

# ANN-Meeting

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Zulkaida

## Propose Method & Procedure

Step	Procedure
1	Established local-fit NN code with reasonable results
2	Established global-fit NN code with reasonable results
3	Established comparison with benchmark (for example least square method) & Truth values
4	Play around to win over benchmark
5	Established the generic-hyperparameter search tools
6	Established the good performance of NN over truth evaluation and validation/testing set
7	Finishing: Optimize the job-submission code

## The modification of the new functions?

1. Overall interference function
2. The jacobian
3. A factor of 2 in BHUU
4. Minus sign in tmin
5. Transverse vector-product definiton

```
// Defurne's Jacobian
```

```
jcob = 2. * PI ;
```

```
// Convert Unpolarized Coefficients to nano-barn and use Defurne's Jacobian
```

```
con_AUUBH = AUUBH * GeV2nb * jcob;
```

```
con_BUUBH = BUUBH * GeV2nb * jcob;
```

```
//Minimum t value
```

```
tmin = -( QQ * ( 1. - sqrt( 1. + gg ) + gg / 2. ) ) / ( x * ( 1. - sqrt( 1. + gg ) + gg / ( 2.* x ) ) );
```

```
// Transverse vector products
```

```
kk_T = TProduct(K,K); kk_T = TProduct(K,K);
```

```
kkp_T = kk_T; kkp_T = kk_T;
```

```
kqp_T = TProduct(K,QP); kqp_T = TProduct(K,QP);
```

```
kd_T = -1.* kqp_T; kd_T = -1.* kqp_T;
```

```
dd_T = TProduct(D,D); dd_T = TProduct(D,D);
```

```
kpqp_T = TProduct(KP,QP); kpqp_T = kqp_T;
```

```
kP_T = TProduct(K,P); kP_T = TProduct(K,P);
```

```
kpP_T = TProduct(KP,P); kpP_T = TProduct(KP,P);
```

```
qpP_T = TProduct(QP,P); qpP_T = TProduct(QP,P);
```

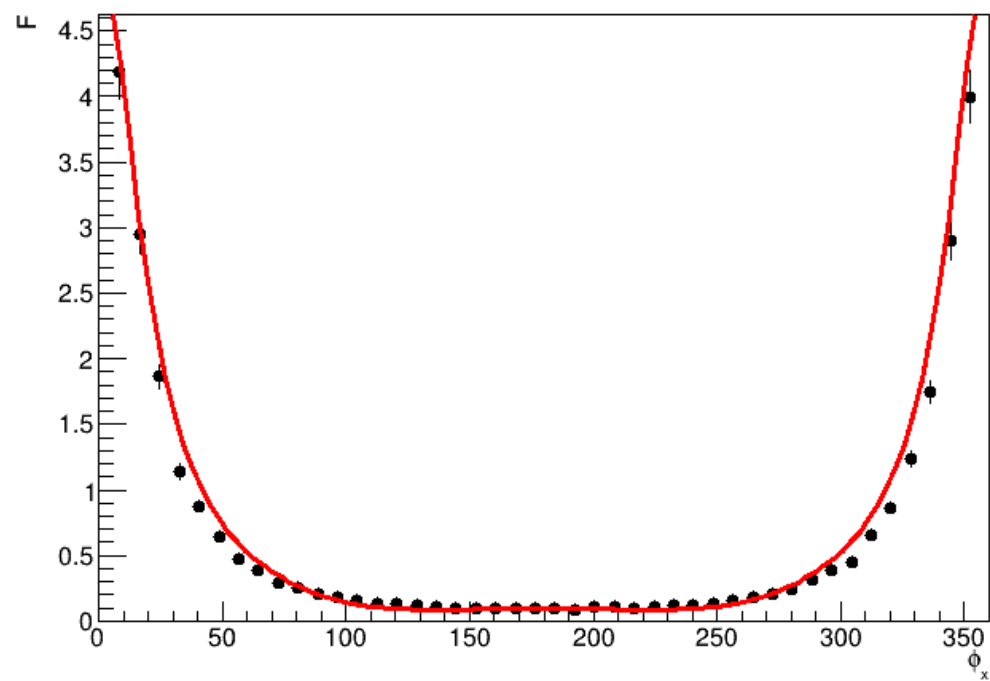
```
kpd_T = TProduct(KP,D); kpd_T = -1.* kqp_T;
```

```
qpd_T = TProduct(QP,D); qpd_T = -1. * dd_T;
```

## First trial on 2 architectures

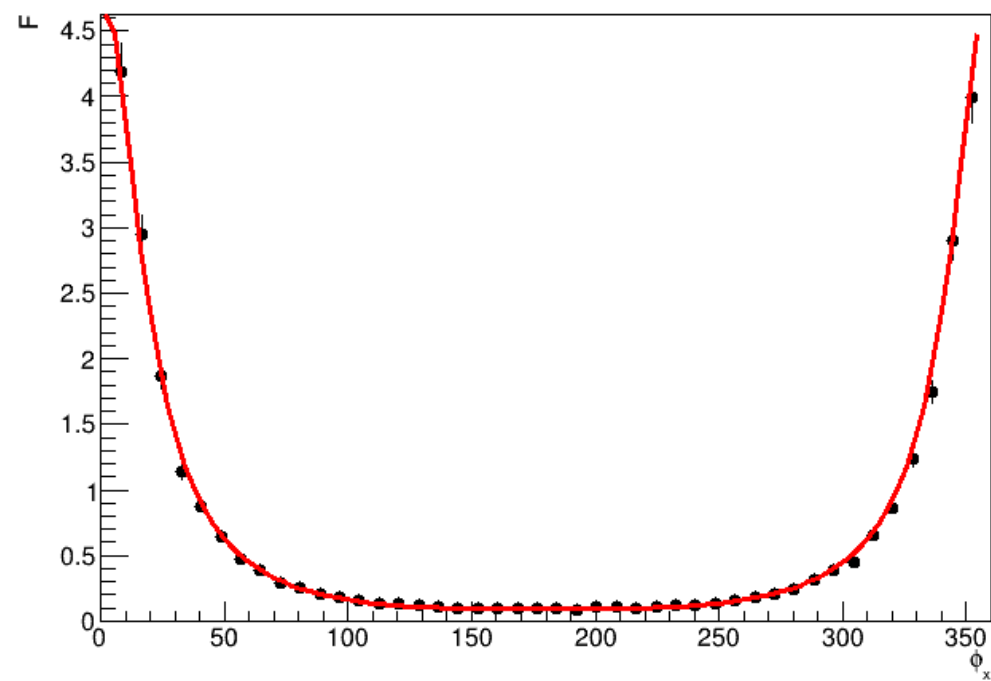
Architecture	1 <sup>st</sup>	2 <sup>nd</sup>
Hidden Layers	3	1
Neurons	100,100,80	512
DropOut	No	No
Epoch	4000	4000
Replica	1000	1000
Loss function	MSE	MSE
Optimized	Adam	Adam
Learning-rate	0.02	0.02
Data Normalization	No	No
Activation function	Tanh/ReLU	ReLU

set 1,  $k=2.750000$ ,  $QQ=1.515760$ ,  $xb=0.369204$ ,  $t=-0.306885$



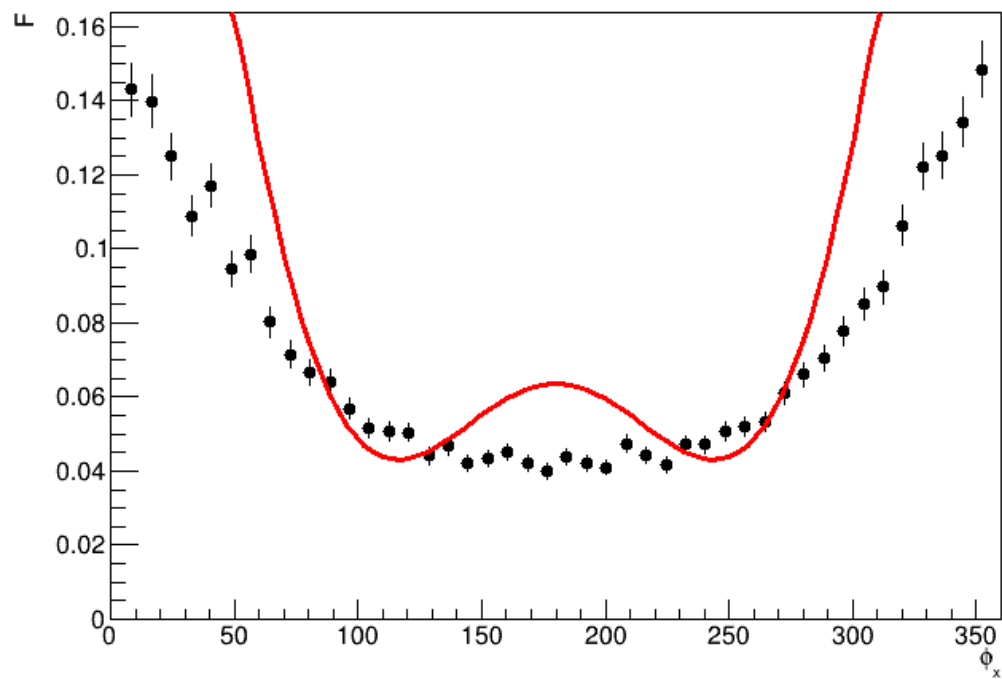
1<sup>st</sup> architecture

set 1,  $k=2.750000$ ,  $QQ=1.515760$ ,  $xb=0.369204$ ,  $t=-0.306885$



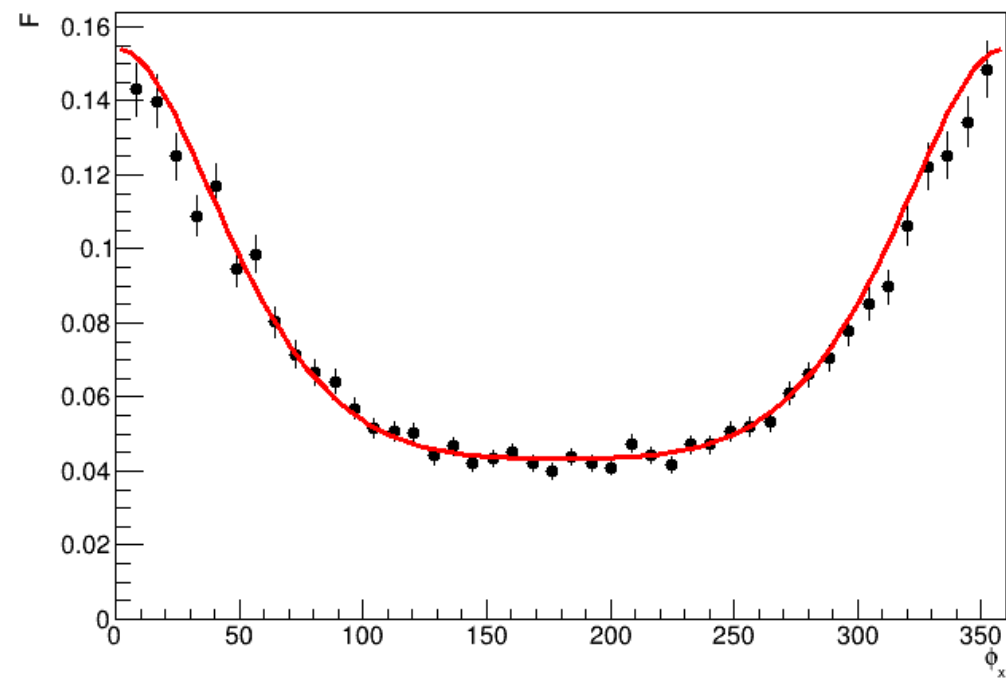
2<sup>nd</sup> architecture

set 2, k=2.750000, QQ=1.867290, xb=0.527413, t=-0.437447



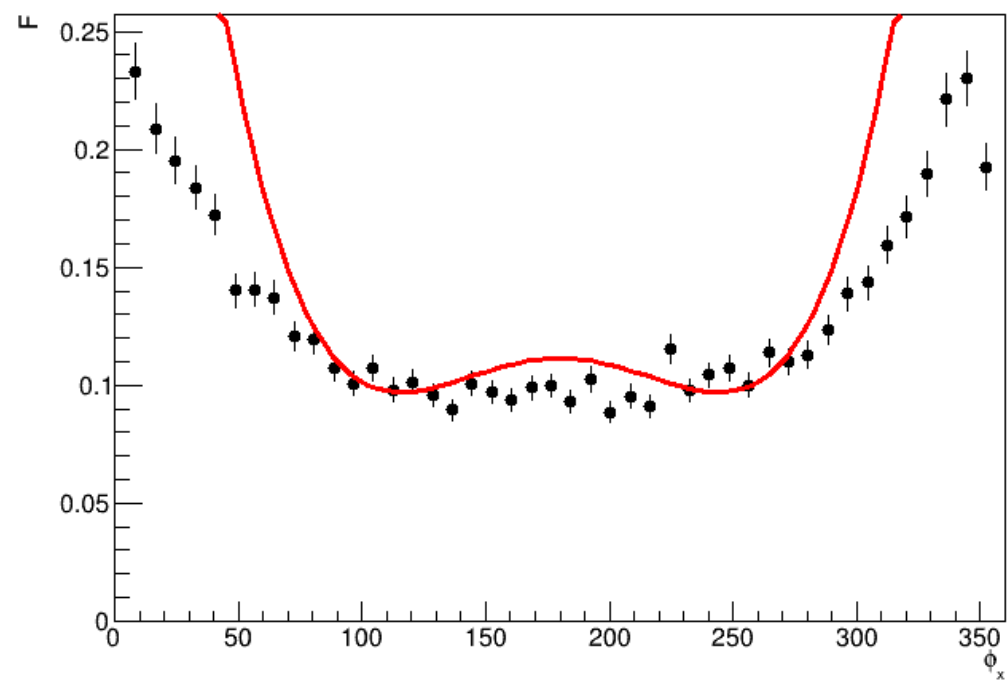
1<sup>st</sup> architecture

set 2, k=2.750000, QQ=1.867290, xb=0.527413, t=-0.437447



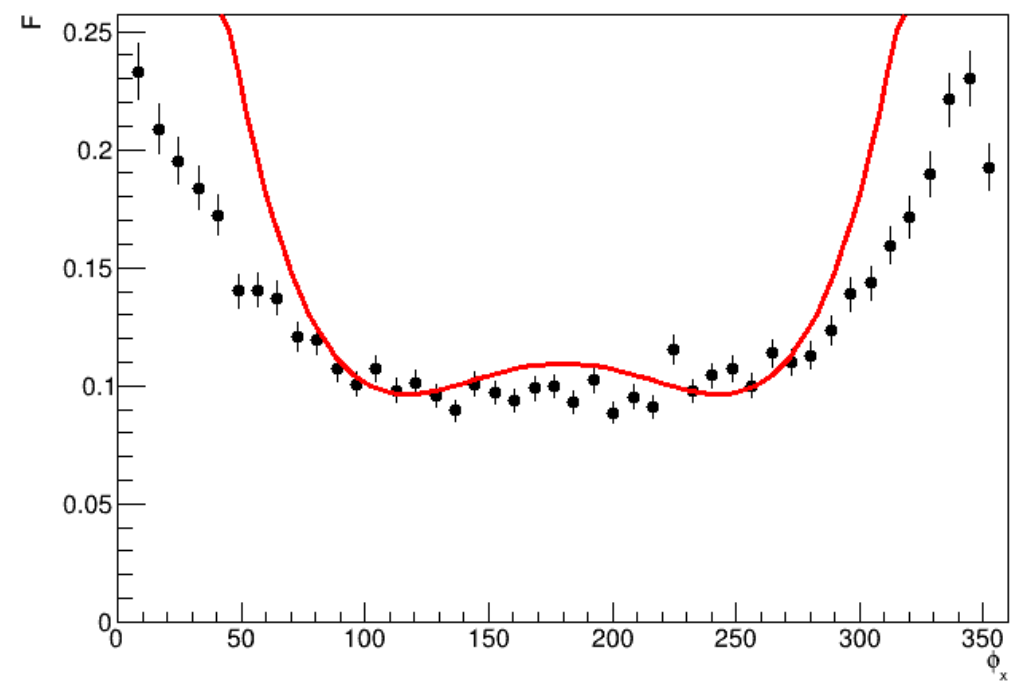
2<sup>nd</sup> architecture

set 3,  $k=2.750000$ ,  $QQ=1.188990$ ,  $xb=0.580204$ ,  $t=-0.479462$



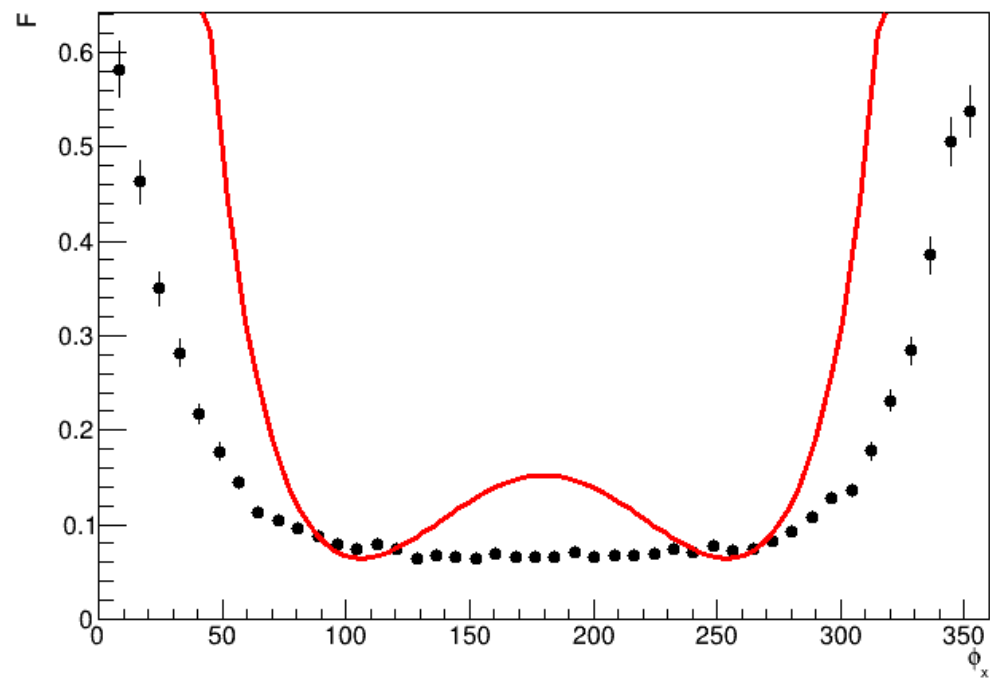
1<sup>st</sup> architecture

set 3,  $k=2.750000$ ,  $QQ=1.188990$ ,  $xb=0.580204$ ,  $t=-0.479462$



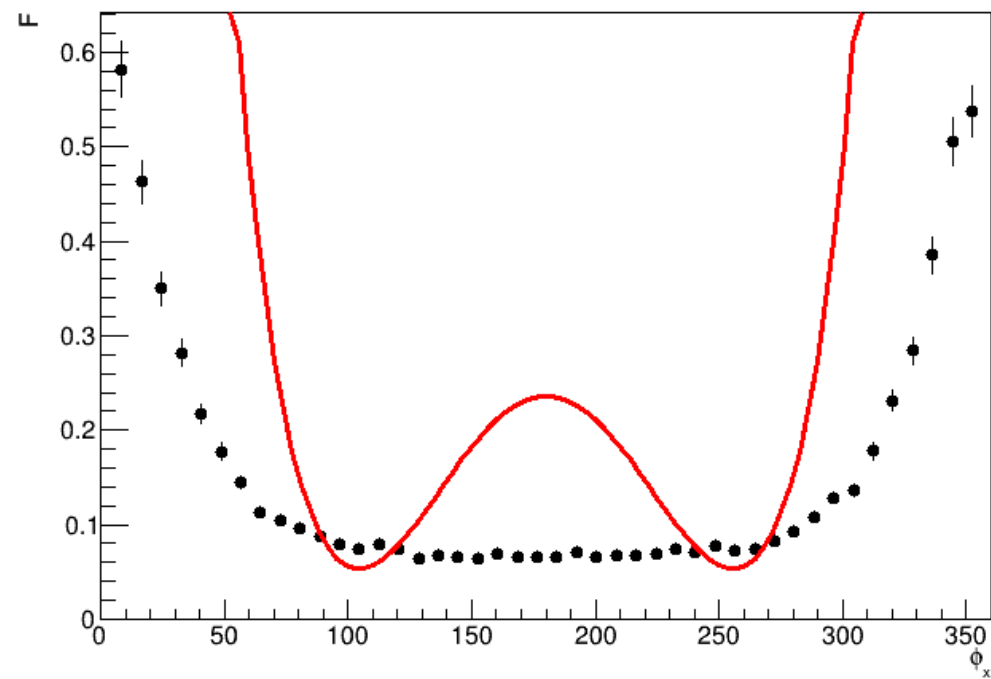
2<sup>nd</sup> architecture

set 4,  $k=2.750000$ ,  $QQ=1.444090$ ,  $xb=0.463857$ ,  $t=-0.483145$



1<sup>st</sup> architecture

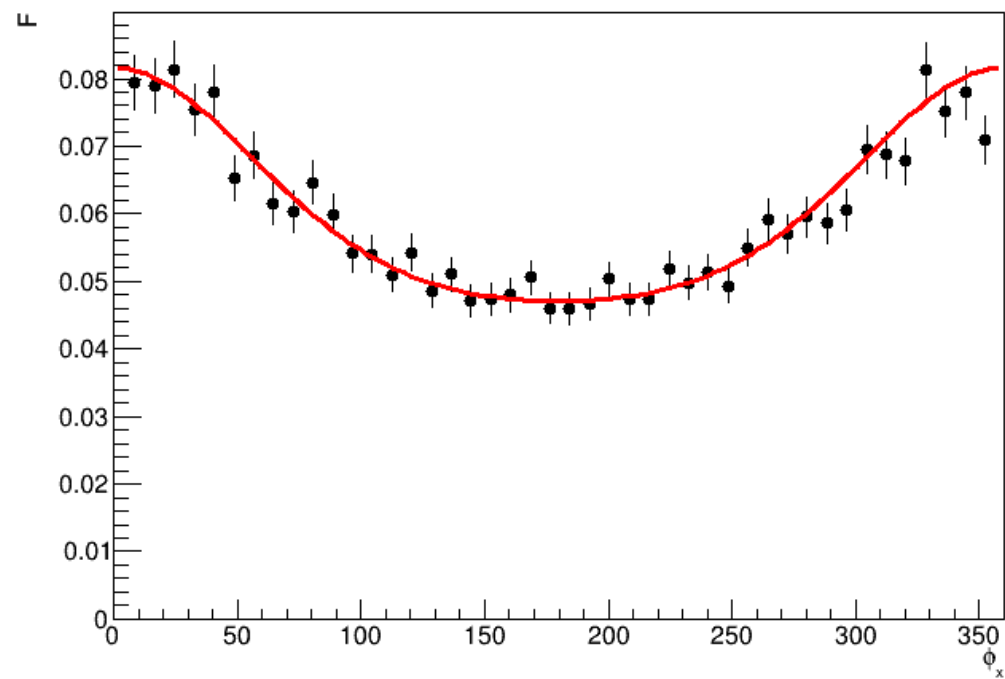
set 4,  $k=2.750000$ ,  $QQ=1.444090$ ,  $xb=0.463857$ ,  $t=-0.483145$



2<sup>nd</sup> architecture

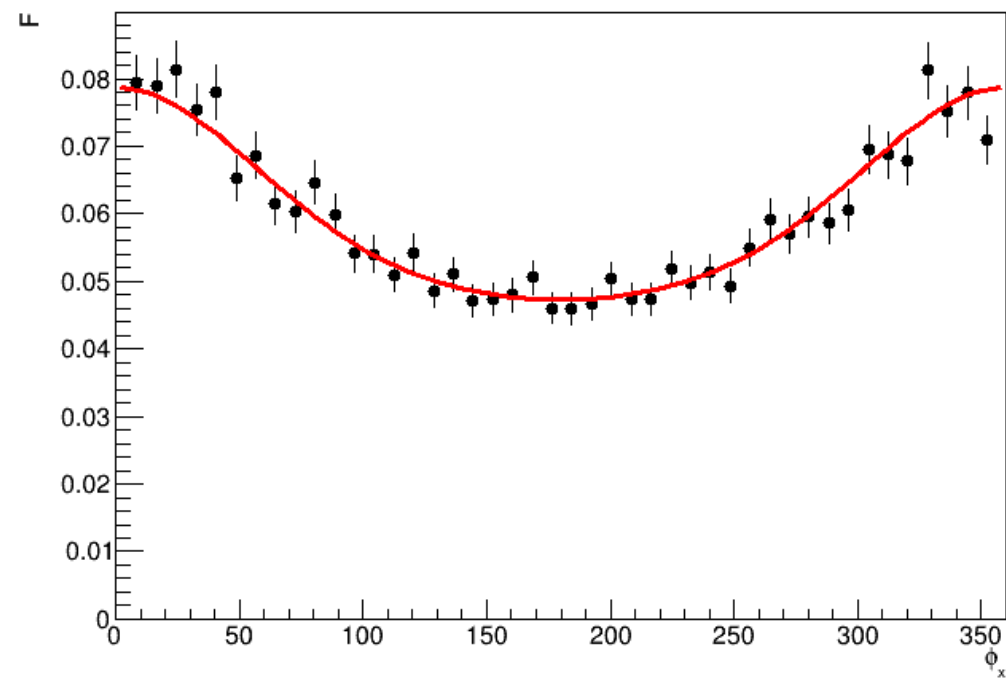


set 5,  $k=2.750000$ ,  $QQ=1.615250$ ,  $xb=0.599496$ ,  $t=-0.491783$



1<sup>st</sup> architecture

set 5,  $k=2.750000$ ,  $QQ=1.615250$ ,  $xb=0.599496$ ,  $t=-0.491783$



2<sup>nd</sup> architecture

# Conclusion

- 1<sup>st</sup> architecture perform reasonable on set 1 and 5
- 2<sup>nd</sup> architecture (less layer but more neurons) perform reasonable on set 1,2 and 5
- Both architecture overfitting on set 3 & 4
- Reducing the number of neurons in 2<sup>nd</sup> architecture from 512 to 256 make it worse
- See github to run the code