

# Deriving Properties of Binary Black Hole from UV + Optical Spectrum with MCMC Simulations

Liu Qing | USTC

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# Finding BBH candidates : *Pros & Cons*

## Variability

vs

## Spectrum

Pros:

1. Simple and popular
2. Exist CRT data

Cons:

1. Time-consuming and painstaking
2. Fraudulent results from other drives (eg: RDW) with intended fitting
2. Some BBHs may not have variabilities

Pros:

1. Novel and intriguing
2. Less elusive and time-saving
3. Avoid complex discussion on the origin of variability

Cons:

1. Not ubiquitous features
2. Probably need further inspection to break degeneracies

— — Spectrum is valuable to investigate in depth.

## What I want to do...

- Conduct an MCMC Fitting like Yan (2015);

## What I need...

- Background on BBHs;
- Knowledge about Stochastic Process;
- Basic Bayesian Inference;
- Techniques about fitting process;
- Methods of model construction;
- Practical operative programming skills;

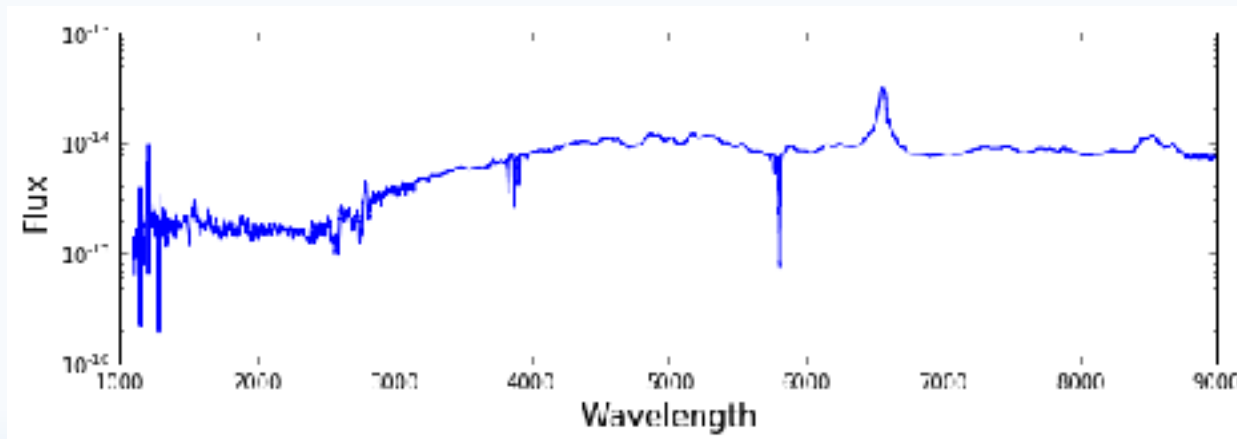


The first week



The second week

➤ Access to Mrk231 spectrum data (kindly provided by Mrs. Yan)



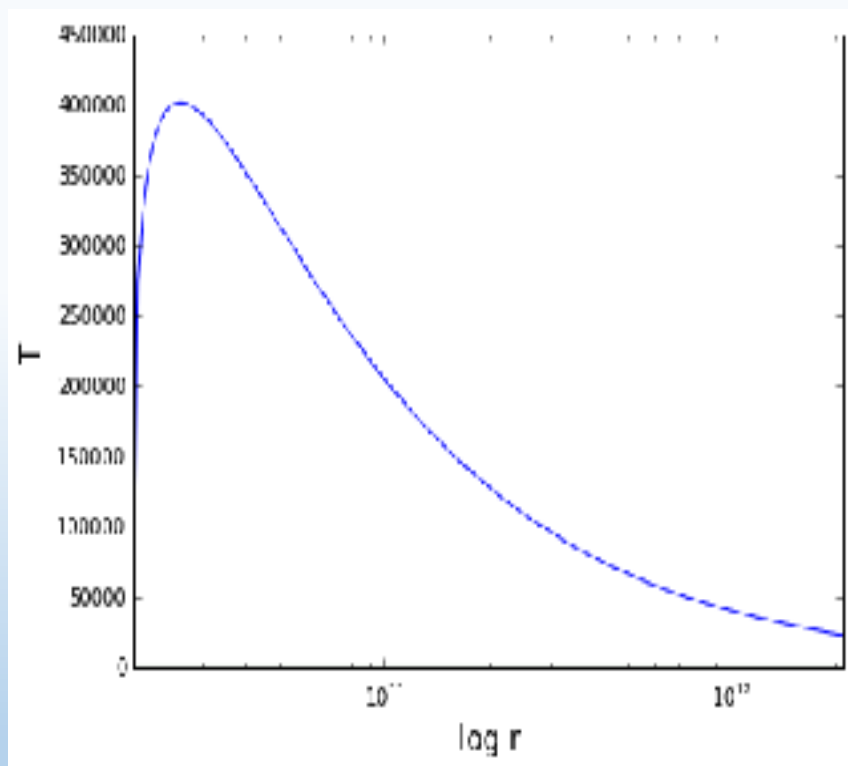
Flat in Optical and FUV  
with a clear drop off  
between 2500-4000 Å

➤ Construct disk SED model

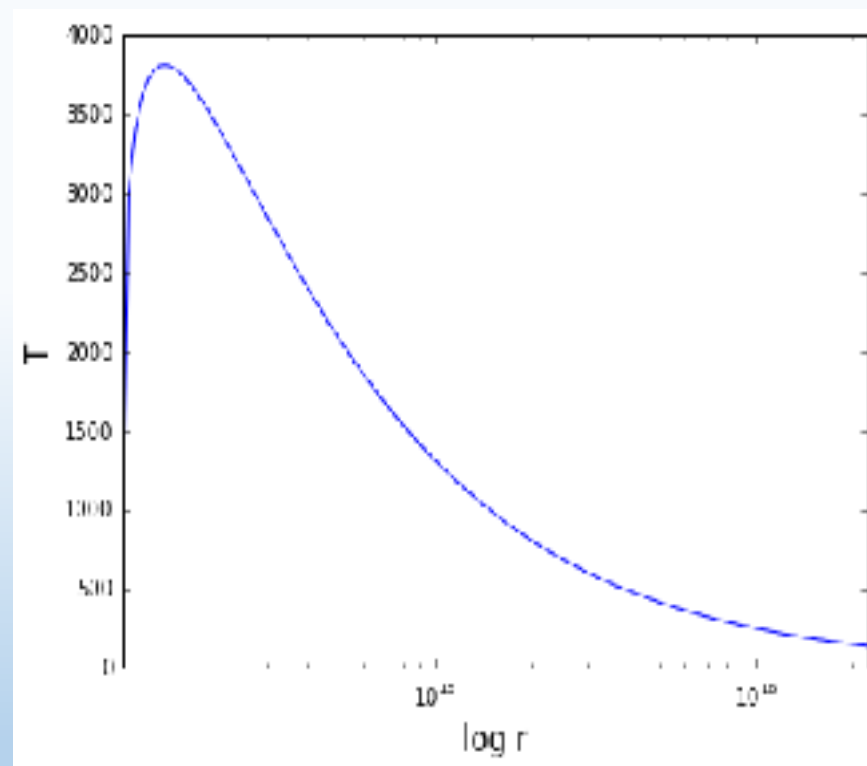
$$T_{\text{eff}}(r) = \left[ \frac{3GM_{\bullet}\dot{M}_{\text{acc}}}{8\pi\sigma_B r^3} \left( 1 - \sqrt{\frac{r_{\text{in}}}{r}} \right) \right]^{1/4}, \quad F_{\lambda} = \int_{r_{\text{in}}}^{r_{\text{out}}} \frac{2hc^2 \cos i / \lambda^5}{\exp[hc / \lambda k_B T_{\text{eff}}(r)] - 1} dr,$$

As things does not change a lot in the case of circumbinary disk model, plus a rather complex issue about viscosity, I take Mrs. Yan's suggestion to use double standard thin disk model to construct the SED.

- Effective temperature distribution of disks

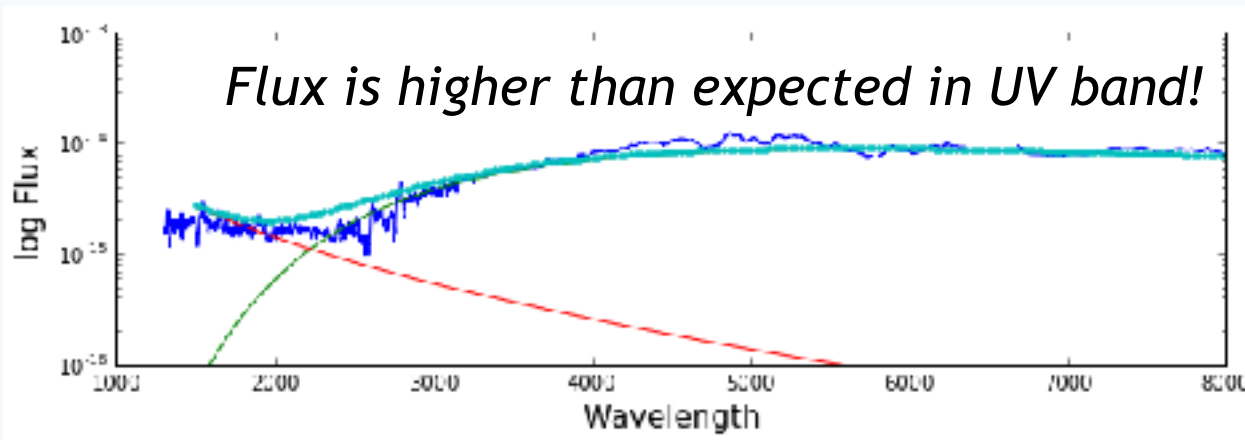


↑ Outer Circumbinary-disk

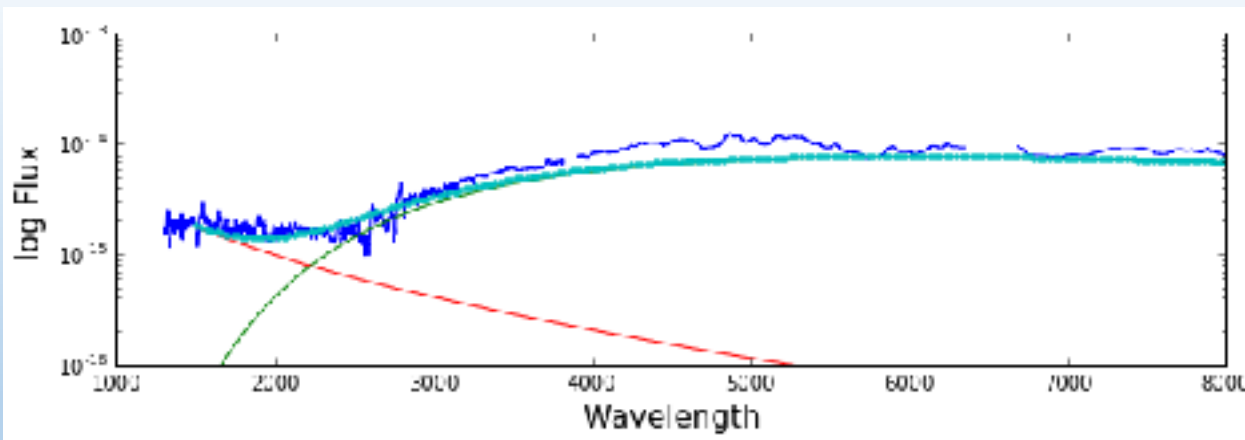


↑ Inner Mini-disk

- Problems during model construction



(Before extinction)



(After extinction)

Conclusion: Extinction matters a lot.

Here I use Calzetti(2000) and mask prominent absorption & emission lines.

# ➤ MCMC Fitting ( with Python )

( † All Python codes are available on Github: [github.com/NGC4676/LIU/tree/Python](https://github.com/NGC4676/LIU/tree/Python) )

## • M-H algorithm

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### Algorithm 6 Metropolis-Hastings 采样算法

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1: 初始化马氏链初始状态  $X_0 = x_0$

2: 对  $t = 0, 1, 2, \dots$ , 循环以下过程进行采样

- 第  $t$  个时刻马氏链状态为  $X_t = x_t$ , 采样  $y \sim q(x|x_t)$
  - 从均匀分布采样  $u \sim Uniform[0, 1]$
  - 如果  $u < \alpha(x_t, y) = \min \left\{ \frac{p(y)q(x_t|y)}{p(x_t)p(y|x_t)}, 1 \right\}$  则接受转移  $x_t \rightarrow y$ ,  
即  $X_{t+1} = y$
  - 否则不接受转移, 即  $X_{t+1} = x_t$
- 

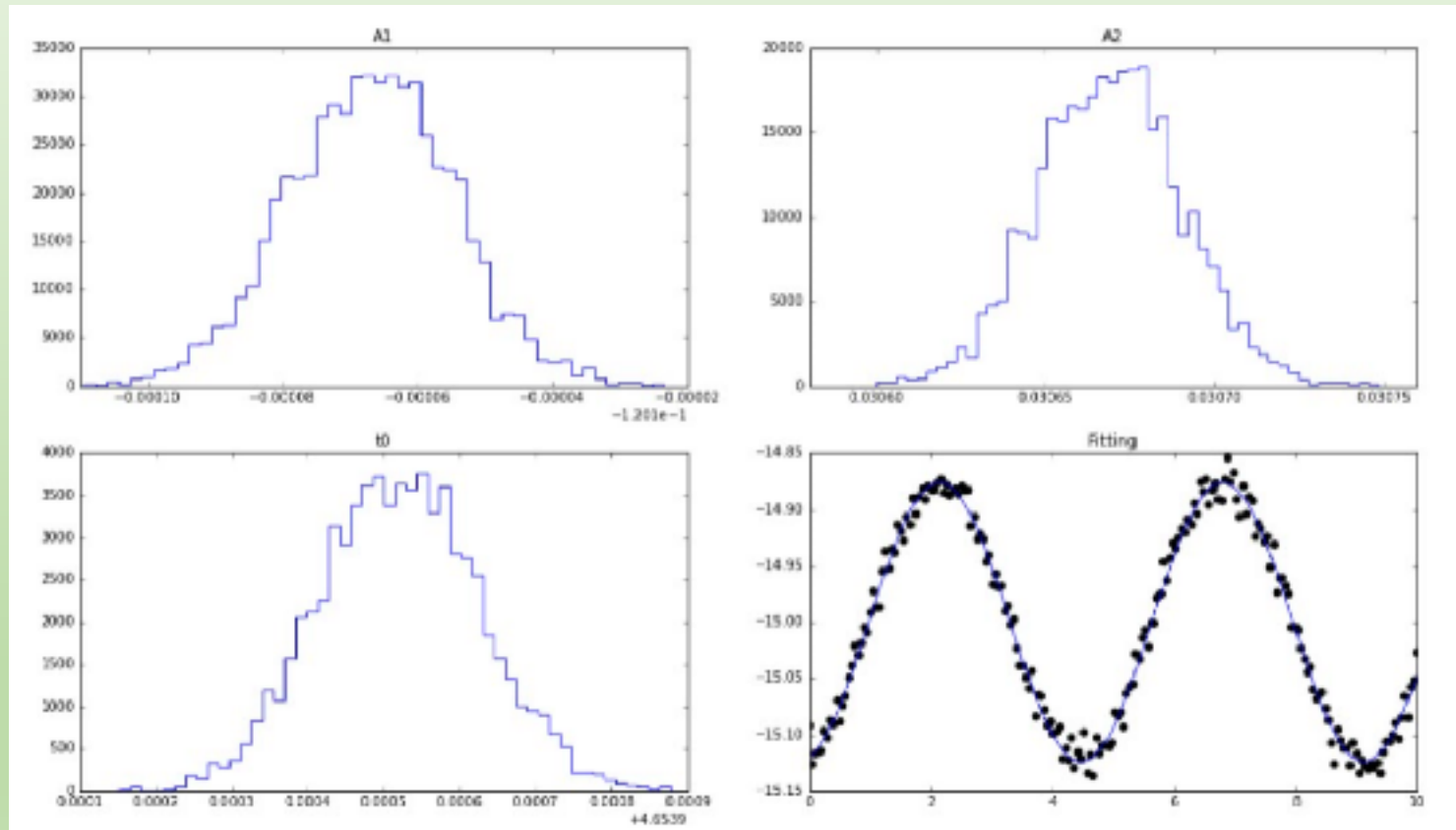
Avoid EXHAUSTIVE Grid method!

— — especially efficient with many parameters

- A petit try on a simple case

3 Parameters ( $A1$ ,  $A2$ ,  $t0$ )

50000 points; Burn-in = 10000; Thinned by 10



Succeed to fit the curve correctly!



# • Practical Fitting on Mrk-231 Spectrum

## 8 Parameters with **uniform Priors**

† Total Mass ( $M_{\bullet}$ );

† Mass Ratio ( $q$ );

† Semi-major Axis ( $a$  BBH);

† The Eddington Ratio of the inner mini-disk ( $f_{\text{Edd},c}$ );

† The Eddington Ratios of the outer circumbinary disk ( $f_{\text{Edd},s}$ );

† Ratio of the outer boundary of the inner disk to the mean Roche Radius ( $f_{r,s}$ );

† Scale Factor ( $A$ );

† Extinction  $E(B-V)$ ;

```
model dict(a_UU1 a_UU1, q q, M_t M_t, M_s M_s,  
           r_in r_in, r_out r_out,  
           r_s_out r_s_out, r_s_in r_s_in,  
           f_s_Ldd f_s_Ldd, f_c_Ldd f_c_Ldd,  
           M_r_acc M_r_acc, M_s_acc M_s_acc,  
           f_lambda f_lambda, r_lambda r_lambda,  
           f_RS=f_RS, E_DV = E_DV, Extinction = Extinction,  
           A = A, S = S, T = T)
```

Easy to  
Check

Easy to  
Track

# Bottom-Top

— Add parameters one by one

- **Problems during MCMC Fitting**

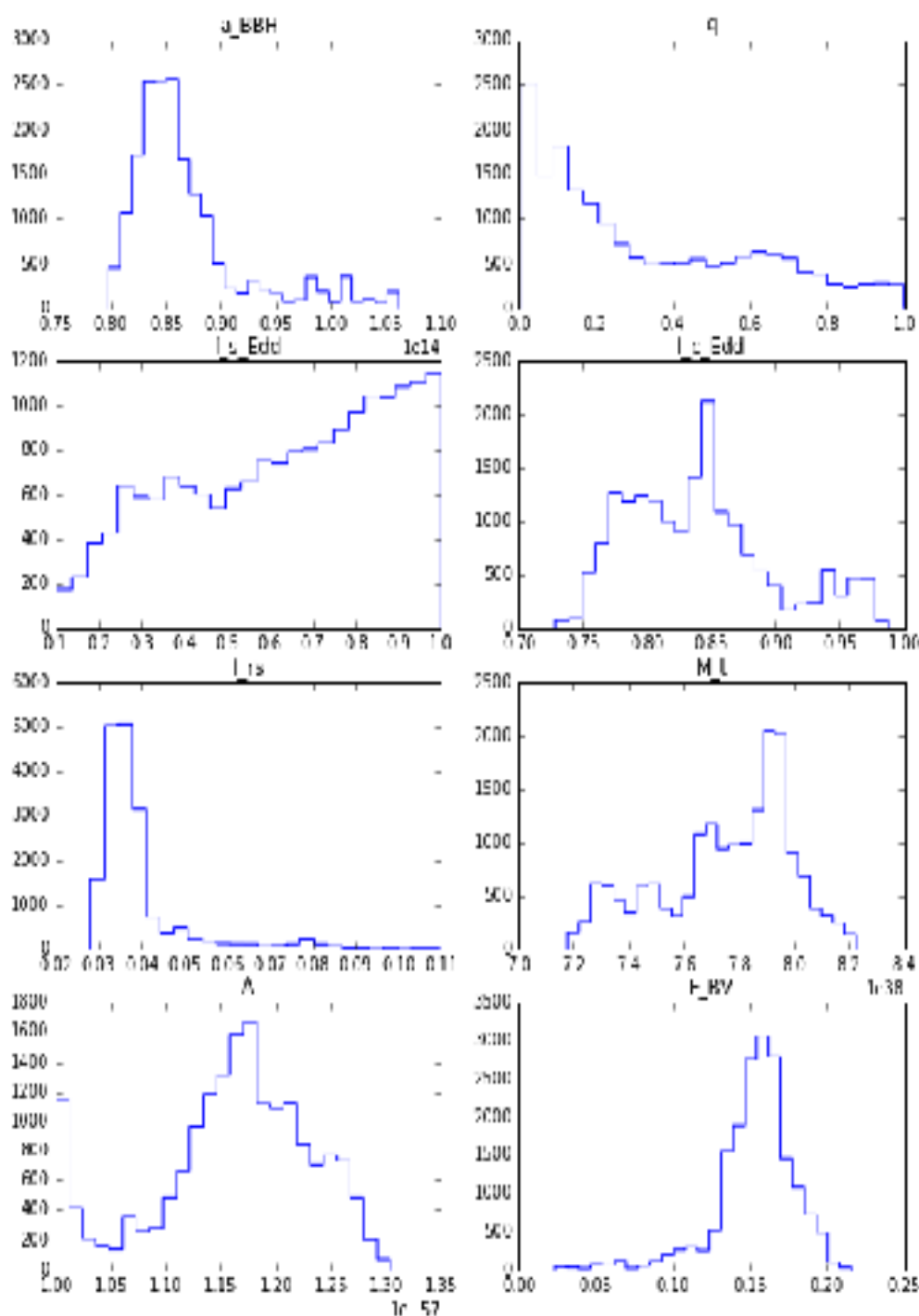
→ Quad method cannot pass a whole list to the integration.

(temperature distribution along radius)

→ Simpson method lacks accuracy and is sensitive to steps.

## ✦ My Solution:

Use Simpson method to fit the spectrum,  
and Quad method to plot the result, sacrificing  
a little accuracy to suffice the speed.



**Run:**

100000 points

Burn-in = 10000

Thinned by 10

5405.7 seconds

**Output:**

Marginalized probability  
distribution function ( pdf ,  
left panel) for 8 Params

## ➤ Limitations & Issues Deserve Attention

— — Do fitting parameters surely be true parameters?

1 Fitting results are sensitive to several non-physical factors:

- p Pick of Priors
- p Precision of fitting
- p Times of iteration
- p Resolution of the input

2 What if I found a multi-modal posterior?

3 How do I know that MCMC didn't get stuck in a local minimum?

4 How do I know that a chain has converged?

**Always Be Careful When Using MCMC !**

## ➤ Potential Improvements

### ◆ A diagnosis for chain convergence

→ Save running time

### ◆ Criteria to tell whether one fitting is better than another

→ Give more convincing result

### ◆ Adapt the fitting for other models

→ Find more details about the source

### ◆ See if the method can be applied in other facets

→ Look for a wider prospect

## ➤ 一些想法和感悟

Given an unfamiliar topic , how to establish the framework in a short time?

新接触一个领域、一项课题，怎么在最短的时间内迅速建立起一个框架，找到自己能做的事情，想做的事情，BrainStorm各种idea ( 这一过程是极为愉悦的 )，构造自己的文献网络，寻找需要的数据，然后有导向地获取需要的知识（各种document）

—— 这是我此行最大的收获。

平时往往没有足够的时间花一整天功夫去想、去写、去改、去挖掘 —— 重要的是静下心。

遇到困难不要沮丧，科研问题总是举步维艰的 —— 坚持就有突破。

不要先想着依赖他人，你在一个问题上绊脚并爬起，将来你就是这个问题的专家。

At Last , Have Fun with Astronomy and Scientific Research !

Finally, appreciating for Prof. Lu 's kindly guidance and Mrs. Yan 's generous help !

( I enjoy those discussions with you very much !  
)

