $T = 1.0 \times 10^4$  sec, continue Fig.(5)

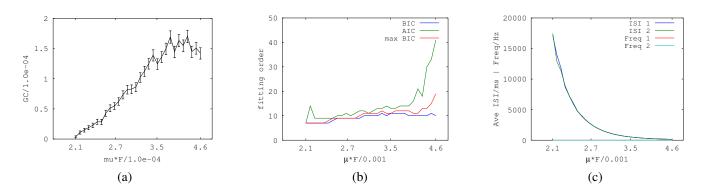


Figure 12: Scan *F*.  $\mu = 0.02$ , S = 0.01,  $T = 1.0 \times 10^4$  sec, continue Fig.(6)

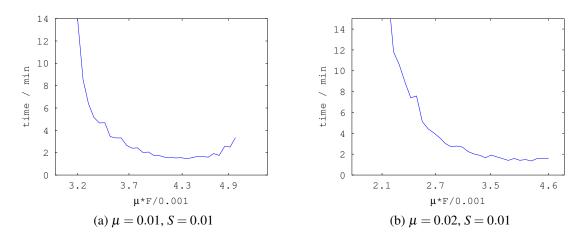


Figure 13: Required time length

## References

[1] http://en.wikipedia.org/wiki/Non-central\_chi-square\_distribution

[2]	Measurement of Linear Dependence and Feedback Between Multiple Time Series, John Geweke, Journal of the American Statistical Association, Vol.77, No.378 (1982)