





Strange distributions and fragmentation functions  
Nucleon tomography (GPDs)  
Quark orbital angular momentum (TMDs)  
Quark hadronization in the nuclear medium  
Hadron spectroscopy and search for exotic mesons

- 1 The NNPDF methodology
  - Monte Carlo sampling, Neural Network parametrization and reweighting
- 2 Unpolarized PDFs: NNPDF2.3
  - with emphasis on the strangeness content of the proton
- 3 Polarized PDFs: NNPDFpol1.0
  - including recent developments
- 4 Conclusions

## DISCLAIMER

Not a systematic review of recent developments in PDF analyses

[\[arXiv:1301.6754\]](#) [\[arXiv:1211.5142\]](#) [\[arXiv:1306.6515\]](#)

Rather, partial view on PDFs from the NNPDF side  
focusing on strangeness

# 1. The NNPDF methodology

# The NNPDF methodology in a nutshell

- 1 Generate  $N_{\text{rep}}$  **Monte Carlo** replicas of the experimental data
  - multi-Gaussian probability distribution
  - take into account all experimental correlations
- 2 Fit a set of Parton Distribution Functions (PDFs) parametrized at initial scale
  - with **Neural Networks**
  - to each replica

Resulting PDF replicas are equally probable members of a **statistical ensemble** which samples the probability density  $\mathcal{P}[f_i]$  in the space of PDFs

$$\langle \mathcal{O} \rangle = \int \mathcal{D}f_i \mathcal{P}[f_i] \mathcal{O}[f_i]$$

**Expectation values** for observables are **Monte Carlo integrals**

$$\langle \mathcal{O}[f_i(x, Q^2)] \rangle = \frac{1}{N_{\text{rep}}} \sum_{k=1}^{N_{\text{rep}}} \mathcal{O}[f_i^{(k)}(x, Q^2)]$$

and corresponding formulae for the estimators of Monte Carlo samples are used to compute uncertainties, correlations, etc.

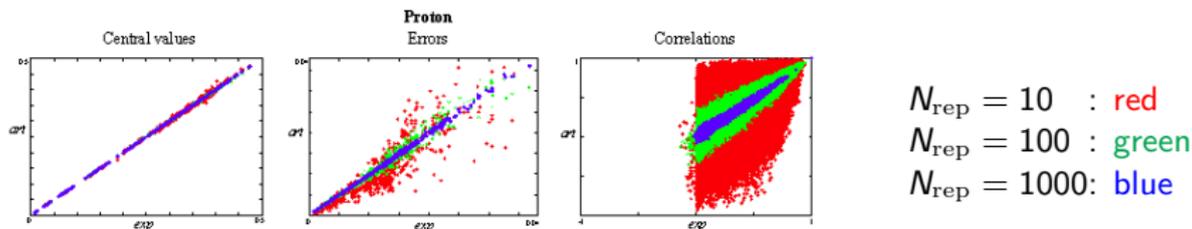
# The NNPDF methodology: Monte Carlo sampling

- Monte Carlo replicas are **generated** according to the distribution

$$\mathcal{O}_i^{\text{art}(k)} = \left(1 + r_N^{(k)} \sigma_N\right) \left[ \mathcal{O}_i^{(\text{exp})} + \sum_{p=1}^{N_{\text{sys}}} r_p^{(k)} \sigma_{i,p} + r_{i,s}^{(k)} \sigma_{i,s} \right]$$

where  $i = 1, N_{\text{dat}}$ ,  $k = 1, N_{\text{rep}}$  and  $r^{(k)}$  are (Gaussianly distributed) random numbers

- Monte Carlo replicas are **validated** against experimental data

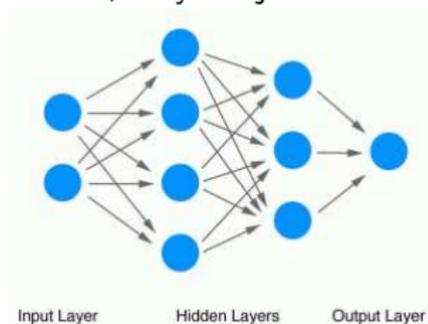


$N_{\text{rep}} \sim 100$  to reproduce central values and errors,  $N_{\text{rep}} \sim 1000$  to reproduce correlations to percent accuracy

**No** need to rely on **linear propagation** of errors  
Possibility to test the impact of **non-Gaussianly** distributed **uncertainties**  
Possibility to test for **non-Gaussian behaviour** of uncertainties of fitted **PDFs**

# The NNPDF methodology: Neural Networks

- **Neural Networks** provide us with a **redundant parametrization** for PDFs
- Only require **smoothness** of the fitted function
- In the end, they are just an other basis of functions



$$\xi_i^{(l)} = g \left( \sum_j^{n_{l-1}} \omega_{ij}^{(l-1)} \xi_j^{(l-1)} - \theta_i^{(l)} \right)$$

$$g(x) = \frac{1}{1 + e^{-x}}$$

- Use **genetic algorithms** for minimization owing to high dimensional parameter space
- Avoid to fit statistical fluctuations (overlearning) with proper **stopping criterion**

Neural Networks **reduce** the theoretical **bias** due to the parametrization  
**Minimization** and **stopping** algorithms ensure **proper PDF fitting**

# The NNPDF methodology: reweighting [arXiv:1012.0836] [arXiv:1108.1758]

- We would like to assess the impact of including a **new data set**  $\{y\} = \{y_1, \dots, y_n\}$  (delivered with  $\sigma_{ij}$ ) in a **prior ensemble** of PDF replicas  $\{f_k\}$ ,  $k = 1, \dots, N_{\text{rep}}$
- We can apply **Bayes theorem** to determine the conditional probability of PDF upon inclusion of the new data and update the probability density in the space of PDFs

$$\mathcal{P}_{\text{new}} = \mathcal{N}_x \mathcal{P}(\chi_k^2 | \{f_k\}) \mathcal{P}_{\text{old}}(\{f_k\}) \quad \mathcal{P}(\chi_k^2 | \{f_k\}) = [\chi_k^2(\{y\}, \{f_k\})]^{\frac{1}{2}(n-1)} e^{-\frac{1}{2}\chi_k^2(\{y\}, \{f_k\})}$$

$$\chi_k^2(\{y\}, \{f_k\}) = \sum_{i,j}^n \{y_i - y_i[f_k]\} \sigma_{ij} \{y_j - y_j[f_k]\}$$

- Replicas are **no longer equally probable**. Expectation values are given by

$$\langle \mathcal{O}[f_i(x, Q^2)] \rangle_{\text{new}} = \sum_{k=1}^{N_{\text{rep}}} w_k \mathcal{O}[f_i^{(k)}(x, Q^2)]$$

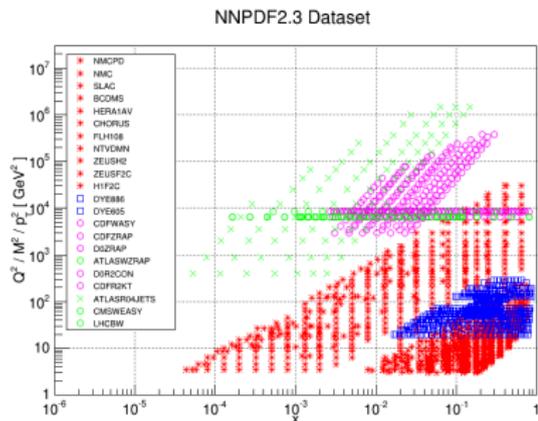
$$w_k \propto [\chi_k^2(\{y\}, \{f_k\})]^{\frac{1}{2}(n-1)} e^{-\frac{1}{2}\chi_k^2(\{y\}, \{f_k\})} \quad \text{with} \quad N_{\text{rep}} = \sum_{k=1}^{N_{\text{rep}}} w_k$$

**Reweighting** allows to incorporate new datasets **without** need of **refitting**

## 2. Unpolarized PDFs: NNPDF2.3

[[arXiv:1207.1303](https://arxiv.org/abs/1207.1303)]

# Experimental data set



Experiment	$N_{\text{dat}}$
Fixed-target DIS	926
Fixed-target neutrino DIS	1026
HERA DIS	834
Fixed-target DY	318
Tevatron W/Z	70
Tevatron Jets	186
LHC W/Z	56
LHC Jets	90

3506 data points

## ● New data in 2.3

- **ATLAS 2010** Inclusive Jets,  $36 \text{ pb}^{-1}$  [arXiv:1112.6297]
- **ATLAS 2010**  $W^\pm/Z$  distributions,  $36 \text{ pb}^{-1}$  [arXiv:1109.5141]
- **LHCb 2010**  $W^\pm$  rapidity distributions,  $36 \text{ pb}^{-1}$  [arXiv:1204.1620]
- **CMS 2011**  $W$  lepton asymmetry,  $840 \text{ pb}^{-1}$  [arXiv:1206.2598]

## ● NNPDF2.3 family

- **NNPDF2.3** - global data set (including LHC data)
- **NNPDF2.3 Collider** - HERA, Tevatron and LHC data sets only
- **NNPDF2.3 QED** - also includes QED corrections and extraction of photon PDF
- ...

# Which experimental data sets can constrain strangeness?

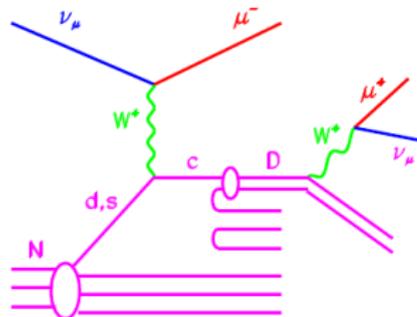
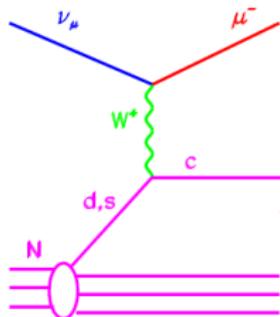
## Data included in NNPDF2.3

- 1 Inclusive neutrino DIS (CHORUS [[Phys. Lett. B632 \(2006\) 65](#)])

$$\nu_\mu(\bar{\nu}_\mu)N \rightarrow \mu^-(\mu^+)X$$

- 2 Dimuon production in DIS (NuTeV [[arXiv:hep-ex/0102049](#)])

$$\nu_\mu(\bar{\nu}_\mu)N \rightarrow \mu^-\mu^+(\mu^+\mu^-)X$$



## Data not included in NNPDF2.3

- 1  $W$  boson production in association with charm at LHC (CMS [[arXiv:1310.1138](#)])

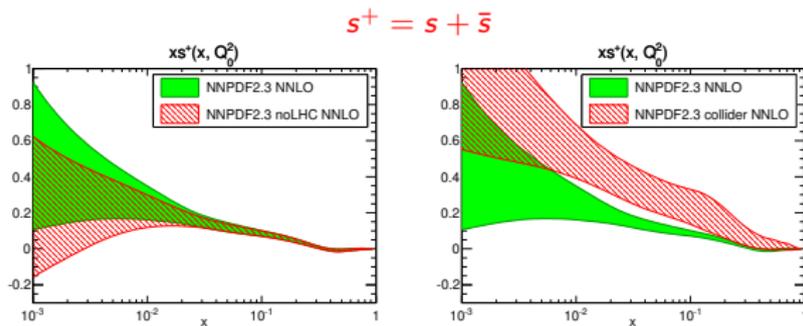
$$pp \rightarrow W + c + X$$

- 2 Kaon electroproduction data from SIDIS on deuteron (HERMES [[arXiv:0803.2993](#)])

see talk by K. Rith

# The strange content of the proton

Both total  $s^+ = s + \bar{s}$  and valence  $s^- = s - \bar{s}$  strangeness are extracted from data  
Only MSTW08 [arXiv:0901.0002] also provides independent parametrization of  $s^-$

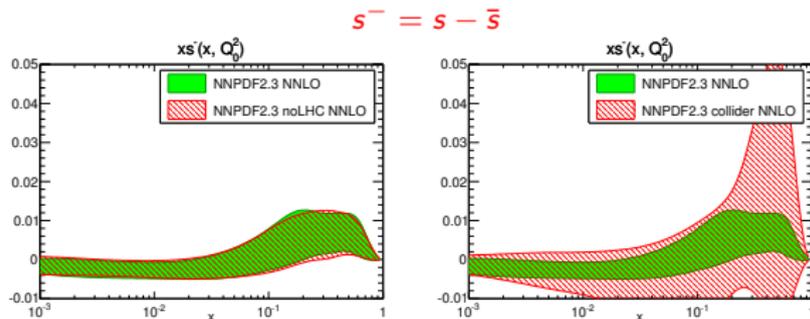


- Moderate impact of LHC data on strangeness (both  $s^+$  and  $s^-$ )
- **NNPDF2.3 Collider** prefers larger values for total strangeness, with larger uncertainties (lack of neutrino DIS data which provide in turn more constraint on strangeness)
- No experimental constraints below  $x \sim 10^{-2}$
- Theoretical constraint from **valence sum rule**  $\int_0^1 dxs^-(x) = 0$   
(does not imply symmetric strangeness  $s(x) = \bar{s}(x)$ )

A **global fit** to all available data is **needed** to determine both  $s^+$  and  $s^-$

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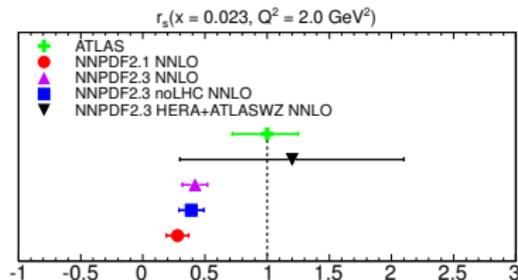
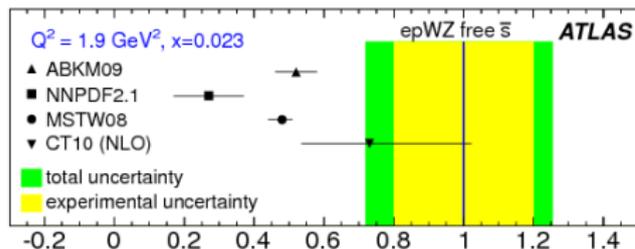


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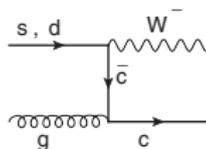
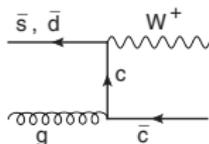
# The strange content of the proton

$$r_s(x, Q^2) = \frac{s(x, Q^2) + \bar{s}(x, Q^2)}{2\bar{d}(x, Q^2)}$$

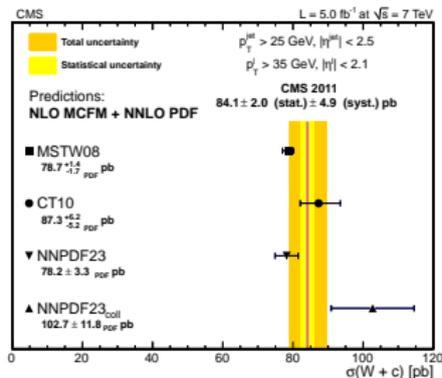
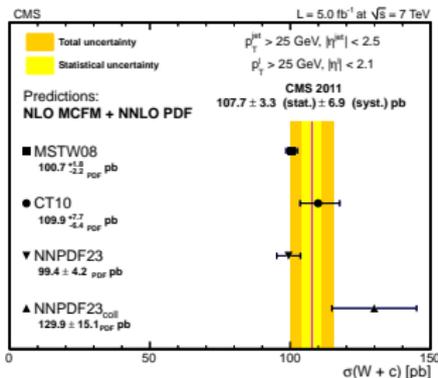


- Evidence for strangeness larger than hitherto thought from ATLAS [[arXiv:1203.4051](https://arxiv.org/abs/1203.4051)]
- This result, based on HERA+ATLAS data, is inconsistent at the  $2\sigma$  level with **NNPDF2.1**
- **NNPDF2.3 HERA+ATLASWZ** finds that these data sets cannot determine strangeness with sufficient accuracy to make any reliable conclusion
- The uncertainty of the ATLAS result is likely to be underestimated
- Strangeness in **NNPDF2.3** is accurately determined, though LHC data have slight impact

# The strange content of the proton



- $W + c$  measurements are directly sensitive to separated  $s$  and  $\bar{s}$ : CMS [[arXiv:1310.1138](https://arxiv.org/abs/1310.1138)]
- Few percent contribution from (Cabibbo-suppressed)  $dg \rightarrow W^- + c$  and  $\bar{d}g \rightarrow W^+ + \bar{c}$

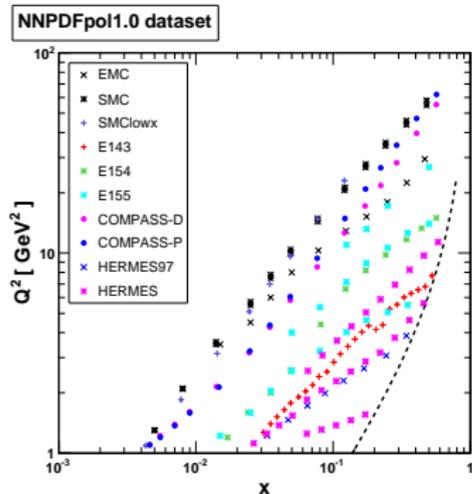


- Nice consistency between measured total cross sections and theoretical expectations
- Larger uncertainties from **NNPDF2.3 Collider** arise from the lack of direct constraints on strangeness in a collider-only fit (no neutrino DIS charm production data are included)

# 3. Polarized PDFs: NNPDFpol1.0

[[arXiv:1303.7236](https://arxiv.org/abs/1303.7236)]

# Experimental data set



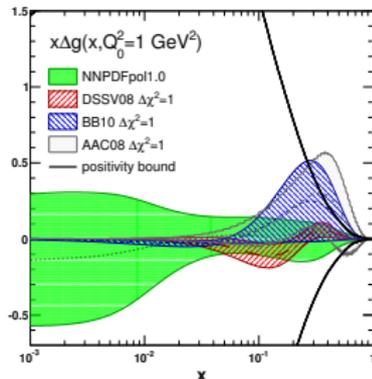
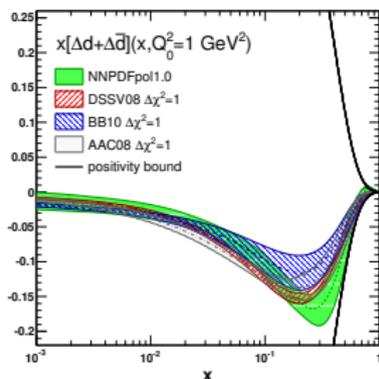
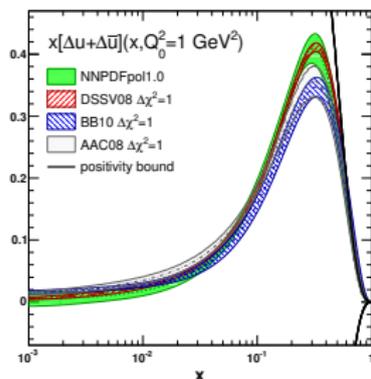
- Include all available data on **inclusive DIS** structure function  $g_1^{p,d,n}$
- **Kinematical cut** imposed to remove sensitivity to dynamical higher-twist contributions  $W^2 = Q^2(1-x)/x \geq 6.25 \text{ GeV}^2$   
no JLab data included (see [arXiv:1310.3734])
- **Initial scale**  $Q_0^2 = 1 \text{ GeV}^2$ , data included down to  $Q^2 > 1 \text{ GeV}^2$

245 data points

Inclusive (NC) DIS does not allow us to disentangle the contributions from  $q$  and  $\bar{q}$   
Choose a basis of **four polarized PDFs** (gluon + linear combinations of light quarks)  
e.g.  $\{\Delta\Sigma; \Delta T_3; \Delta T_8; \Delta g\}$  or  $\{\Delta u + \Delta\bar{u}; \Delta d + \Delta\bar{d}; \Delta s + \Delta\bar{s}; \Delta g\}$

**Positivity** and **integrability** play a substantial role due to loose constraints from data

# The NNPDFpol1.0 parton set at $Q_0^2 = 1 \text{ GeV}^2$



## $\Delta u + \Delta \bar{u}$ and $\Delta d + \Delta \bar{d}$

- Central values in reasonable agreement with those of other parton sets
- Uncertainties slightly larger for NNPDF than for other sets (notice that DSSV08 fit is based on a much wider dataset)
- Where no data/theoretical constraints available, uncertainties are larger

## $\Delta g$

- Central value compatible with zero
- Uncertainty much larger than any other set, especially in the low- $x$  region

**DSSV08**

[arXiv:0904.3821]

DIS+SIDIS+pp  
( $\pi^0$ /jet)

**BB10**

[arXiv:1005.3113]

DIS only

**AAC08**

[arXiv:0808.0413]

DIS+pp ( $\pi^0$ )

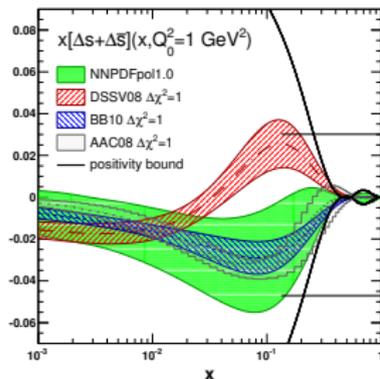
# The NNPDFpol1.0 parton set at $Q_0^2 = 1 \text{ GeV}^2$

NNPDFpol1.0 [arXiv:1303.7236]  
 $\int_0^1 dx [\Delta s + \Delta \bar{s}] = -0.13 \pm 0.09$

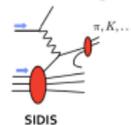
Lattice [arXiv:1112.3354]  
 $\int_0^1 dx [\Delta s + \Delta \bar{s}] = -0.020(10)(1)$

First moment constrained by  
 $a_3 = \int_0^1 dx [\Delta u^+ - \Delta d^+]$   
 $= 1.2701 \pm 0.0025$

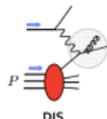
$a_8 = \int_0^1 dx [\Delta u^+ + \Delta d^+ - 2\Delta s^+]$   
 $= 0.585 \pm 0.025$



directly from SIDIS Kaon data



indirectly from DIS scaling violations



## $\Delta s + \Delta \bar{s}$

- Good agreement with BB10 and AAC08, but larger uncertainty for almost all  $x$  values
- Inconsistency at  $2\sigma$  with DSSV08 ( $\Delta\chi^2 = 1$ ) in the medium-small  $x$  region (SIDIS data)

## Is there mounting tension between DIS and SIDIS data?

- Inclusion of SIDIS data requires the knowledge of the fragmentation  $s \rightarrow K$   
 $\rightarrow$  how well do we know the kaon fragmentation function?
- Baryon octet decay constants fix the first moments of non-singlet PDF combinations  
 $\rightarrow$  conservative 30% uncertainty estimation on  $a_8$  to allow for  $SU(3)$  symmetry violation  
 $\rightarrow$  lattice finds large  $SU(3)$  symmetry violation [arXiv:1112.3354]

# May an EIC help to constrain strangeness?

We would like to assess the impact of EIC data on PDFs

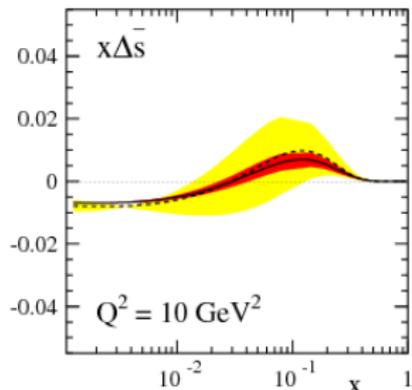
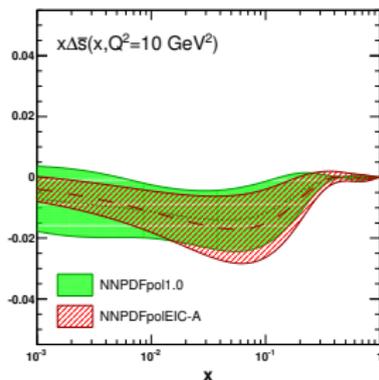
- Include DIS pseudodata at an EIC in our QCD analysis
- Pseudodata generated from DSSV+ global fit
- RHIC option of an EIC:  
 $E_p = 100 - 200$  GeV,  $E_e = 5$  GeV,  $\mathcal{L} = 10 \text{ fb}^{-1}$ , 70% beam polarization

## 1 NNPDF [arXiv:1310.0461]

- Only inclusive DIS was considered
- The shape of the strangeness does not change
- Slight reduction of the uncertainty at low- $x$  values
- Other PDFs show more significant improvement particularly the gluon, which acquires shape dictated by pseudodata

## 2 DSSV [arXiv:1206.6014]

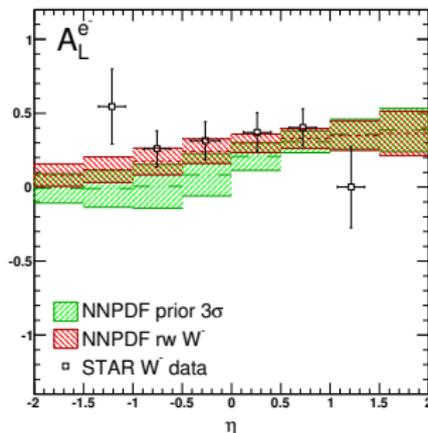
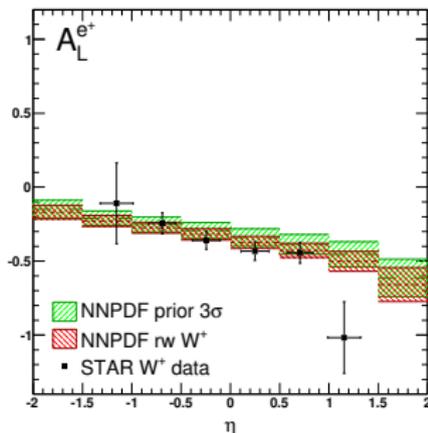
- Semi-inclusive DIS was also considered
- Substantial reduction of the uncertainty  
all uncertainties are shown for  $\Delta\chi^2 = 9$
- The shape is dictated by SIDIS kaon pseudodata  
important role of parton to Kaon fragmentation  
[see talks on Wednesday morning](#)



# Sea quarks from $W$ boson production at RHIC

$$A_L^{W^+} \sim \frac{-\Delta u(x_1)\bar{d}(x_2) + \Delta\bar{d}(x_1)u(x_2)}{u(x_1)\bar{d}(x_2) + \bar{d}(x_1)u(x_2)}$$

$$A_L^{W^-} \sim \frac{-\Delta d(x_1)\bar{u}(x_2) + \Delta\bar{u}(x_1)d(x_2)}{d(x_1)\bar{u}(x_2) + \bar{u}(x_1)d(x_2)}$$

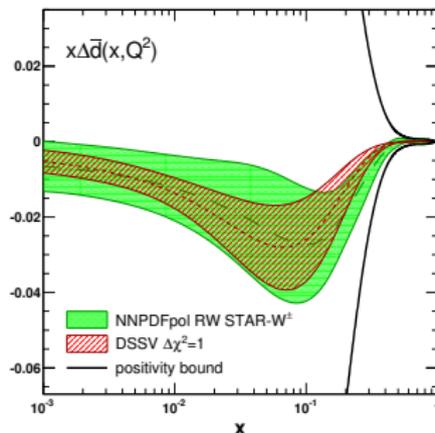
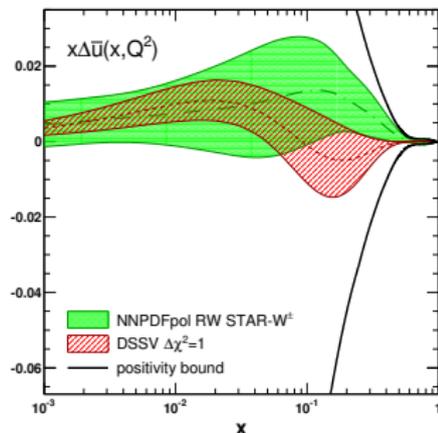


- Provide an handle on  $\Delta u$ ,  $\Delta\bar{u}$ ,  $\Delta d$ ,  $\Delta\bar{d}$  separation, independent from SIDIS (purely weak process coupling  $q_L$  with  $\bar{q}_R$  at partonic level,  $u_L\bar{d}_R \rightarrow W^+$  or  $d_L\bar{u}_R \rightarrow W^-$ )
- Preliminary results from STAR [[arXiv:1302.6639](https://arxiv.org/abs/1302.6639)] included using Bayesian reweighting

# Sea quarks from $W$ boson production at RHIC

$$A_L^{W^+} \sim \frac{-\Delta u(x_1)\bar{d}(x_2) + \Delta\bar{d}(x_1)u(x_2)}{u(x_1)\bar{d}(x_2) + \bar{d}(x_1)u(x_2)}$$

$$A_L^{W^-} \sim \frac{-\Delta d(x_1)\bar{u}(x_2) + \Delta\bar{u}(x_1)d(x_2)}{d(x_1)\bar{u}(x_2) + \bar{u}(x_1)d(x_2)}$$



- $W^\pm$  data are effective in constraining light sea quark distributions  $\Delta\bar{u}$  and  $\Delta\bar{d}$
- Reweighting with separate  $W^+$  and  $W^-$  datasets shows that  $\Delta\bar{u}$  and  $\Delta\bar{d}$  behaviour is driven by experimental information from  $W^-$  and  $W^+$  respectively
- All other PDFs are almost unaffected (including strangeness)

# 4. Conclusions

## 1 Unpolarized PDFs:

- Strangeness is mainly constrained by neutrino DIS data (inclusive and dimuon)
- LHC  $W + c$  data will allow for a direct extraction of separate  $s$  and  $\bar{s}$  distributions
- Ongoing analysis towards NNPDF3.0 including new data sets from LHC

## 2 Polarized PDFs:

- Inclusive DIS data only provide indirect constraint on strangeness
- Further constraints are provided by SIDIS  
(but we need an accurate determination of kaon FF)

## 3 Most updated NNPDF ensembles available within LHAPDF interface or at

<https://nnpdf.hepforge.org/>

together with stand-alone Fortran/C++/Mathematica code for handling NNPDFs

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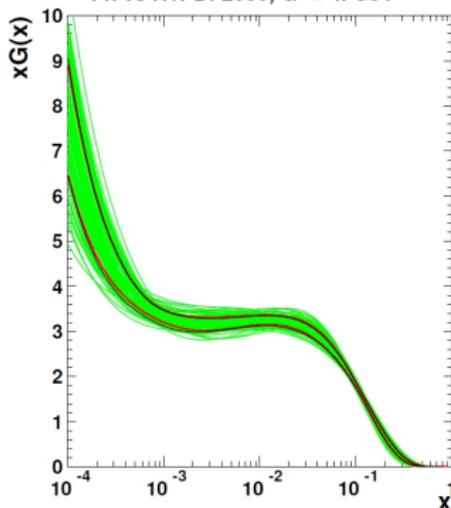
**Thank you for your attention!**

# 5. Backup

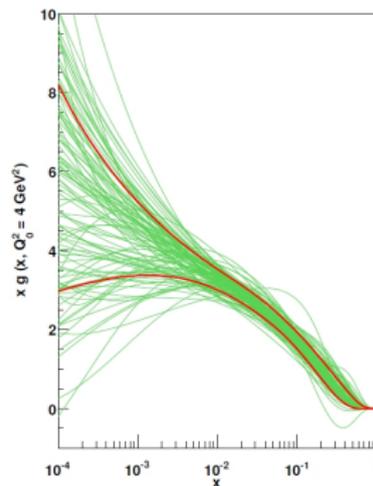
# Simple functional forms vs Neural Networks

## HERA-LHC 2009 PDF benchmarks

Fit vs H1PDF2000,  $Q^2 = 4. \text{ GeV}^2$



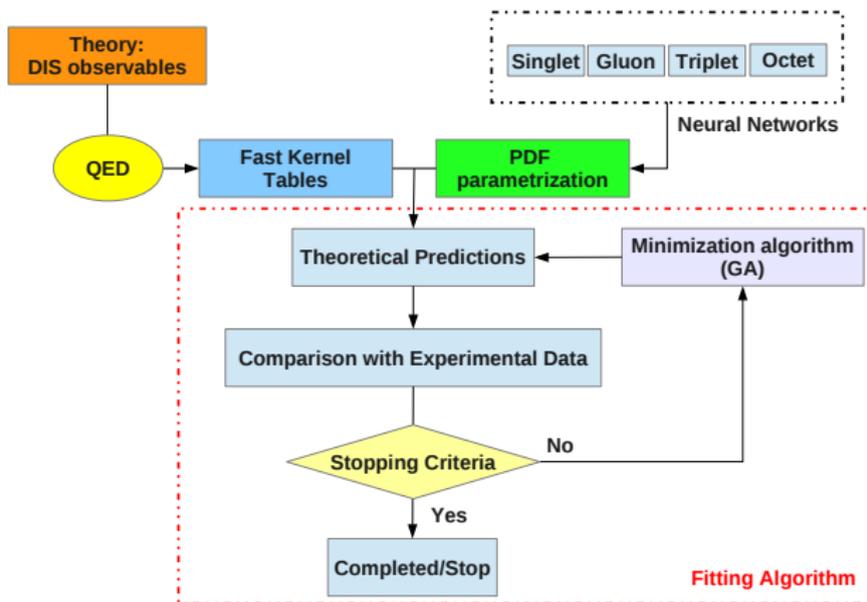
simple functional forms



Neural Networks

- Simple functional forms  $\Delta q(x) = Ax^b(1-x)^c P(x)$   
→ systematic underestimation of uncertainties  $\Rightarrow$  tolerance
- Artificial Neural Networks as universal interpolants  
→ reduce theoretical bias from choice of PDF functional form

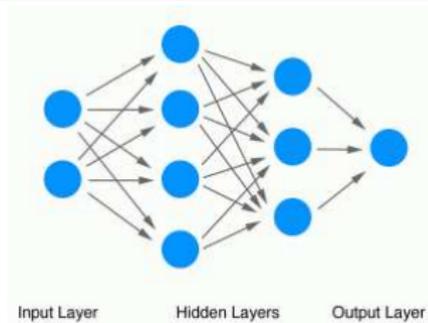
# A general overview on the methodology



Ingredients:

Monte Carlo sampling and Neural Networks

A convenient **functional form**  
providing **redundant** and **flexible** parametrization  
used as a generator of random functions in the PDF space



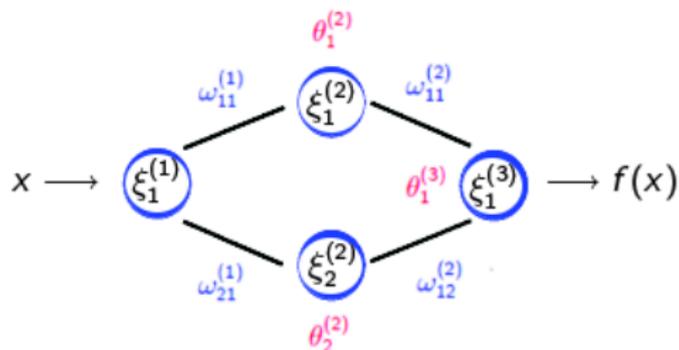
$$\xi_i^{(l)} = g \left( \sum_j^{n_{l-1}} \omega_{ij}^{(l-1)} \xi_j^{(l-1)} - \theta_i^{(l)} \right)$$

$$g(x) = \frac{1}{1 + e^{-x}}$$

- made of neurons grouped into layers (define the architecture)
- each neuron receives input from neurons in preceding layer (feed-forward NN)
- activation determined by parameters (**weights** and **thresholds**)
- activation determined according to a **non-linear function** (except the last layer)

# Neural Networks

## EXAMPLE: THE SIMPLEST 1-2-1 NN



$$f(x) \equiv \xi_1^{(3)} = \left\{ 1 + \exp \left[ \theta_1^{(3)} - \frac{\omega_{11}^{(2)}}{1 + e^{\theta_1^{(2)} - x\omega_{11}^{(1)}}} - \frac{\omega_{12}^{(2)}}{1 + e^{\theta_2^{(2)} - x\omega_{21}^{(1)}}} \right] \right\}^{-1}$$

$$\text{Recall: } \xi_i^{(l)} = g \left( \sum_j^{n_{l-1}} \omega_{ij}^{(l-1)} \xi_j^{(l-1)} - \theta_i^{(l)} \right); \quad g(x) = \frac{1}{1 + e^{-x}}$$

# One more ingredient: minimization and stopping

## GENETIC ALGORITHM

Standard minimization unefficient owing to the large parameter space and non-local  $x$ -dependence of the observables  
Genetic algorithm provides better exploration of the whole parameter space

- Set Neural Network parameters randomly
- Make clones of the parameter vector and mutate them
- Define a **figure of merit** or error function for the  $k$ -th replica

$$E^{(k)} = \frac{1}{N_{\text{rep}}} \sum_{i,j=1}^{N_{\text{rep}}} \left( g_{1,i}^{(\text{art})^{(k)}} - g_{1,i}^{(\text{net})^{(k)}} \right) \left( (\text{cov})^{-1} \right)_{ij} \left( g_{1,j}^{(\text{art})^{(k)}} - g_{1,j}^{(\text{net})^{(k)}} \right)$$

$g_{1,i}^{(\text{art})^{(k)}}$ : generated from Monte Carlo sampling

$g_{1,i}^{(\text{net})^{(k)}}$ : computed from Neural Network PDFs

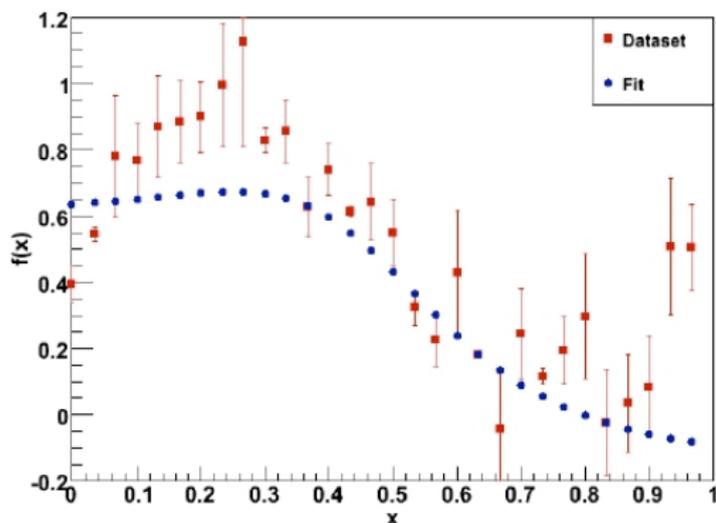
- Select the best set of parameters and perform other manipulations (crossing, mutating, ...) until stability is reached.

# Minimization and stopping

## DRAWBACK

- NN can learn fluctuations owing to their flexibility

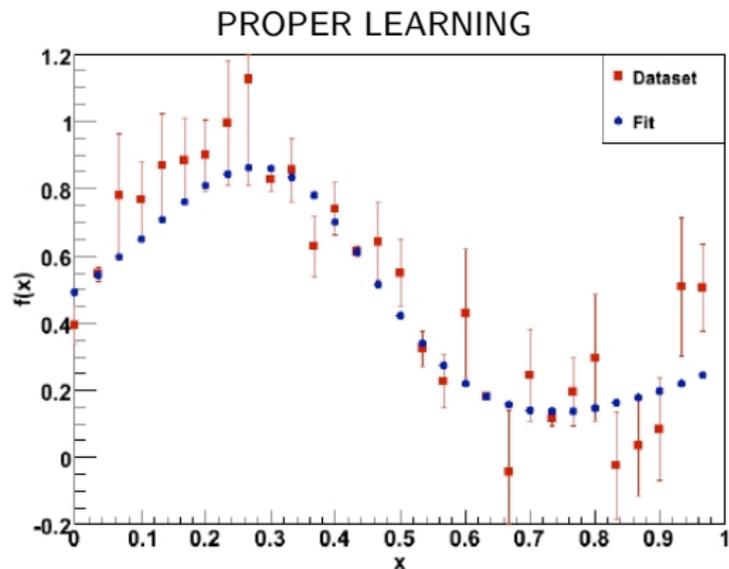
### UNDERLEARNING



# Minimization and stopping

## DRAWBACK

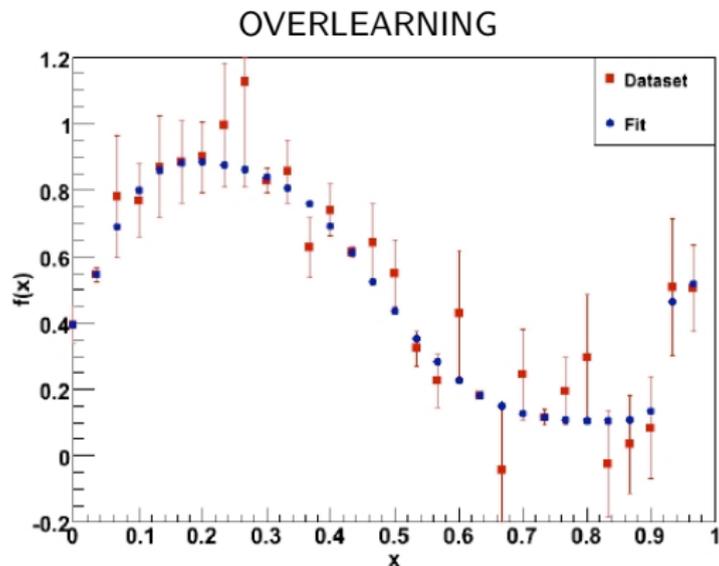
- NN can learn fluctuations owing to their flexibility



# Minimization and stopping

## DRAWBACK

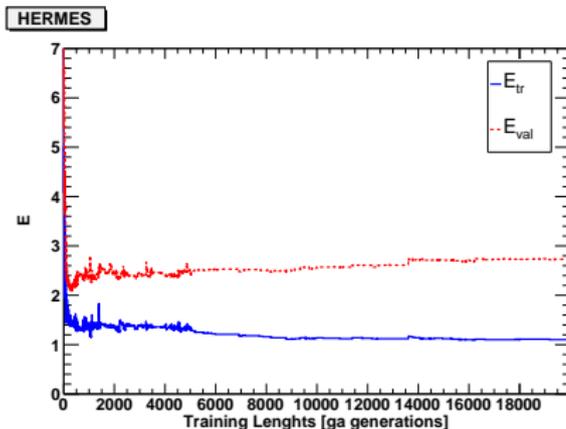
- NN can learn fluctuations owing to their flexibility



# Minimization and stopping

## CROSS-VALIDATION METHOD

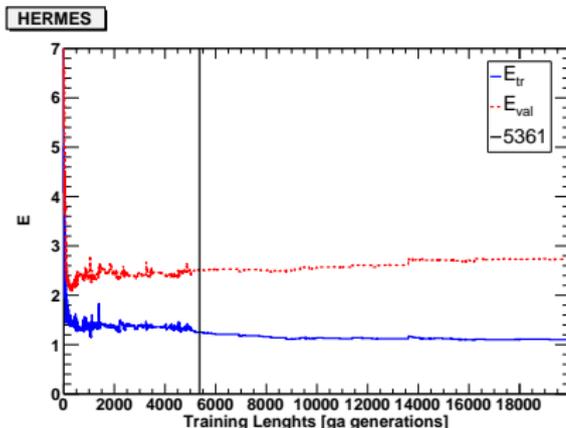
- divide data into two subsets (**training** & **validation**)
- train the NN on training subset and compute  $\chi^2$  for each subset
- stop when  $\chi^2$  of validation subset no longer decreases (NN are learning fluctuations!)



# Minimization and stopping

## CROSS-VALIDATION METHOD

- divide data into two subsets (**training** & **validation**)
- train the NN on training subset and compute  $\chi^2$  for each subset
- stop when  $\chi^2$  of validation subset no longer decreases (NN are learning fluctuations!)



The best fit does not coincide with the  $\chi^2$  absolute minimum

# NNPDF timeline

	2008		2009		2010		2012	
	NNPDF1.0	NNPDF1.2	NNPDF2.0	NNPDF2.1	NNPDF2.3	MSTW08	CT10	
DIS	✓	✓	✓	✓	✓	✓	✓	
Drell-Yan	✗	✗	✓	✓	✓	✓	✓	
Jet	✗	✗	✓	✓	✓	✓	✓	
LHC	✗	✗	✗	✗	✓	✗	✗	
strange	✗	✓	✓	✓	✓	✓	✓	
Heavy Quark mass	✗	✗	✗	✓	✓	✓	✓	
NNLO	✗	✗	✗	✓	✓	✓	✓	

## Seven polarized PDFs (gluon + singlet and non-singlet quark combinations)

- 1 singlet  $\Sigma(x) \equiv \sum_{i=1}^{n_f} q_i(x)$
  - 2 gluon  $g(x)$
  - 3 total valence  $V(x) \equiv u(x) - \bar{u}(x) + d(x) - \bar{d}(x)$
  - 4 triplet  $T_3(x) \equiv u(x) - d(x)$
  - 5 sea asymmetry  $\Delta_S(x) \equiv \bar{d}(x) - \bar{u}(x)$
  - 6 total strangeness  $s^+(x) = s(x) + \bar{s}(x)$
  - 7 valence strangeness  $s^-(x) = s(x) - \bar{s}(x)$
- At **initial scale**  $Q_0^2 = 2\text{GeV}^2$

## Four polarized PDFs (gluon + linear combinations of light quarks)

- 1 singlet  $\Delta\Sigma(x) \equiv \sum_{i=1}^{n_f} \Delta q_i(x)$
- 2 gluon  $\Delta g(x)$
- 3 triplet  $\Delta T_3(x) \equiv \Delta u(x) - \Delta d(x)$
- 4 octet  $\Delta T_8(x) \equiv \Delta u(x) + \Delta d(x) - 2\Delta s(x)$

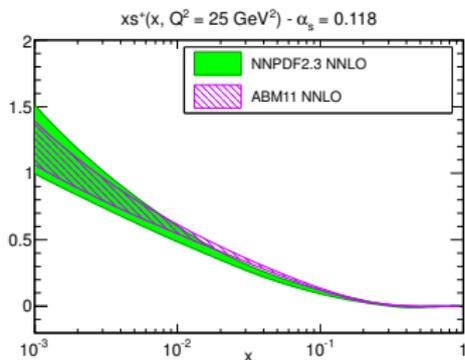
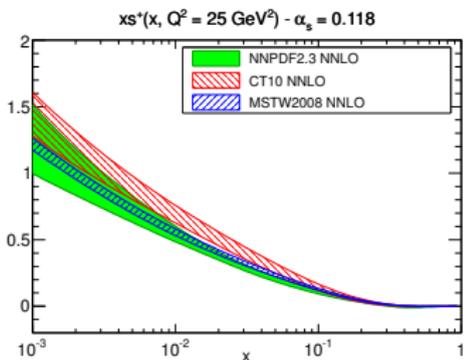
$$\Delta q_i(x, Q^2) = q_i^{\uparrow\uparrow}(x, Q^2) + \bar{q}_i^{\uparrow\uparrow}(x, Q^2) - q_i^{\uparrow\downarrow}(x, Q^2) + \bar{q}_i^{\uparrow\downarrow}(x, Q^2)$$

$$\Delta g(x, Q^2) \equiv g^{\uparrow\uparrow}(x, Q^2) - g^{\uparrow\downarrow}(x, Q^2)$$

Inclusive neutral-current DIS data do not allow disentangling the contributions from  $q$  and  $\bar{q}$ .  
In our notation,  $\Delta q$  takes into account flavor plus anti-flavor contributions.

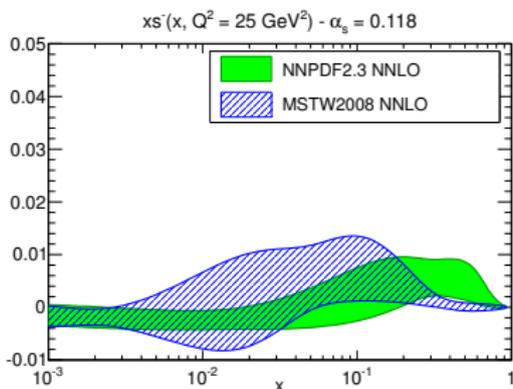
- At **initial scale**  $Q_0^2 = 1 \text{ GeV}^2$
- Assume all heavy quarks are generated radiatively
- Adopt  $\alpha_s(M_Z^2) = 0.119$ ,  $m_c = 1.4 \text{ GeV}$ ,  $m_b = 4.75 \text{ GeV}$

$$s^+ = s + \bar{s}$$



- HERAPDF1.5 does not provide independent strangeness, hence it is not shown (HERA data alone do not allow for disentangling of the strange contribution)
- The CT10 strange distribution is somewhat higher than that of other groups (different non-perturbative parametrization of the PDFs) (differences in the heavy quark treatment of neutrino dimuon data)
- Recall that NNPDF2.3 also includes LHC data (however, they were shown to have slight impact on strangeness)

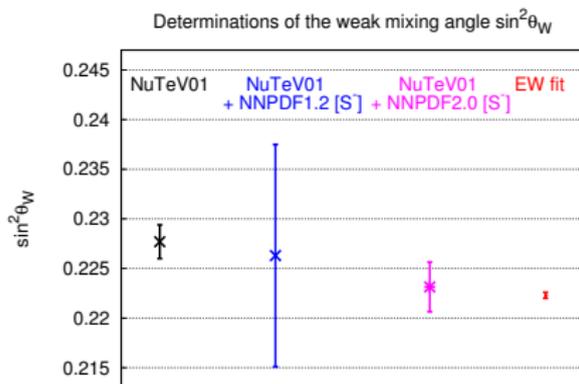
$$s^- = s - \bar{s}$$



- Only MSTW08 and NNPDF2.3 provide independent parametrization of  $s^-$  (though the MSTW08 involves only a few parameters)
- Extractions from both parton sets are in reasonable agreement
- Other parton sets are delivered assuming symmetric strangeness  $s = \bar{s}$
- Recall that NNPDF2.3 also includes LHC data (however, they were shown to have slight impact on strangeness)

# Impact on the NuTeV anomaly

- The accurate determination of  $s^-$  has important phenomenological implications
- The NuTeV anomaly: discrepancy (at  $3\sigma$  level) between indirect (global fit) and direct (NuTeV neutrino scattering) determinations of  $\sin^2\theta_W$
- The magnitude of the strangeness asymmetry  $s^-$  is large enough to remove the NuTeV anomaly



- EW fit  
 $\sin^2\theta_W = 0.2223 \pm 0.0003$
- NuTeV (assumes  $[S^-] = 0$ )  
 $\sin^2\theta_W = 0.2277 \pm 0.0017$
- NuTeV + NNPDF1.2  $[S^-]$   
 $\sin^2\theta_W = 0.2263 \pm 0.0017^{\text{exp}} \pm 0.0107^{\text{PDFs}}$
- NuTeV + NNPDF2.0  $[S^-]$   
 $\sin^2\theta_W = 0.22314 \pm 0.0017^{\text{exp}} \pm 0.00251^{\text{PDFs}}$

# $W^\pm$ asymmetry and the $s^-$ distribution [arXiv:1203.1290]

- The  $W^\pm$  asymmetry is driven mainly by the difference between  $u\bar{d}$  and  $\bar{u}d$  fusion, but also has some sensitivity to the strange quark asymmetry  $s^- = s - \bar{s}$
- The strange  $s$  and  $\bar{s}$  quarks make significant contributions to  $W^+$  and  $W^-$  production at hadron colliders, particularly the LHC, through  $c\bar{s}$  and  $\bar{c}s$  partonic fusion

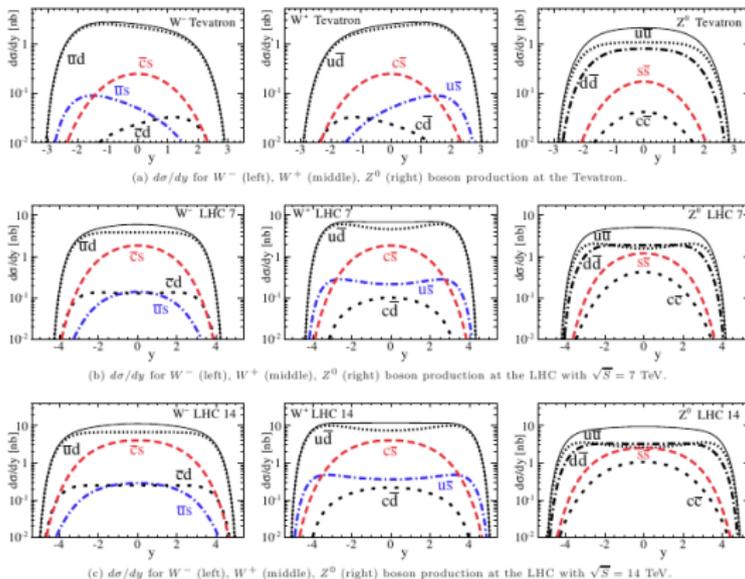
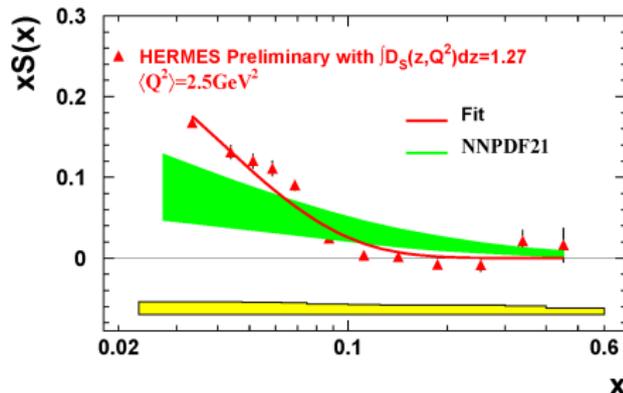
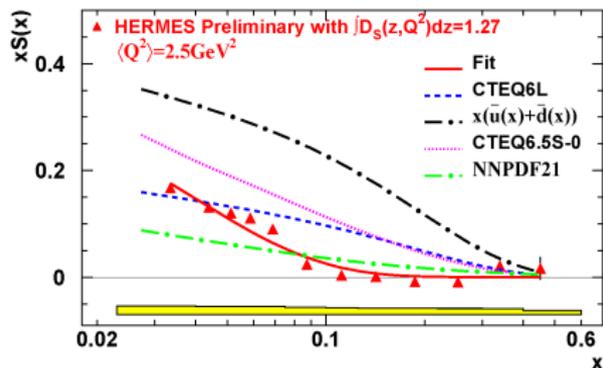


Figure 7: Partonic contributions to the differential cross section of on-shell  $W^\pm/Z$  boson production at LO as a function of the vector boson rapidity. Partonic contributions containing a strange or anti-strange quark are denoted by (red) dashed and (blue) dot-dashed lines. The solid lines show the total contribution.

# Strangeness from charged-kaon production in DIS

- Consider inclusive and semi-inclusive-charged-kaon spin asymmetries for a longitudinally polarized deuteron target
- Extract the LO parton distributions of the strange sea in the proton
- Find that the momentum densities are softer than previously assumed
- Helicity densities are consistent with zero
- The partial moment of the octet axial combination is observed to be substantially less than the axial charge extracted from hyperon decays under the assumption of  $SU(3)$  symmetry



# Polarized PDF Parametrization

$$\Delta\Sigma(x, Q_0^2) = (1-x)^{m_{\Delta\Sigma}} x^{-n_{\Delta\Sigma}} NN_{\Delta\Sigma}(x)$$

$$\Delta g(x, Q_0^2) = (1-x)^{m_{\Delta g}} x^{-n_{\Delta g}} NN_{\Delta g}(x)$$

$$\Delta T_3(x, Q_0^2) = A_{\Delta T_3} (1-x)^{m_{\Delta T_3}} x^{-n_{\Delta T_3}} NN_{\Delta T_3}(x)$$

$$\Delta T_8(x, Q_0^2) = A_{\Delta T_8} (1-x)^{m_{\Delta T_8}} x^{-n_{\Delta T_8}} NN_{\Delta T_8}(x)$$

- 1 Each polarized PDF parametrized with a multi-layer feed-forward NN (2-5-3-1)
- 2 Parametrization supplemented with a preprocessing polynomial:
  - exponents  $m$  and  $n$  randomly chosen in fixed intervals;
  - intervals must be sufficient large not to introduce a bias on the fit
  - check *a posteriori* by studying asymptotic exponents
- 3 Overall normalization constant factored out for triplet and octet.

$$A_{\Delta T_3} = \frac{a_3}{\int_0^1 dx [(1-x)^{m_{\Delta T_3}} x^{-n_{\Delta T_3}} NN_{\Delta T_3}(x)]}$$

$$A_{\Delta T_8} = \frac{a_8}{\int_0^1 dx [(1-x)^{m_{\Delta T_8}} x^{-n_{\Delta T_8}} NN_{\Delta T_8}(x)]}$$

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$$\Delta\Sigma(x, Q_0^2) = (1-x)^{m_{\Delta\Sigma}} x^{-n_{\Delta\Sigma}} NN_{\Delta\Sigma}(x)$$

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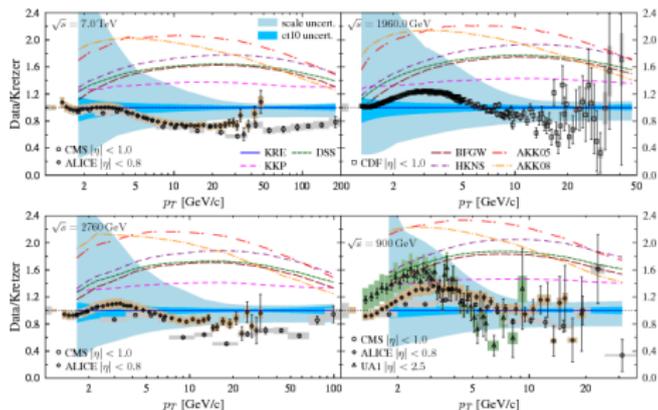
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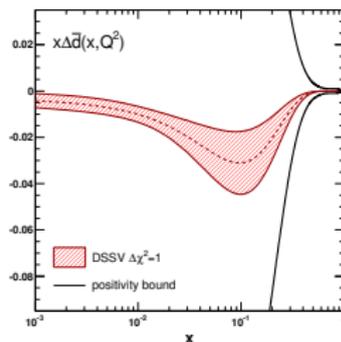
# How well do we know fragmentation functions? [arXiv:1311.1415]

- Examine LHC, Tevatron, Sp $\bar{p}$ S and RHIC data for inclusive unidentified charged-hadron production at  $\sqrt{s} = 0.2 - 7$  TeV against NLO pQCD calculations with different sets of FFs
- Quantify the systematics associated with the scale and PDF uncertainties
- Find large spread among the predictions with different FFs essentially due to sizable mutual differences in the gluon-to-hadron FFs
- None of the existing FF sets can reproduce the experimental results optimally, overshooting the data by up to a factor two for  $p_T > 10$  GeV (modest dependence on scale)
- The best agreement is reached by Kretzer FF (based on SIA data only)
- Need for a global refit including LHC data



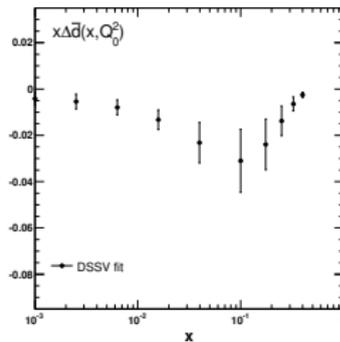
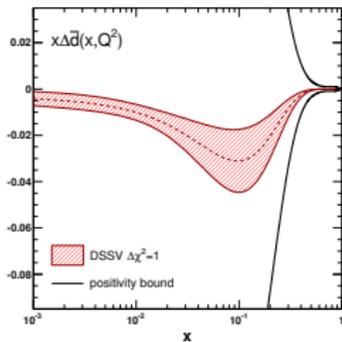
# Construction of a suitable prior PDF ensemble

- 1 Take a polarized parton set which provides separated  $\Delta q - \Delta \bar{q}$  PDFs (from SIDIS)  
→ we choose DSSV
- 2 Suppose they reproduce the *true* physical underlying behaviour
- 3 Sample their  $\Delta \bar{u}$  and  $\Delta \bar{d}$  distributions at a given reference scale  
→ we take 10 points in the range  $10^{-3} \lesssim x \lesssim 0.4$  at  $Q_0^2 = 1 \text{ GeV}^2$  (half logarithmically, half linearly spaced)
- 4 Perform a NN fit to these pseudodata
- 5 Supplement each replica in NNPDFpo11.0 with  $\Delta \bar{u}$  and  $\Delta \bar{d}$  obtained in this way
- 6 Reweight PDFs and check that observables are properly reproduced
- 7 Check that reweighted results are stable upon the choice of different PDF priors  
→ relax item 2, for example by increasing the nominal PDF uncertainty, until independence from the prior is reached  
→ we have considered four different PDF priors:  $1\sigma$ ,  $2\sigma$ ,  $3\sigma$ ,  $4\sigma$



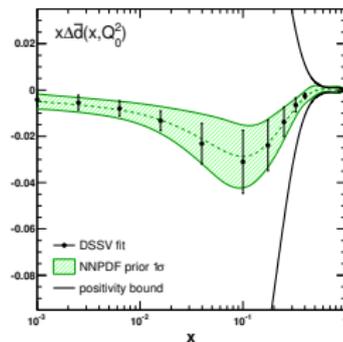
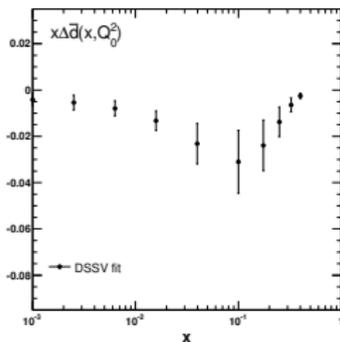
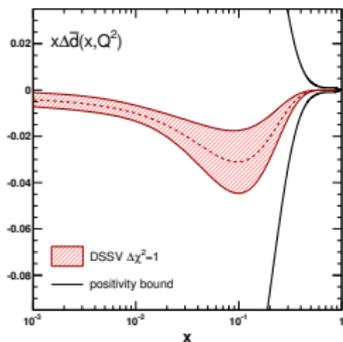
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# Construction of a suitable prior PDF ensemble

- 1 Take a polarized parton set which provides separated  $\Delta q - \Delta \bar{q}$  PDFs (from SIDIS)  
→ we choose DSSV
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→ we have considered four different PDF priors:  $1\sigma$ ,  $2\sigma$ ,  $3\sigma$ ,  $4\sigma$



# Reweighting with $W^\pm$ production at RHIC

New dataset on longitudinal single-spin asymmetry for  $W^\pm$  boson production

$$A_L^{W^+} \sim \frac{-\Delta u(x_1)\bar{d}(x_2) + \Delta\bar{d}(x_1)u(x_2)}{u(x_1)\bar{d}(x_2) + \bar{d}(x_1)u(x_2)} \quad A_L^{W^-} \sim \frac{-\Delta d(x_1)\bar{u}(x_2) + \Delta\bar{u}(x_1)d(x_2)}{d(x_1)\bar{u}(x_2) + \bar{u}(x_1)d(x_2)}$$

## FEATURES

- sensitive to individual quark and antiquark flavours ( $\Delta u, \Delta\bar{u}, \Delta d, \Delta\bar{d}$ )  
(purely weak process coupling  $q_L$  with  $\bar{q}_R$  at partonic level,  $u_L\bar{d}_R \rightarrow W^+$  or  $d_L\bar{u}_R \rightarrow W^-$ )
- no need of fragmentation functions (instead of SIDIS)

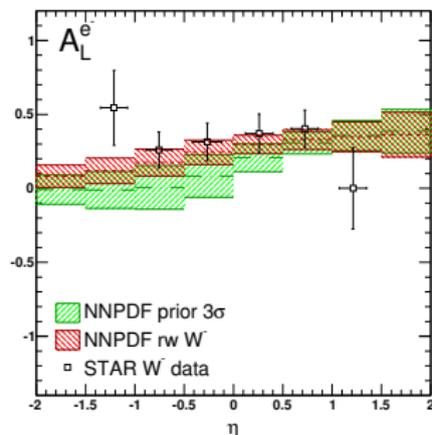
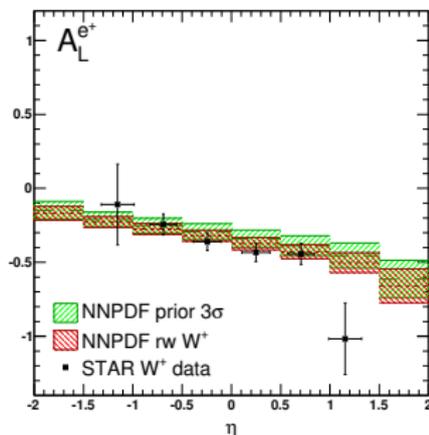
## EXPERIMENTAL MEASUREMENT

- STAR and PHENIX at RHIC [[arXiv:1009.0326](#)] [[arXiv:1009.0505](#)]  
(only preliminary measurements from STAR (2012) [[arXiv:1302.6639](#)] will be considered here)

## THEORETICAL PREDICTION

- NNPDFpo1.0 itself is not good since  $q$  and  $\bar{q}$  contributions are not disentangled  
→ if new data bring in sufficient new information, results will be independent from prior
- the asymmetry is computed at NLO using CHE code [[arXiv:1003.4533](#)]

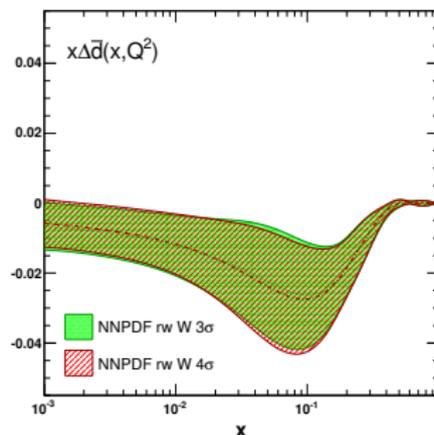
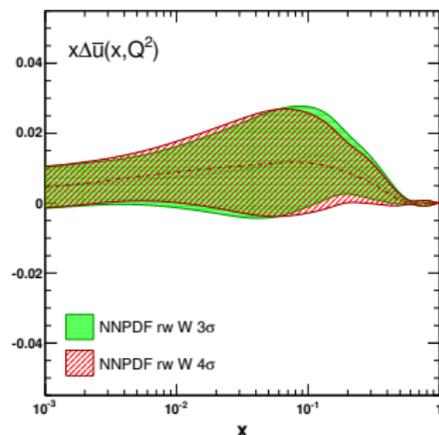
# Sea quarks from W boson production at RHIC



Experiment	Set	$N_{\text{dat}}$	$\chi^2/N_{\text{dat}}$				$\chi_{\text{rw}}^2/N_{\text{dat}}$			
			$1\sigma$	$2\sigma$	$3\sigma$	$4\sigma$	$1\sigma$	$2\sigma$	$3\sigma$	$4\sigma$
STAR		12	2.02	2.08	<b>2.09</b>	1.94	1.48	1.19	<b>1.07</b>	1.07
	STAR- $W^+$	6	1.37	1.37	<b>1.54</b>	1.49	1.21	1.12	<b>1.10</b>	1.10
	STAR- $W^-$	6	2.67	2.79	<b>2.64</b>	2.39	1.75	1.25	<b>1.05</b>	1.04

Overall **good description** of both **data sets** after they are included in the fit via reweighting  
**Stability of reweighted results** reached starting from  $3\sigma$  or  $4\sigma$  prior indifferently

# Sea quarks from W boson production at RHIC



- $W^\pm$  data are effective in constraining light sea quark distributions  $\Delta\bar{u}$  and  $\Delta\bar{d}$
- Reweighting with separate  $W^+$  and  $W^-$  datasets shows that  $\Delta\bar{u}$  and  $\Delta\bar{d}$  behaviour is driven by experimental information from  $W^-$  and  $W^+$  respectively
- All other PDFs are almost unaffected (including strangeness)
- Independence of the reweighted results from the choice of the prior