



Study of the performance of the DeepTau “VSjet” discriminator in the 4tops analysis

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Some nomenclature



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- VSjet discriminator comes in 8 WPs, will use shorthands for them
 - VeryVeryTight $\rightarrow$ **VVT**
 - VeryTight $\rightarrow$ **VT**
 - Tight $\rightarrow$ **T**
 - Medium $\rightarrow$ **M**
 - Loose $\rightarrow$ **L**
 - VeryLoose $\rightarrow$ **VL**
 - VeryVeryLoose $\rightarrow$ **VVL**
 - VeryVeryVeryLoose $\rightarrow$ **VVVL**
- Signal sample: $t\bar{t}t\bar{t}$
- Background samples: QCD, $t\bar{t}$, $t\bar{t}+X$
- HLT_PFJet450 is added to the signal triggers



- **For every WP**, I am looking at the following:
 - Number of signal events, **S**
 - Number of background events, **B**
 - Significance, **Z** = $\sqrt{2[(S+B)\log(1+S/B) - S]}$
 - Significance when a 20% unc. is associated with B, **Z_{syst}**
 - Reconstruction efficiency in $t\bar{t}\tau$, **E(sig)**: number of gen taus that are matched to a reco tau in signal sample
 - Reconstruction efficiency in background, **E(bkg)**: number of gen taus that are matched to a reco tau in background samples
 - Reconstruction purity in $t\bar{t}\tau$, **P(sig)**: number of reco taus that are matched to a gen tau in signal sample
 - Reconstruction purity in background, **P(bkg)**: number of reco taus that are matched to a gen tau in background samples



- To compute Z_{syst} , **two options**:
 - **RooStats**::NumberCountingUtils::BinomialExpZ (Double_t sExp, Double_t bExp, Double_t fractionalBUncertainty) method
 - **Combine** datacard
- First method is fast but seemingly unstable under some conditions
- Second method requires some more work, but stabler
- Decided to go with Combine in the end
- Write the easiest datacard possible:
 - 1 bin (the WP), two processes (S, B)
 - 1 log-normal, systematic uncertainty affecting B
 - Set uncertainty to 20%
- Write 1 datacard for each WP and category, feed it to Combine

The datacard



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```
imax 1   number of channels
jmax 1   number of backgrounds
kmax *   number of nuisance parameters
```

bin WP

observation -1

bin	WP	WP
process	S	B
process	0	1
rate	4.204	2633.654

bkg_unc	lnN	-	1.2
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- To run Combine and have significances printed:

```
combine -M Significance datacard.txt -t -1 --expectSignal=1
```

- Meaning:

- -M Significance: use “Significance” method documented in [Combine documentation](#)
- -t -1: fit an Asimov dataset
- --expectSignal=1: generate Asimov dataset with $\mu = 1$

Before categories



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	VVT	VT	T	M	L	VL	VVL	VVVL
S	14	19	25	32	43	63	89	91
B	89647	145460	232105	393839	709586	1618358	3063950	3115740
Z	0.0462	0.0490	0.0509	0.0509	0.0509	0.0499	0.0510	0.0517
E(sig)	0.2056	0.2674	0.3343	0.4065	0.4894	0.5843	0.6441	0.6528
E(bkg)	0.2556	0.3305	0.4097	0.4965	0.5930	0.7028	0.7670	0.7723
P(sig)	0.9147	0.8822	0.8400	0.7833	0.7028	0.5672	0.4444	0.4201
P(bkg)	0.3897	0.3110	0.2419	0.1729	0.1151	0.0602	0.0351	0.0341



	VVT	VT	T	M	L	VL	VVL	VVVL
S	4	6	8	10	13	20	27	23
B	2634	5619	9460	14165	25807	51725	92553	84908
Z	0.0819	0.0760	0.0772	0.0829	0.0826	0.0863	0.0892	0.0789
Z _{syst}	0.0074	0.0047	0.0037	0.0032	0.0023	0.0017	0.0012	0.0011
E(sig)	0.8851	0.8779	0.8727	0.8588	0.8341	0.7816	0.7113	0.7028
E(bkg)	0.9769	0.9728	0.9681	0.9599	0.9435	0.9088	0.8576	0.8524
P(sig)	0.9068	0.8696	0.8253	0.7606	0.6731	0.5278	0.3990	0.3922
P(bkg)	0.6652	0.4071	0.3030	0.2477	0.1634	0.0968	0.0586	0.0557

1τ 0L yields



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	VVT	VT	T	M	L	VL	VVL	VVVL
$t\bar{t}t\bar{t}$	4	6	8	10	13	20	27	23
$t\bar{t}$	2190	3137	4476	6509	10081	18269	30013	26298
QCD	368	2378	4842	7461	15443	32989	61806	57951
$t\bar{t}+X$	75	103	142	194	283	467	734	660



	VVT	VT	T	M	L	VL	VVL	VVVL
S	3	5	6	8	10	15	21	19
B	839	1209	1728	2528	4028	7422	13164	12221
Z	0.1178	0.1325	0.1451	0.1561	0.1652	0.1793	0.1867	0.1710
Z _{syst}	0.019	0.018	0.016	0.014	0.012	0.0097	0.0075	0.0072
E(sig)	0.8786	0.8751	0.8613	0.8488	0.8222	0.7675	0.6947	0.6868
E(bkg)	0.9415	0.9302	0.9146	0.8903	0.8537	0.7723	0.6694	0.6688
P(sig)	0.9076	0.8720	0.8278	0.7681	0.6781	0.5339	0.4040	0.3984
P(bkg)	0.7461	0.6648	0.5779	0.4801	0.3611	0.2312	0.1405	0.1403



	VVT	VT	T	M	L	VL	VVL	VVVL
S	0.8	1.1	1.4	1.9	2.5	3.8	5.3	4.9
B	20.8	32.2	50.8	85.7	150.2	324.8	602.2	571.1
Z	0.1737	0.1898	0.1991	0.2007	0.2073	0.2120	0.2176	0.2045
Z _{syst}	0.13	0.13	0.12	0.10	0.084	0.061	0.047	0.046
E(sig)	0.8789	0.8693	0.8614	0.8447	0.8171	0.7563	0.6818	0.6782
E(bkg)	0.7366	0.6995	0.6351	0.5574	0.4328	0.2710	0.1727	0.1710
P(sig)	0.9029	0.8697	0.8178	0.7622	0.6679	0.5166	0.3893	0.3868
P(bkg)	0.2685	0.2160	0.1690	0.1256	0.0843	0.0449	0.0257	0.0255



	VVT	VT	T	M	L	VL	VVL	VVVL
S	0.0518	0.0726	0.0932	0.1233	0.1670	0.2720	0.4149	0.3837
B	0.3973	0.5690	0.7806	1.0087	1.5519	2.7000	4.3679	4.1507
Z	0.0805	0.0942	0.1035	0.1204	0.1318	0.1629	0.1955	0.1855
Z _{syst}	0.080	0.094	0.10	0.12	0.13	0.16	0.18	0.17
E(sig)	0.8977	0.8801	0.8700	0.8418	0.8115	0.7507	0.6533	0.6506
E(bkg)	0.8729	0.8589	0.8444	0.8270	0.8009	0.7783	0.6669	0.6673
P(sig)	0.9027	0.8379	0.7927	0.7224	0.6309	0.4638	0.3232	0.3269
P(bkg)	0.5399	0.5134	0.4570	0.4236	0.3378	0.2430	0.1604	0.1584



	VVT	VT	T	M	L	VL	VVL	VVVL
S	0.09	0.17	0.30	0.52	0.97	2.26	4.63	9.77
B	32.50	68.77	133.38	278.95	703.92	2934.12	10111.94	62923.10
Z	0.0166	0.0210	0.0264	0.0313	0.0364	0.0417	0.0461	0.0389
Z_{syst}	0.010	0.010	0.0098	0.0084	0.0063	0.0036	0.0021	0.00066
E(sig)	0.9695	0.9590	0.9634	0.9522	0.9422	0.9030	0.8599	0.7956
E(bkg)	0.9870	0.9881	0.9881	0.9862	0.9813	0.9647	0.9383	0.8970
P(sig)	0.9567	0.9503	0.9199	0.8772	0.8040	0.6715	0.5416	0.4296
P(bkg)	0.7867	0.7318	0.6863	0.6074	0.4672	0.2598	0.1415	0.0708



	VVT	VT	T	M	L	VL	VVL	VVVL
S	0.05	0.08	0.14	0.26	0.49	1.14	2.42	5.27
B	3.86	7.35	15.61	31.38	87.52	339.18	1030.43	3924.37
Z	0.0230	0.0307	0.0362	0.0463	0.0527	0.0620	0.0754	0.0842
Z _{syst}	0.021	0.026	0.027	0.032	0.023	0.015	0.011	0.0063
E(sig)	0.9729	0.9627	0.9575	0.9490	0.9347	0.9003	0.8427	0.7852
E(bkg)	0.8750	0.9012	0.9035	0.9066	0.8659	0.8179	0.7412	0.7030
P(sig)	0.9589	0.9422	0.9070	0.8565	0.7865	0.6564	0.5163	0.4122
P(bkg)	0.2959	0.3835	0.3642	0.3629	0.3130	0.2241	0.1476	0.1241



	VVT	VT	T	M	L	VL	VVL	VVVL
S	0.004	0.008	0.016	0.028	0.053	0.133	0.341	0.827
B	0.105	0.164	0.265	0.452	0.944	3.515	10.357	48.405
Z	0.0133	0.0189	0.0310	0.0419	0.0540	0.0707	0.1052	0.1186
Z _{syst}	0.012	0.019	0.031	0.041	0.052	0.067	0.090	0.073
E(sig)	0.9710	0.9649	0.9537	0.9644	0.9062	0.8898	0.8052	0.7466
E(bkg)	0.7391	0.7370	0.7628	0.7499	0.7859	0.5851	0.3824	0.2246
P(sig)	0.9860	0.9296	0.8657	0.7911	0.7315	0.5904	0.4485	0.3586
P(bkg)	0.2013	0.1699	0.1969	0.1732	0.1694	0.1024	0.0528	0.0257



- **For every WP**, I am looking at the following:
 - **After Hideki's suggestion**: “gen-level efficiencies”
 - Define “gen-level categories” by making requirements on gen tau and gen leptons multiplicities
 - Ask for an equal number of reco taus
 - See how many of these reco taus are matched to the gen taus
 - $\varepsilon_{1m} = N_{1\text{-reco-tau-is-matched}}^{\text{events}} / N_{\text{total}}^{\text{events}}$
 - $\varepsilon_{2m} = N_{2\text{-reco-tau-are-matched}}^{\text{events}} / N_{\text{total}}^{\text{events}}$
 - $\varepsilon_{\text{compl}} = 1 - \varepsilon_{1m} - \varepsilon_{2m}$



	VVT	VT	T	M	L	VL	VVL	VVVL
ε_{1m}	0.9687	0.9570	0.9403	0.9132	0.8707	0.7827	0.6794	0.6690
ε_{2m}	nan	nan	nan	nan	nan	nan	nan	nan
$\varepsilon_{\text{compl}}$	0.0313	0.0430	0.0597	0.0868	0.1293	0.2173	0.3206	0.3310



	VVT	VT	T	M	L	VL	VVL	VVVL
ε_{1m}	0.9674	0.9559	0.9390	0.9182	0.8838	0.8123	0.7267	0.7171
ε_{2m}	nan	nan	nan	nan	nan	nan	nan	nan
$\varepsilon_{\text{compl}}$	0.0326	0.0441	0.0610	0.0818	0.1162	0.1877	0.2733	0.2829



	VVT	VT	T	M	L	VL	VVL	VVVL
ε_{1m}	0.9622	0.9549	0.9395	0.9182	0.8932	0.8468	0.7794	0.7705
ε_{2m}	nan	nan	nan	nan	nan	nan	nan	nan
$\varepsilon_{\text{compl}}$	0.0378	0.0451	0.0605	0.0818	0.1068	0.1532	0.2206	0.2295



	VVT	VT	T	M	L	VL	VVL	VVVL
ε_{1m}	0.9639	0.9484	0.9332	0.9198	0.9062	0.8765	0.8498	0.8532
ε_{2m}	nan	nan	nan	nan	nan	nan	nan	nan
$\varepsilon_{\text{compl}}$	0.0361	0.0516	0.0668	0.0802	0.0938	0.1235	0.1502	0.1468



	VVT	VT	T	M	L	VL	VVL	VVVL
ε_{1m}	0.0117	0.0160	0.0263	0.0477	0.0642	0.1062	0.1424	0.1943
ε_{2m}	0.9883	0.9840	0.9726	0.9501	0.9318	0.8788	0.8204	0.7373
$\varepsilon_{\text{compl}}$	0.0000	0.0000	0.0011	0.0022	0.0040	0.0150	0.0371	0.0684



	VVT	VT	T	M	L	VL	VVL	VVVL
ε_{1m}	0.0343	0.0369	0.0356	0.0610	0.0717	0.0944	0.1306	0.1870
ε_{2m}	0.9634	0.9589	0.9617	0.9350	0.9244	0.8995	0.8574	0.7740
$\varepsilon_{\text{compl}}$	0.0023	0.0043	0.0027	0.0040	0.0039	0.0061	0.0119	0.0390



	VVT	VT	T	M	L	VL	VVL	VVVL
ε_{1m}	0.0668	0.0909	0.1163	0.0906	0.1247	0.1037	0.1241	0.1639
ε_{2m}	0.9332	0.9000	0.8779	0.9094	0.8692	0.8902	0.8586	0.8138
$\varepsilon_{\text{compl}}$	0.0000	0.0091	0.0058	0.0000	0.0061	0.0061	0.0173	0.0223



- Numbers looked odd at the beginning
 - We understand them better now, they seem OK
- **We used to use the Medium VSjet WP**
- In 5 out of 7 categories, the **VVL WP gives the best significance**
 - In the remaining two, VVVL WP gives the best significance
- **It seems we should go looser in VSjet**
 - **ZhangYu is seeing the same**