



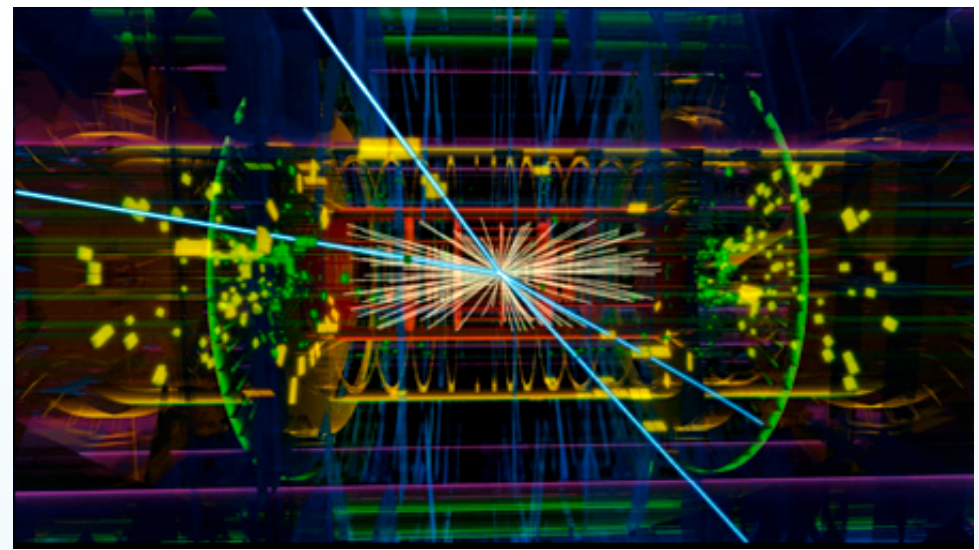
The structure of the proton in the LHC precision era

Juan Rojo

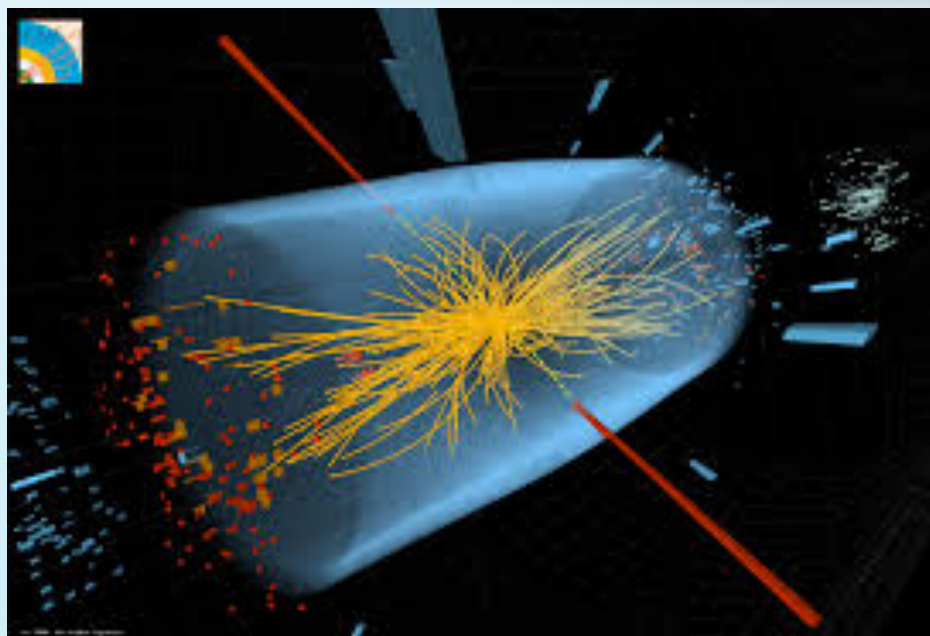
VU Amsterdam & Theory Group, Nikhef

Physikalisches Kolloquium

Department of Physics, University of Bonn, 12/4/2018



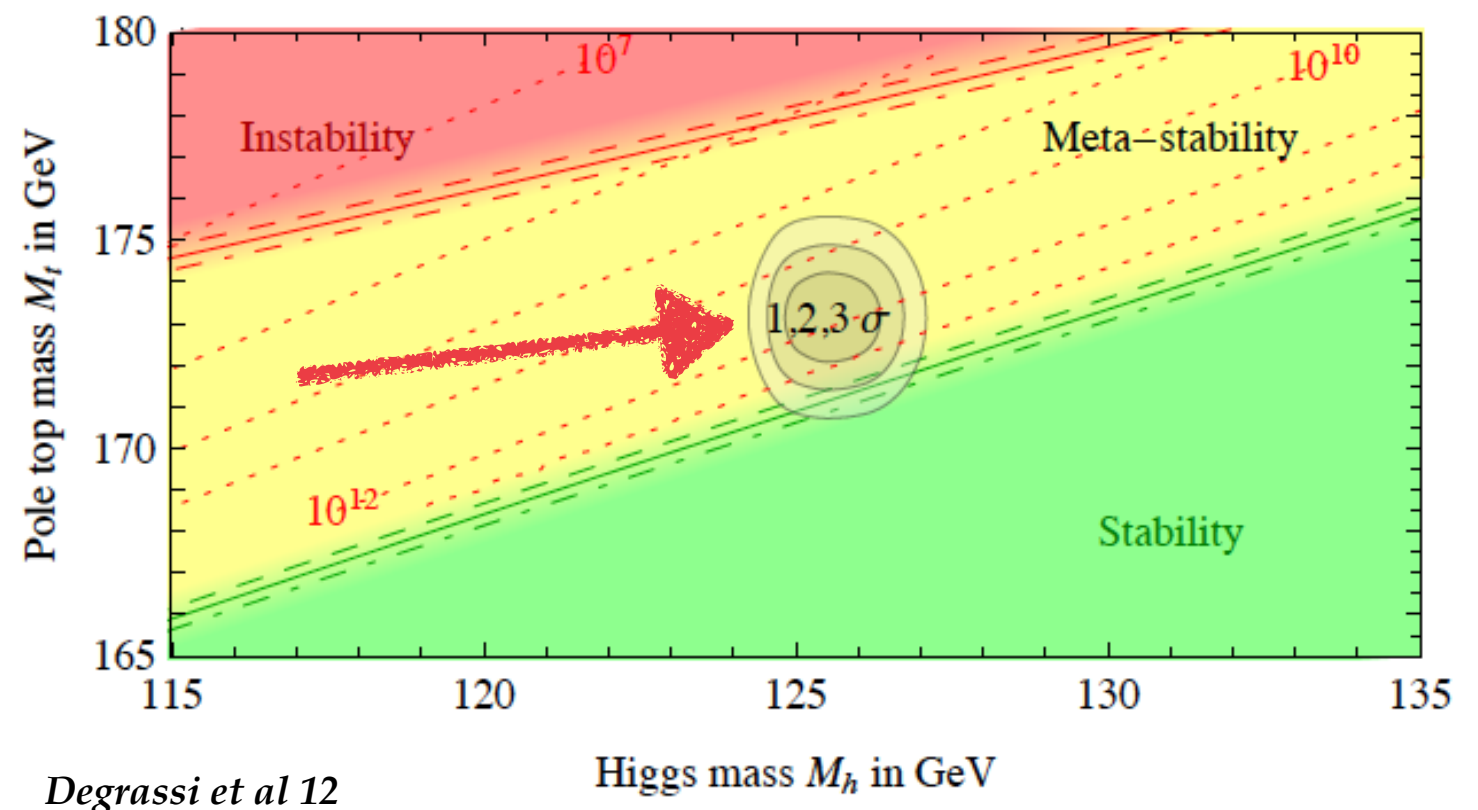
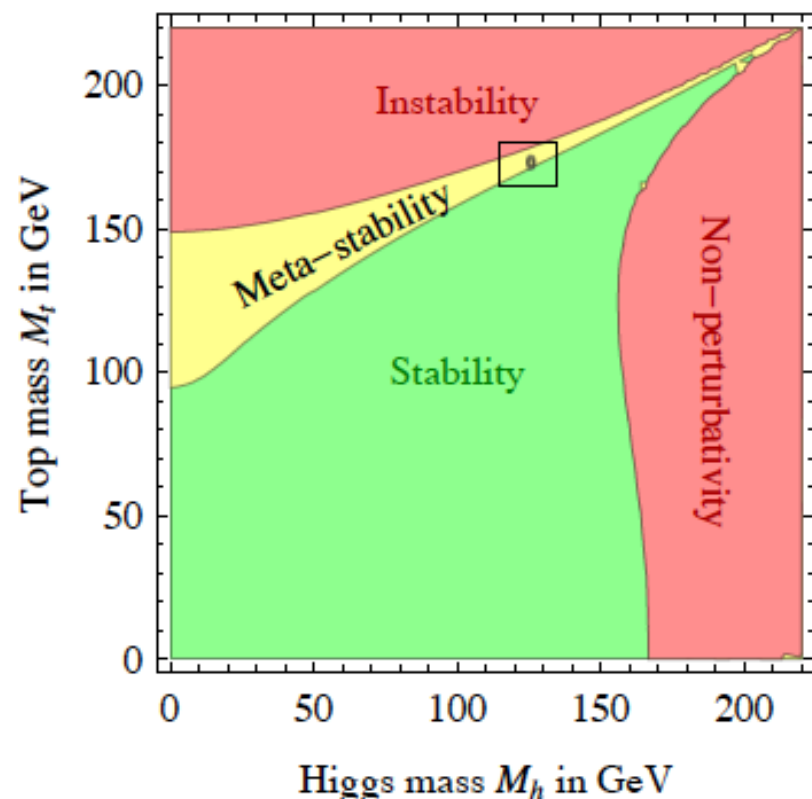
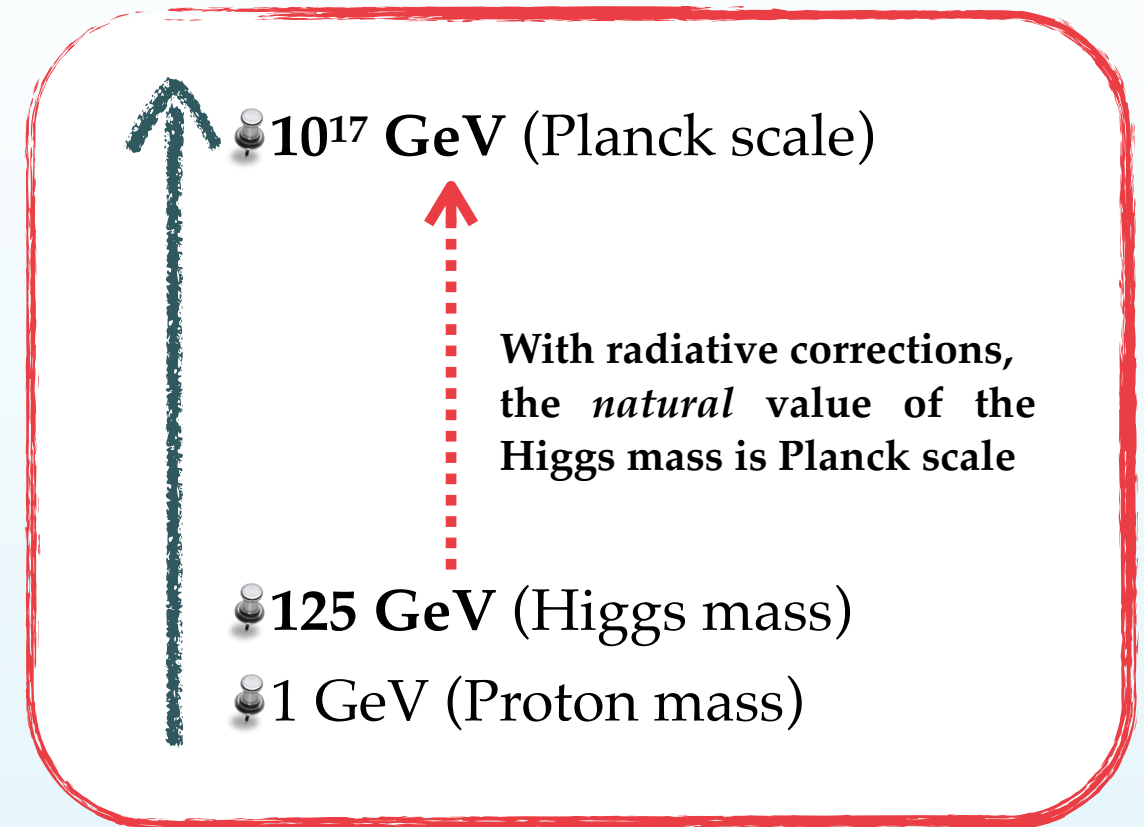
Exploring the high-energy frontier at the Large Hadron Collider



Outstanding questions in Particle Physics

The Higgs boson

- ☑ Huge gap, 10^{17} , between Higgs and Plank scales
- ☑ Elementary or composite? Additional Higgs bosons?
- ☑ Coupling to Dark Matter? Role in cosmological phase transitions?
- ☑ Is the vacuum state of the Universe stable?



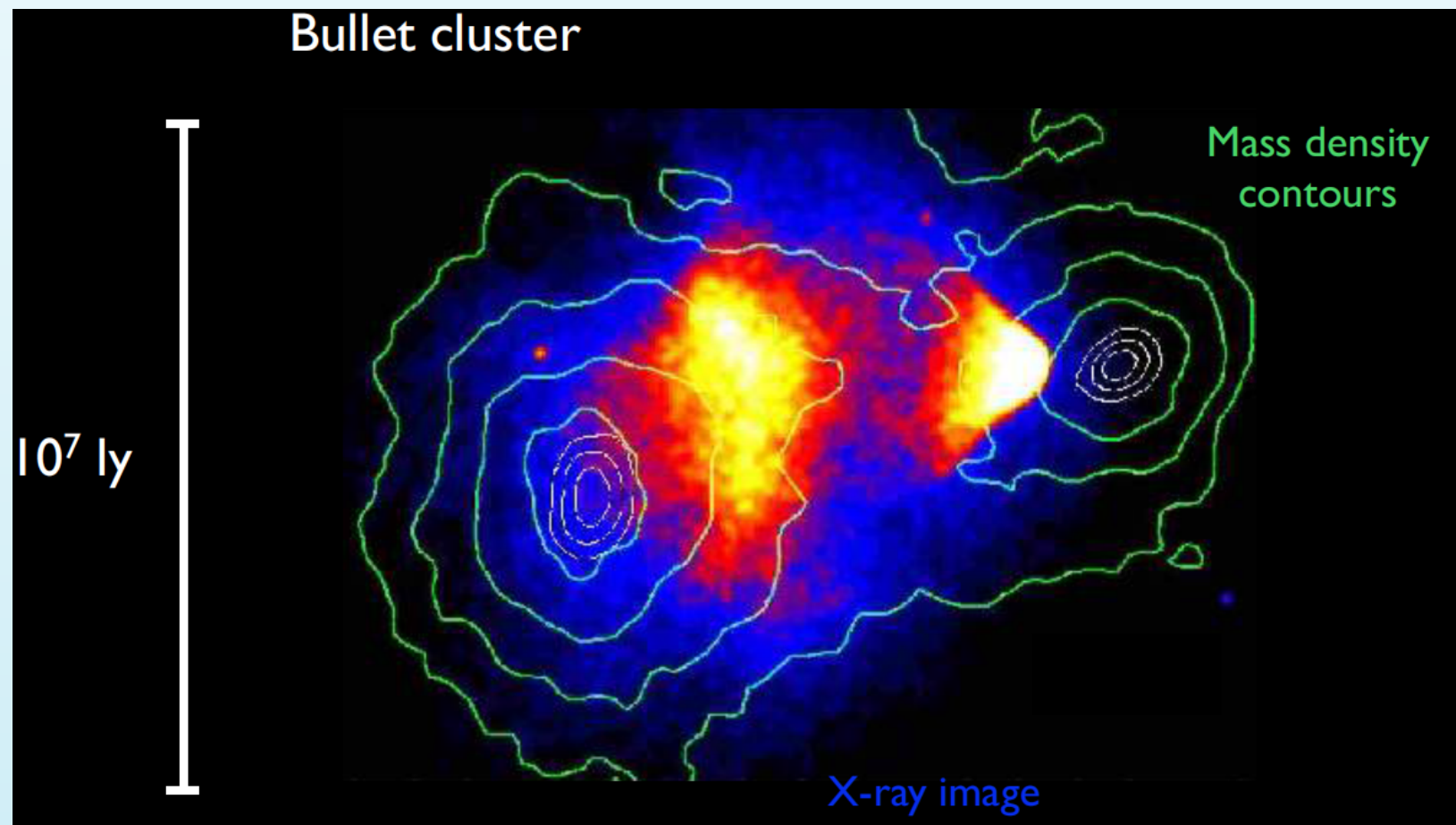
Outstanding questions in Particle Physics

The Higgs boson

- ☑ Huge gap, 10^{17} , between Higgs and Plank scales
- ☑ Elementary or composite? Additional Higgs bosons?
- ☑ Coupling to Dark Matter? Role in cosmological phase transitions?
- ☑ Is the vacuum state of the Universe stable?

Dark Matter

- ☑ Weakly interacting massive particles? Sterile neutrinos? Extremely light particles (axions)?
- ☑ Interactions with Standard Model particles?
- ☑ What is the structure of the Dark Sector? Is Dark Matter self-interacting?



Outstanding questions in Particle Physics

The Higgs boson

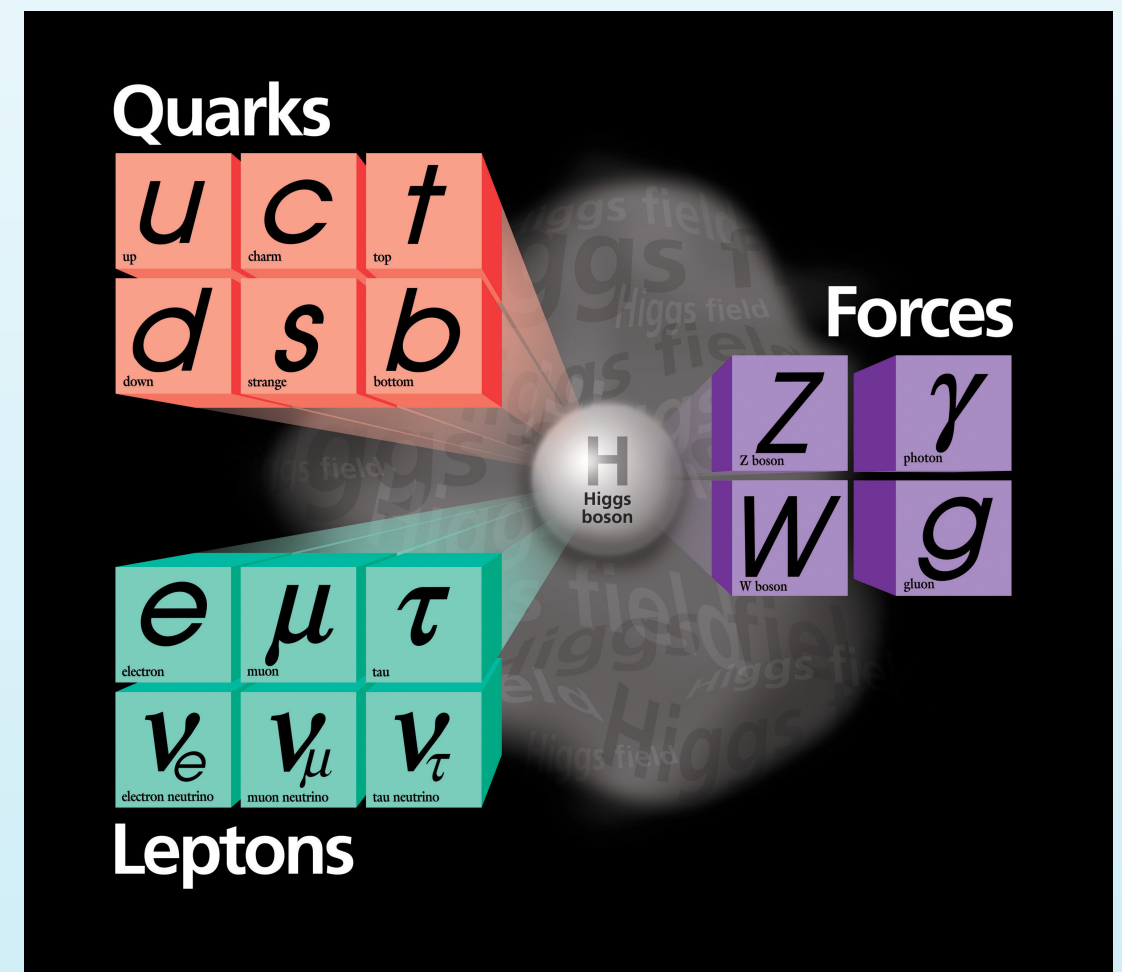
- ✓ Huge gap, 10^{17} , between Higgs and Plank scales
- ✓ Elementary or composite? Additional Higgs bosons?
- ✓ Coupling to Dark Matter? Role in cosmological phase transitions?
- ✓ Is the vacuum state of the Universe stable?

Quarks and leptons

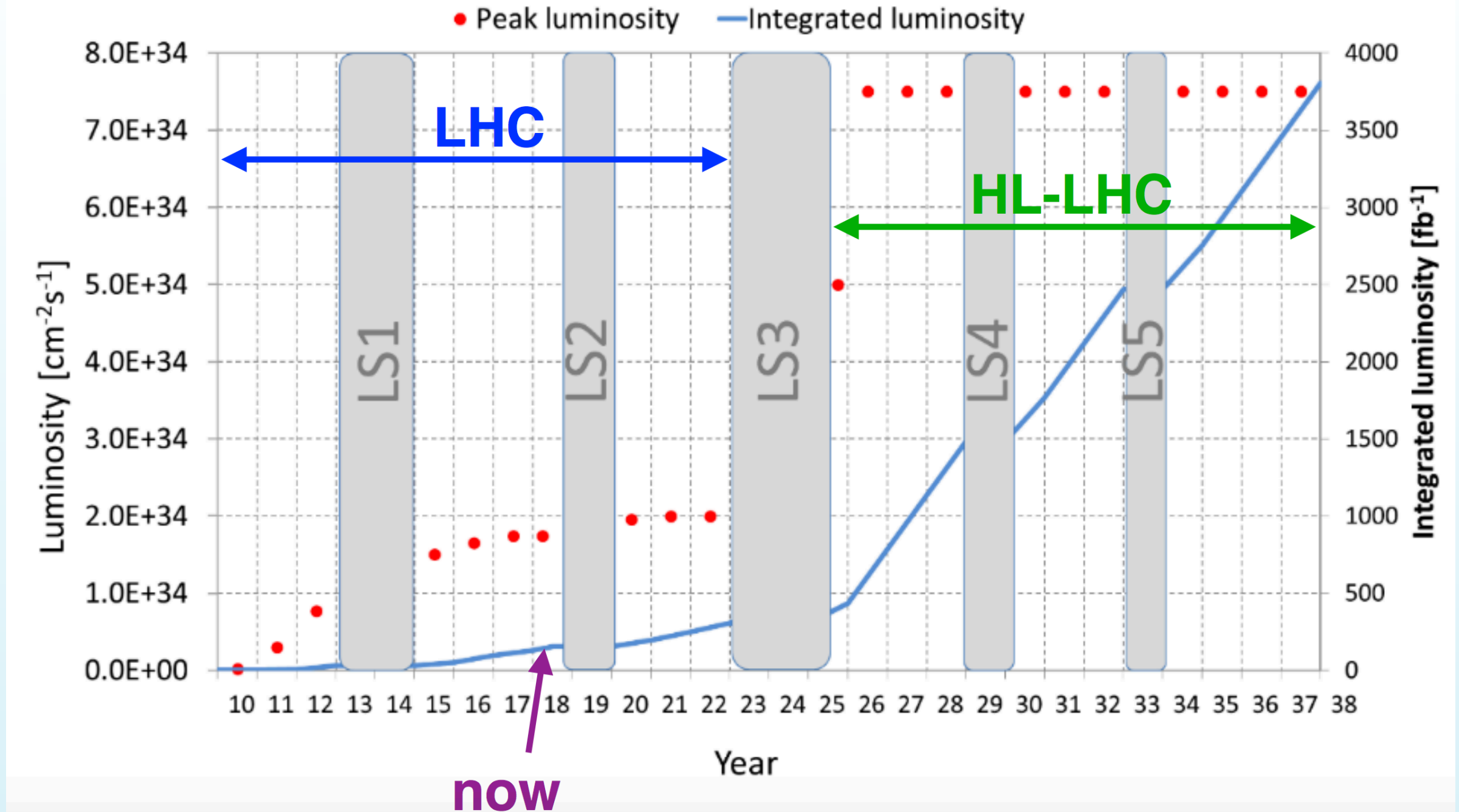
- ✓ Why three families? Can we explain masses and mixings?
- ✓ Origin of Matter-Antimatter asymmetry in the Universe?
- ✓ Are neutrinos Majorana or Dirac? CP violation in the lepton sector?

Dark Matter

- ✓ Weakly interacting massive particles? Sterile neutrinos? Extremely light particles (axions)?
- ✓ Interactions with Standard Model particles?
- ✓ What is the structure of the Dark Sector? Is Dark Matter self-interacting?



Outstanding questions in Particle Physics

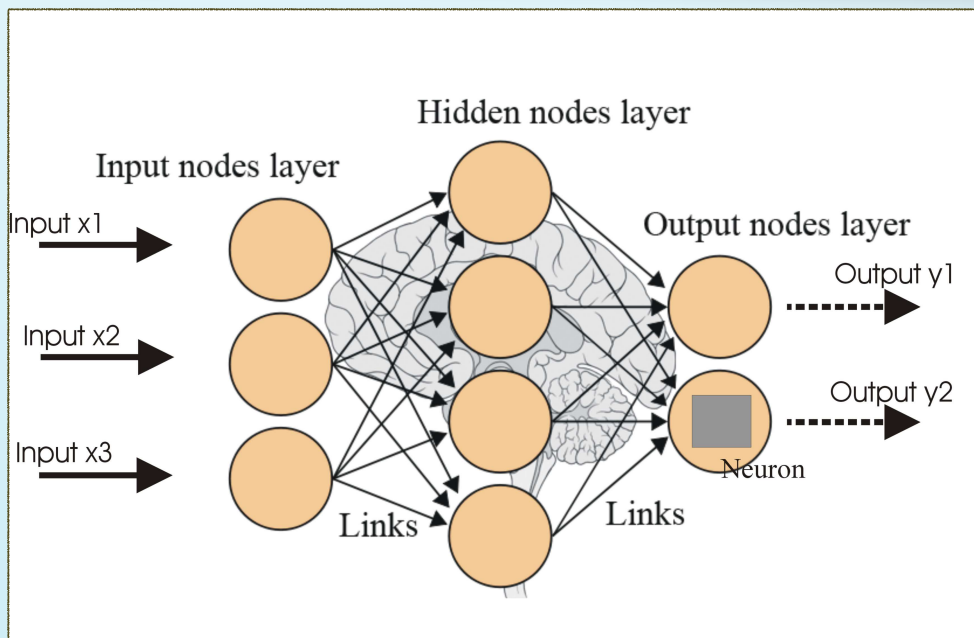


Many of these crucial questions can be addressed at the Large Hadron Collider

For the next 20 years, LHC will be the forefront of the exploration of the high-energy frontier



Machine Learning and Artificial Neural Networks at the LHC



Juan Rojo

Machine Learning at the LHC

- 📍 By **Machine Learning** we usually denote those families of computer algorithms that **learn how to excel on a task** based on a **large sample of examples**, rather than on some a priori fixed rules
- 📍 ML algorithms are nowadays ubiquitous, from **driverless cars** to **Amazon's purchase suggestions**, to **automated medical imaging recognition** to beating the words best players at Go and chess
- 📍 ML tools rely on the **efficient exploitation of immense datasets**. And the **LHC** has a lot of data!

The Big Data Universe, 2016

Amount of data stored in Petabytes
(1 Petabyte = 1 000 000 GB)

Share



Human brain
2.5 PB

Ebay
90 PB

Spotify
10 PB

Facebook
300 PB

Google
15,000 PB
(estimated)

LHC data analysis: 30 PB/year!



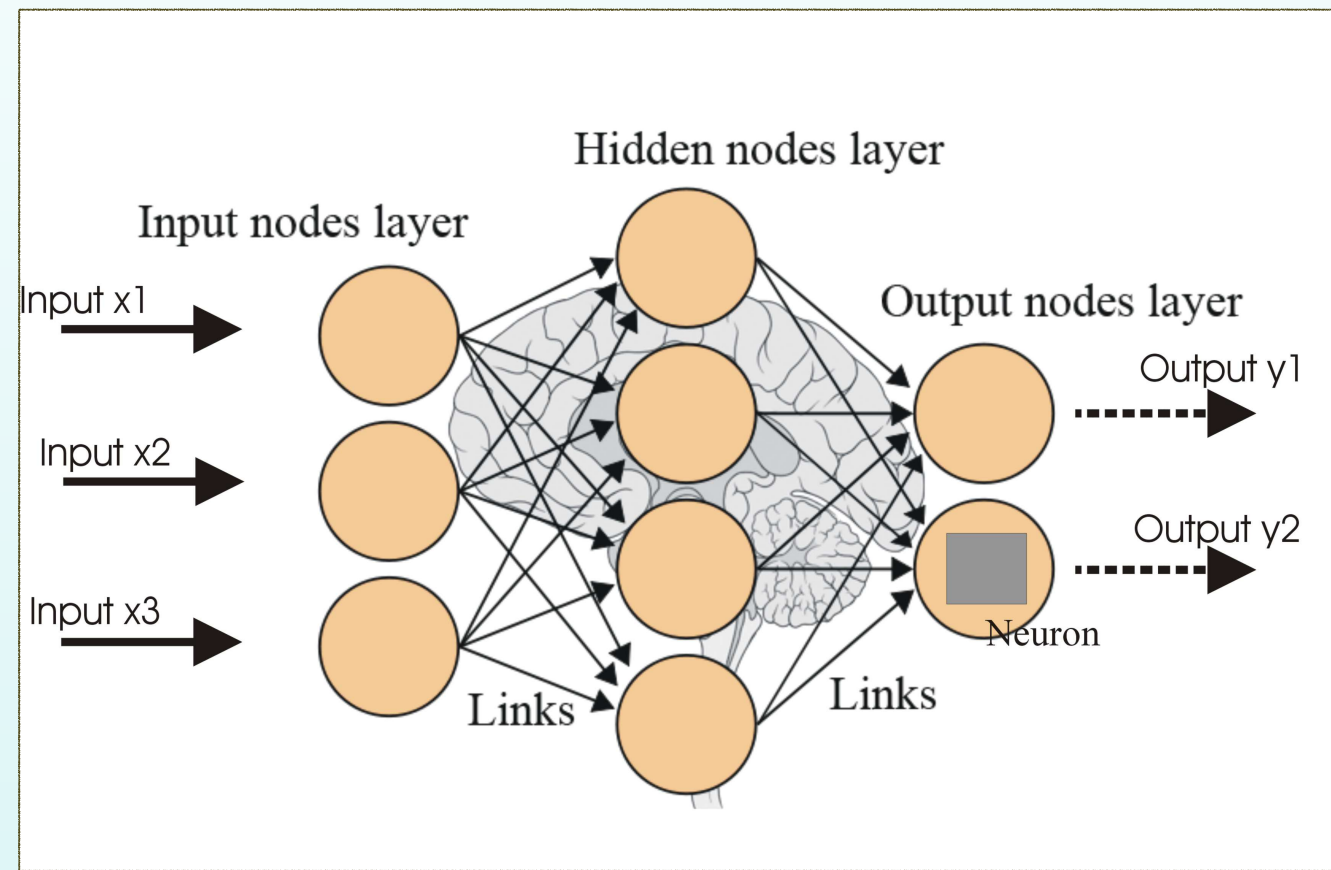
Machine Learning in high-energy physics

- 📌 **Huge, fast growing field**, with new applications being proposed every day
- 📌 Here restrict ourselves to **two representative examples**: if you want to learn more about other applications, don't hesitate to ask!
- 📌 For further **overviews of ML applications to high-energy physics** and related fields please see
 - ☑ *Big data tools in Physics and Astronomy* (Amsterdam, <https://indico.cern.ch/event/622093/>)
 - ☑ *Machine learning for Phenomenology* (Durham, <https://conference.ippp.dur.ac.uk/event/660/>)
 - ☑ *Inter-Experimental LHC Machine Learning WG* (<https://iml.web.cern.ch/>)
 - ☑ *Accelerating searches for Dark Matter with Machine Learning* (<https://indico.cern.ch/event/664842/>)
 - ☑ *CERN Data Science seminars* (<https://indico.cern.ch/category/9320/>)

Artificial Neural Networks

Inspired by **biological brain models**, **Artificial Neural Networks (ANNs)** are **mathematical algorithms** widely used in a wide range of applications, from **HEP** to **targeted marketing** and **finance forecasting**

From biological to artificial neural networks



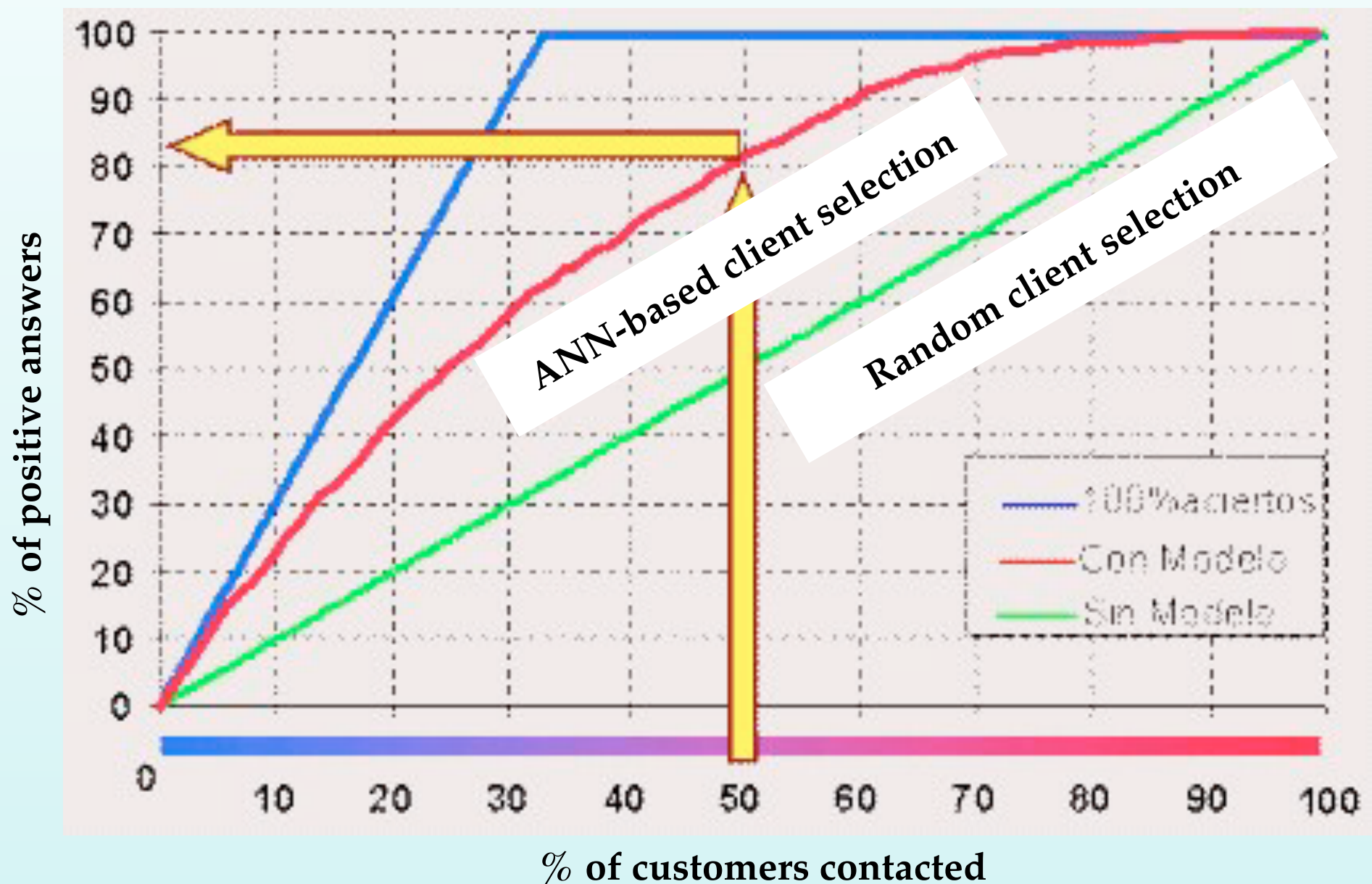
Artificial neural networks aim to excel where domains as their **evolution-driven counterparts** **outperforms traditional algorithms** in tasks such as **pattern recognition**, **forecasting**, **classification**, ...

ANNs - a marketing example

A bank wants to offer a new credit card to their clients. Two possible strategies:

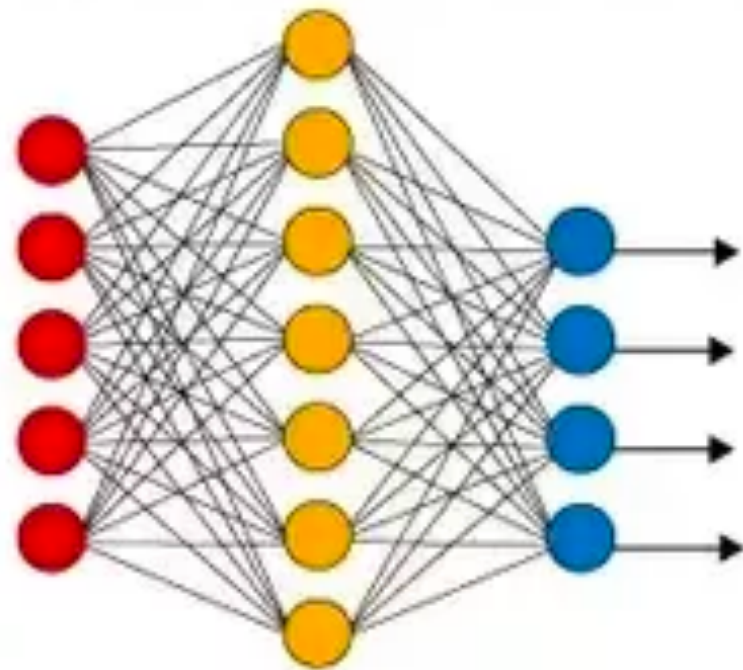
- 📌 **Contact all customers:** slow and costly
- 📌 Contact **5%** of the customers, **train a ANN with their input** (savings, income, loans) and **their output** (yes/no) and use the information to **contact only clients likely to accept the product**

Cost-effective method to improve marketing performance!

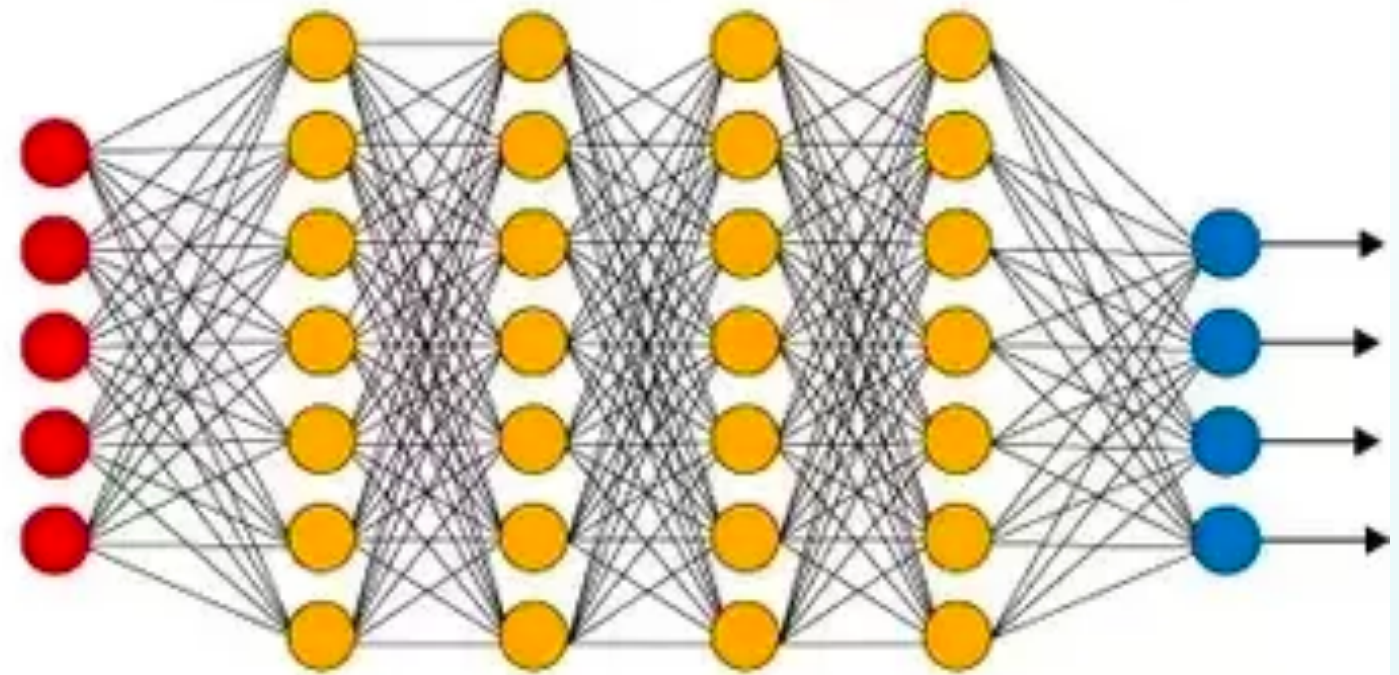


Deep Neural Networks

Simple Neural Network



Deep Learning Neural Network



● Input Layer

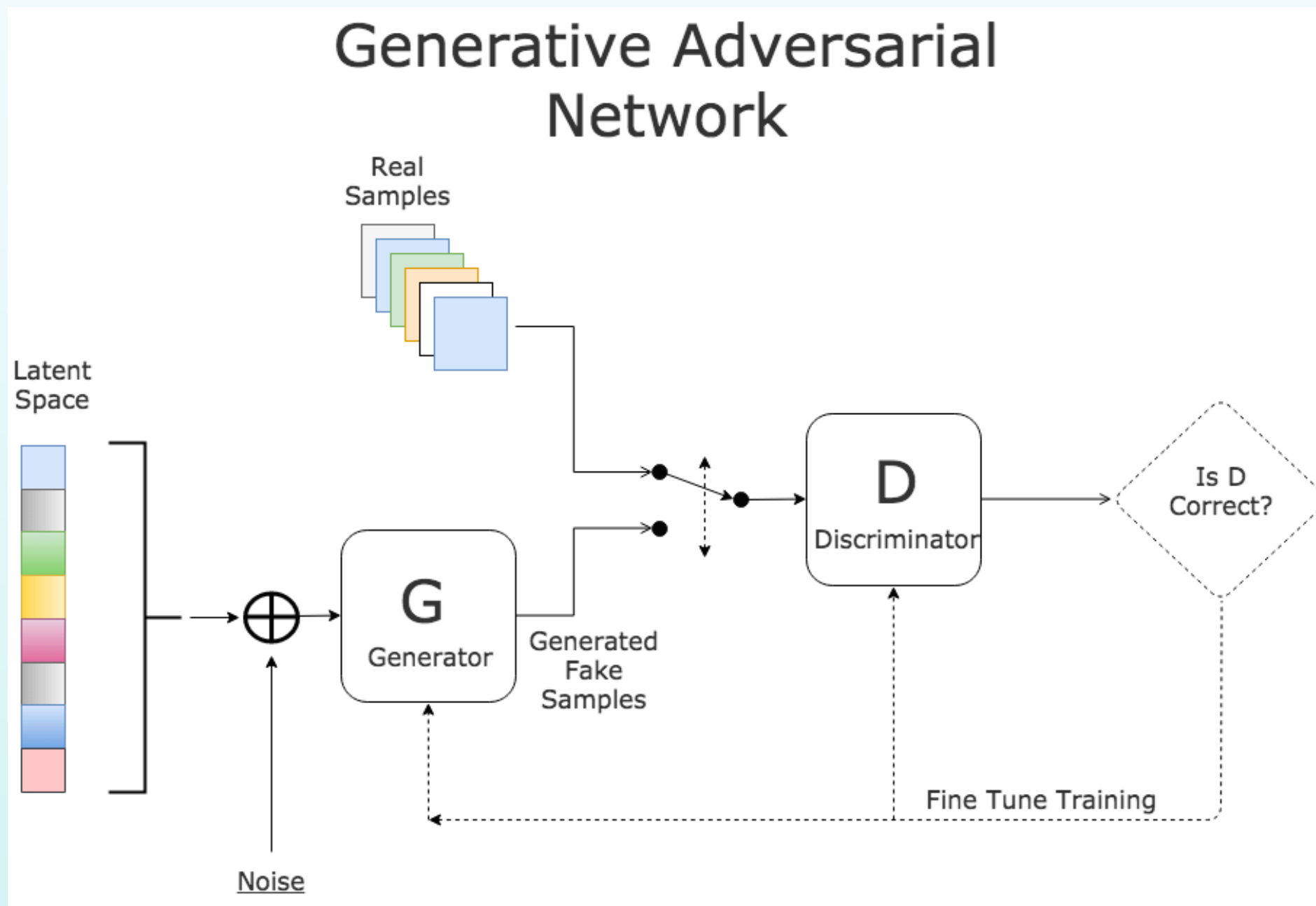
● Hidden Layer

● Output Layer

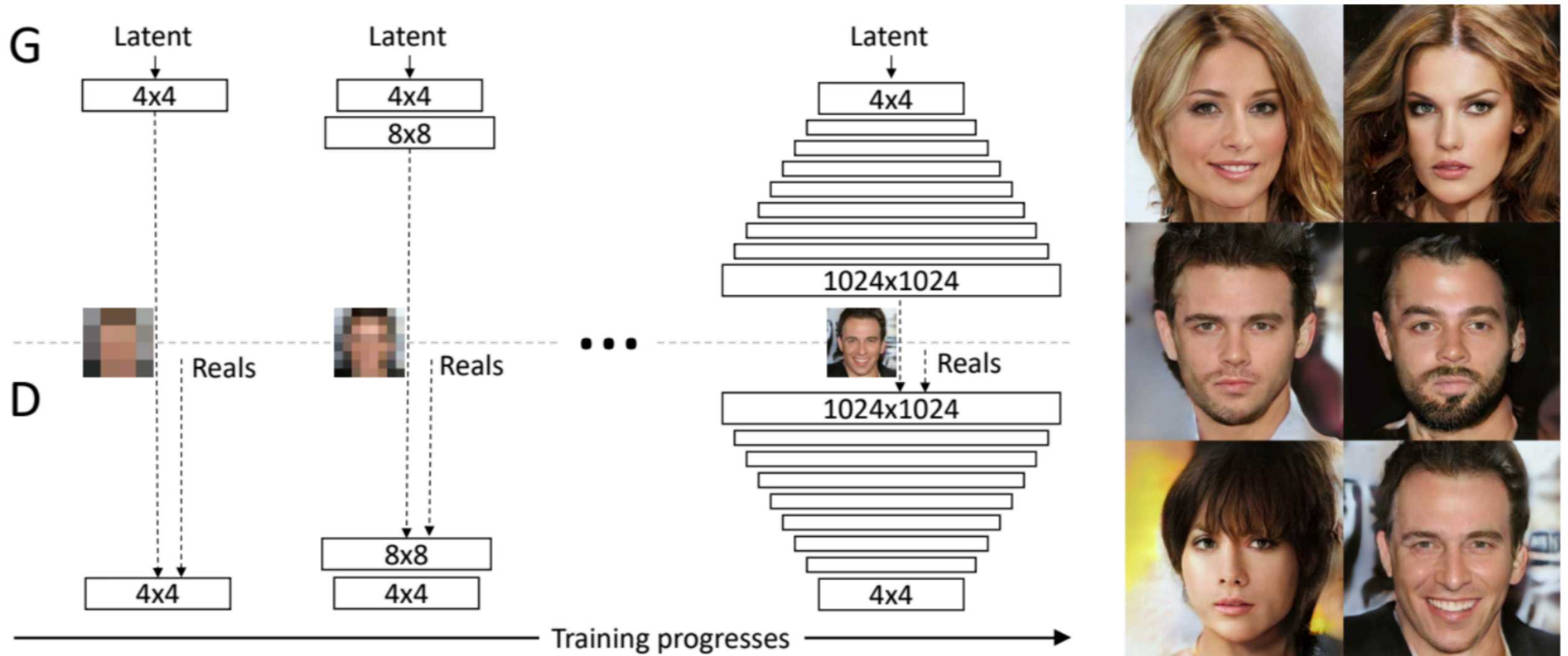
- 📌 A **Deep Neural Network (DNN)** is a standard multi-layer feed-forward perceptron with a large number of internal layers
- 📌 All types of neural nets eg **Recursive, Convolutional, Parametrised** etc can be made “deep” by adding more hidden layers
- 📌 For several applications, the **increased complexity** achieved this way leads to a significant improvement in performance

Generative Adversarial Networks

- New architecture for an **unsupervised neural network training** (unlabelled samples)
- Based on two **independent nets** that work separately and act as adversaries:
 - the **Discriminator (D)** undergoes training and plays the role of classifier, and
 - the **Generator (G)** and is tasked to generate random samples that **resemble real samples** with a twist rendering them as fake samples.



The many uses of GANs

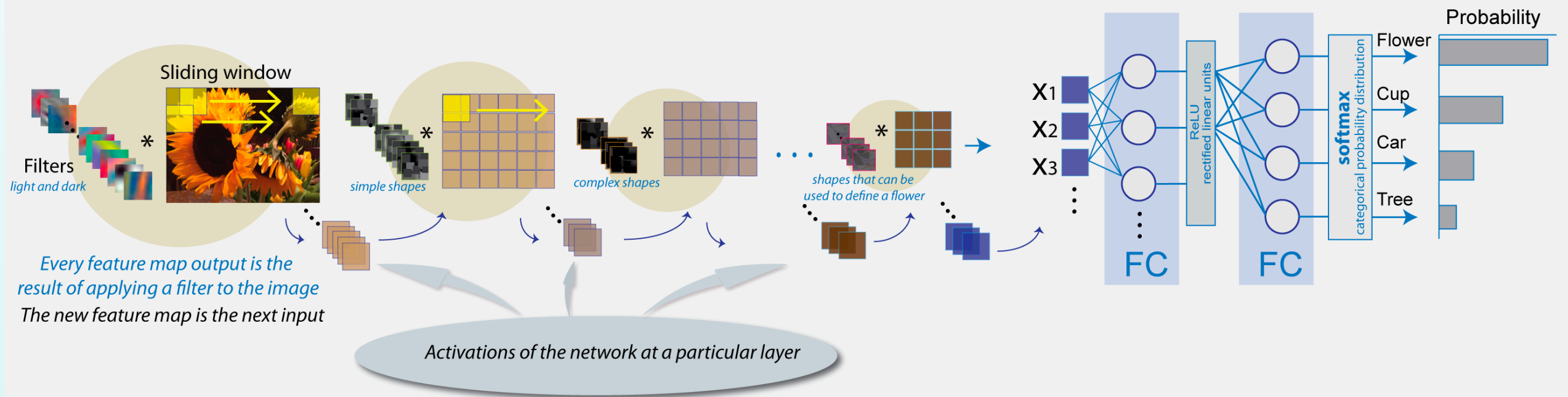
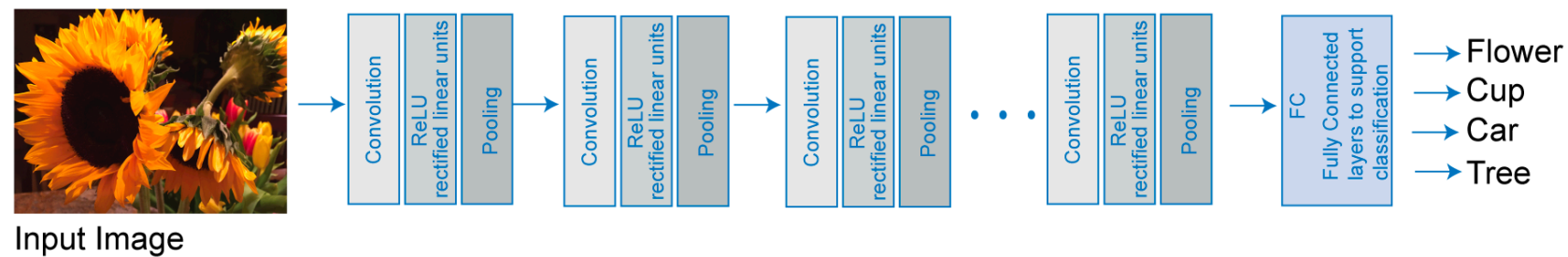


arXiv:1710.10196

Which one of these images are real and which ones are fake (generated by the GANs)?

Convolutional Neural Networks

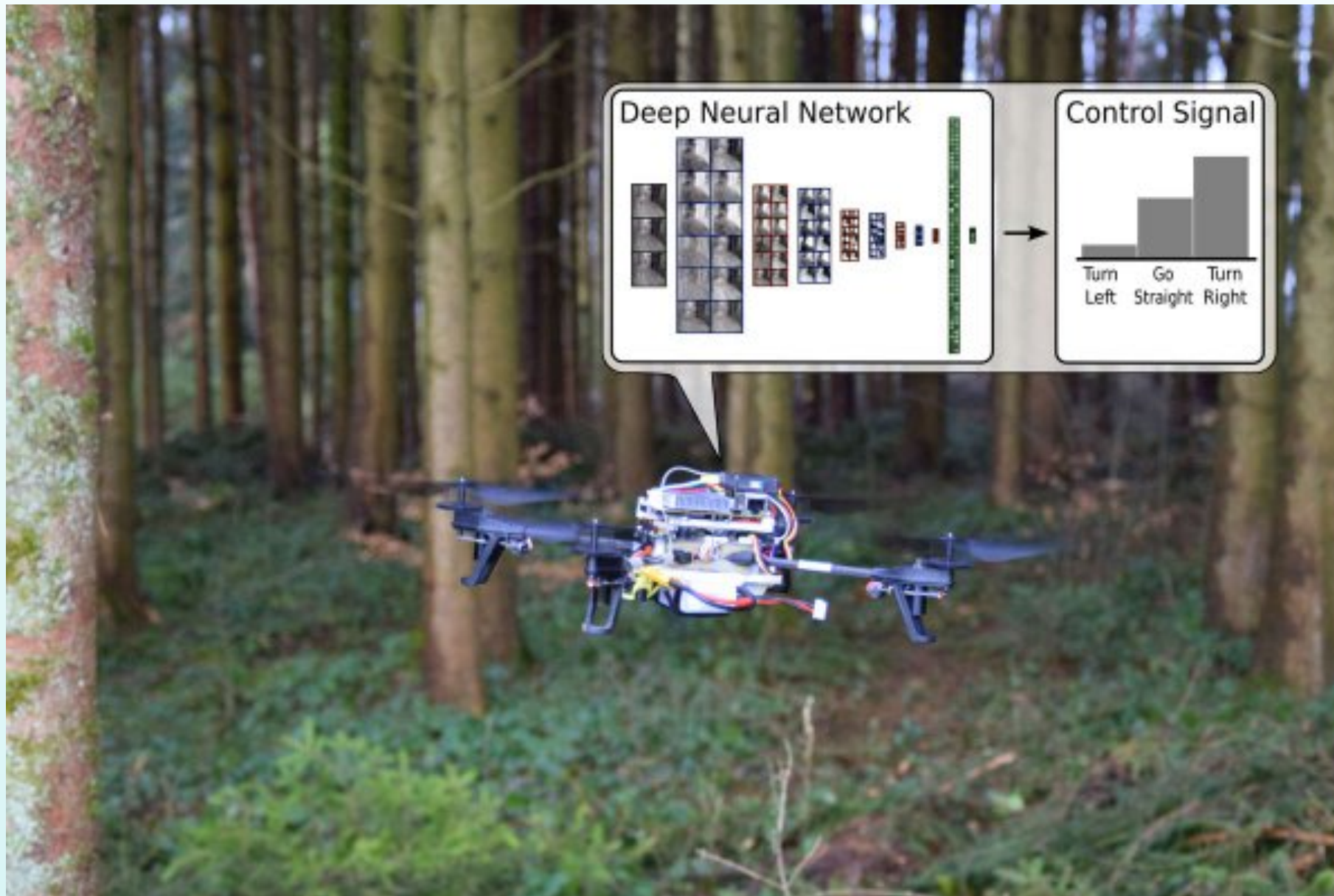
- Convolutional Neural Networks (CNNs) have convolutional layers based on **filters**
- Each **filter** maps a group of numbers into a number, reducing the dimensionality of the data
- Specially useful for **pattern recognition** (eg for self-driving vehicles)



mathworks.com

Convolutional Neural Networks

- ANNs can enable an **autonomous vision-control drone** to recognise and follow forest trails
- Image classifier operates directly on **pixel-level image intensities**
- If a trail is visible, the **software steers the drone** in the corresponding direction

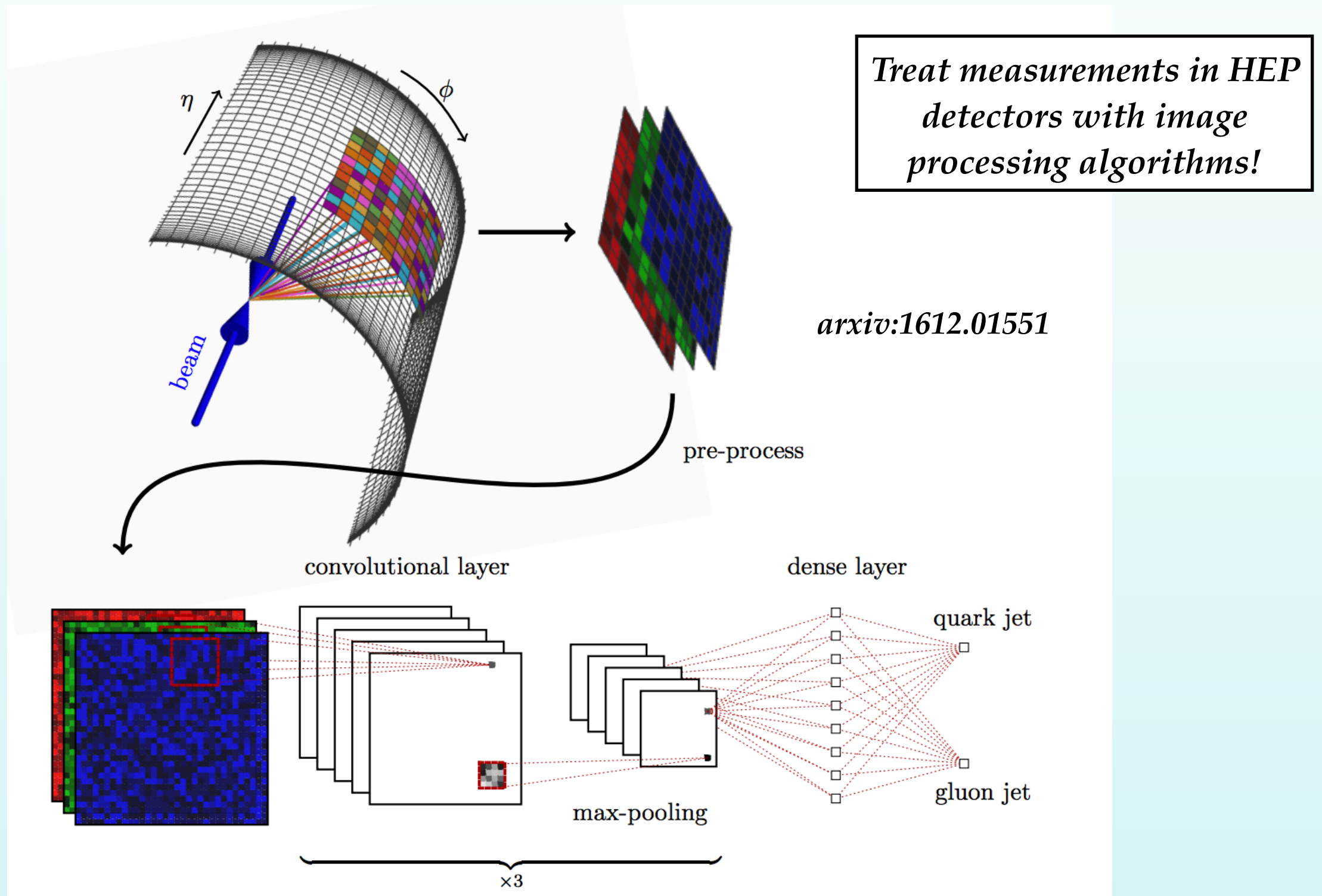


Giusti et al, IEEE Robotics and Automation Letters, 2016

Similar algorithms at work in self-driving cars!

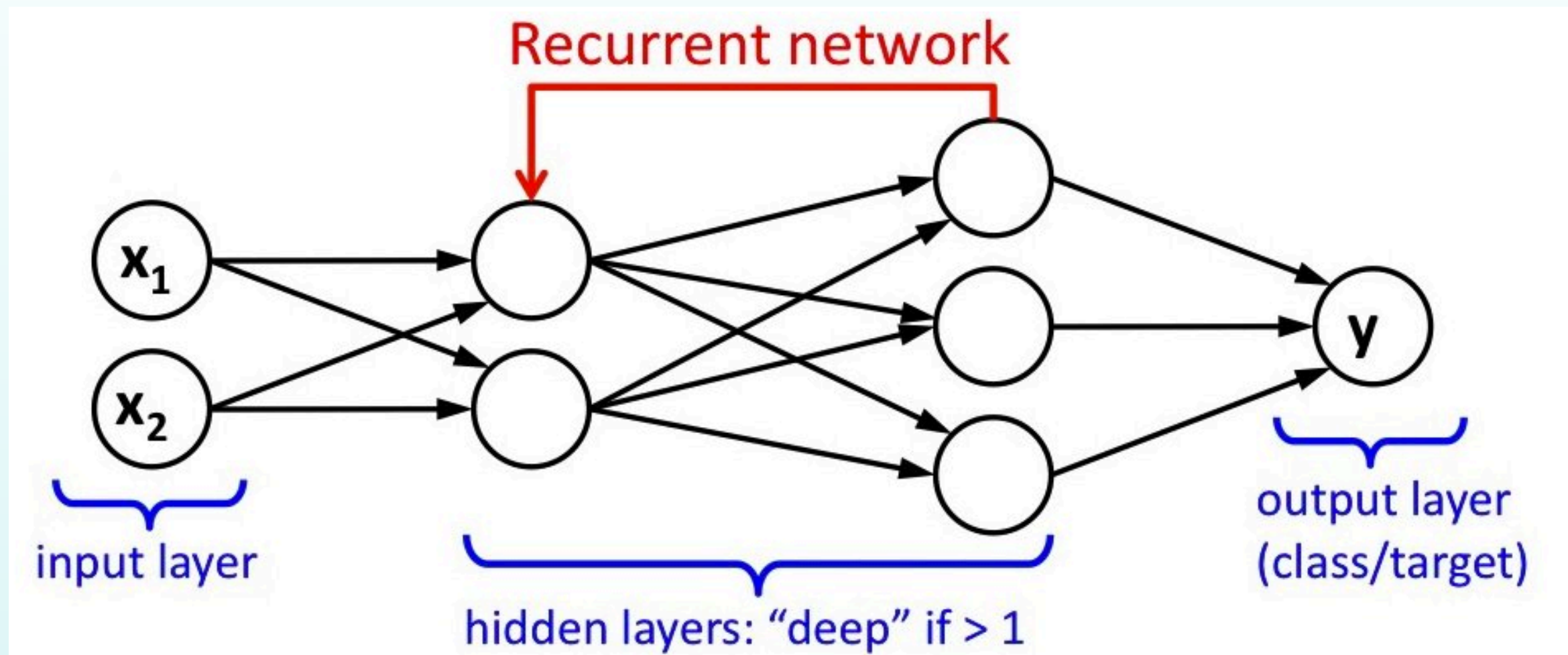
Convolutional Neural Networks

The results of the collisions of high-energy particles can be treated analogously to image processing using Convolutional Neural Networks



Recurrent Neural Networks

RNNs use as inputs not just the current “training examples” but also **what they have perceived previously**: they have a **built-in notion of time ordering** useful for time-dependent functions



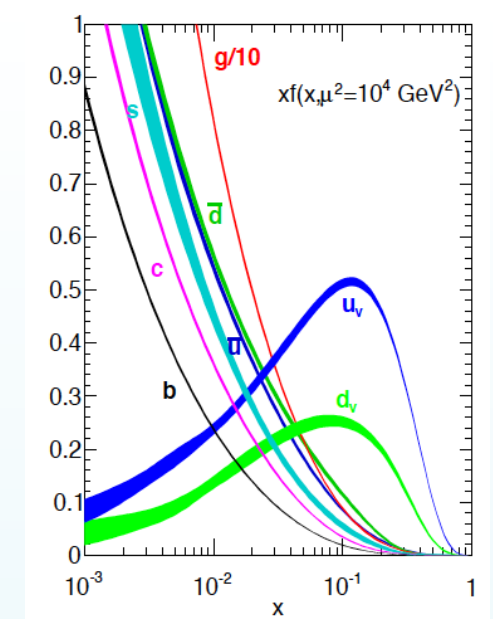
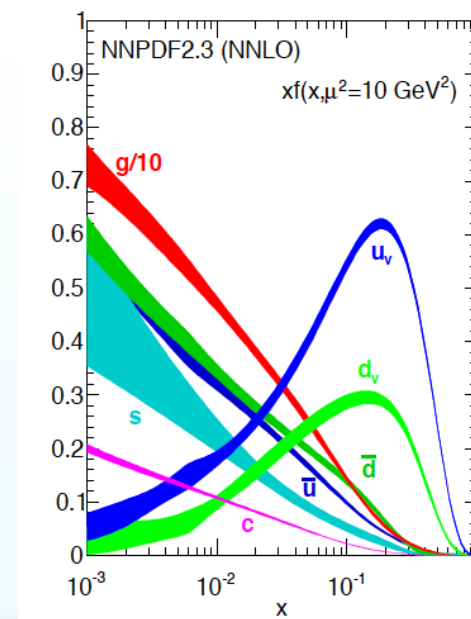
The output of a RNN at time t , $y(t)$, depends both on the current input example $x(t)$ as well as of its previous output $y(t-1)$ (or activation states of hidden neurons at $t-1$)

Recurrent Neural Networks

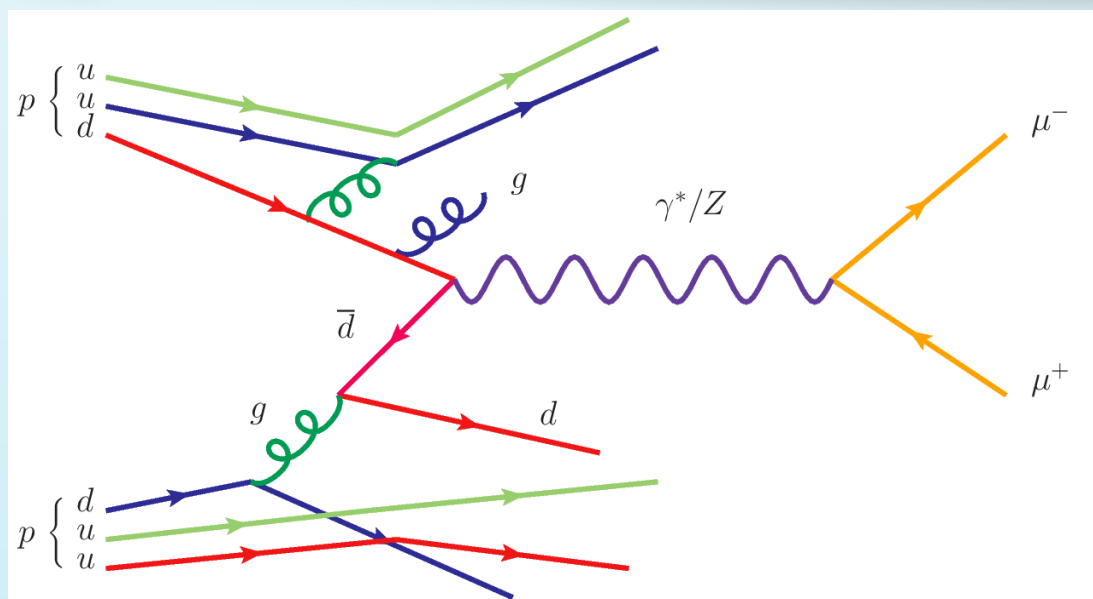
Lead to truly game-changer applications, such as **random generation of country song lyrics**

Tied right now
I got life now he never thought I got by the all
Going up like a house four boy
Nothing his thing out of hands
No one with the danger in the world
I love my black fire as I know
But the short knees just around me
Fun the heart couldnes fall to back
I see a rest of my wild missing far
When I was missing to wait
And if I think
It's a real tame
I say I belong is every long night
Maybe lovin' you

<http://www.mattmoocar.me/blog/RNNCountryLyrics/>

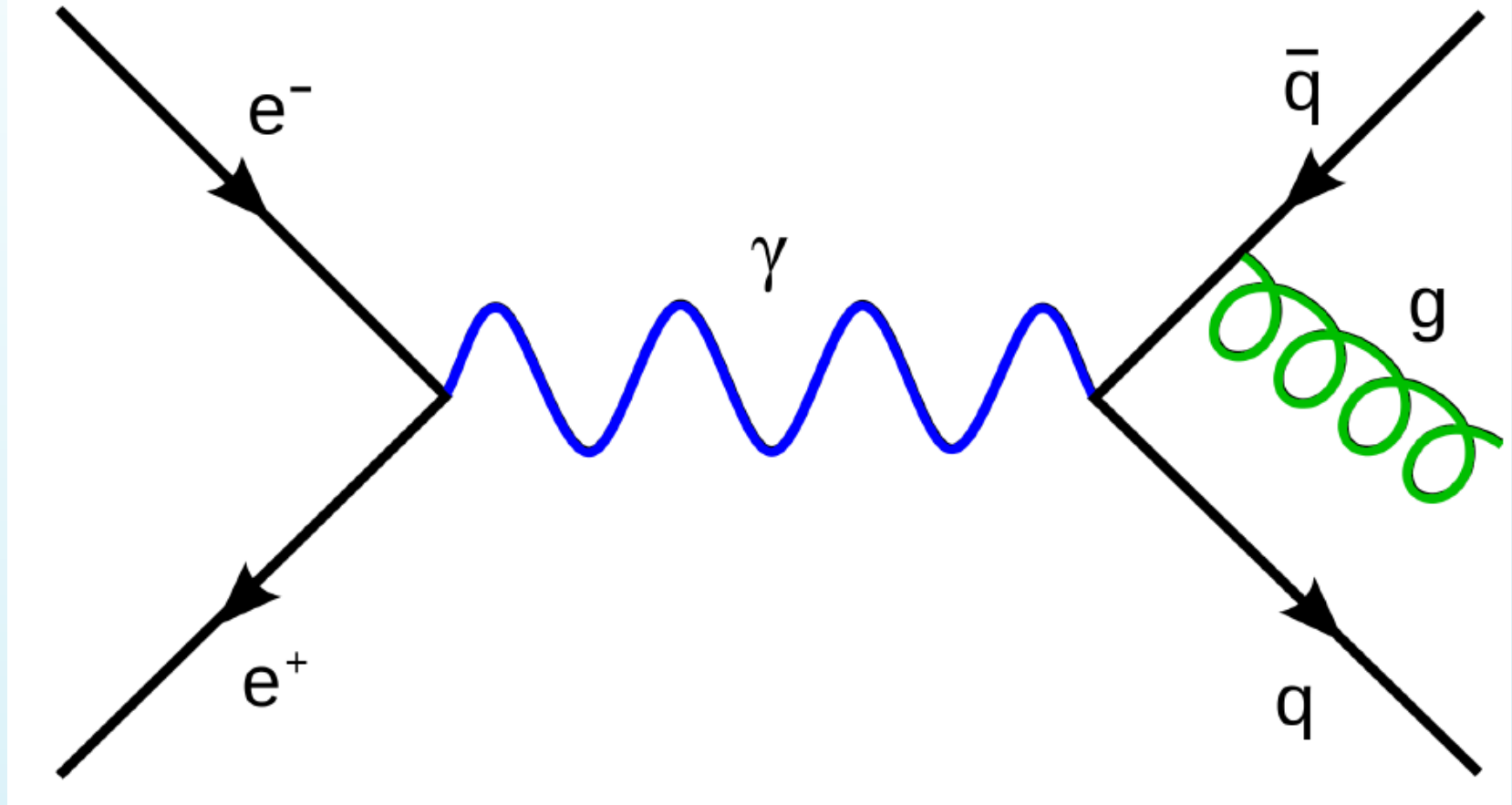
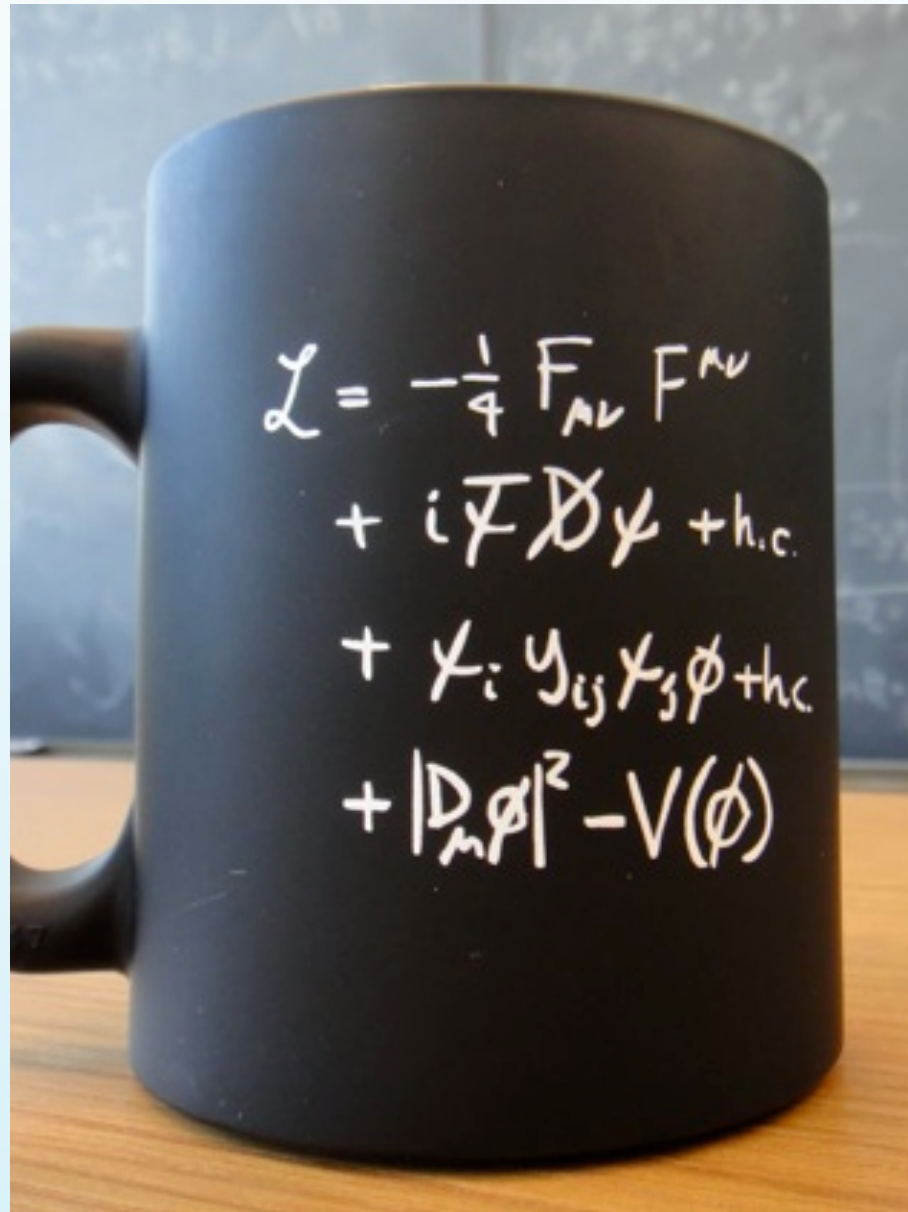


A crash course on parton distributions



Lepton vs Hadron Colliders

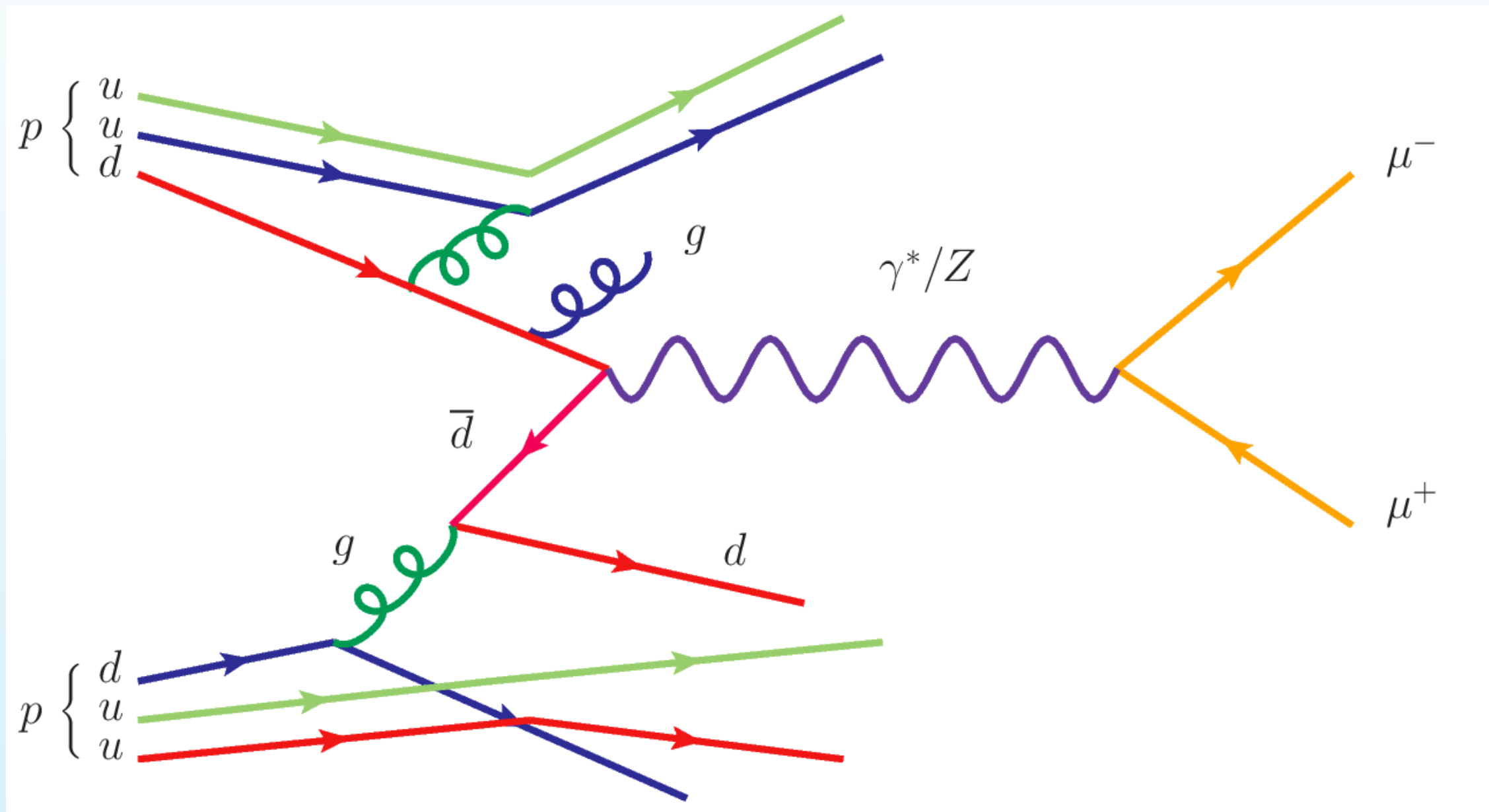
In high-energy lepton colliders, such as the **Large Electron-Positron Collider (LEP)** at CERN, the collisions involve **elementary particles** without substructure



Cross-sections in lepton colliders can be computed in perturbation theory using the Feynman rules of the **Standard Model Lagrangian**

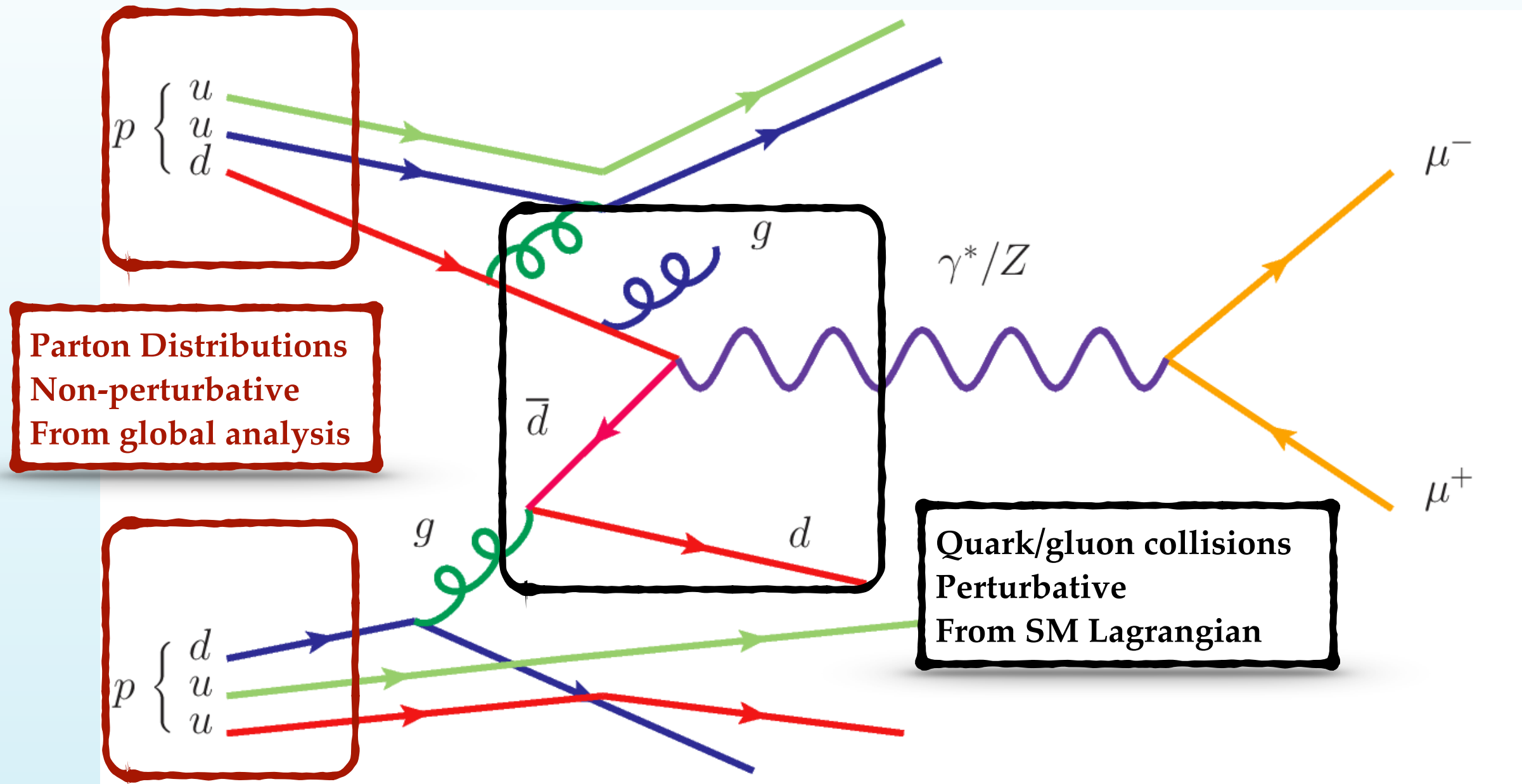
Lepton vs Hadron Colliders

In high-energy **hadron colliders**, such as the LHC, the collisions involve **composite particles** (protons) with internal structure (quarks and gluons)

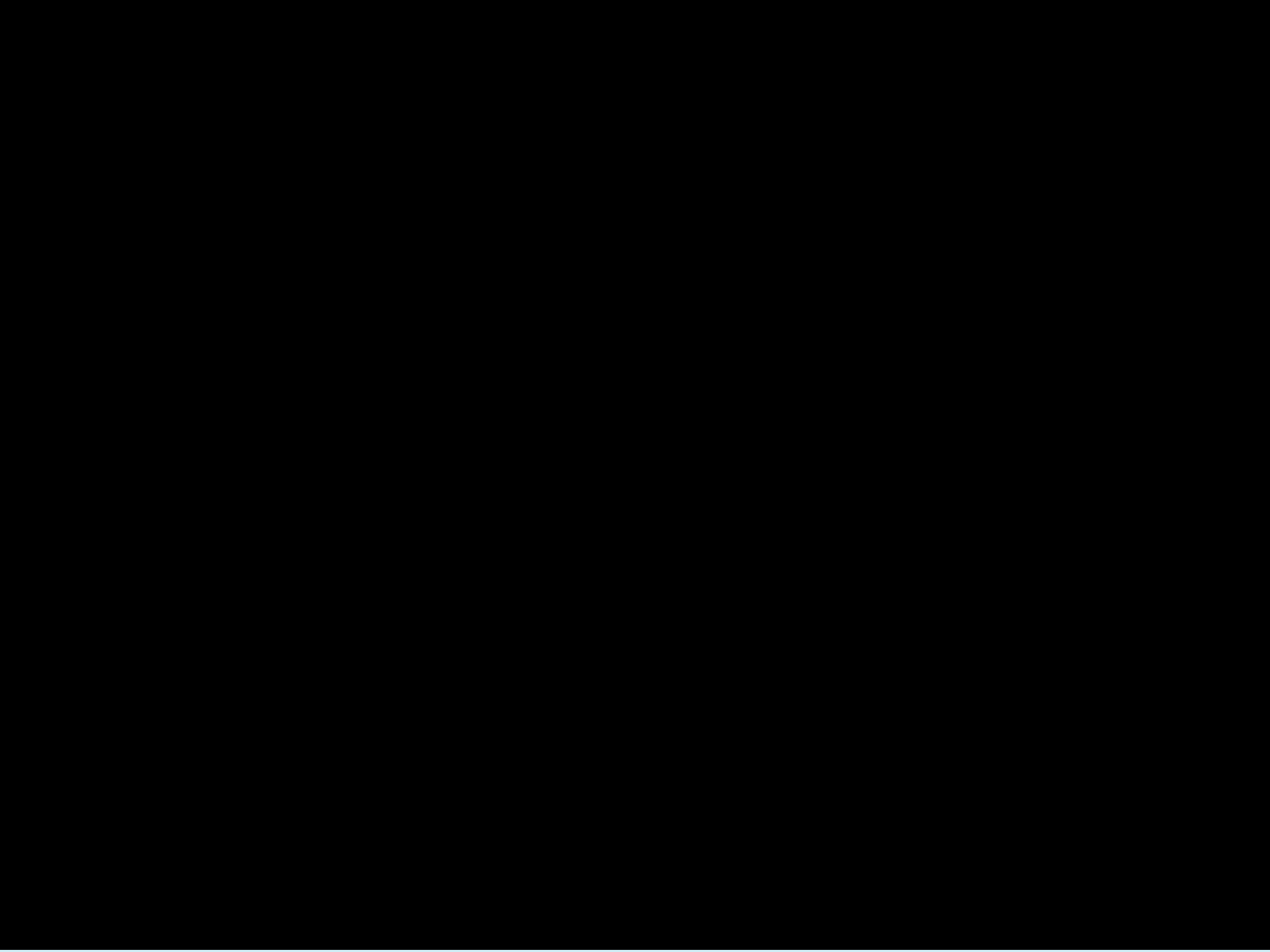


Lepton vs Hadron Colliders

In high-energy **hadron colliders**, such as the LHC, the collisions involve **composite particles** (protons) with internal structure (quarks and gluons)



Calculations of **cross-sections** in hadron collisions require the combination of **perturbative, quark/gluon-initiated processes**, and **non-perturbative, parton distributions**, information



Parton Distributions

The distribution of energy that quarks and gluons carry inside the proton is quantified by the **Parton Distribution Functions (PDFs)**

$$g(x, Q)$$

Q : Energy of the quark/gluon collision
Inverse of the resolution length

$g(x, Q)$: Probability of finding a gluon inside a proton, carrying a fraction x of the proton momentum, when probed with energy Q

x : Fraction of the proton's momentum

Parton Distributions

The distribution of energy that **quarks and gluons** carry inside the **proton** is quantified by the **Parton Distribution Functions (PDFs)**

$$g(x, Q)$$

Q : Energy of the quark/gluon collision
Inverse of the resolution length

$g(x, Q)$: Probability of finding a gluon inside a proton, carrying a fraction x of the proton momentum, when probed with energy Q

x : Fraction of the proton's momentum

PDFs are determined by non-perturbative QCD dynamics, cannot be computed from first principles, and need to be **extracted from experimental data** with a **global analysis**

Parton Distributions

The distribution of energy that **quarks and gluons** carry inside the **proton** is quantified by the **Parton Distribution Functions (PDFs)**

$$g(x, Q)$$

Q: Energy of the quark/gluon collision
Inverse of the resolution length

$g(x, Q)$: Probability of finding a gluon inside a proton, carrying a fraction x of the proton momentum, when probed with energy Q

x : Fraction of the proton's momentum

PDFs are determined by **non-perturbative QCD dynamics**, cannot be computed from first principles, and need to be **extracted from experimental data** with a **global analysis**

📌 **Energy conservation**

$$\int_0^1 dx \left(g(x, Q) + \sum_q q(x, Q) \right) = 1$$

📌 **Dependence with quark/gluon collision energy Q determined in perturbation theory**

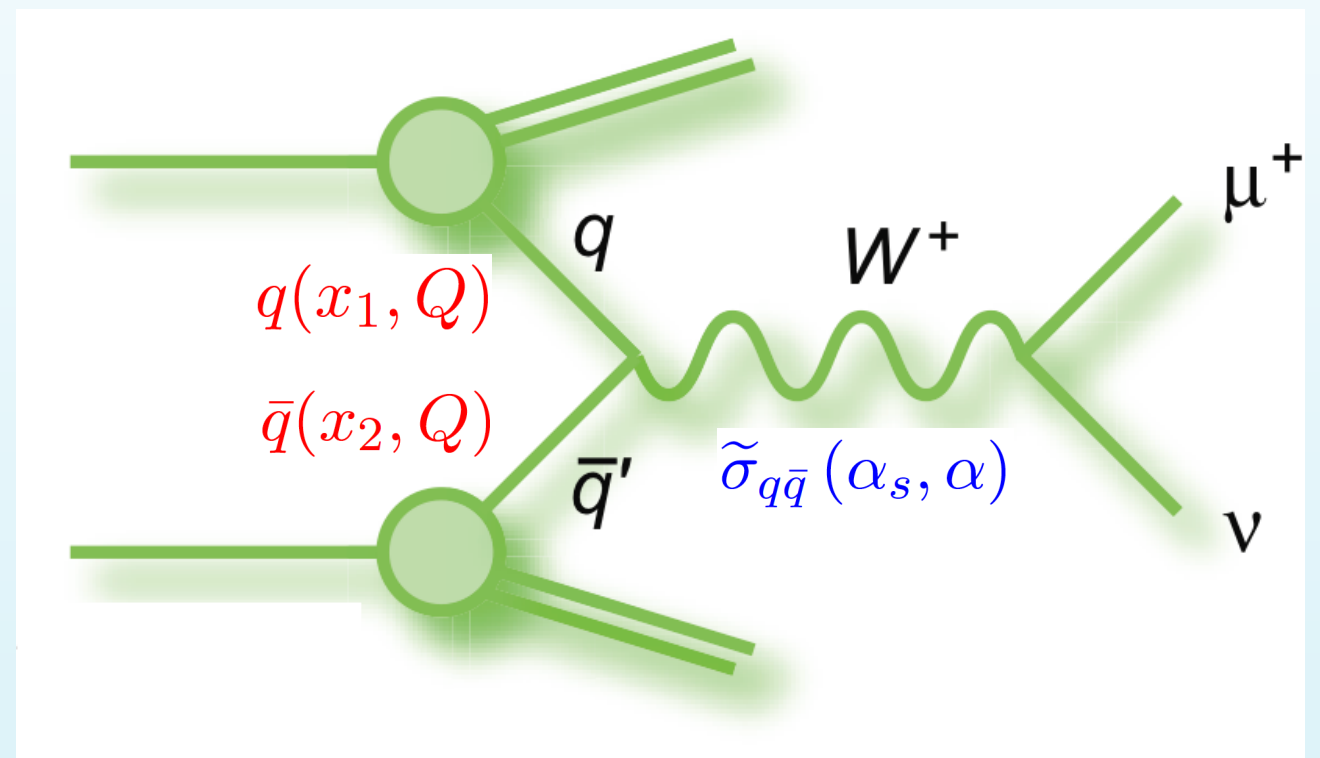
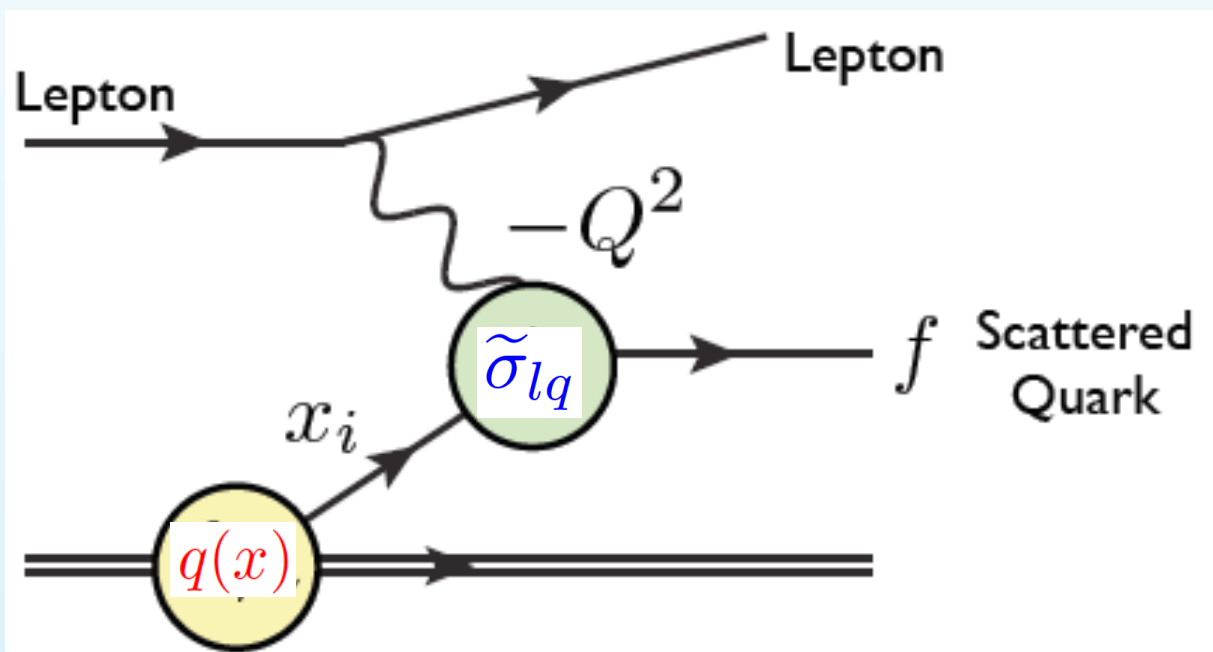
$$\frac{\partial g(x, Q)}{\partial \ln Q} = P_g(\alpha_s) \otimes g(x, Q) + P_q(\alpha_s) \otimes q(x, Q)$$

The QCD Factorization Theorem

The QCD factorization theorem guarantees PDF universality: extract them from a subset of process and use them to provide pure predictions for new processes

$$\sigma_{lp} \simeq \tilde{\sigma}_{lq}(\alpha_s, \alpha) \otimes q(x, Q)$$

$$\sigma_{pp} \simeq \tilde{\sigma}_{q\bar{q}}(\alpha_s, \alpha) \otimes q(x_1, Q) \otimes \bar{q}(x_2, Q)$$

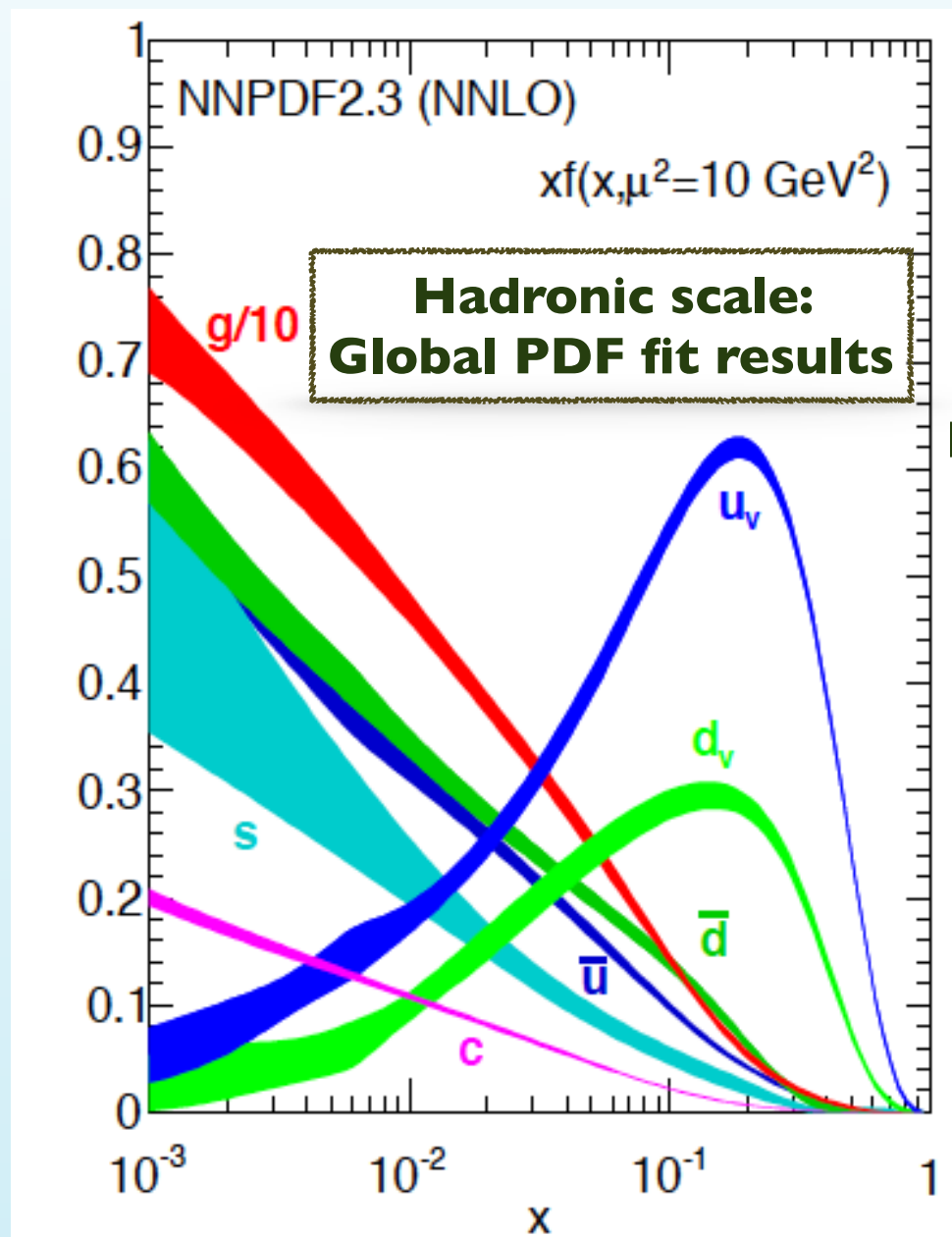


Determine PDFs in lepton-proton collisions

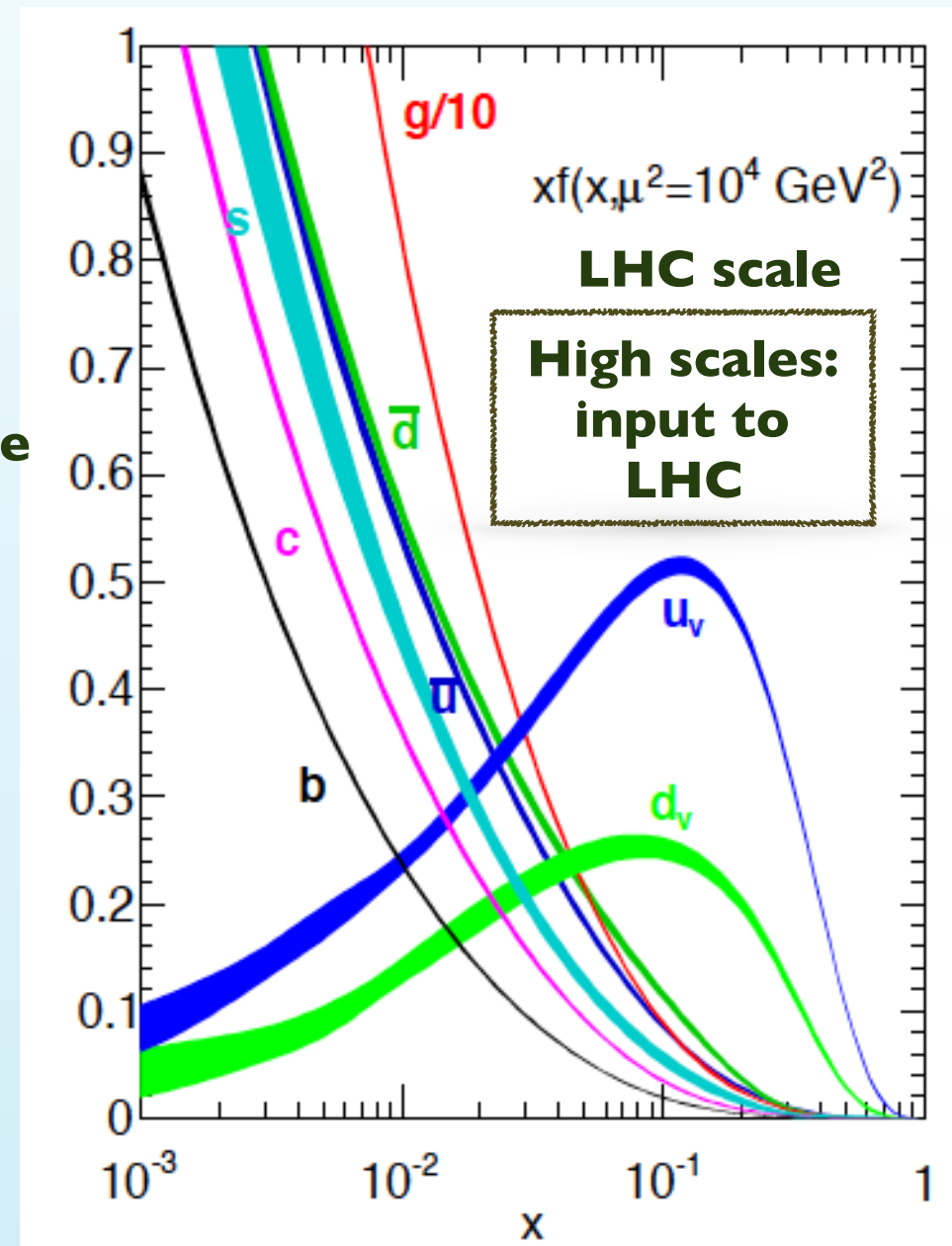
And use them to compute cross-sections in proton-proton collisions at the LHC

The global PDF analysis

- Combine **state-of-the-art theory calculations**, the constraints from PDF-sensitive measurements from different processes and colliders, and a **statistically robust fitting methodology**
- Extract **Parton Distributions** at hadronic scales of a few GeV, where non-perturbative QCD sets in
- Use **perturbative evolution** to compute PDFs at high scales as **input to LHC predictions**



Perturbative Evolution
→



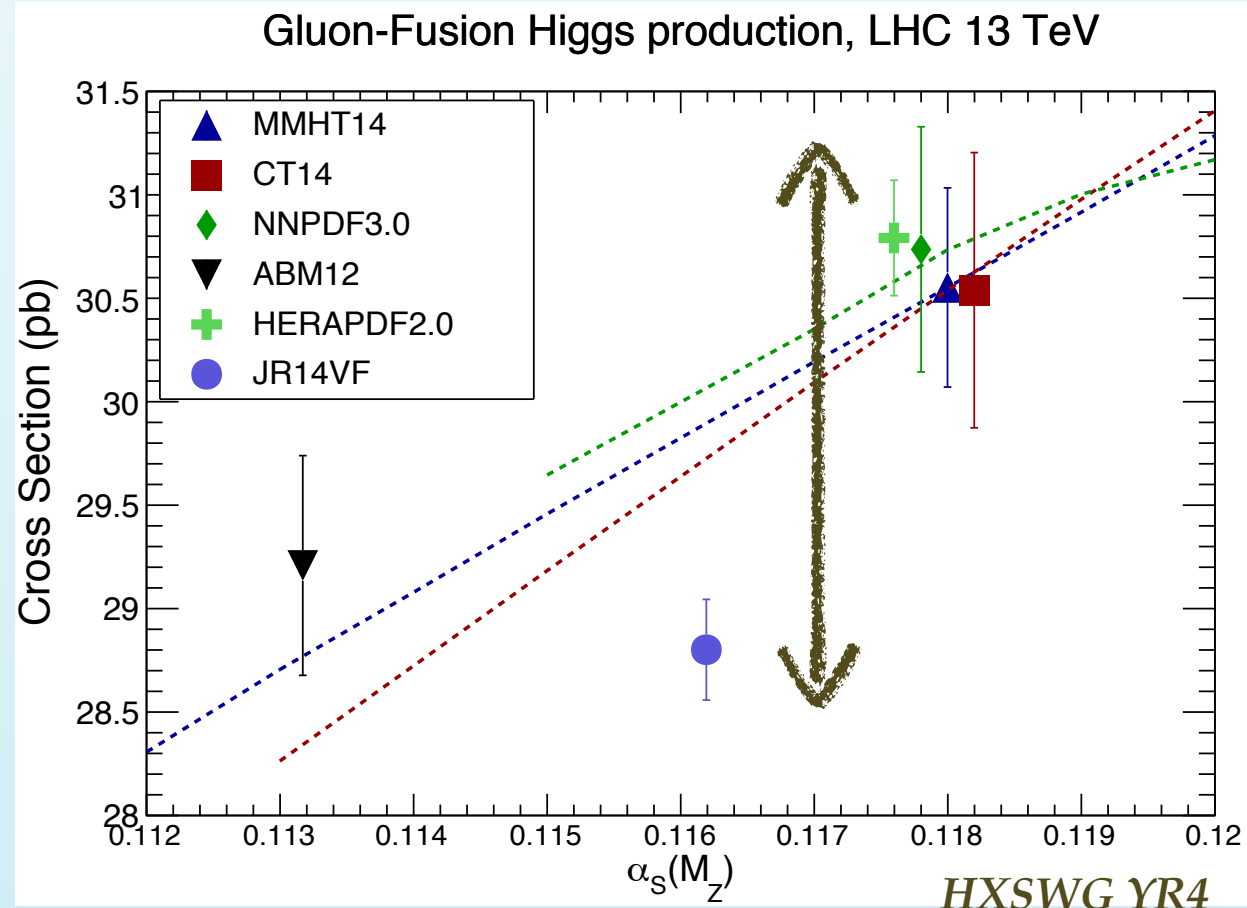
Why better PDFs?

Dominant TH unc for M_W measurements at LHC

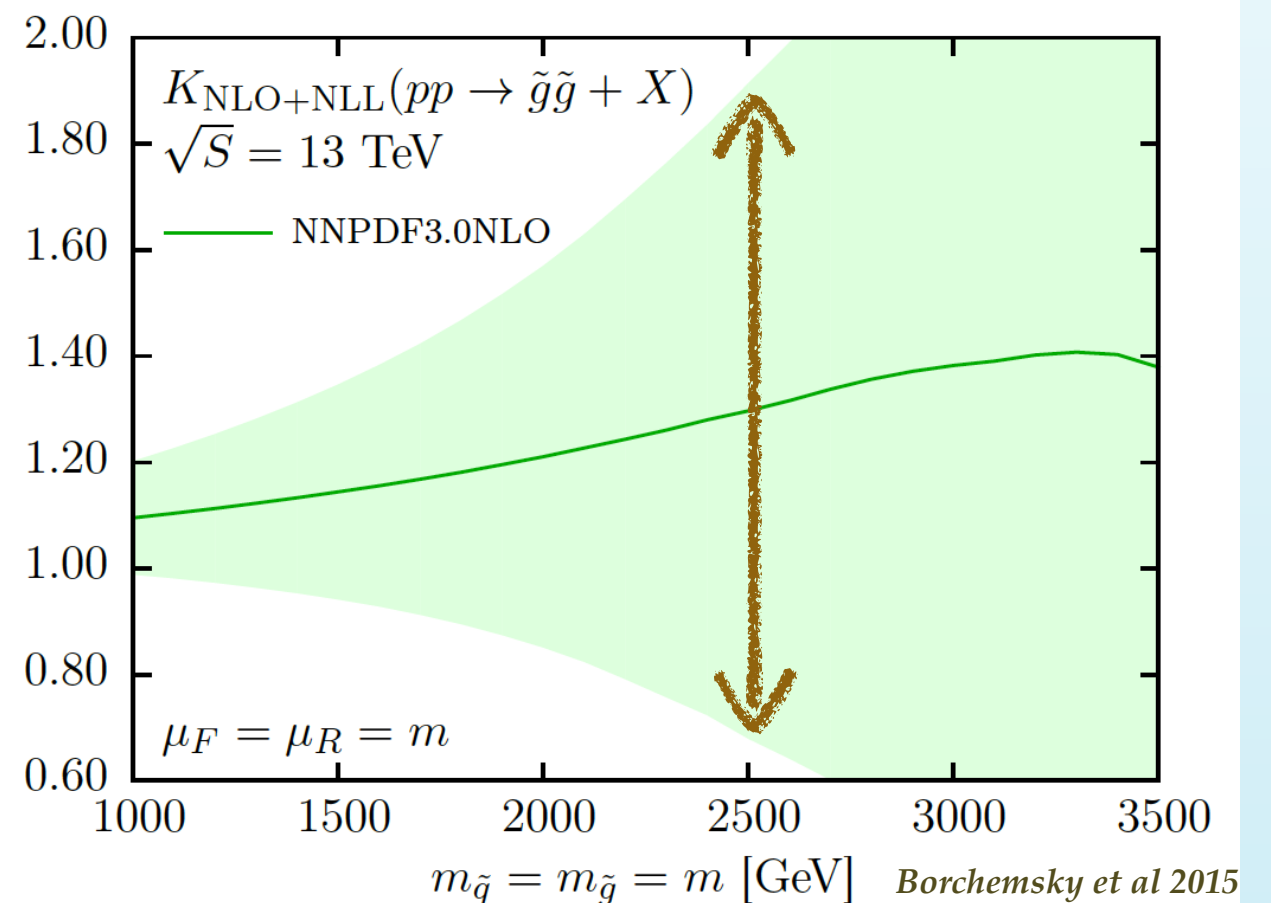
ATLAS 2017

Channel	$m_{W^+} - m_{W^-}$ [MeV]	Stat. Unc.	Muon Unc.	Elec. Unc.	Recoil Unc.	Bckg. Unc.	QCD Unc.	EW Unc.	PDF Unc.	Total Unc.
$W \rightarrow e\nu$	-29.7	17.5	0.0	4.9	0.9	5.4	0.5	0.0	24.1	30.7
$W \rightarrow \mu\nu$	-28.6	16.3	11.7	0.0	1.1	5.0	0.4	0.0	26.0	33.2
Combined	-29.2	12.8	3.3	4.1	1.0	4.5	0.4	0.0	23.9	28.0

Higgs coupling measurements

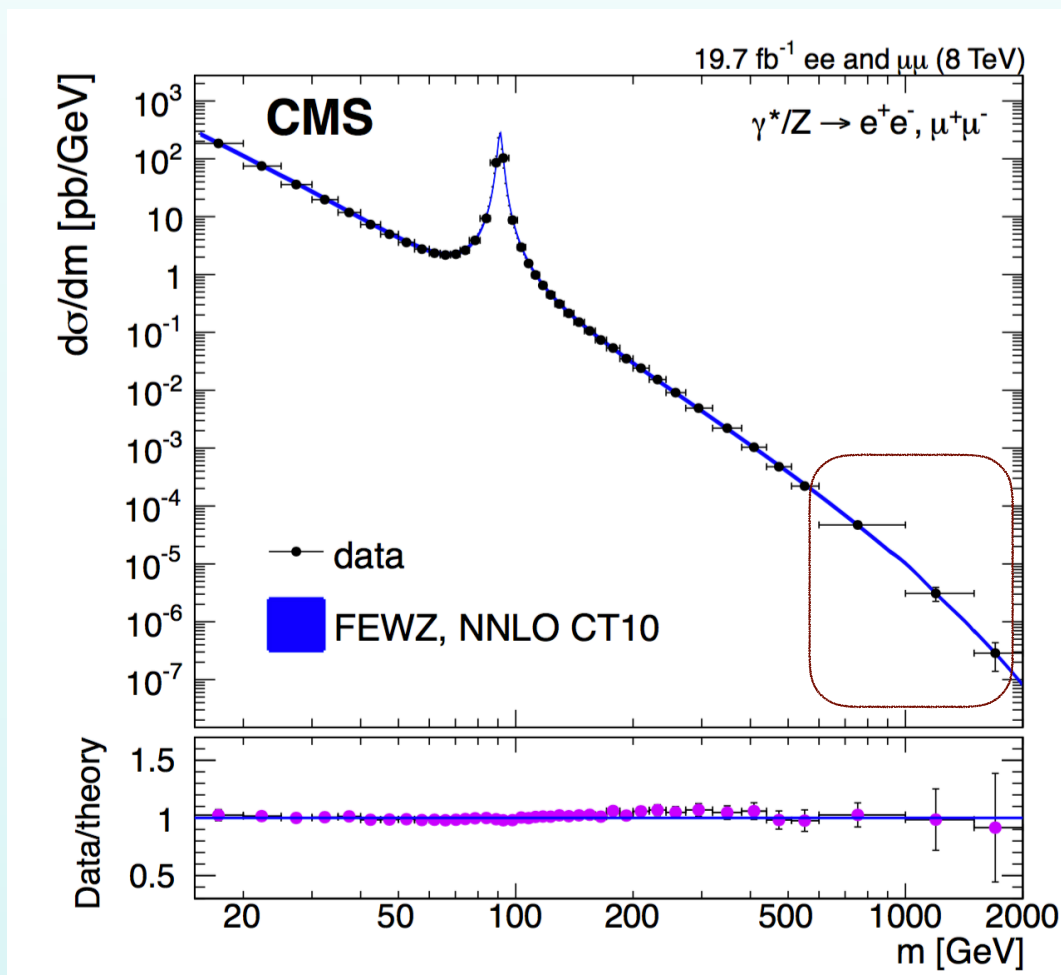


High-mass BSM cross-sections



Why better PDFs?

- BSM physics could manifest as **subtle deviations** wrt to the Standard Model predictions
- Even for high-mass resonances, **PDF uncertainties degrade or limit many BSM searches**
- The robustness of **global stress-tests of the SM** (electroweak fit, SM Effective Field Theory analysis) relies crucially in high-precision theoretical calculations



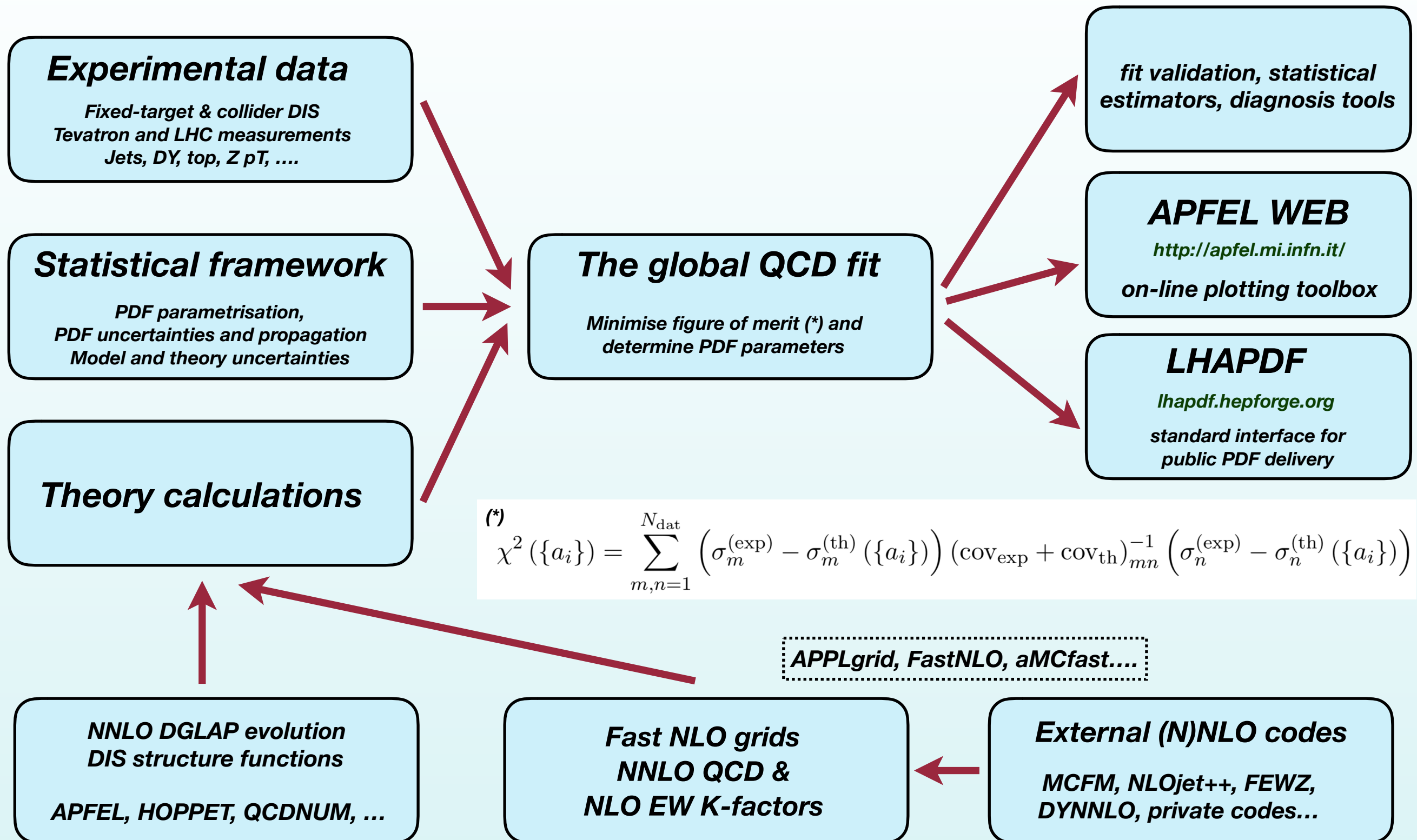
SMEFT expansion

$$\sigma(E) = \sigma_{SM}(E) \left(1 + \epsilon \frac{m_{SM}^2}{m_W^2} + \epsilon \frac{E^2}{m_W^2} + \dots \right)$$

For $E \simeq 1 \text{ TeV}$, a measurement with $\delta\sigma/\sigma \simeq 10\%$ is sensitive to $\epsilon \simeq \mathcal{O}(0.1\%)$!

BSM physics might very well hiding itself in the tails of LHC distributions, but need to make sure first that PDF uncertainties are under control

The global QCD fit machinery



The NNPDF approach

A novel approach to PDF determination, improving the limitations of the traditional PDF fitting methods with the use of advanced statistical techniques such as machine learning and multivariate analysis

Non-perturbative PDF parametrization

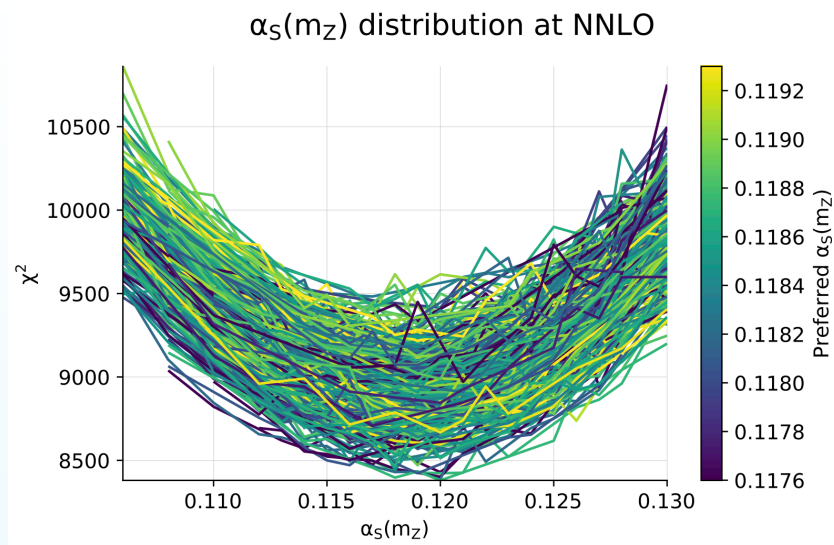
- Traditional approach: based on restrictive functional forms leading to strong theoretical bias
- NNPDF solution: use Artificial Neural Networks as universal unbiased interpolants

PDF uncertainties and propagation to LHC calculations

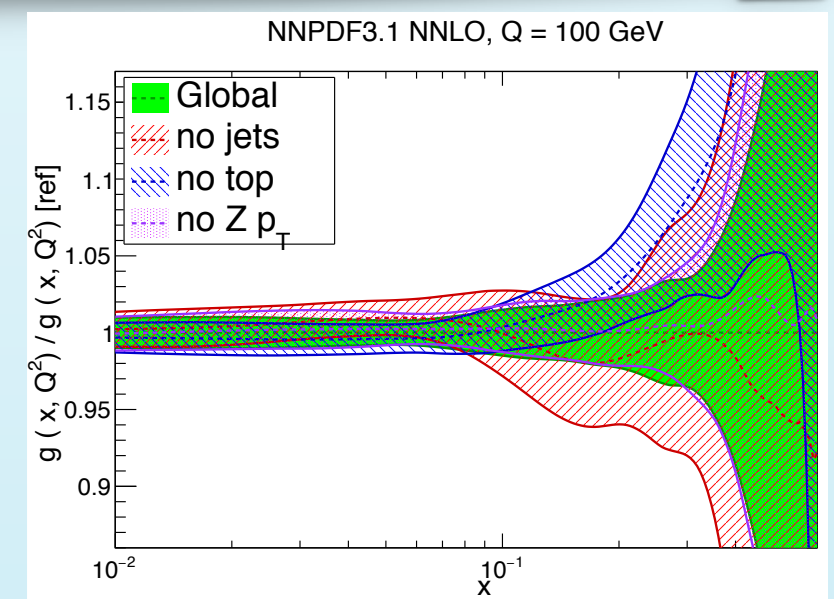
- Traditional approach: limited to Gaussian/linear approximation
- NNPDF solution: based on the Monte Carlo replica method to create a probability distribution in the space of PDFs. Specially critical in extrapolation regions (i.e. high- x) for New Physics searches

Fitting technique

- Traditional approach: deterministic minimization of χ^2 , flat directions problem
- NNPDF solution: Genetic Algorithms to explore efficiently the vast parameter space, with cross-validation to avoid fitting stat fluctuations

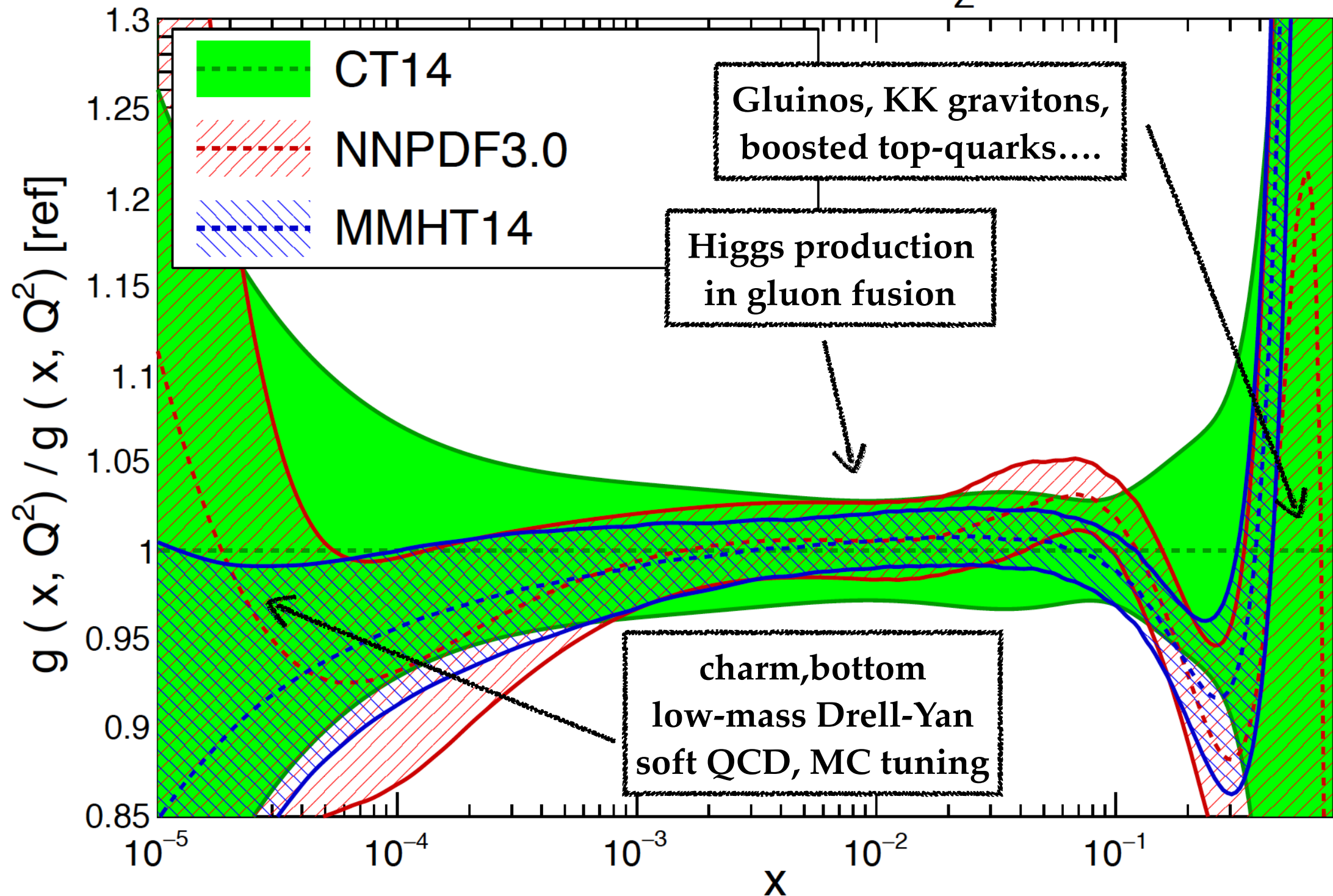


The structure of the proton and LHC phenomenology



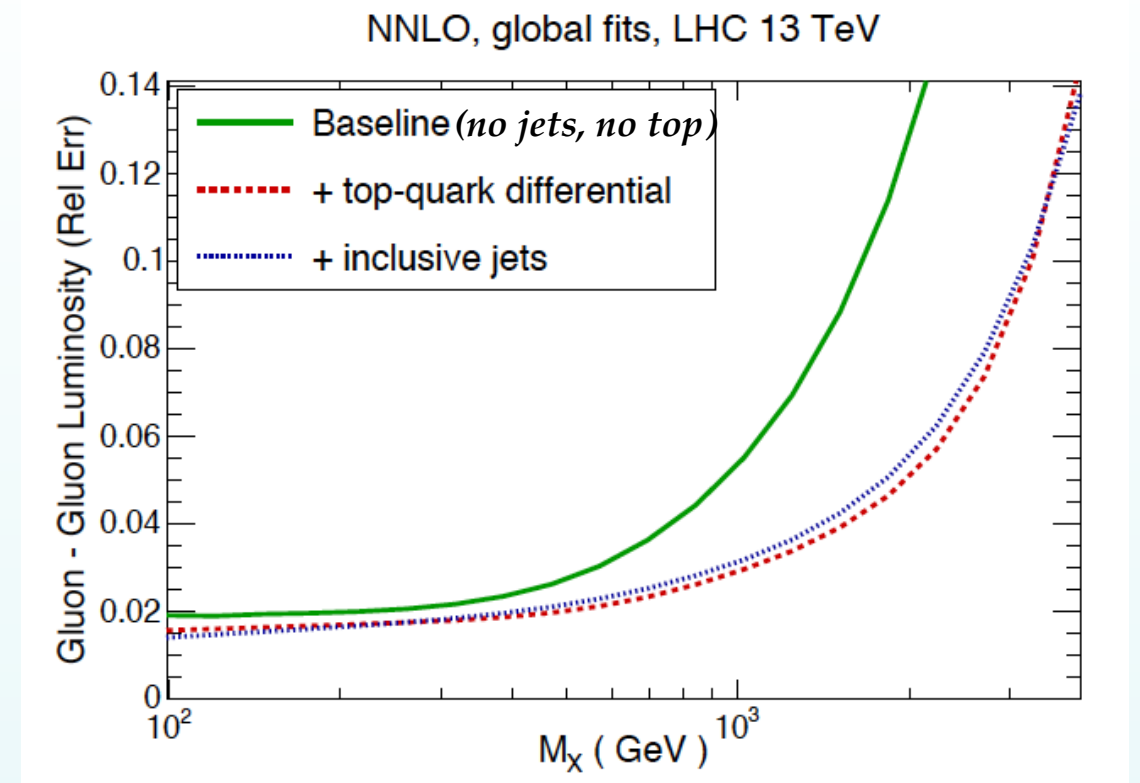
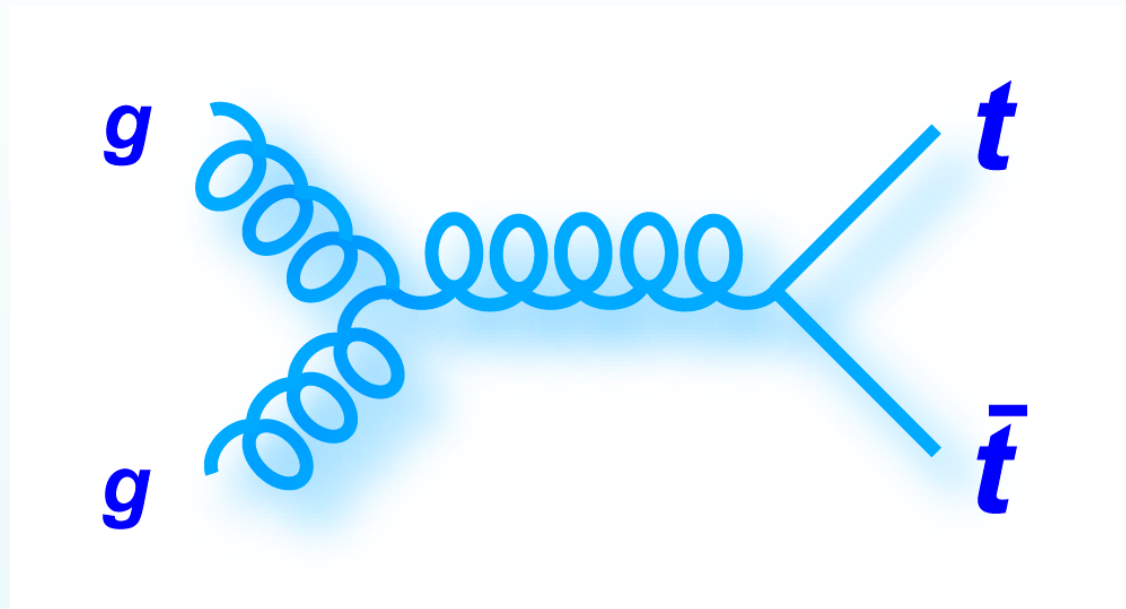
One glue to bind them all

NNLO, $Q^2=100 \text{ GeV}^2$, $\alpha_S(M_Z)=0.118$

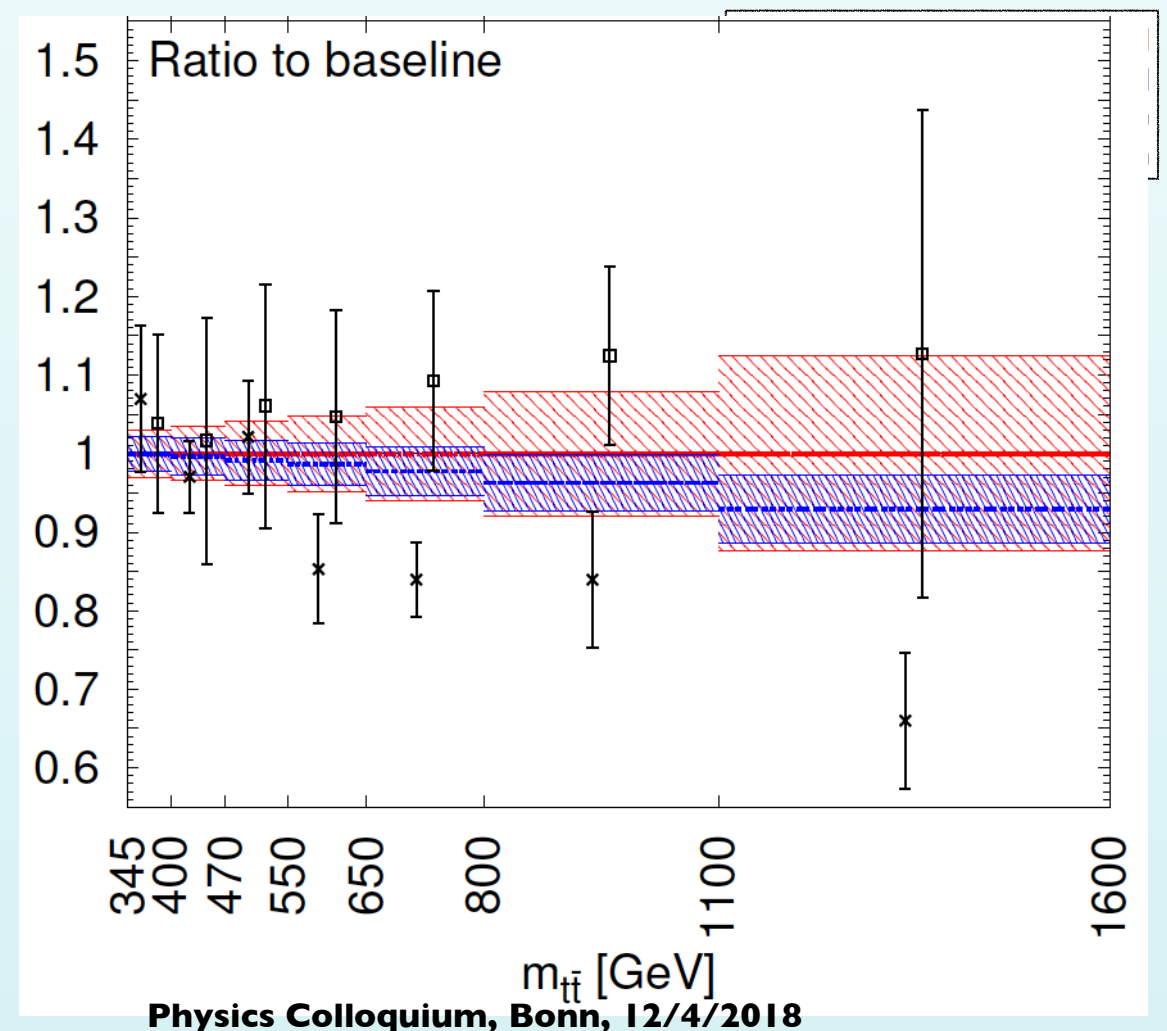


At the LHC, precise knowledge of the gluon is required from small- x to large- x

The large- x gluon from differential top quarks

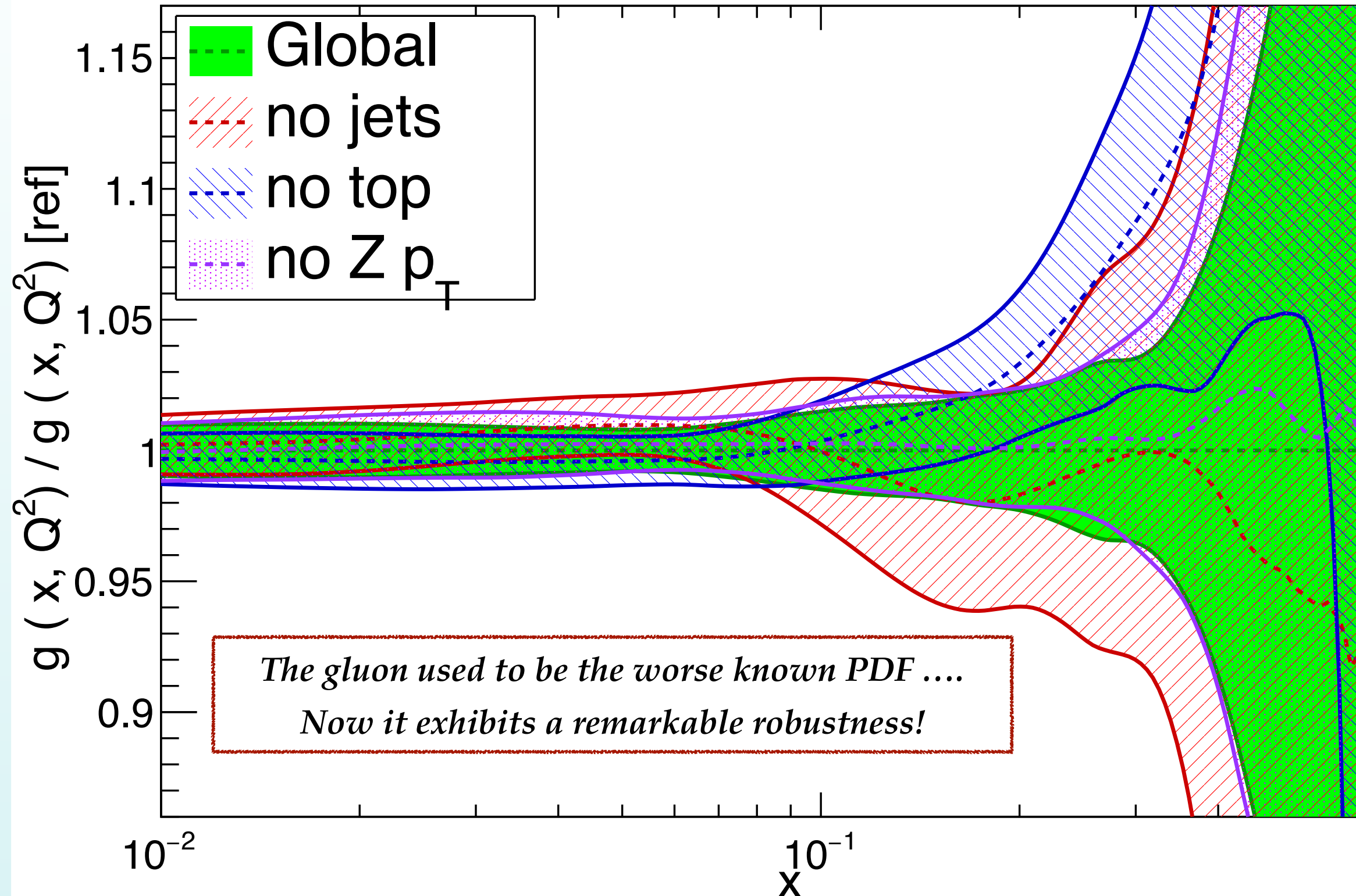


- Top-quark production driven by **gluon-gluon scattering**
- NNLO calculations for stable top quarks available
Czakon, Mitov et al 2015-2017
- Data from ATLAS and CMS at 8 TeV available with breakdown of systematic uncertainties
- Included **differential top data into NNPDF3.0**: constraints on the **large- x gluon** comparable to those of inclusive jet production *Czakon et al 2017*
- Improved theory uncertainties in **regions crucial for BSM searches**, *i.e.*, $m_{t\bar{t}} > 1$ TeV (while fitting only y_t and $y_{t\bar{t}}$)



One (upgraded) glue to bind them all

NNPDF3.1 NNLO, $Q = 100$ GeV



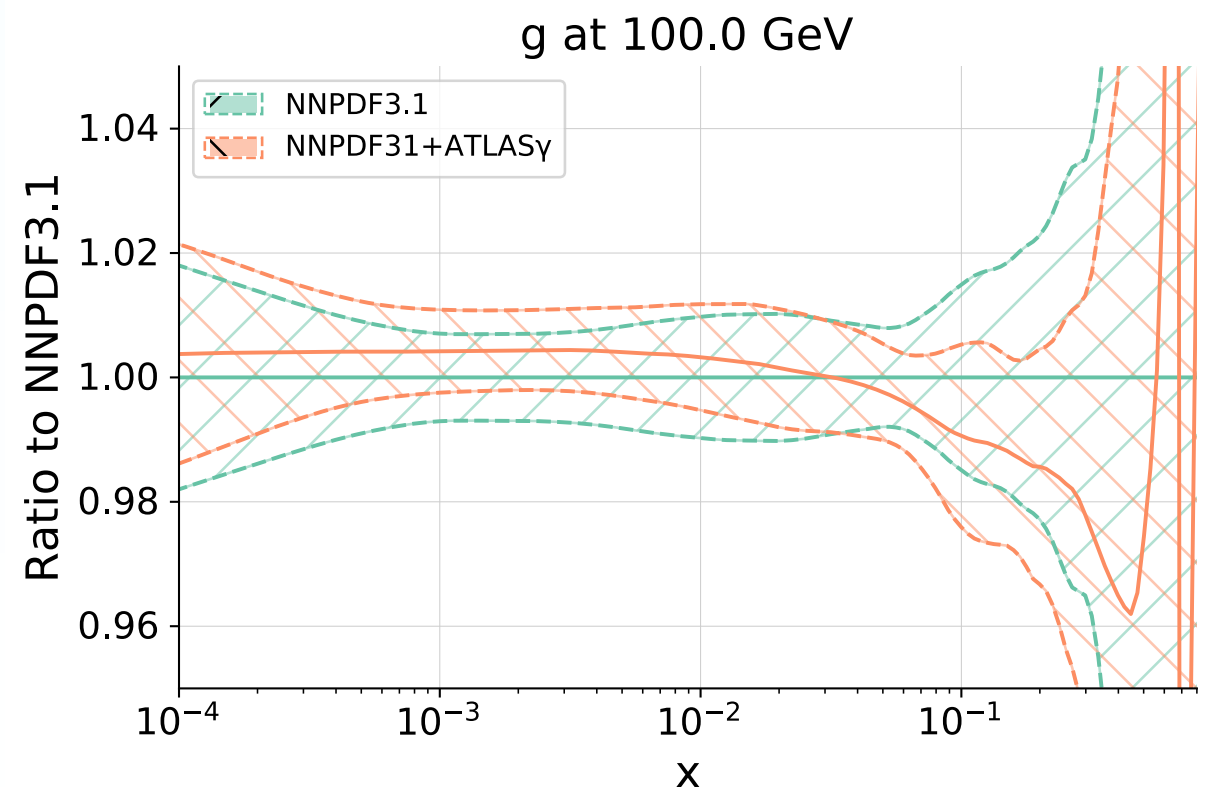
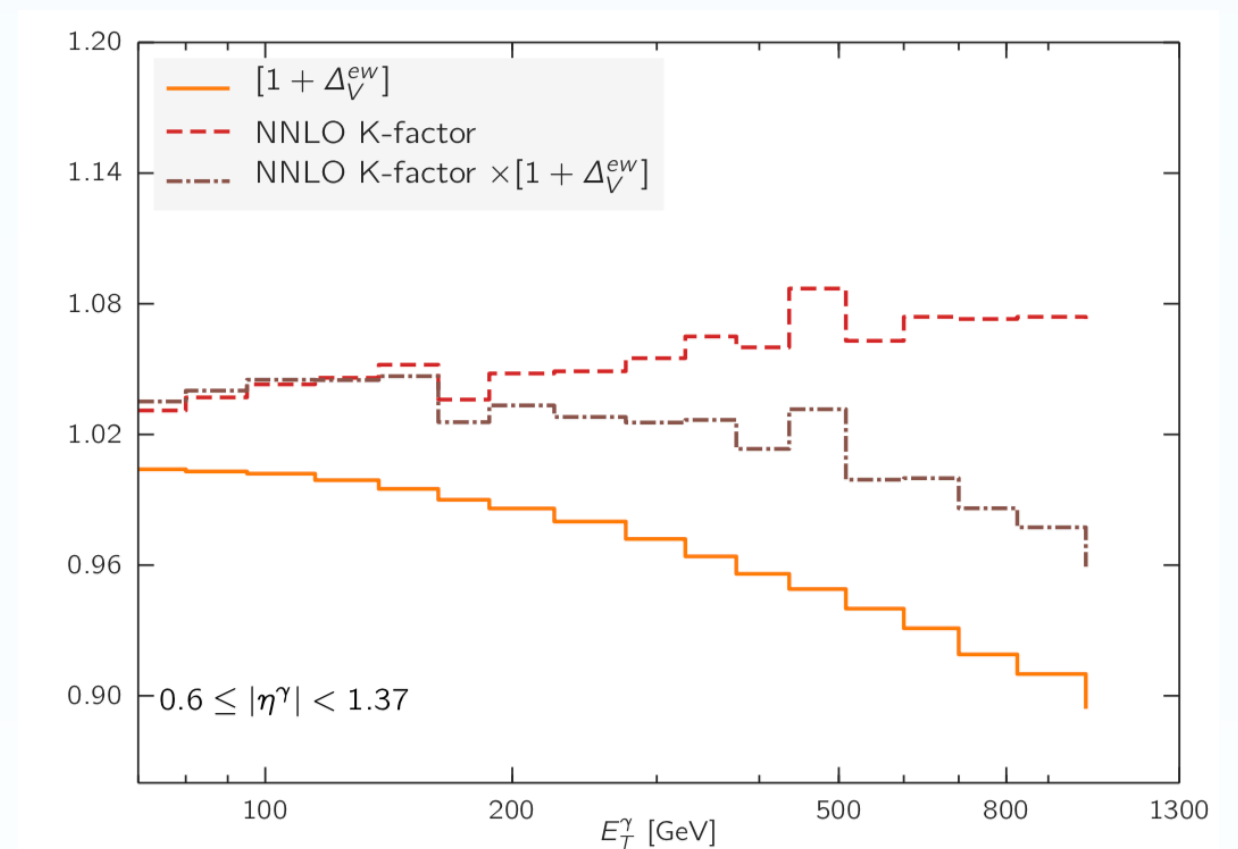
Direct photon production and PDF fits

- Recently we have revisited the impact of LHC photon data into the global PDF fit, specifically the ATLAS 8 TeV data

Campbell, Rojo, Slade, Williams 18

- Theory based on **NNLO QCD** and **LL electroweak** calculations
- Moderate impact on the medium- x gluon, consistent with previous studies at NLO
- Good consistency with the rest of gluon-sensitive experiments in NNPDF3.1

	NNPDF3.1	NNPDF3.1+ATLAS γ
Fixed-target lepton DIS	1.207	1.203
Fixed-target neutrino DIS	1.081	1.087
HERA	1.166	1.169
Fixed-target Drell-Yan	1.241	1.242
Collider Drell-Yan	1.356	1.346
Top-quark pair production	1.065	1.049
Inclusive jets	0.939	0.915
$Z p_T$	0.997	0.980
Total dataset	1.148	1.146



The small-x gluon from forward charm production

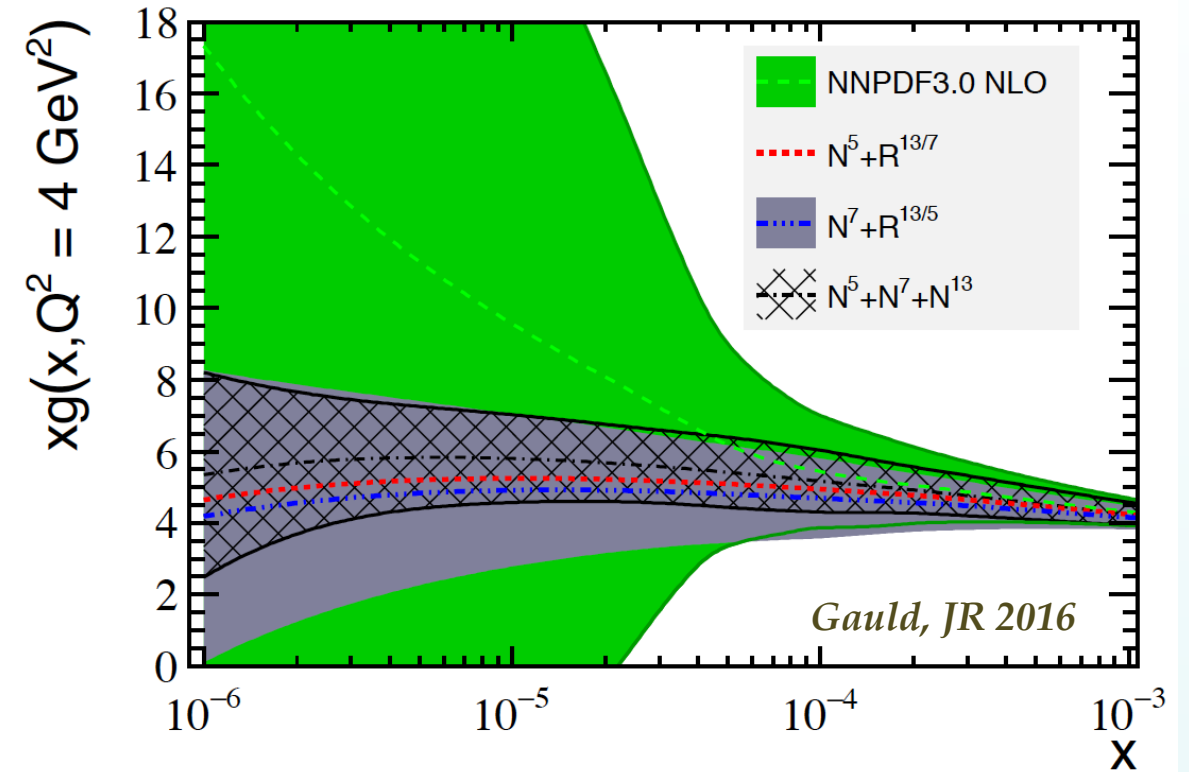
• **D and B meson production from LHCb** allow accessing the gluon down to $x \approx 10^{-6}$, well below the HERA coverage

PROSA 2015, Gauld et al 2015

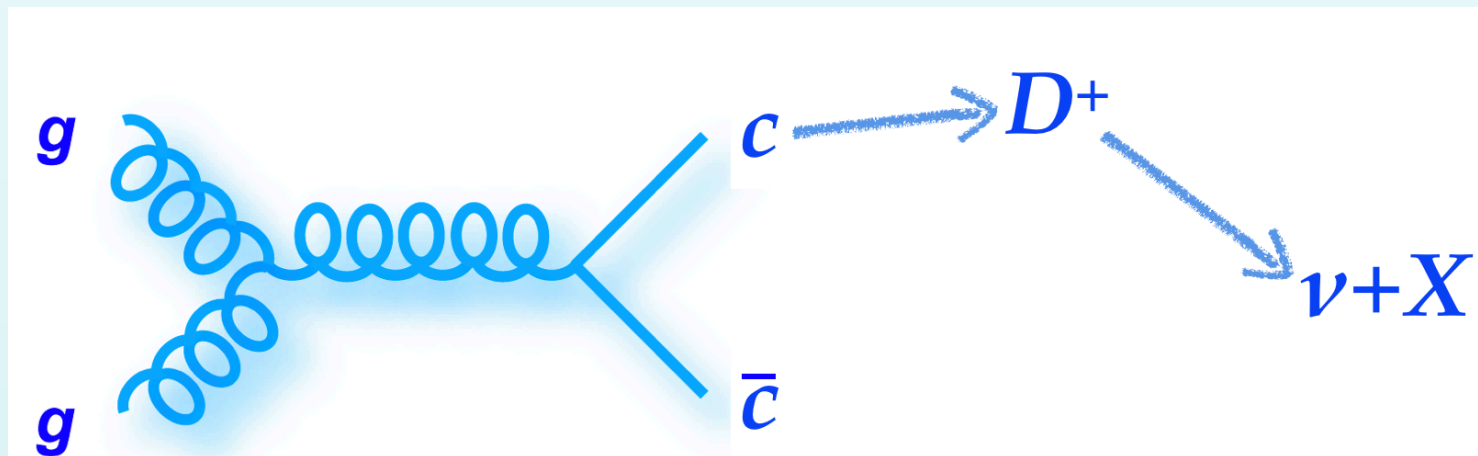
• Gluon PDF errors reduced by up to a factor 10!

• Allows robust estimate for the *prompt neutrino flux*, the main background for astrophysical neutrinos at IceCube

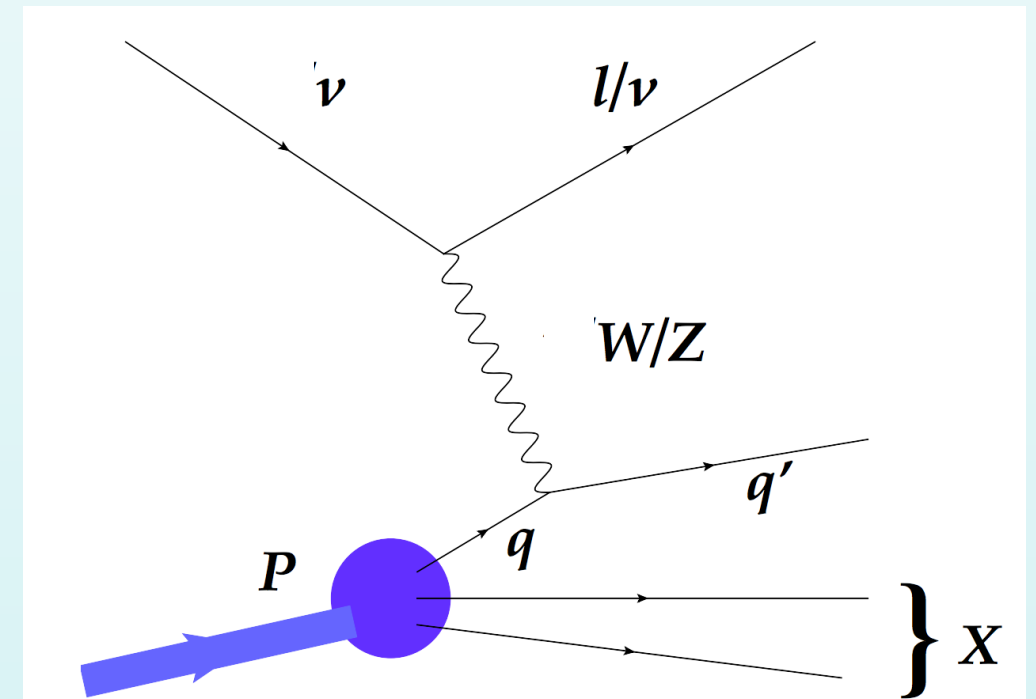
• Precision calculation of the **UHE neutrino-nucleus cross-section**, with few-percent TH errors up to $E_\nu = 10^{12}$ GeV



Prompt neutrino flux at IceCube



UHE neutrino-nucleus xsecs



The small-x gluon from forward charm production

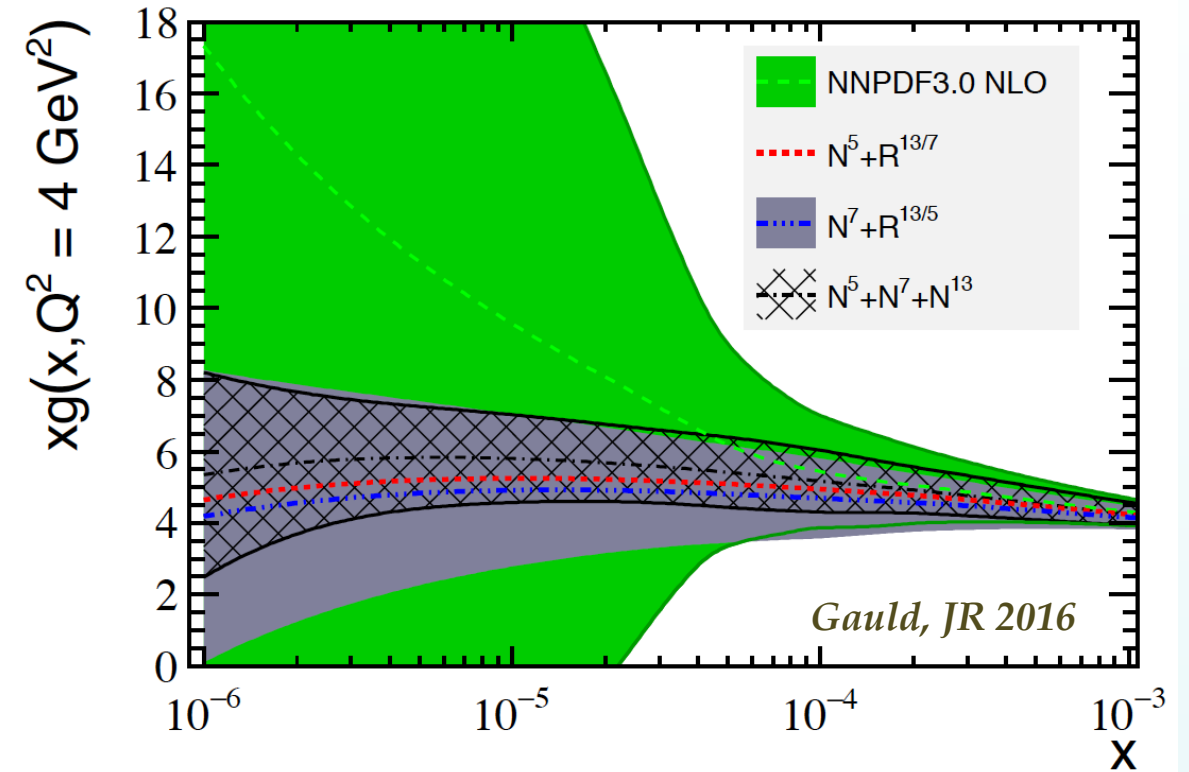
• **D and B meson production from LHCb** allow accessing the gluon down to $x \approx 10^{-6}$, well below the HERA coverage

PROSA 2015, Gauld et al 2015

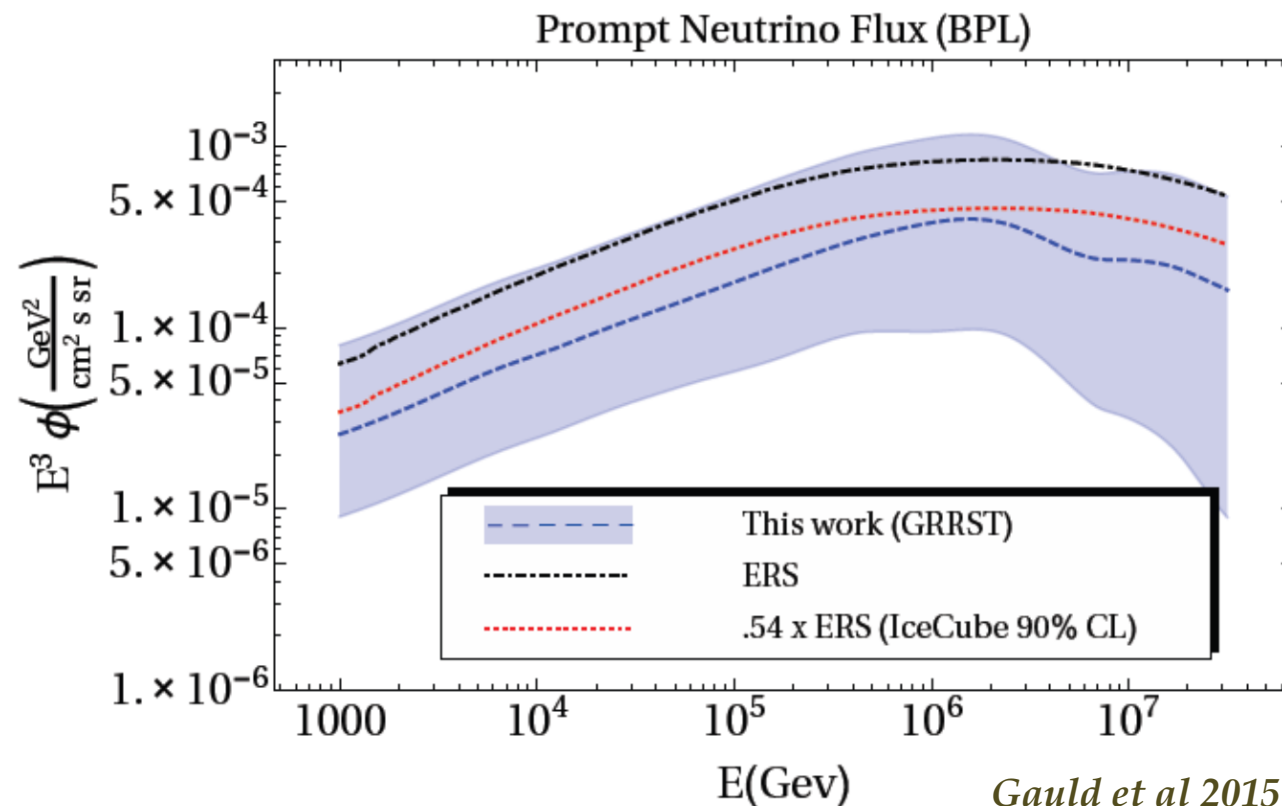
• Gluon PDF errors reduced by up to a factor 10!

• Allows robust estimate for the *prompt neutrino flux*, the main background for astrophysical neutrinos at IceCube

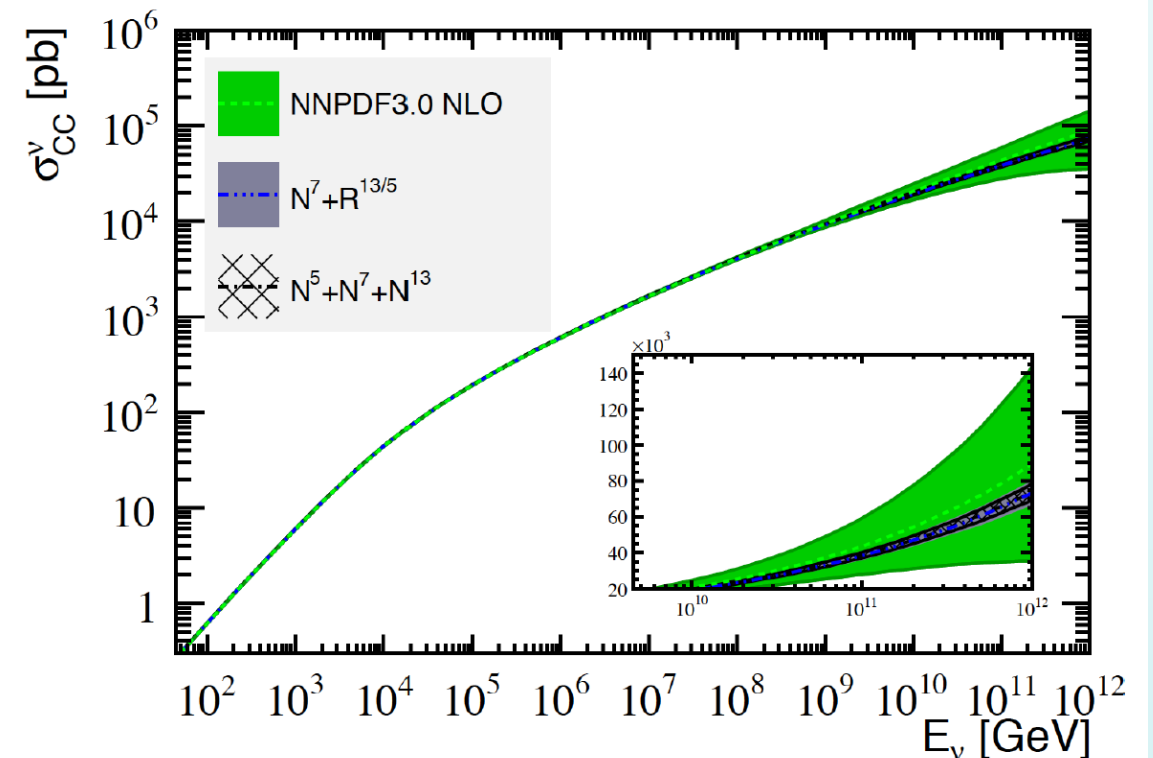
• Precision calculation of the **UHE neutrino-nucleus cross-section**, with few-percent TH errors up to $E_\nu = 10^{12}$ GeV



Prompt neutrino flux at IceCube



UHE neutrino-nucleus xsecs



How bright is the proton?

- The calculation of **QED and electroweak corrections to hadron collider processes** requires by consistency to introduce the PDF of the photon in the proton, $\gamma(x, Q)$
- The first model-independent determination of $\gamma(x, Q)$ from LHC W,Z data was **NNPDF2.3QED**, which however affected by **large uncertainties**, due to the limited experimental information
- Recently, $\gamma(x, Q)$ **computed** in terms of the well-known **inclusive structure functions F_2 and F_L** : the resulting photon PDF, exhibits now **few-percent uncertainties**

$$x\gamma(x, \mu) = \frac{1}{2\pi\alpha(\mu)} \int_x^1 \frac{dz}{z} \left\{ \int_{Q_{\min}^2}^{\mu^2/(1-z)} \frac{dQ^2}{Q^2} \alpha^2(Q^2) \left[-z^2 F_L(x/z, Q^2) + \left(zP_{\gamma q}(z) + \frac{2x^2 m_p^2}{Q^2} \right) F_2(x/z, Q^2) \right] - \alpha^2(\mu) z^2 F_2(x/z, \mu^2) \right\} + \mathcal{O}(\alpha\alpha_s, \alpha^2)$$

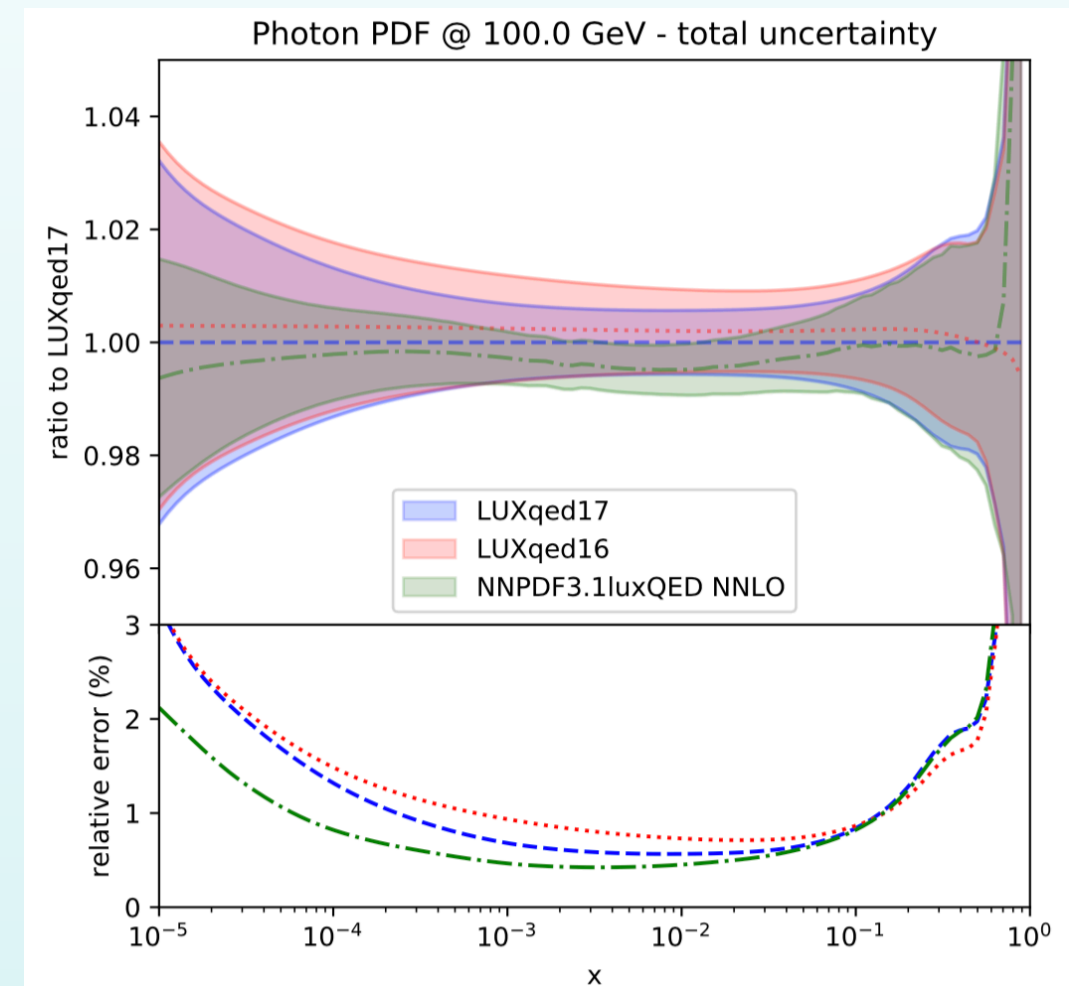
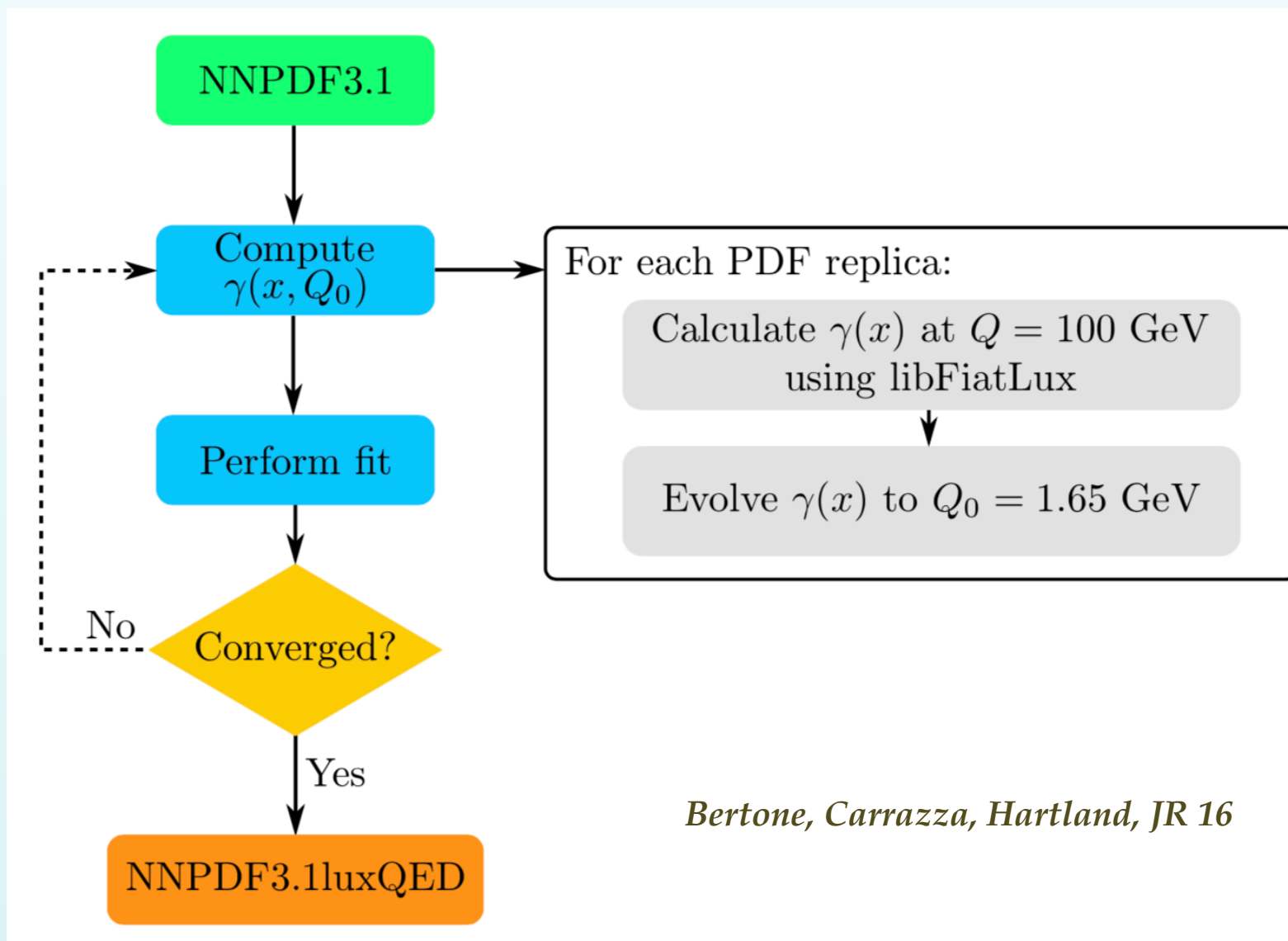
pp \rightarrow H W ⁺ (\rightarrow l ⁺ v) + X at 13 TeV	
non-photon induced contributions	91.2 \pm 1.8 fb
photon-induced contri (NNPDF23)	6.0 ^{+4.4} _{-2.9} fb
photon-induced contri (LUXqed)	4.4 \pm 0.1 fb

Manohar, Nason, Salam, Zanderighi, 16-17

*Crucial implications for LHC pheno:
high-precision determination of photon-
initiated (PI) contributions*

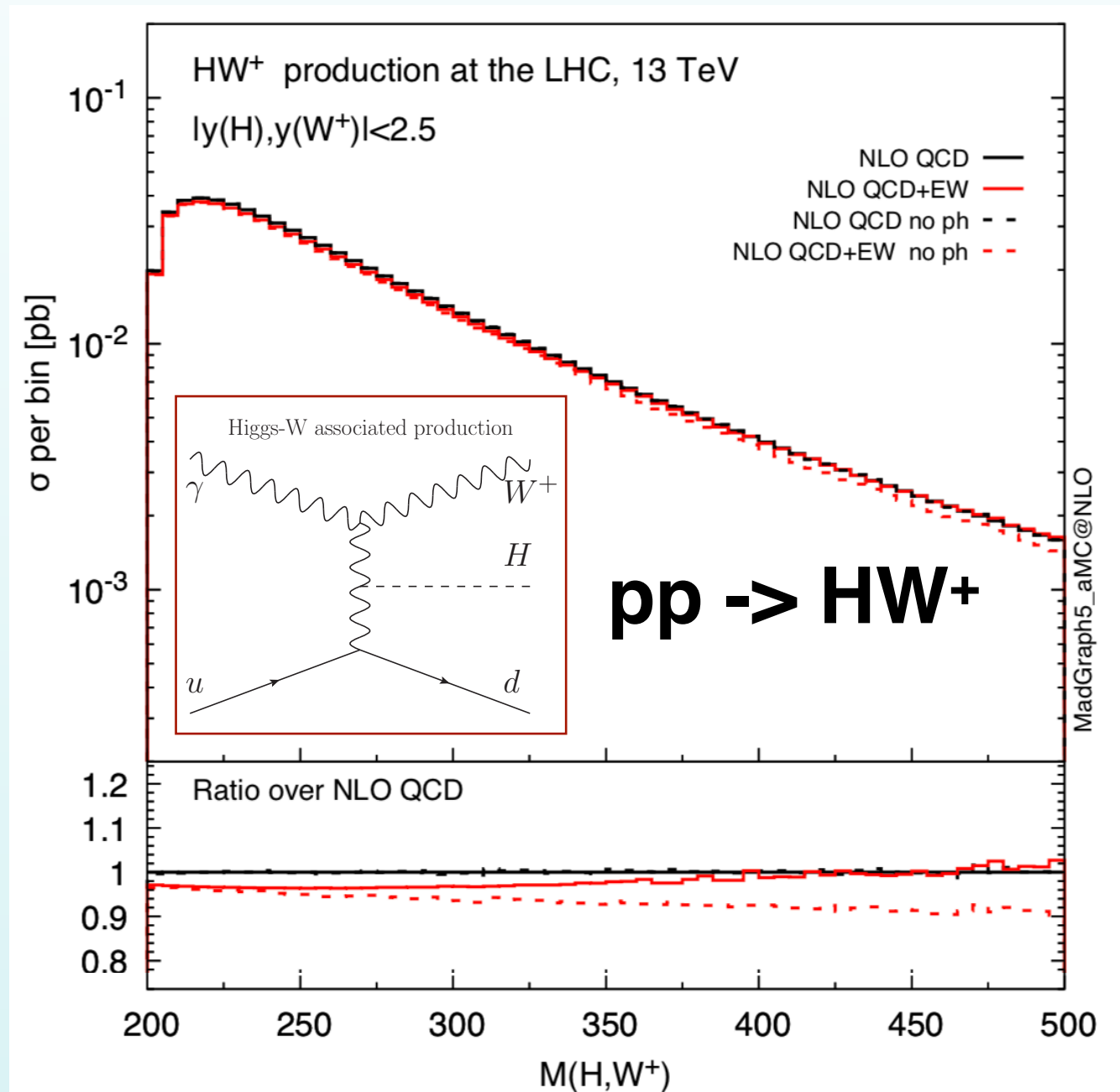
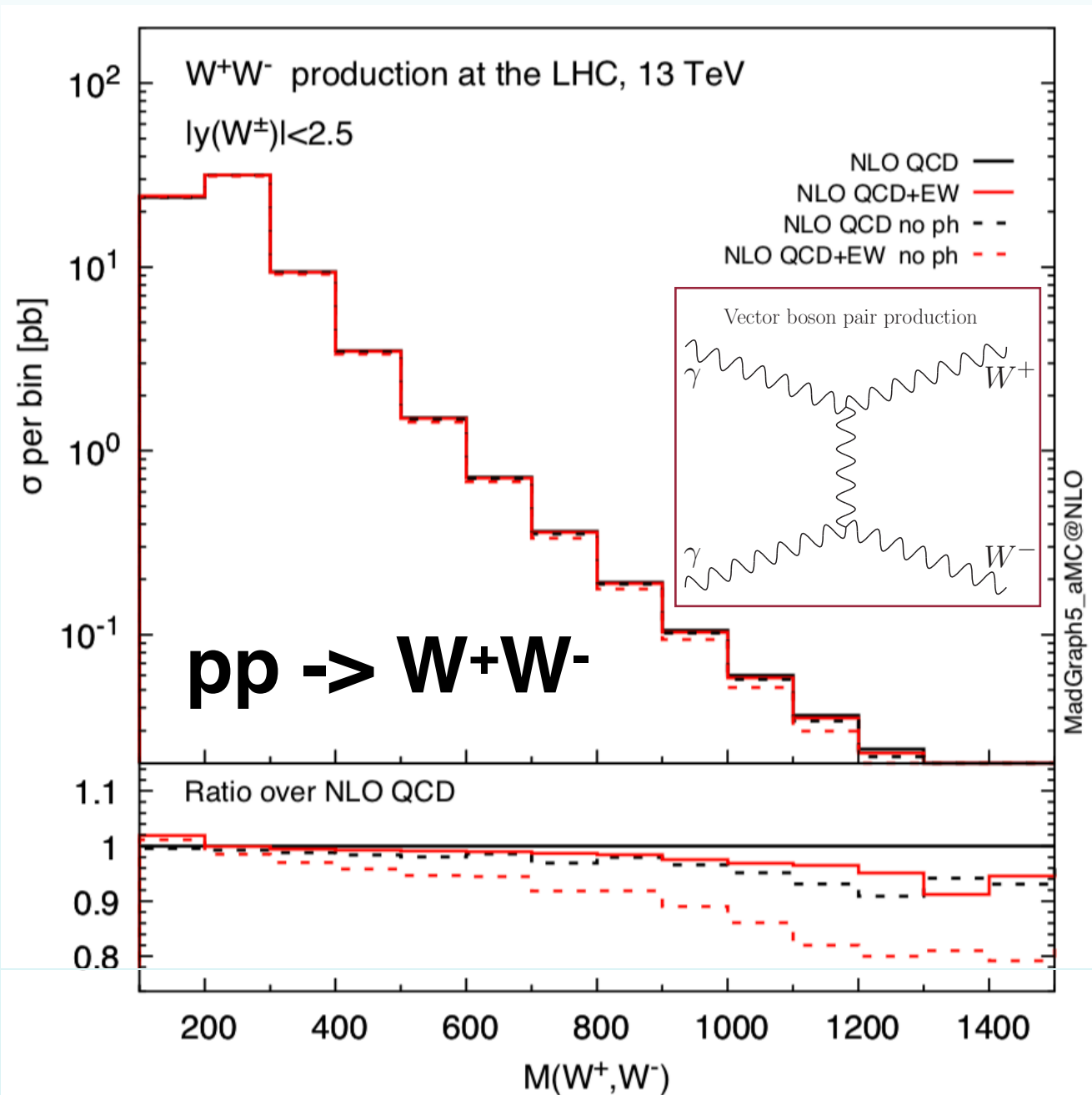
Illuminating the photon content of the proton

- **NNPDF3.1luxQED**: variant of the NNPDF3.1 global analysis supplemented by
 - the LUXqed theoretical constraints,
 - NLO QED corrections to DGLAP evolution, and
 - NLO QED coefficients functions in DIS
- Iterative procedure: photon PDF recomputed at each iteration until convergence is achieved



Photon-initiated processes at the LHC

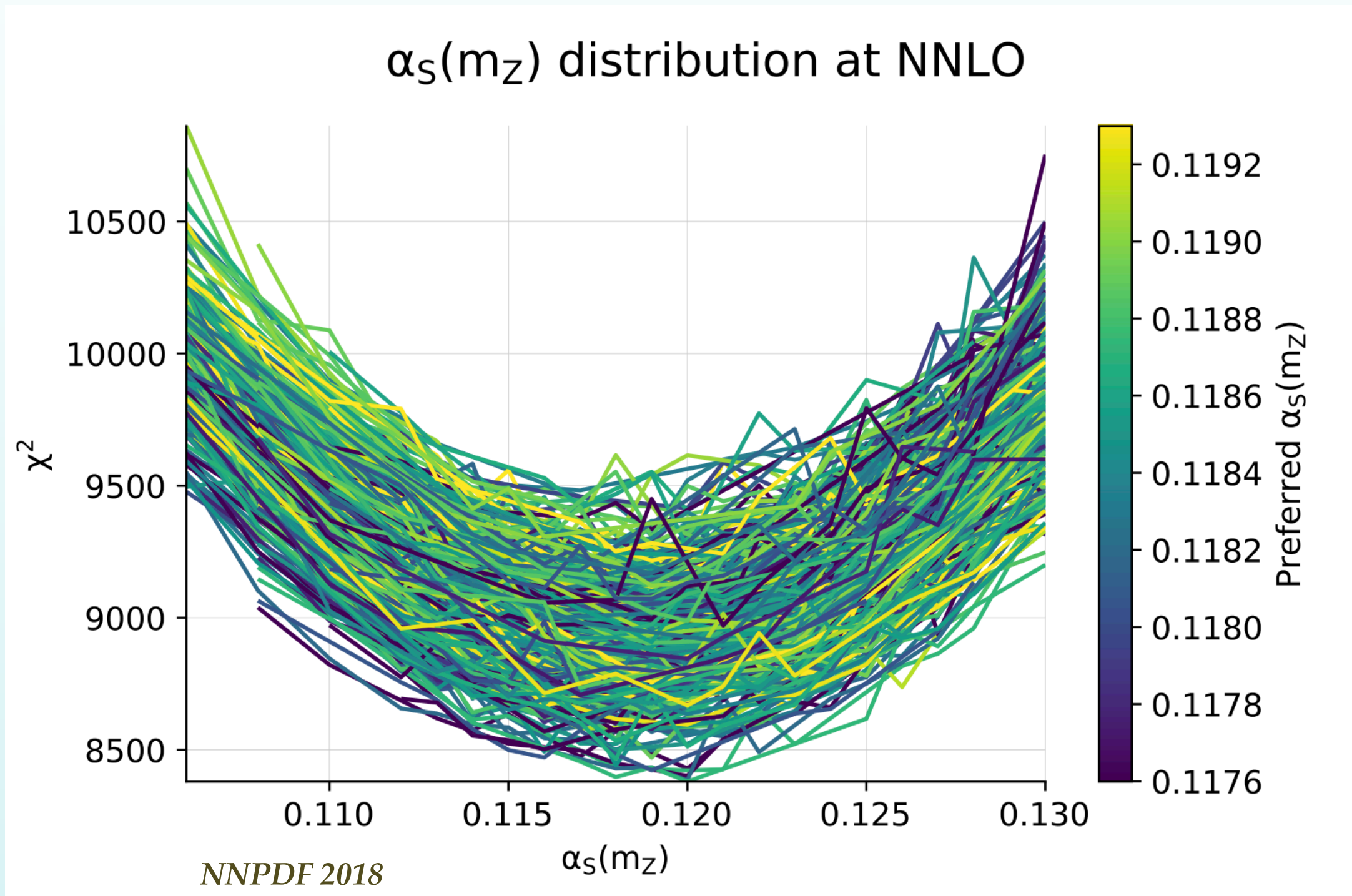
- At high mass **PI contributions** and **NLO EW corrections** have **opposite sign**, and thus in general one expects a partial cancellation among them
- This seems to be the case for many processes: **once PI effects included, NLO EW corrections rather small**



Bertone, Carrazza, Pagani, Rojo, Vicini, Zaro (in preparation)

Precision determination of $\alpha_s(m_Z)$

- Determination of the strong coupling constant using the new Monte Carlo correlated replica method
- In a nutshell, carry out an **independent $\alpha_s(m_Z)$ fit for each MC replica** to determine its complete probability distribution accounting for all correlations with the PDFs



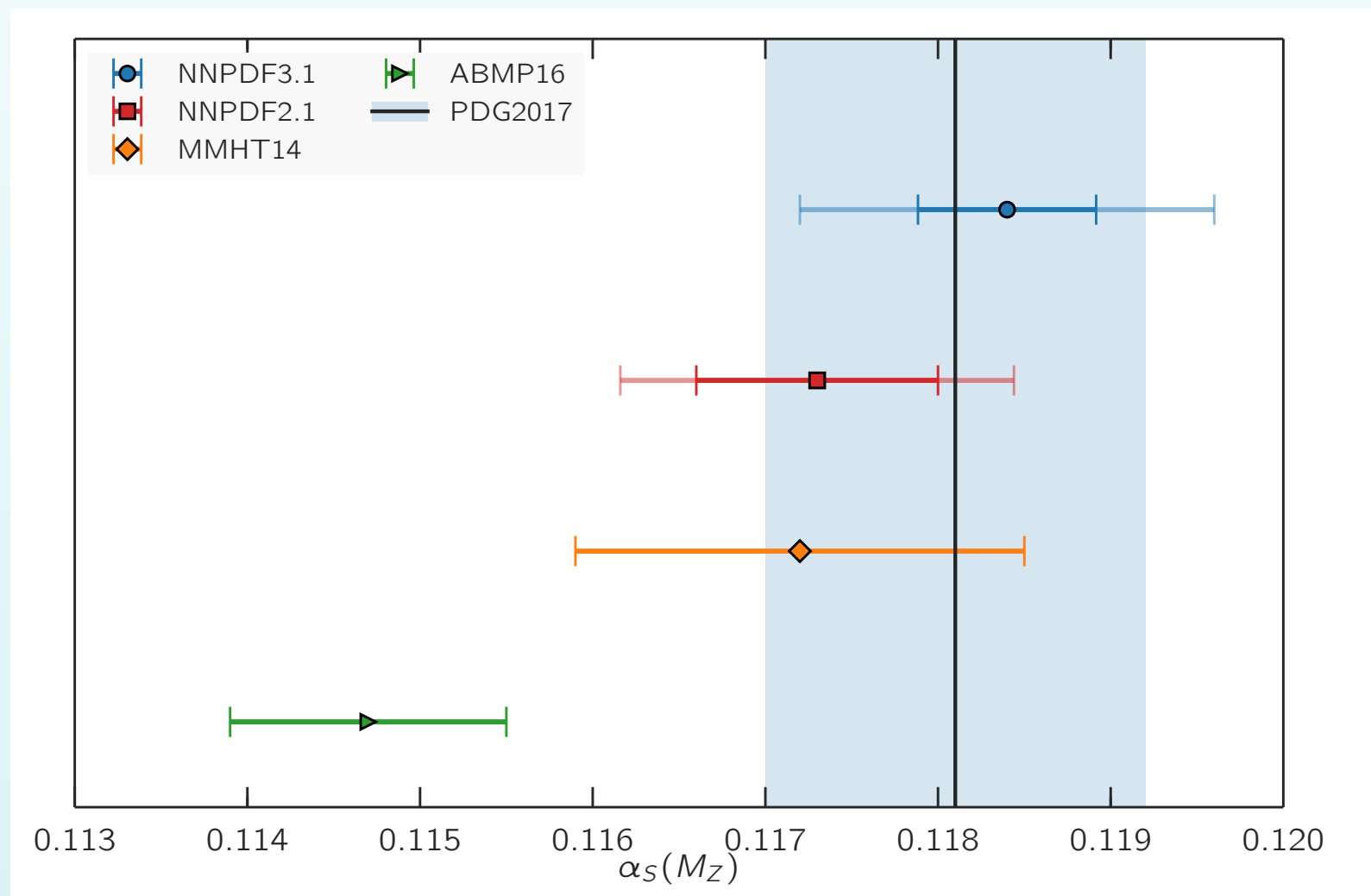
Precision determination of $\alpha_s(m_Z)$

- Fit of $\alpha_s(m_Z)$ based NNPDF3.1NNLO, where all collider processes are treated using exact NNLO theory
- First ever determination of the strong coupling based on a **global NNLO analysis** with inclusive jet, top quark pair production, Z transverse momentum distributions: various **complementary handles!**

$$\alpha_s^{\text{NNLO}}(m_Z) = 0.1185 \pm 0.0005^{\text{exp}} \quad (0.4\%)$$

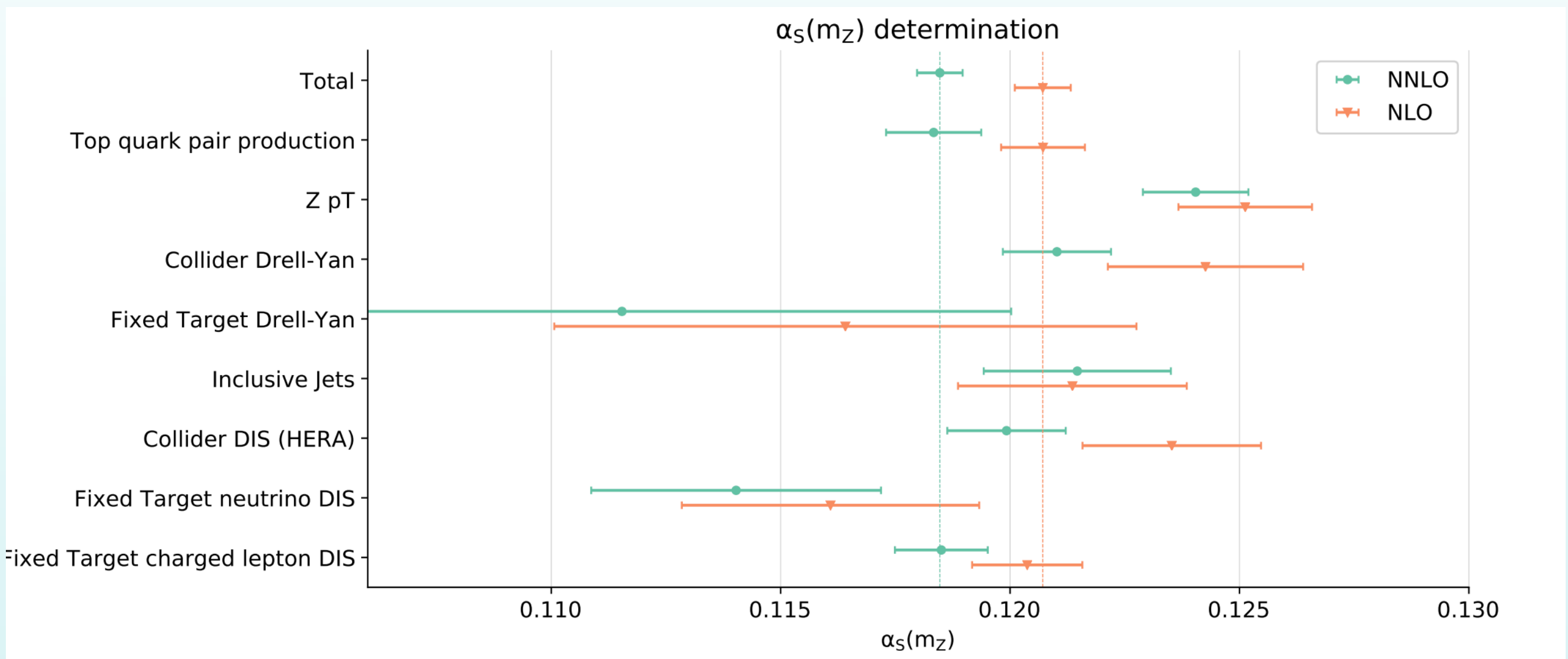
$$\Delta\alpha_s^{\text{pert}} \equiv |\alpha_s^{\text{NNLO}} - \alpha_s^{\text{NLO}}| = 0.0022$$

$$\delta\alpha_s^{\text{mhou}} \simeq 0.0011$$



Precision determination of $\alpha_s(m_Z)$

- Significant constraints from top quark pair, $Z p_T$, and inclusive jets
- But also from fixed-target NC DIS structure functions and from collider DIS and DY data
- Main limitation of the fit is the **poor control over the MHOUs**: once these are accounted for, the global PDF fit would become **one of the most competitive methods** to extract $\alpha_s(m_Z)$



Parton distributions with BFKL resummation

- **Perturbative fixed-order QCD calculations** have been extremely successful in describing a wealth of data from proton-proton and electron-proton collisions
- There are theoretical reasons that eventually we need to go beyond DGLAP: at small- x , **logarithmically enhanced terms in $1/x$ become dominant** and need to be resummed to all orders
- **BFKL/high-energy/small- x resummation** can be matched to the **DGLAP collinear framework**, and thus be included into a standard PDF analysis

DGLAP
Evolution in Q^2

$$\mu^2 \frac{\partial}{\partial \mu^2} f_i(x, \mu^2) = \int_x^1 \frac{dz}{z} P_{ij} \left(\frac{x}{z}, \alpha_s(\mu^2) \right) f_j(z, \mu^2),$$

BFKL
Evolution in x

$$-x \frac{d}{dx} f_+(x, \mu^2) = \int_0^\infty \frac{d\nu^2}{\nu^2} K \left(\frac{\mu^2}{\nu^2}, \alpha_s \right) f_+(x, \nu^2)$$

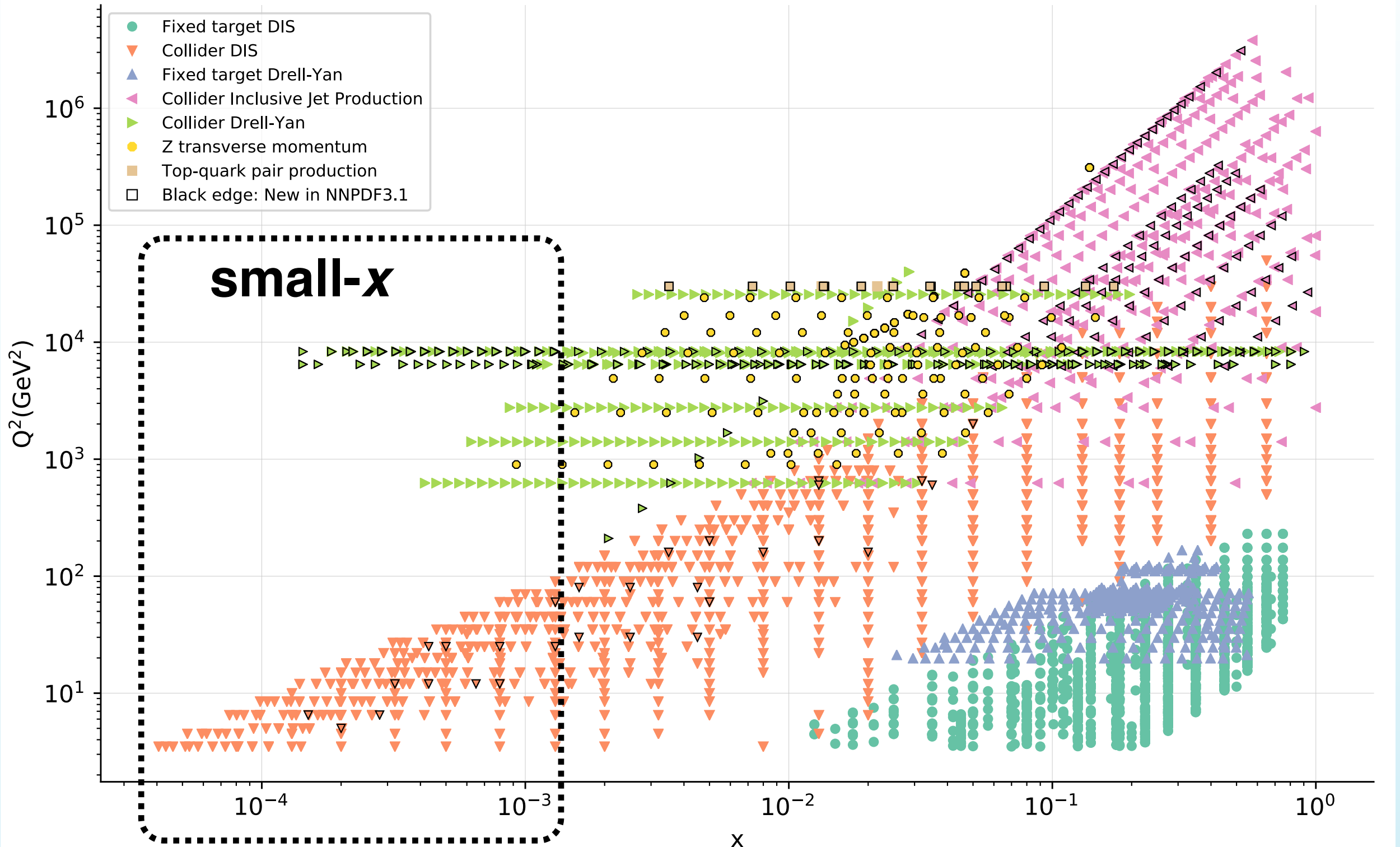
Within small- x resummation, the $N^k\text{LO}$ fixed-order DGLAP splitting functions are complemented with the $N^h\text{LL}x$ contributions from BKFL

ABE, CCSS, TW + others, 94-08

$$P_{ij}^{N^k\text{LO}+N^h\text{LL}x}(x) = P_{ij}^{N^k\text{LO}}(x) + \Delta_k P_{ij}^{N^h\text{LL}x}(x),$$

A new world at small-x

Kinematic coverage

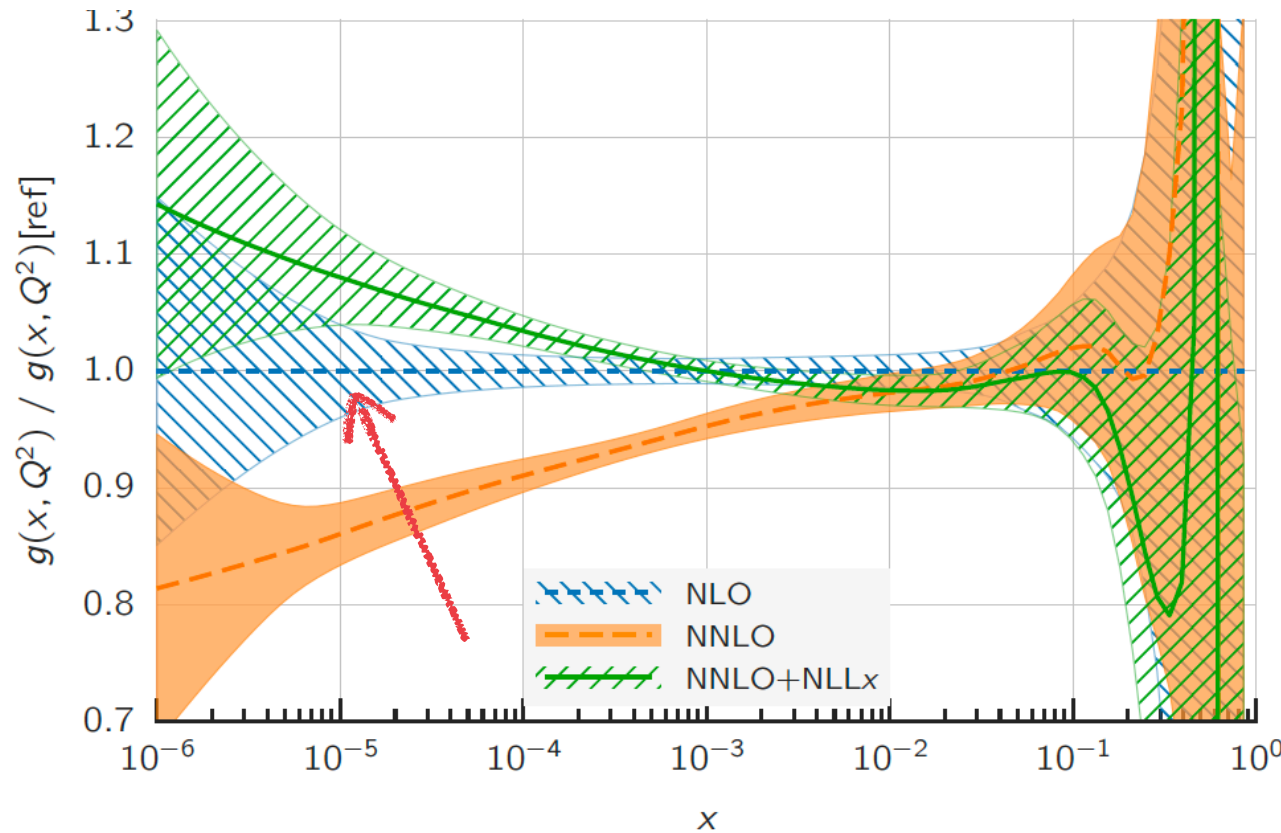


A new world at small-x

- Ultimately, the need for (or lack of) BKFL resummation in ep and pp collider data can only be assessed by performing a **global PDF analysis based on (N)NLO+NLLx theory**
- Theoretical tools are now available: **HELL for NLLx resummation**, interfaced to **APFEL**
- NNPDF3.1 (N)NLO+NLL fits **stabilize the perturbative PDF expansion at small-x**, in particular for the gluon, and markedly improve the **fit quality to the small-x HERA data**

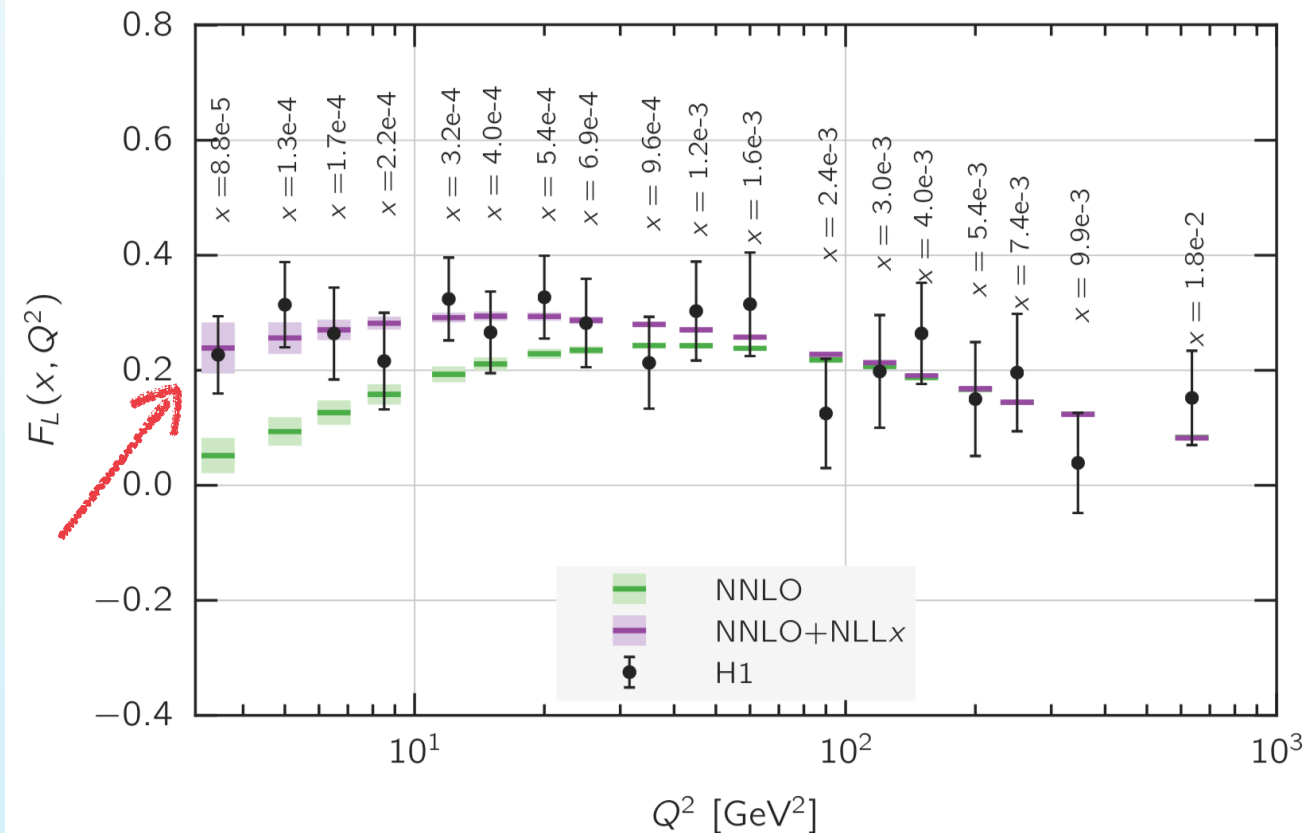
gluon

NNPDF31sx DIS only, $Q = 100$ GeV



$F_L(x, Q)$

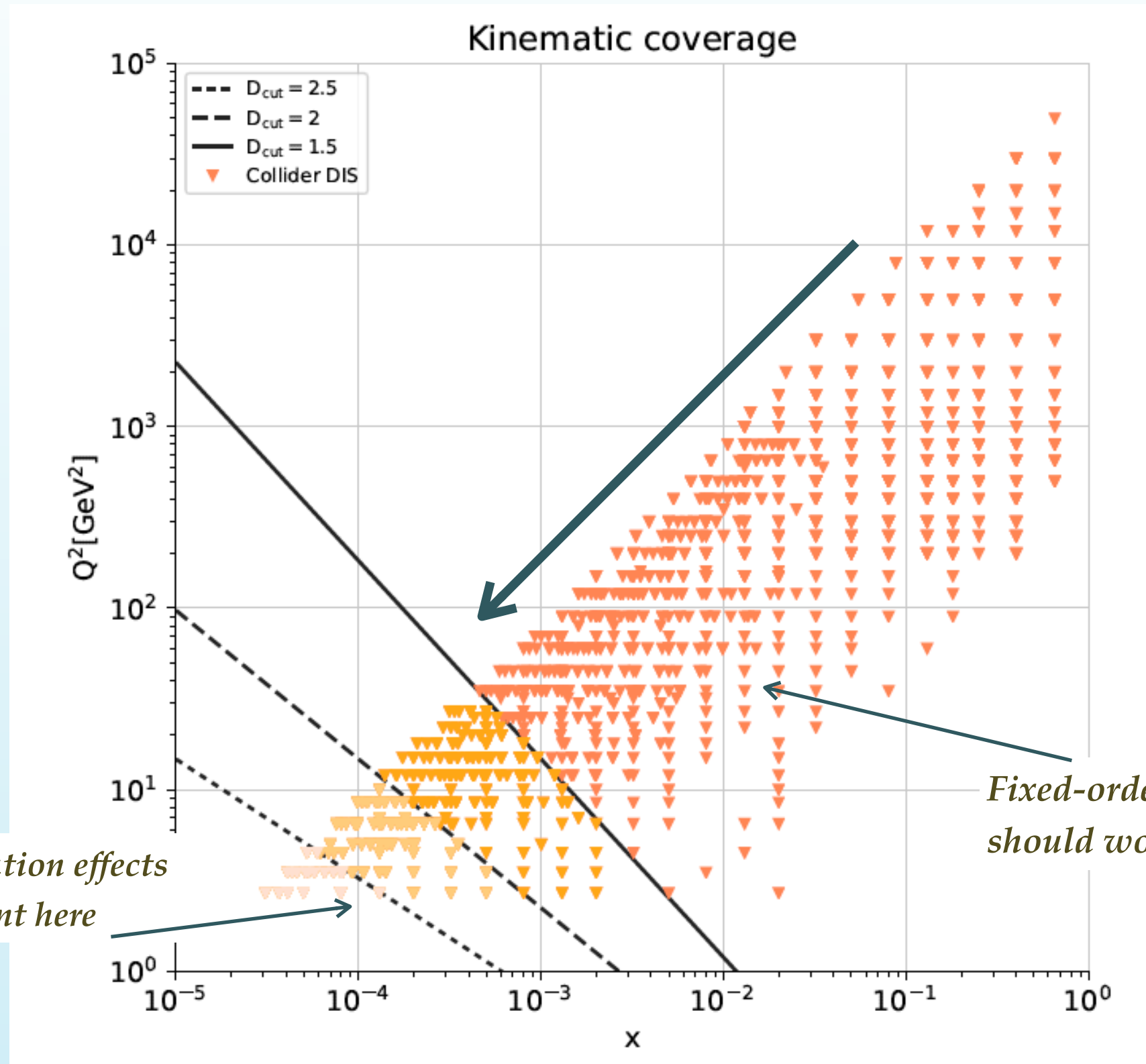
NNPDF3.1sx



Ball, Bertone, Bonvini, Marzani, JR, Rottoli 16

Evidence for BFKL dynamics in HERA data

In order to assess the impact of small- x resummation for the description of the small- x and Q^2 HERA data, compute the χ^2 removing data points in the region where resummation effects are expected



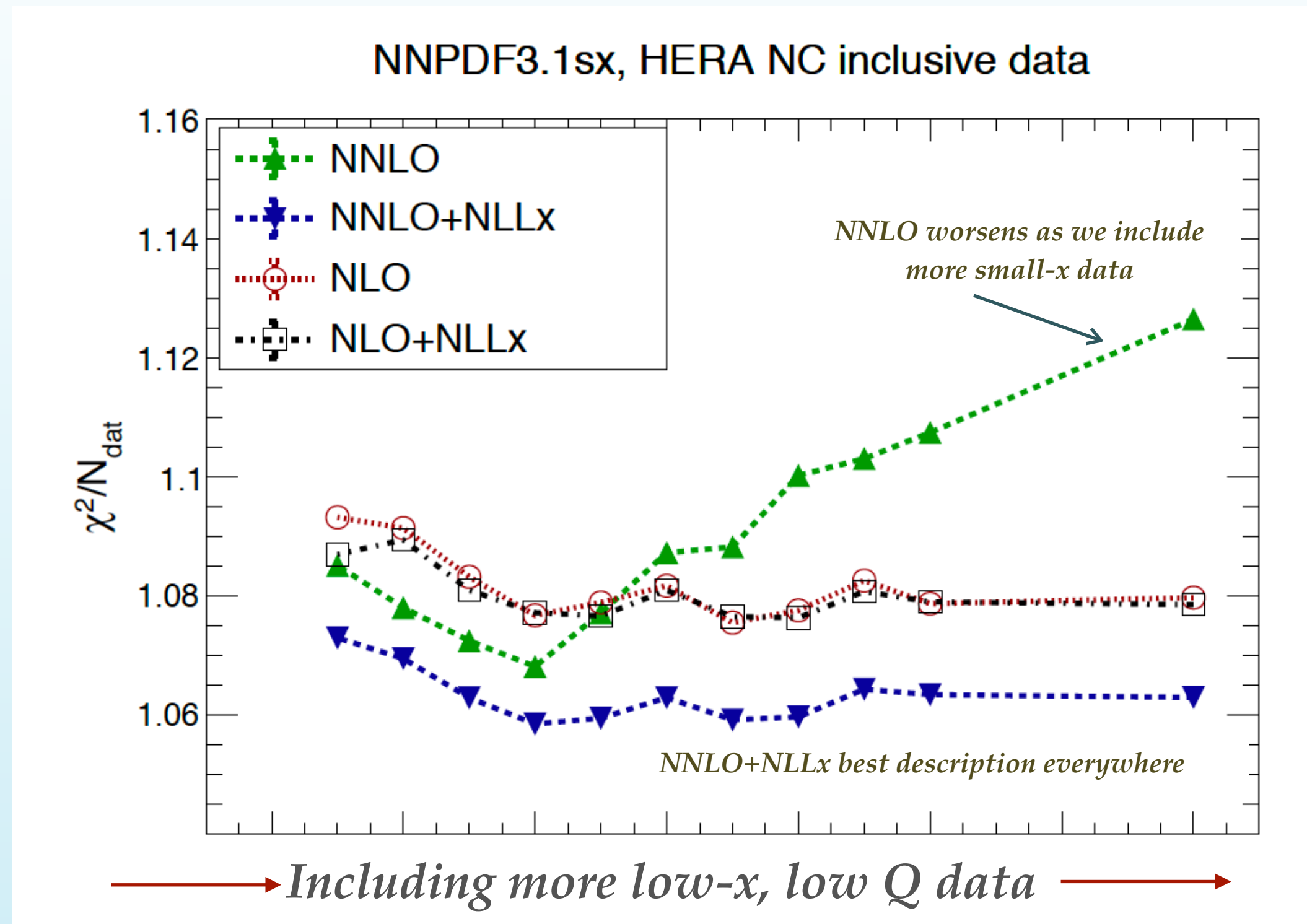
Small- x resummation effects could be important here

Fixed-order theory should work fine here

Evidence for BFKL dynamics in HERA data

Using NNLO+NLLx theory, the NNLO instability at small-x of the χ^2 disappears

Excellent fit quality to inclusive and charm HERA data achieved in the entire (x, Q^2) region



Nunca es tarde si la dicha es buena

Science
Life and Physics

After 40 years of studying the strong nuclear force, a revelation

This was the year that analysis of data finally backed up a prediction, made in the mid 1970s, of a surprising emergent behaviour in the strong nuclear force

Jon Butterworth

🐦 @jonmbutterworth

Thu 28 Dec 2017 17.30 GMT



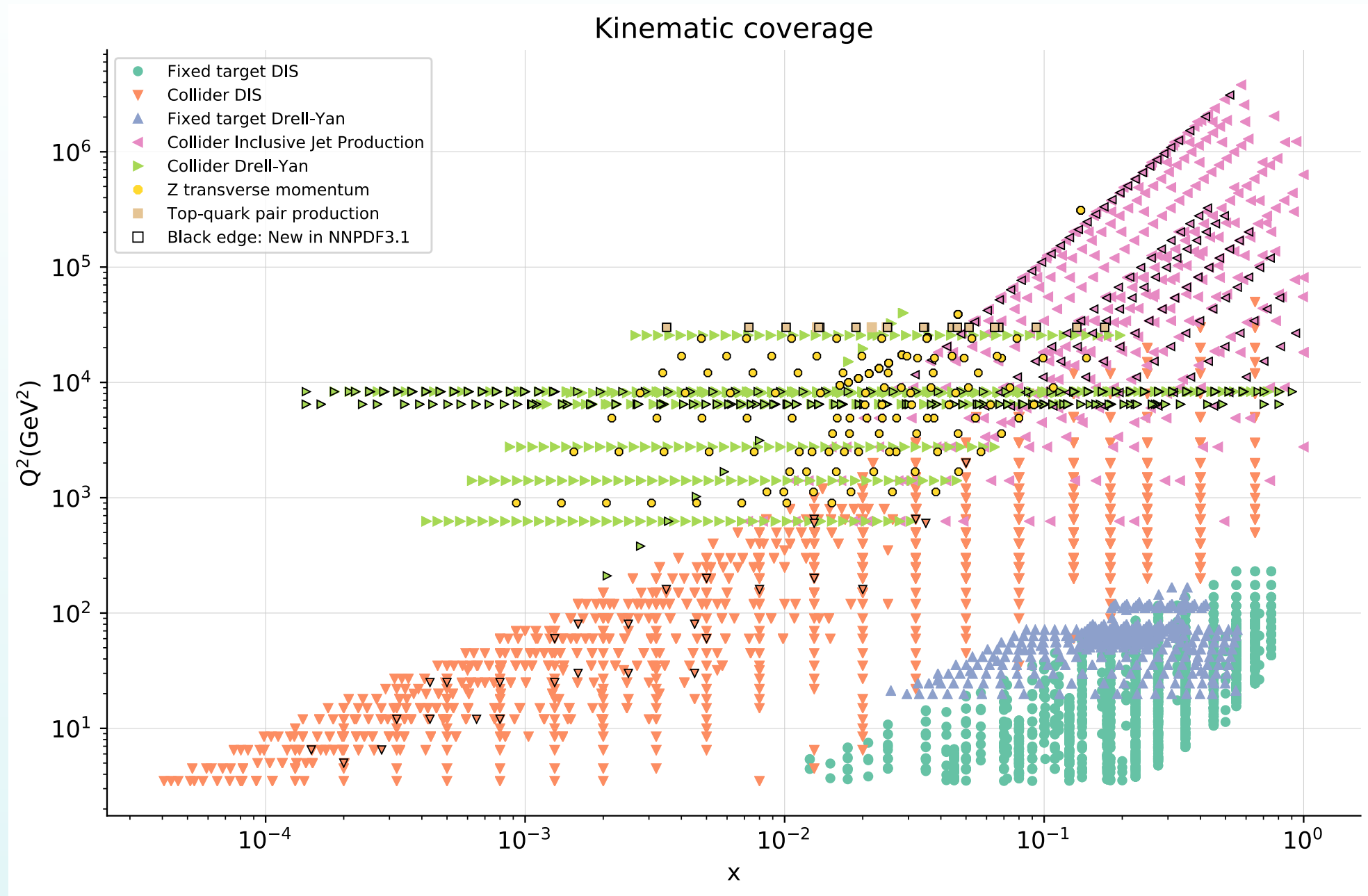
🔗 529 | 💬 59

*Jon Butterworth,
the Guardian, 28/12/2018*



In the mid 1970s, four Soviet physicists, Batlisky, Fadin, Kuraev and Lipatov, made some predictions involving the strong nuclear force which would lead to their initials entering the lore. “BFKL” became a shorthand for a difficult-to-

The global QCD fit machinery



Global PDF fits: highly non-trivial validation of the QCD factorisation framework:

i) including $O(5000)$ data points,

ii) from $O(50)$ experiments,

iii) several of them with $\approx 1\%$ errors,

yet still manage to achieve $\chi^2/N_{\text{dat}} \approx 1$!

Summary and outlook

📌 Recent developments in our understanding of the **quark and gluon structure of the proton** have been driven by a combination of:

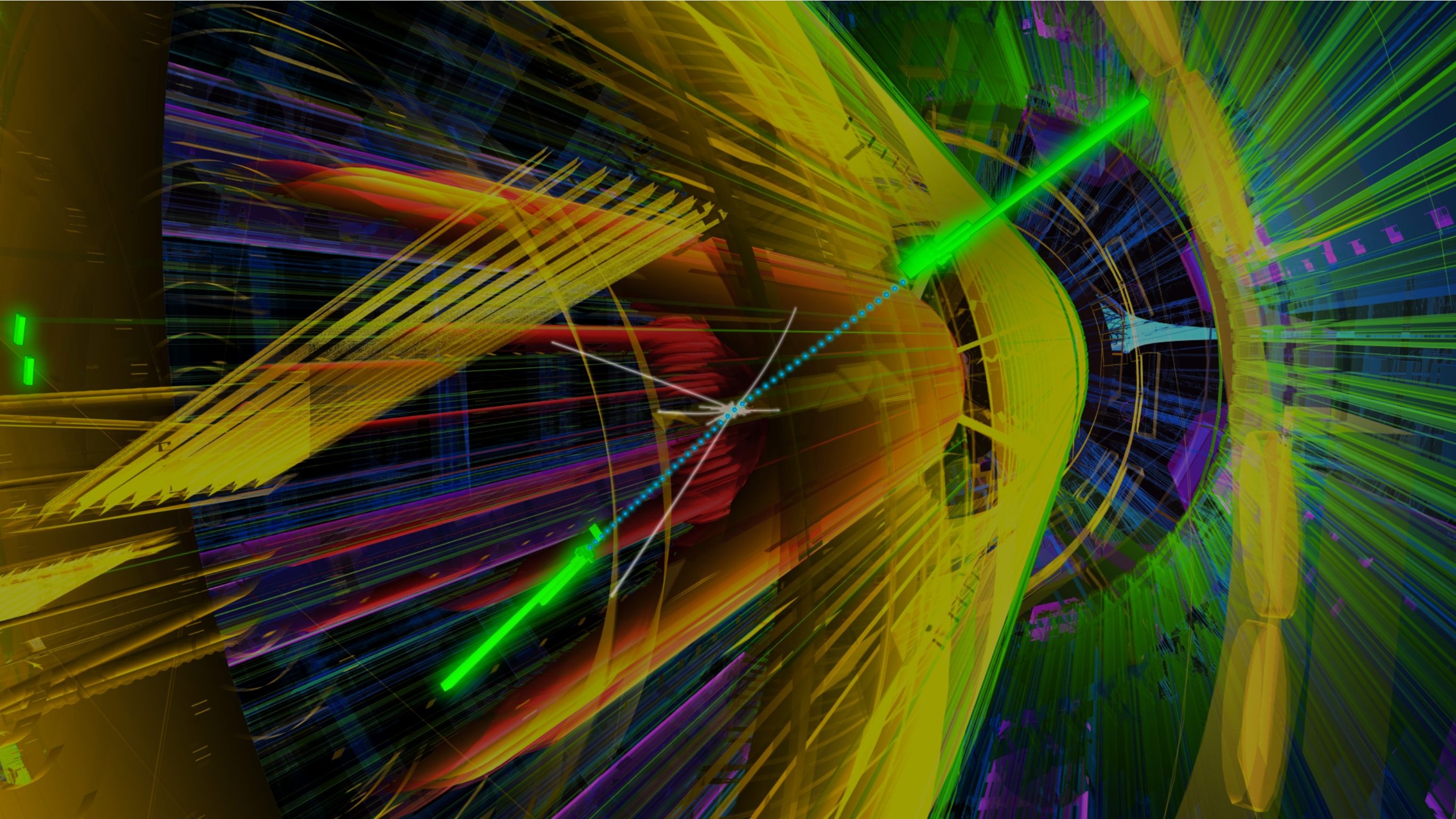
☑ **Theory:** Progress in NNLO QCD and NLO EW calculations for many collider processes: differential top pairs, inclusive jets, the Z transverse momentum ... Also the calculation of the photon PDF in terms of DIS structure functions

☑ **Data:** a wealth of high-precision measurements from HERA, Tevatron, ATLAS, CMS and LHCb, in several cases with sub-percent uncertainties.

☑ **Methodology:** fitted charm PDF, combination/reduction methods for different PDFs, new software for PDF fits, fast NLO/NNLO interfaces,

📌 Improvements for **many Run II analysis:** Higgs couplings, M_W measurements, heavy BSM particle production, precision SM studies, SMEFT fits, MC validation, ...

Fascinating times ahead at the high-energy frontier!



Stay posted for news from the LHC!

Fascinating times ahead at the high-energy frontier!

An abstract, colorful background featuring a dense network of overlapping lines and streaks in various colors including yellow, green, blue, red, and purple. The lines are oriented in different directions, creating a sense of dynamic movement and complexity. A prominent diagonal streak of bright green is visible in the upper right quadrant. The overall effect is reminiscent of particle tracks or data visualization in a high-energy physics context.

Thanks for your attention!

Stay posted for news from the LHC!