Measurement of the top quark mass in the all-hadronic mode at CDF

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CDF Collaboration

A measurement of the top quark mass ($M_{\text{top}}$) in the all-hadronic decay channel is presented. It uses $5.8 \, \text{fb}^{-1}$ of $p\bar{p}$ data collected with the CDF II detector at the Fermilab Tevatron Collider. Events with six to eight jets are selected by a neural network algorithm and by the requirement that at least one of the jets is tagged as a $b$-quark jet. The measurement is performed with a likelihood fit technique, which simultaneously determines $M_{\text{top}}$ and the jet energy scale (JES) calibration. The fit yields a value of $M_{\text{top}} = 172.5 \pm 1.4(\text{stat}) \pm 1.0(\text{JES}) \pm 1.1(\text{syst}) \text{ GeV}/c^2$.
The mass of the top quark ($M_{\text{top}}$) is a fundamental parameter of the standard model (SM) and its large value makes the top quark contribution dominant in loop corrections to many observables, like the $W$ boson mass $M_W$. Precise measurements of $M_W$ and $M_{\text{top}}$ allow one to set indirect constraints on the mass of the, as yet unobserved, Higgs boson [1].

In this Letter we present a measurement of $M_{\text{top}}$ using proton-antiproton collision events at a center-of-mass energy of 1.96 TeV. Top quarks are produced at the largest rate in pairs ($t\bar{t}$), with each top quark decaying immediately into a $W$ boson and a $b$ quark nearly 100% of the time [2]. In this analysis events where both the $W$’s decay to a quark–antiquark pair ($t\bar{t} \rightarrow W^+bW^-\bar{b} \rightarrow qg\bar{q}g\bar{d}d$) are considered. This all-hadronic final state has the largest branching ratio among the possible decay channels (46%), but it is overwhelmed by the QCD multijet background processes, which surpass $t\bar{t}$ production by three orders of magnitude even after a dedicated trigger requirement. Nevertheless, it will be shown how this difficult background can be successfully controlled and significantly suppressed with a properly optimized event selection. The fundamental analysis technique is the same exploited to obtain the previous result from CDF, and is described in details in [3]. However, improvements in the event selection and a larger dataset allow us to decrease the total uncertainty on $M_{\text{top}}$ by 21%. The additional dataset has been acquired at higher instantaneous luminosity, which results in a higher number of background events in the data sample. Despite this fact, the introduction of significant improvements to the analysis results in the world best measurement of $M_{\text{top}}$ in the all-hadronic channel so far, also entering with the third largest weight in the $M_{\text{top}}$ world average calculation [4,5].

The data correspond to an integrated luminosity of 5.8 fb$^{-1}$. They have been collected between March 2002 and February 2010 by the CDF detector, a general-purpose apparatus designed to study $pp$ collisions at the Tevatron and described in detail in [6]. Events used in this measurement are selected by a multijet trigger [3], and retained only if they are well contained in the detector acceptance, have no well-identified energetic electron or muon, and have a missing transverse energy$^{32} \cancel{E}_T$ satisfying $\cancel{E}_T/\sqrt{\sum E_T} < 3$ GeV$^2$, where $\sum E_T$ is the sum of the transverse energy $E_T$ of all jets. Candidate events are also required to have six to eight “tight” ($E_T \geq 15$ GeV and $|\eta| \leq 2.0$) jets. After this preselection, a total of about 5.7 M events is observed in the data, with less than 9 thousand expected from $t\bar{t}$ events. To improve the signal-to-background ratio ($S/B$) a $b$-tagging algorithm [7] is used to identify (“$b$-tag” or simply “tag”) jets that most likely resulted from the fragmentation of a $b$ quark. Only events with one to three tagged jets are then retained, improving the $S/B$ by a factor of 6. In order to further increase the signal purity, a multivariate algorithm is implemented. An artificial neural network, based on a set of kinematic and jet shape variables [3], is used to take advantage of the distinctive features of signal and background events. The neural network was trained using simulated $t\bar{t}$ events generated by Pythia [8] and propagated through the CDF detector simulation. At this level of selection the fraction of signal events is still negligible so that the data can be used to represent the background. The value of the output node, $N_{\text{out}}$, is used as a discriminant between signal and background, providing a gain in $S/B$ by an additional factor of about 30.

The background for the $t\bar{t}$ multijet final state comes mainly from QCD production of heavy-quark pairs ($bb$ and $cc$) and events with false tags from light–quark and gluon jets. Given the large theoretical uncertainties on the QCD multijet production cross section, the background prediction is obtained from the data themselves. The probability of tagging a jet in a background event ($P^+$) is evaluated using data with five tight jets and passing the preselection ($S/B \approx 1/2000$). This “tag” rate is parametrized in terms of a few relevant jet variables and is then used to estimate the probability that a candidate event belongs to the background and contains a given number of tagged jets. As described in detail in [3] this allows to predict the expected amount of background events in the selected samples as well as their distributions. For example, the average number of background 1-tag events is estimated by

$$\sum_{i=1}^{N_{\text{jets}}} C^{1}_{\text{tag}} \cdot P^{+} \prod_{k \neq i} (1 - P^{+})$$

where the outer sum runs over all events selected just before the $b$-tagging requirement, and the inner one over the jets of the event. The factor $C^{1}_{\text{tag}}$ represents a correction to take into account correlations among jets within the same event [3], and it is parametrized as a function of the same variables used for the tag rate.

The analysis employs the template method to measure $M_{\text{top}}$ with simultaneous calibration of the jet energy scale (JES) [3,9], allowing a strong reduction of the associated systematic uncertainty. Distributions of variables sensitive to the “true” values of $M_{\text{top}}$ and JES, obtained by Monte Carlo (MC) events, are used as a reference (“template”) in the measurement. A maximum likelihood fit is performed to define the values that best reproduce the same distributions as observed in the data. An usual choice is to consider the distributions of the event-by-event reconstructed top quark mass, $m^{\text{reco}}_t$, and $W$ boson mass, $m^{\text{reco}}_W$ as the reference templates. The JES is a multiplicative factor representing a correction to the event-by-event reconstructed jet energy ($E^{\text{reco}}_T$), so that its corrected energy $E_{T} = \text{JES} \cdot E^{\text{raw}}_T$, is a better estimate of the energy of the underlying parton [10]. Discrepancies between data and simulation result in an uncertainty on the JES value to be applied in MC events to reproduce the data, and, as a consequence, on the

The missing $E_T$ vector, $\vec{\cancel{E}}_T$, is defined by $\vec{\cancel{E}}_T = -\sum E_T \hat{r}_j$, where $\hat{r}_j$ is the unit vector in the x-y plane pointing from the primary interaction vertex to a given calorimeter tower $i$, and $E_{T,\perp}$ is the $E_T$ measured in that tower. Finally $\vec{\cancel{E}}_T = |\vec{\cancel{E}}_T|$. 

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measurements of $M_{\text{top}}$. Nevertheless, this value can be calibrated “in situ”, using $m_{\text{rec}}$ as a template. This represents a well-tested technique, first applied in [9] and now used to obtain the most precise top quark mass measurements at the Tevatron [4,5].

The templates are built as follows [3]. For each selected event, each of the six highest-$p_T$ jets is assigned in turn to one of the six quarks of a $t\bar{t}$ all-hadronic final state. Then, for each combination the jets are arranged in two triplets (the top quarks), each including a doublet (corresponding to the $W$ boson) and a $b$ quark. To reduce the possible number of permutations, $b$-tagged jets are assigned to $b$ quarks only, resulting in $30, 6$ or $18$ permutations for events with one, two or three tagged jets, respectively.33

For each permutation $m_{\text{rec}}$ is obtained through a constrained fit based on the minimization of the following $\chi^2$-like function:

$$
\chi^2_i = \frac{(m_{1j}^{(1)} - M_W)^2}{\Gamma_W^2} + \frac{(m_{1j}^{(2)} - M_W)^2}{\Gamma_W^2} + \frac{(m_{2j}^{(1)} - m_{\text{rec}}^{(1)})^2}{\Gamma_1^2} + \frac{(m_{2j}^{(2)} - m_{\text{rec}}^{(2)})^2}{\Gamma_1^2} + \sum_{i=1}^{6} \frac{(p_{\text{fit}, i} - p_{\text{meas}, i})^2}{\sigma_i^2}
$$

where $m_{1j}^{(1,2)}$ are the invariant masses of the two pairs of jets assigned to light flavor quarks, $m_{2j}^{(1,2)}$ are the invariant masses of the triplets including one pair and one jet assigned to a $b$ quark, $M_W = 80.4$ GeV/c$^2$ and $\Gamma_W = 2.1$ GeV are the measured mass and natural width of the $W$ boson [2], and $\Gamma_1 = 1.5$ GeV is the assumed natural width of the top quark [11]. The jet transverse momenta are constrained in the fit to the measured values, $p_{\text{fit}, i}$, within their known resolutions, $\sigma_i$. The fit is performed with respect to $m_{\text{rec}}$ and the transverse momenta of the jets $p_{\text{fit}, i}$, and, among all the permutations, the one which gives the lowest value for the minimized $\chi^2_i$ is selected. The variable $m_{\text{rec}}^i$ is reconstructed by the same procedure considered for $m_{\text{rec}}^{(1)}$, but with a $\chi^2$ function, $\chi^2_W$, where also the $W$ mass is left free to vary in the fit. The selected values of $m_{\text{rec}}$ and $m_{\text{rec}}^i$ enter the respective distributions, built separately for events with exactly one or $\geq 2$ tags.

Signal templates are built using MC events with $M_{\text{top}}$ values from 160 to 185 GeV/c$^2$, with steps of 1.0 GeV/c$^2$, and, for each value, moving the JES by $\pm 1\sigma$ from the default. Here $\sigma$ is the absolute uncertainty on the JES [10] and $\Delta$JES is a dimensionless number. Values of $\Delta$JES between $-2$ and $+2$, in steps of 0.5, have been used, and in the following we refer to this parameter to denote variations of the JES. To construct the background templates we apply the fitting technique to the data passing the neural network selection cut, omitting the $b$-tagging requirement (“pretag” sample) [3]. The weight of each value of $m_{\text{rec}}$ and $m_{\text{rec}}^i$ is given by the probability of the event to belong to the background and to contain tagged jets, evaluated by the tag rates of jets, as outlined above.

Sets of simulated experiments (“pseudo-experiments”, PEs) have been performed to optimize the requirements on the values of $N_{\text{cut}}$, $\chi^2_W$ and $\chi^2_W$ in order to minimize the statistical uncertainty on the $M_{\text{top}}$ measurement. As an improvement with respect to [3], two different sets of events, denoted by $S_{\text{JES}}$ and $S_{\text{Mtop}}$, are used to build the $m_{\text{rec}}^i$ and $m_{\text{rec}}$ templates, respectively. The set $S_{\text{JES}}$ is selected by using cuts on $N_{\text{cut}}$ and $\chi^2_W$, while $S_{\text{Mtop}}$ is selected by a further requirement on $\chi^2_W$, so that $S_{\text{Mtop}}$ corresponds to a subset of $S_{\text{JES}}$. This new procedure contributes in reducing the final total uncertainty on $M_{\text{top}}$ with respect to [3] by about 12%. Tables 1 and 2 report the flow of the event selection for 1-tag and $\geq 2$-tag events, respectively. As the final requirements are optimized separately for the two tagging categories, the $b$-tag requirement is included in the flow just after the preselection.

In order to measure $M_{\text{top}}$ with the simultaneous calibration of the JES, a fit is performed in which an unbinned extended likelihood with respect to $\Delta$JES is performed. Here $\sigma$ is the absolute uncertainty on the JES [10] and $\Delta$JES is a dimensionless number. Values of $\Delta$JES between $-2$ and $+2$, in steps of 0.5, have been used, and in the following we refer to this parameter to denote variations of the JES. To construct the background templates we apply the fitting technique to the data passing the neural network selection cut, omitting the $b$-tagging requirement (“pretag” sample) [3]. The weight of each value of $m_{\text{rec}}$ and $m_{\text{rec}}^i$ is given by the probability of the event to belong to the background and to contain tagged jets, evaluated by the tag rates of jets, as outlined above.

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In order to measure $M_{\text{top}}$ with the simultaneous calibration of the JES, a fit is performed in which an unbinned extended likelihood function is maximized to find the values of $M_{\text{top}}$, $\Delta$JES, the number of signal ($n_s$) and background ($n_b$) events for each tagging category which best reproduce the observed distributions of $m_{\text{rec}}$ and $m_{\text{rec}}^i$ [3]. The likelihood depends on the probability density functions (p.d.f.’s) of $m_{\text{rec}}$ and $m_{\text{rec}}^i$ expected for signal ($s$) and background ($b$), $P_s(m_{\text{rec}}^i|M_{\text{top}}, \Delta$JES), $P_s(m_{\text{rec}}^i|M_{\text{top}}, \Delta$JES), $P_s(m_{\text{rec}})$, and $P_b(m_{\text{rec}})$. The notation points out that the shapes of the signal p.d.f.’s are functions of the fit parameters $M_{\text{top}}$ and $\Delta$JES. This dependence is obtained by fitting the whole set of templates, initially built as histograms, Fig. 1 shows examples of signal and background templates for the $\geq 2$-tag sample, with the corresponding p.d.f.’s superimposed.

The presence of the different sets $S_{\text{JES}}$ and $S_{\text{Mtop}}$ requires the generalizations of some of the terms of the likelihood with respect to [3]. The function can be divided into three parts:

$$
L = L_{\text{1 tag}} \times L_{\geq 2 tags} \times L_{\Delta$JES$\text{const}}
$$

where $L_{\Delta$JES$\text{const}}$ is a Gaussian term constraining the JES to the nominal value (i.e. $\Delta$JES to 0) within its uncertainty:

$$
L_{\Delta$JES$\text{const}} = e^{-\frac{(\text{JES} - \text{JESconst})^2}{2\Delta$JES$^2}}
$$

Terms $L_{\text{1 tag}}$ and $L_{\geq 2 tags}$ are in turn defined as:

$$
L_{\text{1 tag}} = L_{\Delta$JES} \times L_{\text{Mtop}} \times L_{\text{Evts}} \times L_{\text{Mrec}}
$$

where, omitting the dependences on $M_{\text{top}}$ and $\Delta$JES.
Fig. 1. Templates of $m_{\text{rec}}$ for events with $\geq 2$ tags and corresponding probability density functions superimposed. Top plot: the signal p.d.f., $P_s$, for various values of $M_{\text{top}}$ and $\Delta$JES = 0. Bottom plot: the background p.d.f., $P_b$.

\[
\mathcal{L}_{\text{A}\text{JES}} = \prod_{i=1}^{N_{\text{obs}}} \frac{n_i P_s^{\text{rec}}(m_{W,i}) + n_b P_b^{\text{rec}}(m_{W,i})}{n_i + n_b},
\]

\[
\mathcal{L}_{\text{M}\text{top}} = \prod_{i=1}^{N_{\text{obs}}} \left( \lambda_i n_i P_s^{\text{rec}}(m_{t_i}) + \lambda_b n_b P_b^{\text{rec}}(m_{t_i}) \right) \prod_{i=1}^{N_{\text{obs}}} \lambda_i n_i + \lambda_b n_b,
\]

\[
\mathcal{L}_{\text{evts}} = \sum_{r_s+r_b=N_{\text{obs}}} P(r_s, n_s) \cdot P(r_b, n_b) 
\cdot \left[ \sum_{t_s+t_b=N_{\text{obs}}} B(t_s, r_s, \lambda_s) \cdot B(t_b, r_b, \lambda_b) \right]
\]

and

\[
\mathcal{L}_{\text{nbkg}} = e^{-\frac{(n_b-n_{b,\text{exp}})^2}{2n_{b,\text{exp}}}}.
\]

In the first term the probability to observe the set $m_{W,i}$ ($i = 1, \ldots, N_{\text{obs}}$) of $m_{W}$ values reconstructed in the data is calculated by the signal and background expected distributions, $P_s^{\text{rec}}$ and $P_b^{\text{rec}}$ respectively, as a function of the free parameters of the fit $M_{\text{top}}$, $\Delta$JES, $n_s$, and $n_b$. In the second the same is done for the distributions of the observed reconstructed top masses, $m_{t,i}$ ($i = 1, \ldots, N_{\text{obs}}$), and the $m_{rec}$ probability density functions. The factors $\lambda_i$ for signal and background, respectively (i.e., the fraction of events selected by the requirements on $\chi^2$ only). For the signal the value accepted as a function of the fit parameters $M_{\text{top}}$ and $\Delta$JES. The third term, $\mathcal{L}_{\text{evts}}$, gives the probability to observe simultaneously the number of events selected in the data in the $S_{\text{JES}}$ and the $S_{\text{Mtop}}$ sample, given the assumed values for the average number of signal ($n_s$) and background ($n_b$) events to be expected in $S_{\text{JES}}$ and the acceptance $\lambda_i$ for signal and background, respectively. It depends on the Poisson ($P$) and Binomial ($B$) probabilities

\[
P(r, n) = \frac{e^{-n} \cdot n^r}{r!},
\]

\[
B(t, r, \lambda) = \binom{r}{t} \cdot \lambda^t \cdot (1 - \lambda)^{r-t}.
\]

In the last term, $\mathcal{L}_{\text{nbkg}}$, the parameter $n_b$ is constrained by a Gaussian to the a priori background estimate i.e. $n_{b,\text{exp}} = 3652 \pm 181$ for 1-tag events and $n_{b,\text{exp}} = 718 \pm 14$ for $\geq 2$-tag events.

The possible presence of biases in the values returned by the likelihood fit has been investigated. Pseudo-experiments are performed assuming specific values for $M_{\text{top}}$ and $\Delta$JES and “pseudo-data” are therefore extracted from the corresponding signal and background templates. The results of these PEs have been compared to the input values, and calibration functions to be applied to the output from the fit have been defined in order to obtain, on average, a more reliable estimate of the true values and uncertainties.

Finally, the likelihood fit is applied to data. After the event selection described above, we are left with 4368 and 1196 events with one and $\geq 2$ tags (147 have 3 tags), respectively, in the $S_{\text{JES}}$ sample. The corresponding expected backgrounds amount to $3652 \pm 181$ and $718 \pm 14$ for $\geq 2$-tag events.

For these events the variables $m_{W}$ and $m_{rec}$ have been reconstructed and used as the data inputs to the likelihood fit. Once the calibration procedure has been applied, the measurements of $M_{\text{top}}$ and $\Delta$JES are

\[
M_{\text{top}} = 172.5 \pm 1.4 \text{(stat)} \pm 1.0 \text{(JES)} \text{GeV/c}^2,
\]

$\Delta$JES = $0.1 \pm 0.3 \text{(stat)} \pm 0.3 \text{(Mtop)}$.

Fig. 2 shows the measured values together with the negative log-likelihood contours whose projections correspond to one, two, and three $\sigma$ uncertainties on the values of $M_{\text{top}}$ and $\Delta$JES as obtained from the likelihood fit.

Fig. 3 shows the $m_{\text{rec}}$ and $m_{W}$ distributions for the data compared to the expected background and the signal for $M_{\text{top}}$ and
Fig. 2. Negative log-likelihood contours for the likelihood fit performed for the $M_{\text{top}}$ and $\Delta\text{JES}$ measurement. The minimum is shown along with the contours whose projections correspond to one, two, and three $\sigma$ uncertainties on the $M_{\text{top}}$ and $\Delta\text{JES}$ measurements.

Table 3

Sources of systematic uncertainty affecting the $M_{\text{top}}$ and $\Delta\text{JES}$ measurements. The total uncertainty is obtained by the quadrature sum of each contribution.

<table>
<thead>
<tr>
<th>Source</th>
<th>$\delta M_{\text{top}}$ (GeV/c$^2$)</th>
<th>$\delta \Delta\text{JES}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Residual bias</td>
<td>0.2</td>
<td>0.03</td>
</tr>
<tr>
<td>Calibration</td>
<td>0.1</td>
<td>0.01</td>
</tr>
<tr>
<td>Generator</td>
<td>0.5</td>
<td>0.21</td>
</tr>
<tr>
<td>Initial/final state radiation</td>
<td>0.1</td>
<td>0.04</td>
</tr>
<tr>
<td>$b$-jet energy scale</td>
<td>0.2</td>
<td>0.05</td>
</tr>
<tr>
<td>$b$-tag</td>
<td>0.1</td>
<td>0.01</td>
</tr>
<tr>
<td>Residual JES</td>
<td>0.4</td>
<td>–</td>
</tr>
<tr>
<td>Parton distribution functions</td>
<td>0.2</td>
<td>0.04</td>
</tr>
<tr>
<td>Multiple $p\bar{p}$ interactions</td>
<td>0.1</td>
<td>0.04</td>
</tr>
<tr>
<td>Color reconnection</td>
<td>0.3</td>
<td>0.12</td>
</tr>
<tr>
<td>Statistics of templates</td>
<td>0.3</td>
<td>0.05</td>
</tr>
<tr>
<td>Background</td>
<td>0.6</td>
<td>0.11</td>
</tr>
<tr>
<td>Trigger</td>
<td>0.2</td>
<td>0.04</td>
</tr>
<tr>
<td>Total</td>
<td>1.1</td>
<td>0.29</td>
</tr>
</tbody>
</table>

$\Delta\text{JES}$ corresponding to the measured values. The signal and background distributions are normalized to the respective yields as fitted to the data, with the 1-tag and $\geq 2$-tag contributions summed together.

Various sources of systematic uncertainties affect the $M_{\text{top}}$ and $\Delta\text{JES}$ measurements, as described in [3]. They are evaluated by performing PEs using templates built by signal samples where effects due to systematic uncertainties have been included. The differences in the average values of $M_{\text{top}}$ and $\Delta\text{JES}$ with respect to the PEs performed with default templates are then taken into account. Possible residual biases existing after the calibration, and uncertainties on the parameters of the calibration functions are also taken into account. The largest contributions come from uncertainties on the modeling of the background, on the simulation of $t\bar{t}$ events, and on the individual corrections which JES depends on [10]. Table 3 shows a summary of all the systematic uncertainties.

In summary, we have presented a measurement of the top quark mass in the all-hadronic channel, using $p\bar{p}$ collision data corresponding to an integrated luminosity of 5.8 fb$^{-1}$. An optimized event selection, based mainly on a neural network and a $b$-tagging algorithm, allows us to select candidate event samples with $S/B$ close to 1 in spite of the huge background still existing at trigger level. The simultaneous calibration of the jet energy scale, following a well-established technique, allows to reduce down to 1 GeV/c$^2$ the systematic uncertainty due to this source. The value obtained for the JES is in agreement both with the default value [10] and with the results obtained by other measurements of the top quark mass performed by the CDF Collaboration using the in situ calibration technique [4,5]. The measured value of the top quark mass is

$$M_{\text{top}} = \ldots$$
$M_{\text{top}} = 172.5 \pm 1.4 \text{(stat)} \pm 1.0 \text{(JES)} \pm 1.1 \text{(syst)} \text{ GeV}/c^2$, with a total uncertainty of 2.0 GeV/$c^2$. This result complements and is consistent with the most recent measurements obtained in other channels by the CDF and D0 Collaborations, and also represents the most accurate all-hadronic measurement at the Tevatron so far.

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