

light\_data\_3.10/result/3.13:

(1) /rand\_bias0.3: 采样率 10M, 接收速率 60M, 均匀分布, 偏置电流 0.3A。

1. /mix\_amp: 混合幅度数据作为训练数据, 且数据归一化。发送信号是均匀分布的随机信号, 采样率为 10M, 接收速率 150M, 偏置电流 0.3A。

1.1 /Twononlinear1:

相关符号数为 3, 即  $h\_order=3*rate\_times$ 。L=2, U=25

```
Twononlinear ,
ini learningRate = 1.000000e-02 ,
min batch size = 400 ,
DropPeriod = 5 ,
DropFactor = 0.100000 ,
amp begin = 2 , amp end = 26 , amp step = 1
data_num = 100
validationFrequency is floor(numel(xTrain)/miniBatchSize/4)
H order = 18
Hidden Units = 25
```

1.2 /Twononlinear2:

相关符号数为 5, 即  $h\_order=5*rate\_times$ 。L=2, U=25

```
Twononlinear ,
ini learningRate = 1.000000e-02 ,
min batch size = 400 ,
DropPeriod = 5 ,
DropFactor = 0.100000 ,
amp begin = 2 , amp end = 26 , amp step = 1
data_num = 100
validationFrequency is floor(numel(xTrain)/miniBatchSize/4)
H order = 30
Hidden Units = 25
```

1.3 /Twononlinear3:

相关符号数为 8, 即  $h\_order=8*rate\_times$ 。L=2, U=25

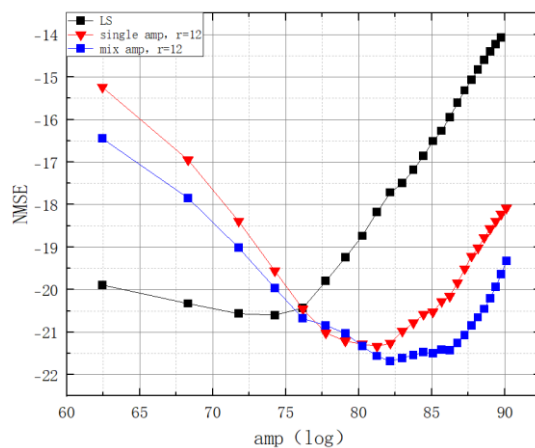
```
Twononlinear ,
ini learningRate = 1.000000e-02 ,
min batch size = 400 ,
DropPeriod = 5 ,
DropFactor = 0.100000 ,
amp begin = 2 , amp end = 26 , amp step = 1
data_num = 100
validationFrequency is floor(numel(xTrain)/miniBatchSize/4)
H order = 48
Hidden Units = 25
```

#### 1.4 /Twononlinear4:

相关符号数为 12, 即  $h\_order=12*rate\_times$ 。L=2, U=25

```
Twononlinear ,
ini learningRate = 1.000000e-02 ,
min batch size = 400 ,
DropPeriod = 5 ,
DropFactor = 0.100000 ,
amp begin = 2 , amp end = 26 , amp step = 1
data_num = 100
validationFrequency is floor(numel(xTrain)/miniBatchSize/4)
H order = 72
Hidden Units = 25
```

#### 1.5 /Twononlinear5:



基于此图, 决定先训练线性的部分, 看能不能把线性部分的性能训练到和 LS 持平。即  $amp\_begin=2$ ,  $amp\_end=6$ 。相关符号数为 12, 即  $h\_order=12*rate\_times$ 。L=2, U=25

```
Twononlinear ,
ini learningRate = 1.000000e-02 ,
min batch size = 400 ,
DropPeriod = 5 ,
DropFactor = 0.100000 ,
amp begin = 2 , amp end = 6 , amp step = 1
data_num = 100
validationFrequency is floor(numel(xTrain)/miniBatchSize/4)
H order = 72
Hidden Units = 25
```

#### 1.6 /Onenonlinear1:

基于上图, 决定先训练线性的部分, 看能不能把线性部分的性能训练到和 LS 持平。即  $amp\_begin=2$ ,  $amp\_end=6$ 。相关符号数为 12, 即  $h\_order=12*rate\_times$ 。L=1, U=25

```
Onenonlinear ,
ini learningRate = 1.000000e-02 ,
min batch size = 400 ,
DropPeriod = 5 ,
DropFactor = 0.100000 ,
amp begin = 2 , amp end = 6 , amp step = 1
data_num = 100
validationFrequency is floor(numel(xTrain)/miniBatchSize/4)
H order = 72
Hidden Units = 25
```

### 1.7 /Linear1:

基于上图，决定先训练线性的部分，看能不能把线性部分的性能训练到和 LS 持平。即 [amp\\_begin=2, amp\\_end=6](#)。相关符号数为 12，即  $h\_order=12*rate\_times$ 。一层线性隐藏层，隐藏层点数=25

```
linear ,
ini learningRate = 1.000000e-02 ,
min batch size = 400 ,
DropPeriod = 5 ,
DropFactor = 0.100000 ,
amp begin = 2 , amp end = 6 , amp step = 1
data_num = 100
validationFrequency is floor(numel(xTrain)/miniBatchSize/4)
H order = 72
Hidden Units = 25
```

### 1.8 /Linear2:

基于上图，决定先训练线性的部分，看能不能把线性部分的性能训练到和 LS 持平。即 [amp\\_begin=2, amp\\_end=6](#)。相关符号数为 12，即  $h\_order=12*rate\_times$ 。一层线性隐藏层，隐藏层点数=40

```
linear ,
ini learningRate = 1.000000e-02 ,
min batch size = 400 ,
DropPeriod = 5 ,
DropFactor = 0.100000 ,
amp begin = 2 , amp end = 6 , amp step = 1
data_num = 100
validationFrequency is floor(numel(xTrain)/miniBatchSize/4)
H order = 72
Hidden Units = 40
```

### 1.9 /Linear3:

基于上图，决定先训练线性的部分，看能不能把线性部分的性能训练到和 LS 持平。即 [amp\\_begin=2, amp\\_end=6](#)。相关符号数为 12，即  $h\_order=12*rate\_times$ 。一层线性隐藏层，隐藏层点数=60

```
linear ,
ini learningRate = 1.000000e-02 ,
min batch size = 400 ,
DropPeriod = 5 ,
DropFactor = 0.100000 ,
amp begin = 2 , amp end = 6 , amp step = 1
data_num = 100
validationFrequency is floor(numel(xTrain)/miniBatchSize/4)
H order = 72
Hidden Units = 60
```

#### 1.10 /Linear4:

基于上图，决定先训练线性的部分，看能不能把线性部分的性能训练到和 LS 持平。即 `amp_begin=2, amp_end=6`。相关符号数为 12，即 `h_order=12*rate_times`。两层线性隐藏层，隐藏层点数=40

```
linear ,
ini learningRate = 1.000000e-02 ,
min batch size = 400 ,
DropPeriod = 5 ,
DropFactor = 0.100000 ,
amp begin = 2 , amp end = 6 , amp step = 1
data_num = 100
validationFrequency is floor(numel(xTrain)/miniBatchSize/4)
H order = 72
Hidden Units = 40
```

#### 1.11 /Linear5:

基于上图，决定先训练线性的部分，看能不能把线性部分的性能训练到和 LS 持平。即 `amp_begin=2, amp_end=6`。相关符号数为 12，即 `h_order=12*rate_times`。两层线性隐藏层，隐藏层点数=60

```
linear ,
ini learningRate = 1.000000e-02 ,
min batch size = 400 ,
DropPeriod = 5 ,
DropFactor = 0.100000 ,
amp begin = 2 , amp end = 6 , amp step = 1
data_num = 100
validationFrequency is floor(numel(xTrain)/miniBatchSize/4)
H order = 72
Hidden Units = 60
```

light\_data\_3.11/result/3.13:

(1) /rand\_bias0.3: 采样率 10M, 接收速率 150M, 均匀分布, 偏置电流 0.3A。

1. /norm\_LS: 用 LS 算法, 求出各个幅度信号的 NMSE。

2. /single\_amp: 单一幅度数据作为训练数据, 且数据归一化。发送信号是均匀分布的随机信号, 采样率为 10M, 接收速率 150M, 偏置电流 0.3A。

2.1 /Threenonlinear1:

相关符号数为 5, 即  $h\_order=5*rate\_times$ 。L=3, U=60

```
Threenonlinear ,  
ini learningRate = 1.000000e-02 ,  
min batch size = 200 ,  
DropPeriod = 12 ,  
DropFactor = 0.100000 ,  
amp begin = 1 , amp end = 101 , amp step = 1  
data_num = 100  
validationFrequency is floor(size(xTrain{1},2)/miniBatchSize  
origin rate = 1.000000e+07 , receive rate = 1.500000e+08  
H order = 75  
Hidden Units = 60
```

3. /mix\_amp: 混合幅度数据作为训练数据, 且数据归一化。发送信号是均匀分布的随机信号, 采样率为 10M, 接收速率 150M, 偏置电流 0.3A。

3.1 /Threenonlinear1:

相关符号数为 5, 即  $h\_order=5*rate\_times$ 。L=3, U=60

```
Threenonlinear ,  
ini learningRate = 1.000000e-02 ,  
min batch size = 400 ,  
DropPeriod = 5 ,  
DropFactor = 0.100000 ,  
amp begin = 1 , amp end = 101 , amp step = 2  
data_num = 25  
validationFrequency is floor(numel(xTrain)/miniBatchSize/4)  
H order = 75  
Hidden Units = 60
```