

# Weekly report

Qiyu Sha

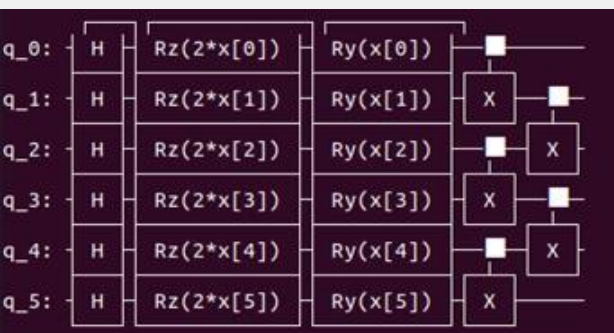
# Quantum machine learning

200 events(100 sig&100 bkg), 6 qubits(6 variables)

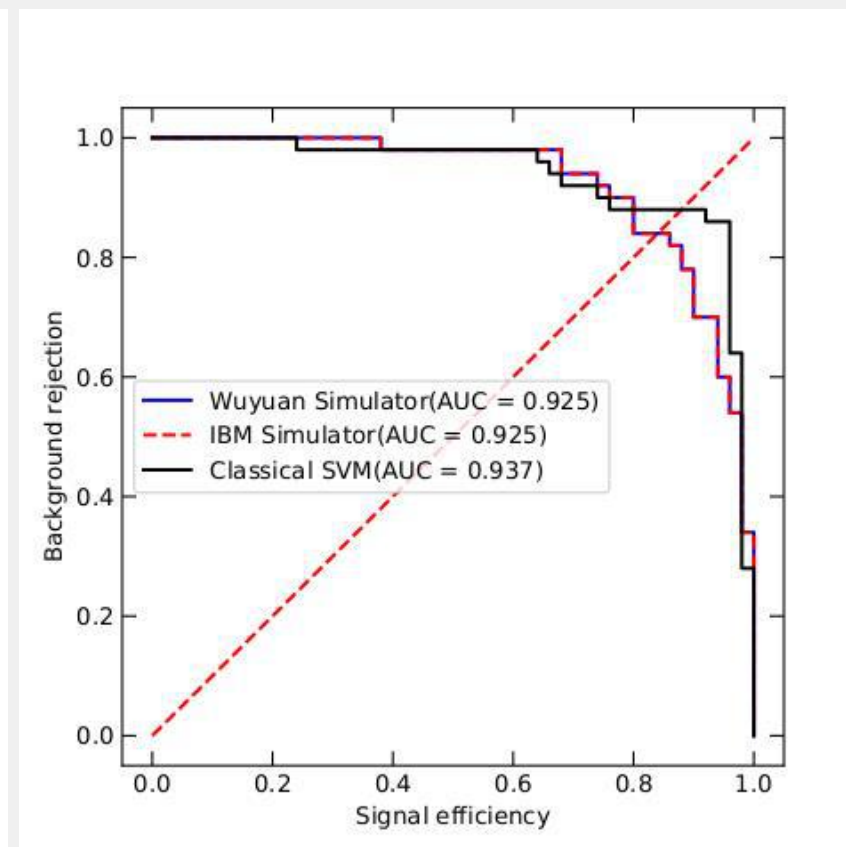
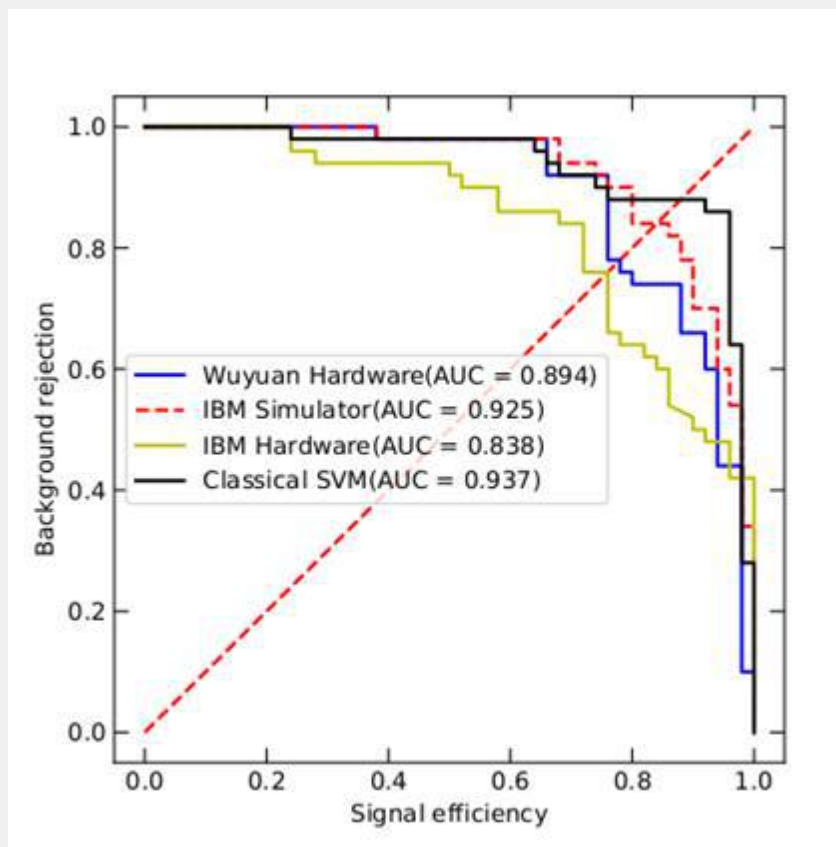
Svm&QSVM(IBM/wuyuan simulator/hardware) compare:

IBM Machine: ibm\_Nairobi  
Simulator: use statevector.

Feature map for QSVM:



Method	AUC value
SVM	0.937
IBM simulator	0.925
wuyuan simulator	0.925
IBM hardware	0.838
Wuyuan hardware	0.894

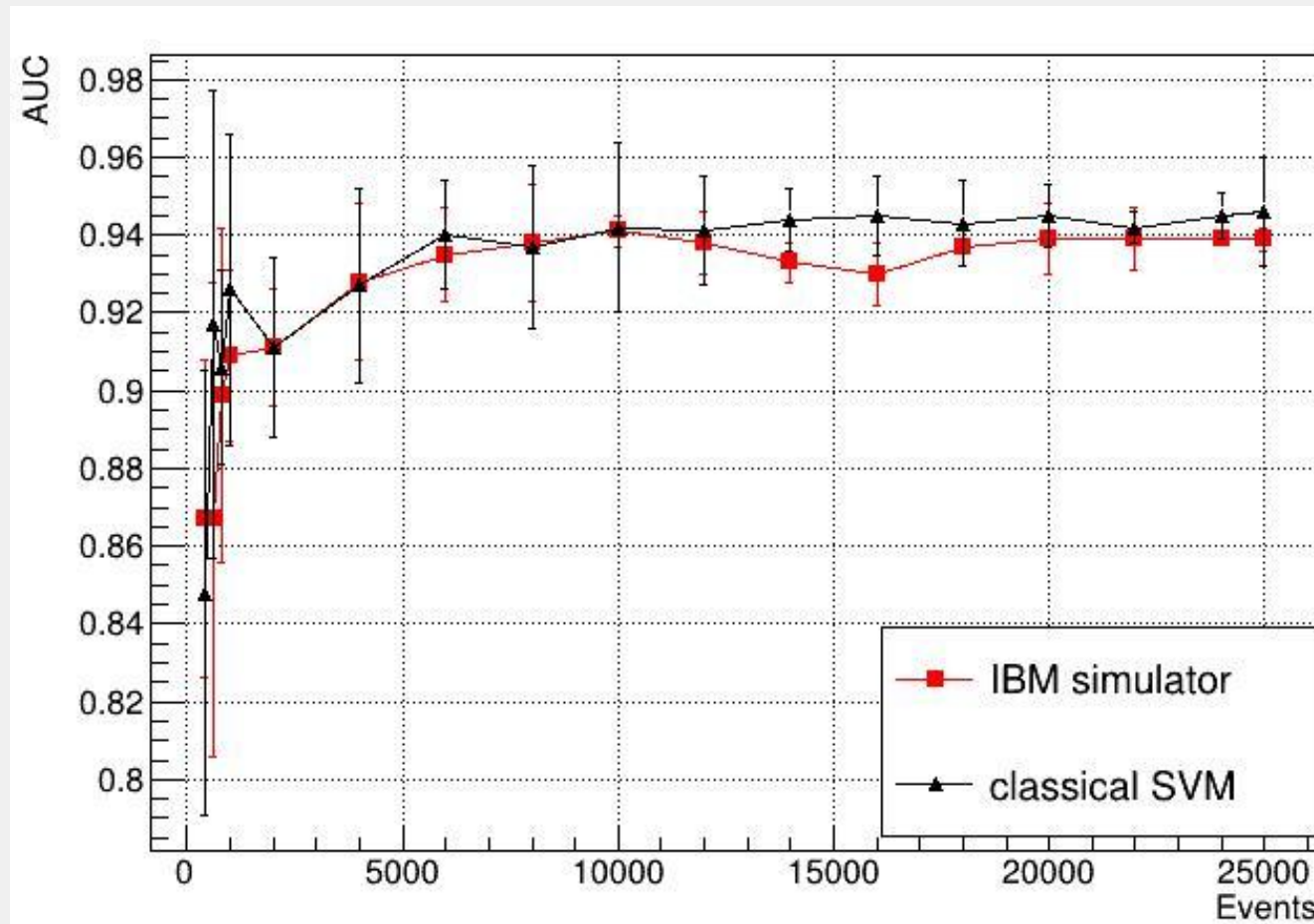


# Quantum machine learning

Scan for best hyper parameters(use cross validation, 3 folds)

Svm: rbf kernel. Best C and value: (Half of events for training, and then apply these best parameters in the test dataset.)

scan	cv=3	
	SVM	
	C	gamma
100	9	0.1
200	5	0.1
400	24	0.1
600	3	0.1
800	8	0.05
1000	18	0.1
2000	36	0.1
4000	6	0.1
6000	4	0.1
8000	18	0.1
10000	9	0.1
12000	6	0.1
14000	8	0.1
16000	16	0.1
18000	22	0.1
20000	6	0.1
22000	24	0.1
24000	18	0.1
25000	14	0.1

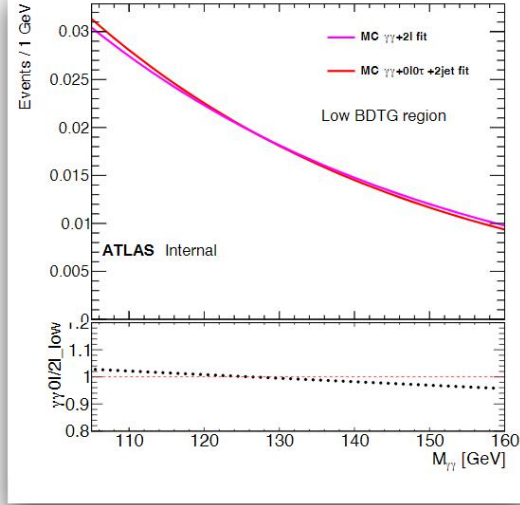
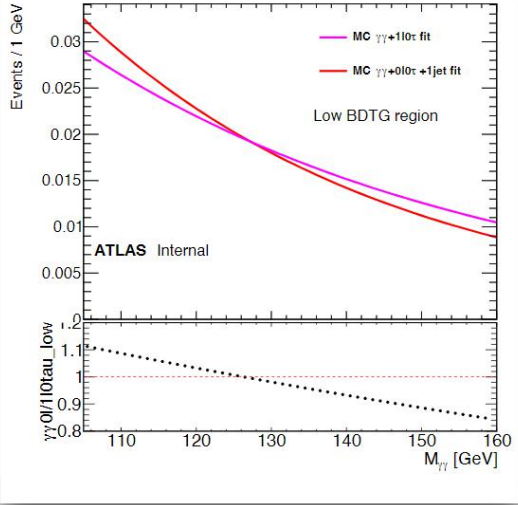


SVM finish the scan.  
Working with QSVM  
scan now.

# HH $\gamma\gamma$ +multilepton-bkg modeling

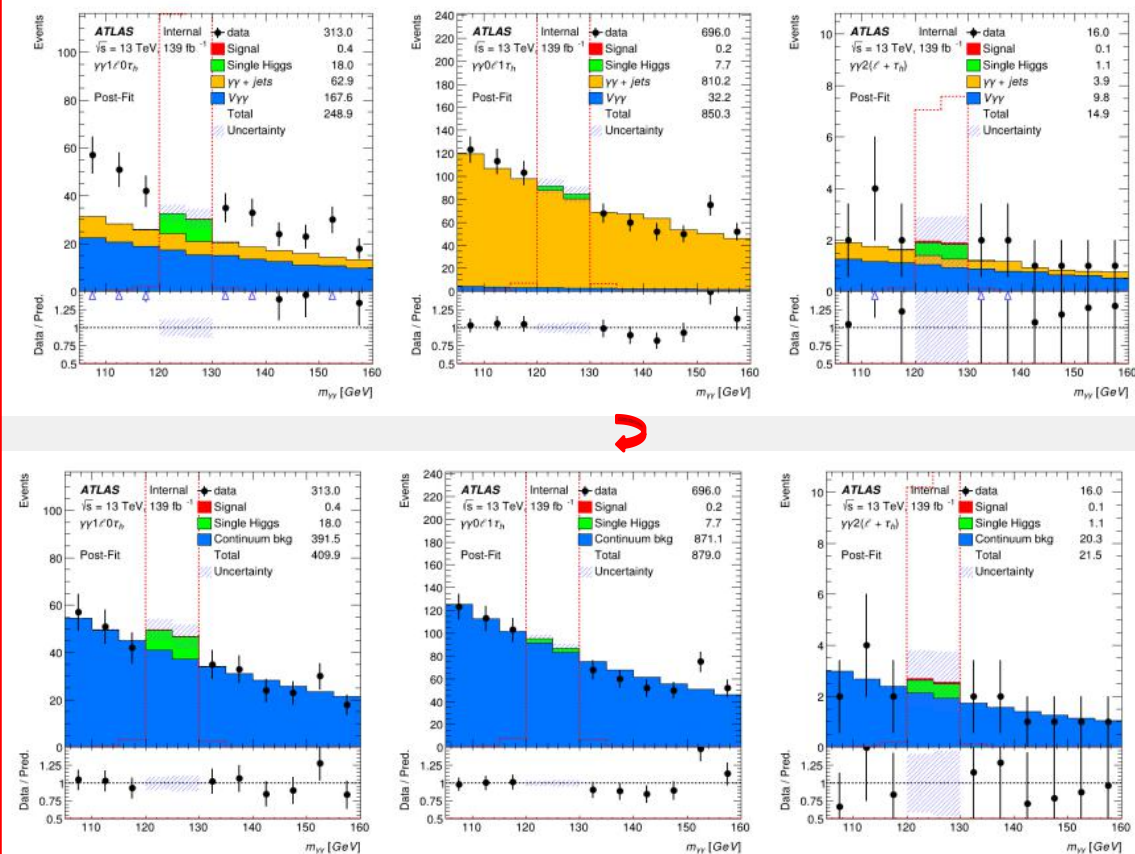
Use  $\gamma\gamma 0\ell 0\tau + 1/2$  jets shape as  $\gamma\gamma 1\ell/\gamma\gamma 2\ell$  continue background (The exponential function with a 2th polynomial function is chosen).

The shape different as a systematic uncertainty.



Channel	Low BDTG region	Medium BDTG region	Tight BDTG region
$\gamma\gamma+1\ell 0\tau_{\text{had}}$	$\pm 2.34\%$	$\pm 2.76\%$	$\pm 2.09\%$
$\gamma\gamma+0\ell 1\tau_{\text{had}}$	$\pm 1.83\%$	$\pm 2.77\%$	$\pm 2.34\%$
$\gamma\gamma+2L$	$\pm 1.74\%$	$\pm 2.55\%$	$\pm 2.68\%$

## Before bkg modeling & After bkg modeling:



# HH $\gamma\gamma$ +multilepton-spurious signal

- Use signal+background model to fit a background only sample.
- Use histpdf to get the signal model and use a exponential function with a second-order polynomial as the background model.
- This fit can get the  $S_{spur}$ . Then do the test which shows below.

- $S_{spur} < 10\%N_{s,exp}$  where  $N_{s,exp}$  is the expected number of signal events within the SM in category ( $\mu_{sp} = S_{spur}/N_{s,exp}$ ),
- $S_{spur} < 20\%\sigma_{bkg}$ , where  $\sigma_{bkg}$  is the statistical uncertainty on the fitted number of signal events, when fitting the signal+background model to a background only Asimov dataset ( $Z_{sp} = S_{spur}/\sigma_{bkg}$ ).

# HH yy+multilepton

	Function	Ndof	$\mu_{sp}[\%]$	$Z_{sp}[\%]$	P(chi2) [%]	Result
1l0tau low	ExpPoly2	2	0.44	1.76	27.67%	pass
1l0tau medium	ExpPoly2	2	0.34	0.93	51.17%	pass
1l0tau tight	ExpPoly2	2	0.54	1.36	49.91%	pass
	Function	Ndof	$\mu_{sp}[\%]$	$Z_{sp}[\%]$	P(chi2) [%]	Result
0l1tau low	ExpPoly2	2	1.75	8.30	15.80%	pass
0l1tau medium	ExpPoly2	2	0.47	1.27	39.06%	pass
0l1tau tight	ExpPoly2	2	0.17	0.39	41.27%	pass
	Function	Ndof	$\mu_{sp}[\%]$	$Z_{sp}[\%]$	P(chi2) [%]	Result
2l low	ExpPoly2	2	0.037	0.078	97.18%	pass
2l medium	ExpPoly2	2	0.004	0.007	96.08%	pass
2l tight	ExpPoly2	2	0.045	0.037	97.12%	pass

# To do

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- Scan best parameters for QSVM.
- Working with theory uncertainty.
- Try to use function to fit the signal shape(Use Hgam core and then redo the spurious signal.) (no urgent.)