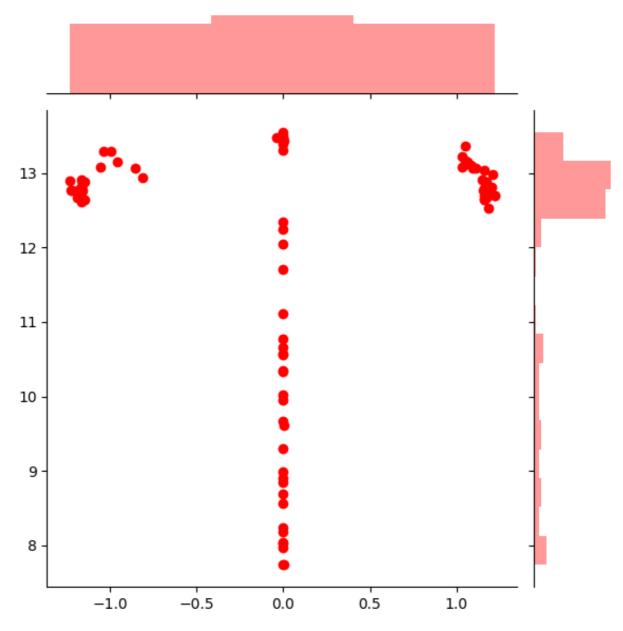
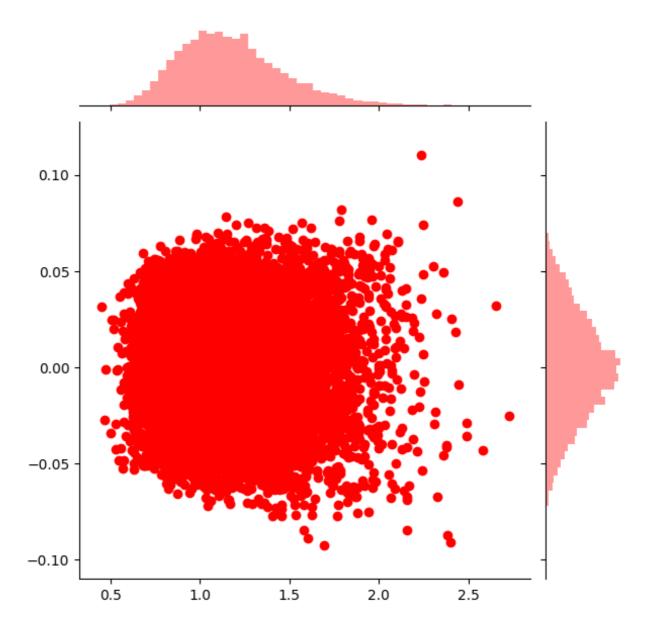
## **Resample in 3-layer NN**

- · network architecture:
- $x \in \mathbb{R}$ ,  $h_1 \in \mathbb{R}^{n_1}$  and  $h_2 \in \mathbb{R}^{n_2}$ :  $n_1 = 100$ ,  $n_2 = 10000$ ,
- assume  $\lambda_3 >> \lambda_2 >> \lambda_1$
- step1: could learn a concentrated distribution of u and  $\theta_2$ 
  - continuous version:  $\theta_1 \in \mathbb{R}$ ,  $\theta_2 \in L^2(\mathbb{R})$ 
    - define  $\omega_1(\theta_1) = \int \theta_2(\theta_1) \mu(d\theta_2)$ , we should have  $\mathrm{Var}_{\mu(\theta_1)}(\omega_1) \approx 0$
    - define  $\omega_2(\theta_2) = \int u\mu(d\theta_2, du)$ , we should have  $\text{Var}_{\mu(\theta_2)}(|\omega_2|) \approx 0$
  - $\quad \text{o discrete version: } W_1 \in \mathbb{R}^{n_1 \times 1} \text{, } W_2 \in \mathbb{R}^{n_2 \times n_1} \text{, } W_3 \in \mathbb{R}^{1 \times n_2}$ 
    - we could witness that the variances of the norm of different columns of  $W_2$  and  $W_1$  are near zero.
  - experiment results:
    - [1] the variance of  $\omega_1(\theta_1)$  increases (when  $\theta_1 \approx 0$ ,  $\omega_1(\theta_1)$  varies: x is  $theta_1$ , y is  $\omega_1(\theta_1)$ )



• [2]  $\omega_2(\theta_2)$  doesn't concentrate (x is  $\|\theta_2\|_2$ , y is  $\omega_2(\theta_2)$ ):



## o other results:

- I found the problem of [2] is originated from the facts that the batch norm in layer 2 is not effective.
  - std of each neuron in layer 1 is ranged from [0.64, 1.44]: 1.03 +/- 0.09
  - std of each neuron in layer 2 is ranged from [0.067, 3.919]: 1.04 +/- 0.32
  - when I conduct experiments in 2-layer nn, it is about 1.0 +/- 0.01