# PDFSEnse:* Mapping the sensitivity of hadronic experiments to nucleon structure 

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Figure 1: For the full CTEQ-TEA data-set (CT14HERA2), we show the absolute correlation $\left|C_{f}\right|$ and sensitivity $\left|S_{f}\right|$ associated with the 14 TeV Higgs production cross section $\sigma_{H^{0}}(14 \mathrm{TeV})$. Points with significant magnitudes of $\left|C_{f}\right|$ and $\left|S_{f}\right|$ are highlighted with color; the thresholds are chosen to highlight the same effective number of points (310) in both plots. When the $\left|C_{f}\right|$ plot at left is considered, only a very small sub-population of jet production data (diagonal open circles and closed squares with $\mu \gtrsim 100 \mathrm{GeV}$ ) exhibits significant correlations, as well as some HERA DIS and $t \bar{t}$ production data points. Conversely, the sensitivity in the right panel reveals a broader range of points constraining the Higgs cross section. Here, a larger fraction of jet production points are important (especially CMS measurements), as well as processes at smaller $\mu$, particularly DIS data from HERA and fixed-target experiments (e.g., BCDMS, NMC, CDHSW, and CCFR). Although its cumulative impact is comparatively modest, ATLAS $t \bar{t}$ production data register significant per-point sensitivities, as do CCFR $F_{p}^{2}$ measurements and CMS $7 \mathrm{TeV} A_{\mu}$ data; similarly, some of the high- $p_{T}$ $Z$ production information from ATLAS provide modest constraints.

## 1. Introduction

The determination of the nucleon's collinear parton distribution functions (PDFs) is becoming an increasingly precise discipline with the advent of high-luminosity experiments at both colliders and fixed-target facilities, and several research groups are involved in the rich research domain of modern PDF analysis. PDFs provide a description of hadronic structure, and are an essential ingredient in perturbative QCD computations. ${ }^{1}$ Since the start of the Large Hadron Collider Run II (LHC Run II), the volume of experimental data pertinent to the PDFs is growing with such speed that isolating measurements of greatest impact presents a significant challenge for PDF fitters. To help address this challenge, we present a new analysis method for identifying high-impact experiments which constrain the PDFs and the resulting Standard Model (SM) predictions that depend on them; this method is complementary to other frameworks like Hessian profiling techniques [2] and Bayesian reweighting [3, 4].

## 2. Correlations

The notion of using correlations between the PDF uncertainties of two physical observables was proposed in Refs. [5, 6] as a means of quantifying the degree to which these quantities were

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Figure 2: Two representations of the sensitivity $\left|S_{g}\right|\left(x_{i}, \mu_{i}\right)$ to the gluon PDF $g(x, \mu)$ of the experimental measurements making up the augmented CTEQ-TEA data-set; in the left panel we plot a histogram showing the distribution of sensitivities for 4021 physical measurements. In the right panel we show the $\left\{x_{i}, \mu_{i}\right\}$ map corresponding to these data within the full data-set.
related based upon their underlying PDFs. The PDF-mediated correlation $C_{f}$ can determine whether there may exist a predictive relationship between the PDF $f$ and goodness of fit to the $i^{t h}$ data point; it is defined as: ${ }^{2}$

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\begin{equation*}
C_{f} \equiv \operatorname{Corr}\left[f, r_{i}\right]=\frac{\vec{\nabla} f \cdot \vec{\nabla} r_{i}}{\Delta f \Delta r_{i}} . \tag{2.1}
\end{equation*}
$$

We have suggestively inserted $f$ and $r_{i}$ as arguments of the correlation function where $f$ is a PDF and $r_{i}$ is the residual constructed as $r_{i}=\left[T_{i}-D_{i}^{s h}\right] / s_{i}$. We take $T_{i}$ as the theory prediction, $D_{i}^{s h}$ is the datum shifted by the systematic uncertainties and $s_{i}$ is the uncorrelated uncertainty; see Ref. [1] for a compete definition including the details of correlated uncertainties.

The Hessian correlation was deployed extensively in Ref. [7] to explore implications of the CTEQ6.6 PDFs for envisioned LHC observables; in this context it proved to be instrumental for identifying the specific PDF flavors and $x$ ranges most tied to the PDF uncertainties for $W, Z, H$, and $t \bar{t}$ production cross sections as well as other processes. At the same time, $C_{f}$ alone does not fully encode the potential impact of measurements on improving PDF determinations in terms of uncertainty reduction, particularly since the correlation of Eq. (2.1) does not significantly depend on the size of experimental errors.

## 3. Sensitivity

As a remedy to these limitations, we introduce a generalization of the PDF-mediated correlations called the sensitivity $S_{f}$; this object better identifies those experimental data points that tightly constrain PDFs both by merit of their inherent precision and their ability to discriminate among PDF error fluctuations. Such an approach can aid in identifying regions of $\{x, \mu\}$ in which the PDFs are particularly constrained by physical observables.

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Figure 3: Distributions of displaced residuals $\vec{\delta}_{i}$ from the CTEQ-TEA analysis obtained by dimensionality reduction methods. Left: a 3-dimensional projection of a 10 -dimensional manifold constructed by principal component analysis (PCA). Right: a distribution from the 3-dimensional t -SNE clustering method. Blue, orange, and red colors indicate data points from DIS, vector boson production, and jet $/ t \bar{t}$ production processes, respectively.

Thus, we define the sensitivity $S_{f}$ to the $\operatorname{PDF} f$ of the $i^{\text {th }}$ point in experiment $E$ to be:

$$
\begin{equation*}
S_{f} \equiv \frac{\vec{\nabla} f \cdot \vec{\nabla} r_{i}}{\Delta f\left\langle r_{0}\right\rangle_{E}}=\frac{\Delta r_{i}}{\left\langle r_{0}\right\rangle_{E}} C_{f}, \tag{3.1}
\end{equation*}
$$

where $\Delta r_{i}$ represents the variation of the residuals across the set of Hessian error PDFs, and we normalize it to the r.m.s. residual for the experiment $E$ data-set, $\left\langle r_{0}\right\rangle_{E}$, to reduce the impact of random fluctuations in the data values.

This definition has the benefit of encoding not only the correlated relationship of $f$ with $r_{i}$, but also the comparative size of the experimental uncertainty with respect to the PDF uncertainty. For example, if new experimental data have reported uncertainties that are much tighter than the present PDF errors, these data would then register as high-sensitivity points.

In fact, in the numerical approach the user can quantify the sensitivity of data not only to individual PDF flavors, but even to specific physical observables, including the modifications due to correlated systematic uncertainties in every experiment of the expanded CTEQ-TEA analysis. For example, for 14 TeV Higgs boson production via gluon fusion $(g g \rightarrow H)$ at the LHC, the short-distance cross sections are known up to $\mathrm{N}^{3} \mathrm{LO}$ with a scale uncertainty of about $3 \%$ [8]. It has been suggested that $t \bar{\tau}$ production and high- $p_{T} Z$ boson production already provide comparable constraints on the gluon PDF in the $x$ region sensitive to LHC production, and that these are comparable to the constraints from LHC and Tevatron data [9, 10]. Verifying the degree to which this hypothesis is true has been difficult without actually including all these data in a fit.

As an alternative to doing a full global fit, we can critically assess this supposition using the Hessian correlations and sensitivities, $\left|C_{f}\right|$ and $\left|S_{f}\right|$, associated with the Higgs production cross section $\sigma_{H^{0}}$, in the context of the updated CTEQ-TEA set that includes the CT14HERA2 [11]
points and newer LHC Run I data.
Fig. 1 shows $\left|C_{f}\right|$ and $\left|S_{f}\right|$ distributions that we obtain in $\{x, \mu\}$ space. The data points that have large values of $\left|C_{f}\right|$ and $\left|S_{f}\right|$, and hence constrain the PDF dependence of $\sigma_{H^{0}}$, are highlighted with color according to the conventions described in Ref. [1]. The sensitivity measure generally identifies a different outlay of data providing constraints on $\sigma_{H^{0}}$ than the correlation, as can be seen by comparing highlighted data points in the left and right panels. Note, the thresholds are chosen to highlight the same effective number of points in both plots. In the left panel, only a very small subgroup of the inclusive jet production points, select HERA Neutral Current (NC) DIS measurements, and several ATLAS $p_{T}^{Z}$ data, show the most significant correlations, taken to have $\left|C_{f}\right|>0.42$ in this comparison. With our improved definition for the sensitivity, however, the corresponding plot in the right panel demonstrates that a different collection of points has large sensitivity to $\sigma_{H^{0}}$, with $\left|S_{f}\right|>0.21$. These data include most of the analyzed jet production data, $t \bar{t}$ and high- $p_{T} Z$ production, as well as various DIS experiments. From this comparison, one would conclude that efforts to constrain PDF-based SM predictions for Higgs production relying only on a few points of $t \bar{t}$ data would be significantly handicapped by the neglect of high-energy jet production points.

## 4. Manifold learning and dimensionality reduction

Lastly, we illustrate a possible analysis technique carried out with the help of the TensorFlow Embedding Projector software for the visualization of high-dimensional data [1,12]. We operate on a table of 4021 vectors $\vec{\delta}_{i}$ defined from theory-data residuals according to $\vec{\delta}_{i}=\delta_{i, k}=\left[r_{i}\left(f_{k}\right)-\right.$ $\left.r_{i}\left(f_{0}\right)\right] /\left\langle r_{0}\right\rangle_{E}$ with $\left\langle r_{0}\right\rangle_{E}$ representing the $r m s$-averaged residual of the central fit evaluated over experiment $E$. This quantity can be computed for the CTEQ-TEA data-set (corresponding to our total number of raw data points) and is generated by our package PDFSENSE and uploaded to the Embedding Projector website. As variations along many eigenvector directions result only in small changes to the PDFs, the 56-dimensional $\vec{\delta}_{i}$ vectors can in fact be projected onto an effective manifold spanned by fewer dimensions. Specifically, the Embedding Projector approximates the 56-dimensional manifold by a 10-dimensional manifold using principal component analysis (PCA). In practice, this 10 -dimensional manifold is constructed out of the 10 components of greatest variance in the effective space, such that the most variable combinations of $\delta_{i, l}$ are retained, while the remaining 46 components needed to fully reconstruct the original 56-dimensional $\vec{\delta}_{i}$ are discarded. However, because the 10 PCA-selected components describe the bulk of the variance of $\delta_{i, l}$, the loss of these 46 components results in only a minimal relinquishment of information, and in fact provides a more efficient basis to study $\delta_{i, l}$ variations.

In the 10 -dimensional PCA representation, some directions result in efficient separation of residuals of different types. For example, the left panel of Fig. 3 shows a 3-dimensional projection of the $\vec{\delta}_{i}$ that separates clusters of DIS, vector boson production, and jet $t \bar{t}$ production residuals. In this example, the jet $/ t \bar{t}$ cluster, shown in red, is roughly orthogonal to the blue DIS cluster and intersects it. This separation is remarkable, as it is based only on numerical properties of the $\vec{\delta}_{i}$ vectors, and not on the meta-data about the types of experiments that is entered after the PCA is completed.

As an alternative, the Embedding Projector can organize the $\vec{\delta}_{i}$ vectors into clusters according to their similarity using $t$-distributed stochastic neighbor embedding (t-SNE). A representative

3-dimensional distribution of the vectors obtained by t-SNE is displayed in the right panel of Fig. 3. In this case, we find that such algorithms can again sort data into clusters according to the experimental process, values of $x$ and $\mu$, and even the experiment itself; for other examples, including animations, see Ref. [12].

## 5. Conclusions

We have confronted the modern challenge of a rapidly growing set of global QCD data with new statistical methodologies for quantifying and exploring the impact of this information. These novel methodologies are realized in a new analysis tool PDFSENSE[1], which allows the rapid exploration of the impact of both existing and potential data on PDF determinations. Crucial to this analysis is introduction of the sensitivity $S_{f}$ which serves as a particularly powerful discriminator; both this and the correlation $C_{f}$ allow us to visualize PDF constraints provided by data across a wide range in $\{x, \mu\}$. While we have demonstrated these techniques in the context of CT14 family of global fits, they are of sufficient generality that one could readily repeat our analysis using alternative PDF sets. These various tools collectively suggest a number of possible avenues to advance PDF knowledge in the coming years.

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[^0]:    *The webpage for the PDFSense tool is: https://metapdf.hepforge.org/PDFSense/
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[^1]:    ${ }^{1}$ See Ref. [1] for additional details and a complete set of references.

[^2]:    ${ }^{2}$ Here, the gradients $\vec{\nabla} f$ and r.m.s. sums $\Delta f$ are computed in the PDF parameter space.

