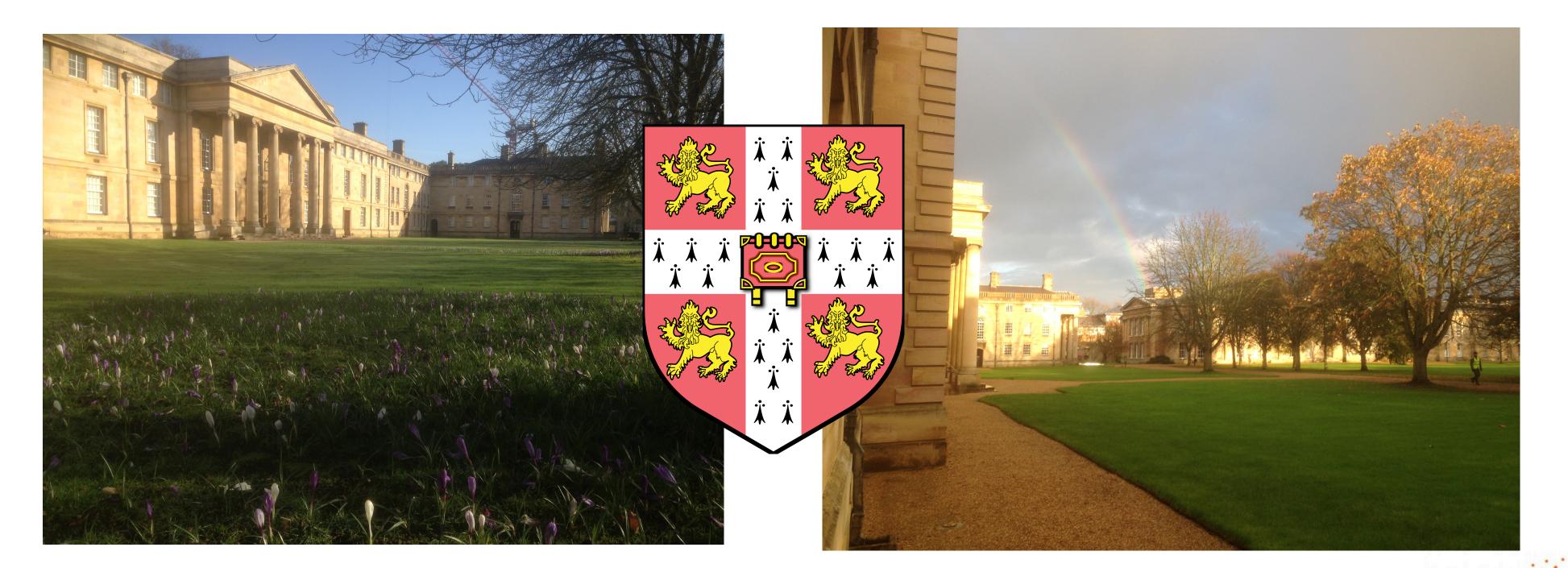
### The top quark legacy of the LHC Run II for PDF and SMEFT analyses

for the University of Cambridge Department of Applied Mathematics and Theoretical Physics, May 2023



#### James Moore, University of Cambridge





**European Research Council** 

Established by the European Commission

# **PBSP: Physics Beyond the Standard Proton**

- The PBSP group is based at the University of Cambridge, and is headed by Maria Ubiali; the project is ERC-funded.
- The aim is to investigate interplay between BSM physics and proton structure - the subject of the rest of this talk!
- The team members are:
  - Postdocs: Zahari Kassabov, Maeve Madigan, Luca Mantani
  - PhD students: Mark Costantini, Shayan Iranipour (former), Elie Hammou, **James Moore**, Manuel Morales, Cameron Voisey (former)





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#### Talk overview

**1. PDFs: a lightning introduction** 

#### **2. PDF fitting**

#### **3. Joint PDF-SMEFT fits**

#### 4. The SIMUnet methodology

5. The top quark legacy of the LHC Run II for PDF and **SMEFT** analyses



# 1. - PDFs: a lightning introduction

# Hadron structure through PDFs

 Hadrons are QCD bound states - they are strongly-coupled, nonperturbative objects.

$$\mathscr{L} = -\frac{1}{4}G^a_{\mu\nu}G^{a,\mu\nu} + \sum_{q}\overline{q}$$

# $\bar{q}(i\gamma_{\mu}D^{\mu} - m_{q})q \longrightarrow \text{hadrons?}$



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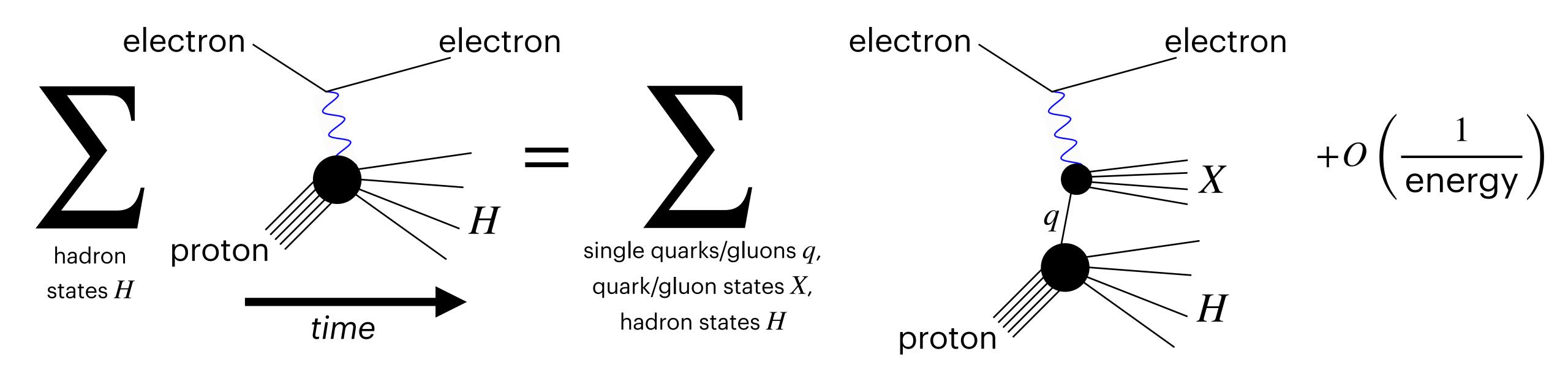
$$\mathscr{L} = -\frac{1}{4}G^{a}_{\mu\nu}G^{a,\mu\nu} + \sum_{q} \overline{q}(i\gamma_{\mu}D^{\mu} - m_{q})q \longrightarrow \text{hadrons}$$

- **Solution:** package all non-perturbative elements into unknown functions, called **parton distribution functions (PDFs)**.

But we still want to make predictions for experiments involving hadrons!

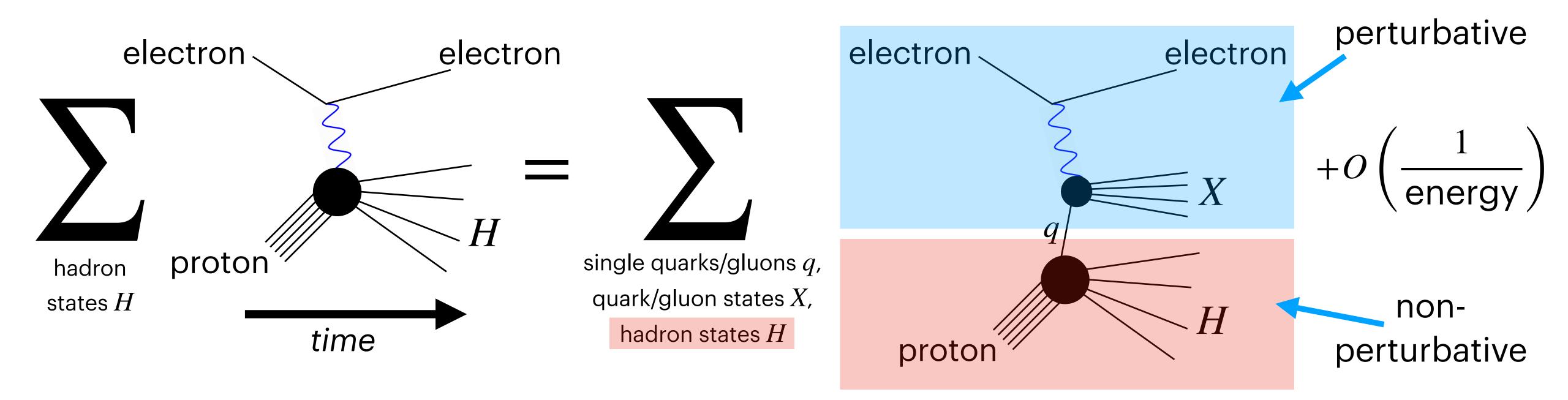


- This is formalised through factorisation theorems.



#### • Model case: deep inelastic scattering, $e^-$ + proton $\rightarrow e^-$ + any hadron.

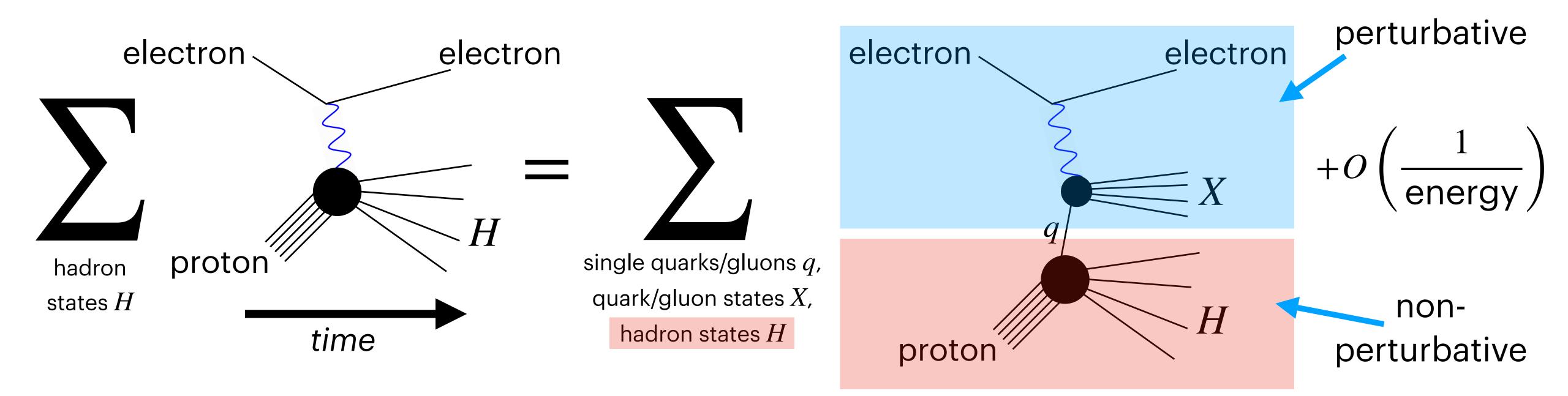
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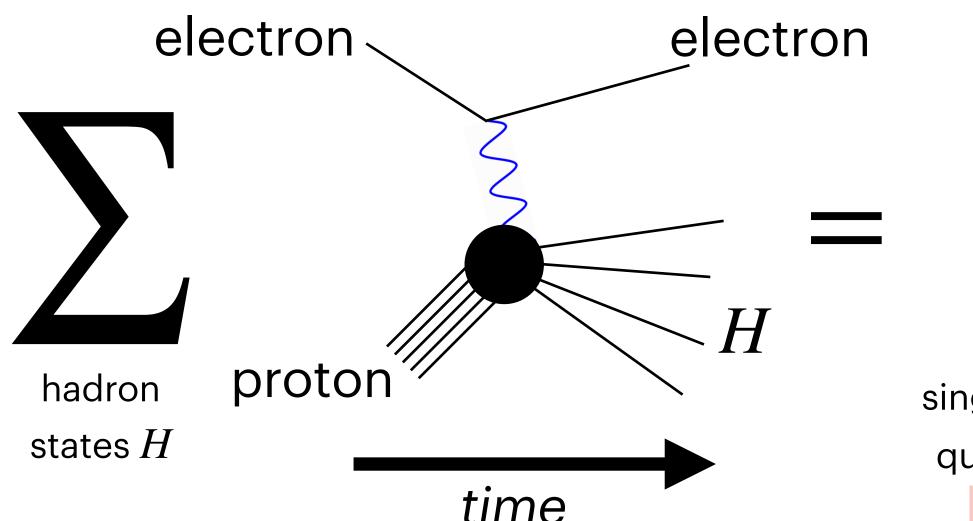


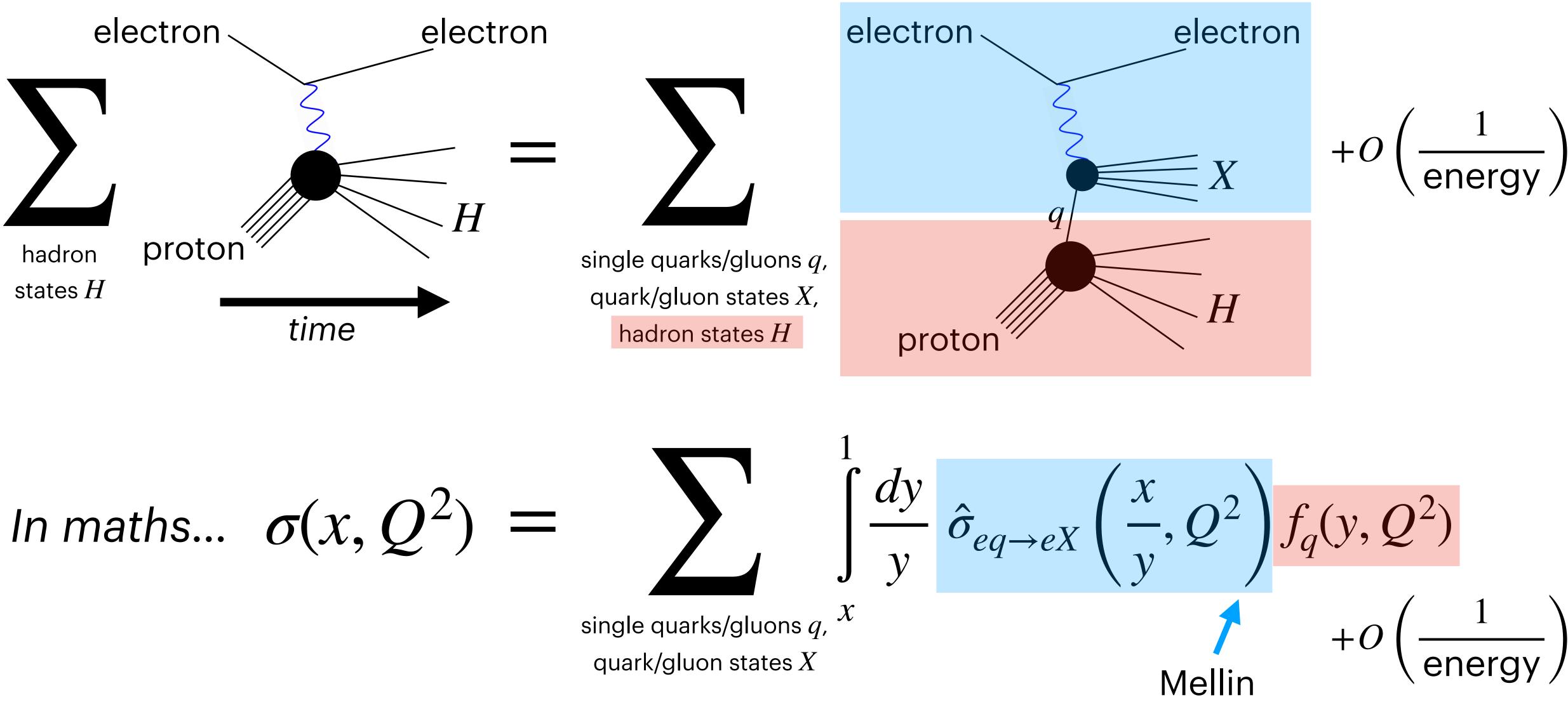
a non-perturbative, BUT universal, parton distribution function.



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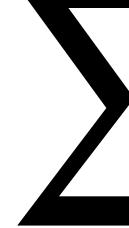
# The calculation is split into a perturbative process-dependent part, and





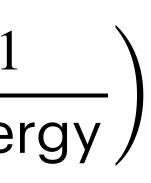
convolution

# **Factorisation theorems** In maths... $\sigma(x, Q^2) = \int \int \frac{dy}{v} \hat{\sigma}_{eq \to eX} \left(\frac{x}{v}, Q^2\right) f_q(y, Q^2)$ single quarks/gluons q, $^{\mathcal{X}}$



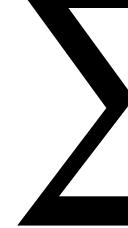
quark/gluon states X

#### • Loosely speaking, the PDFs $f_q(x, Q^2)$ capture the probability that a certain constituent will be **ejected** in a collision.



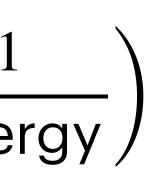
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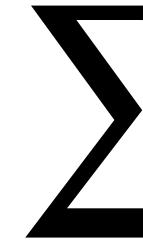


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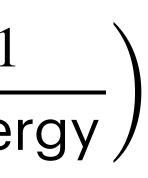
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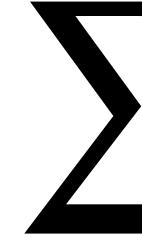
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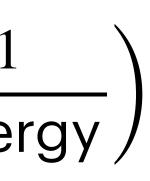
  - The fact we are colliding protons if we started with a neutron, we would need different PDFs

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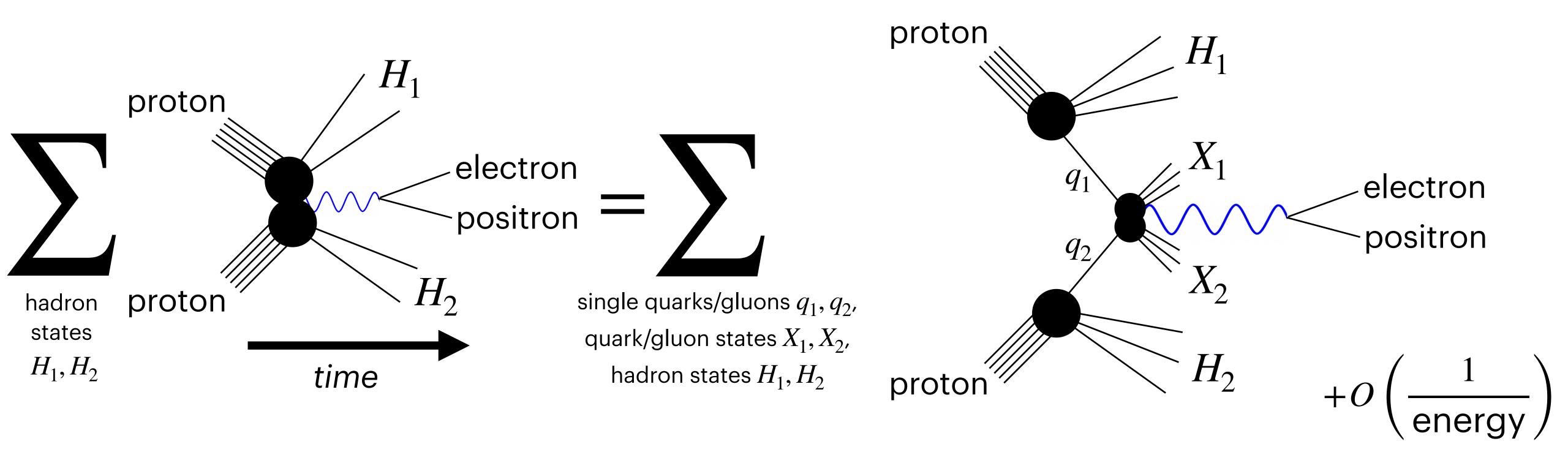
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# **Universality of PDFs**

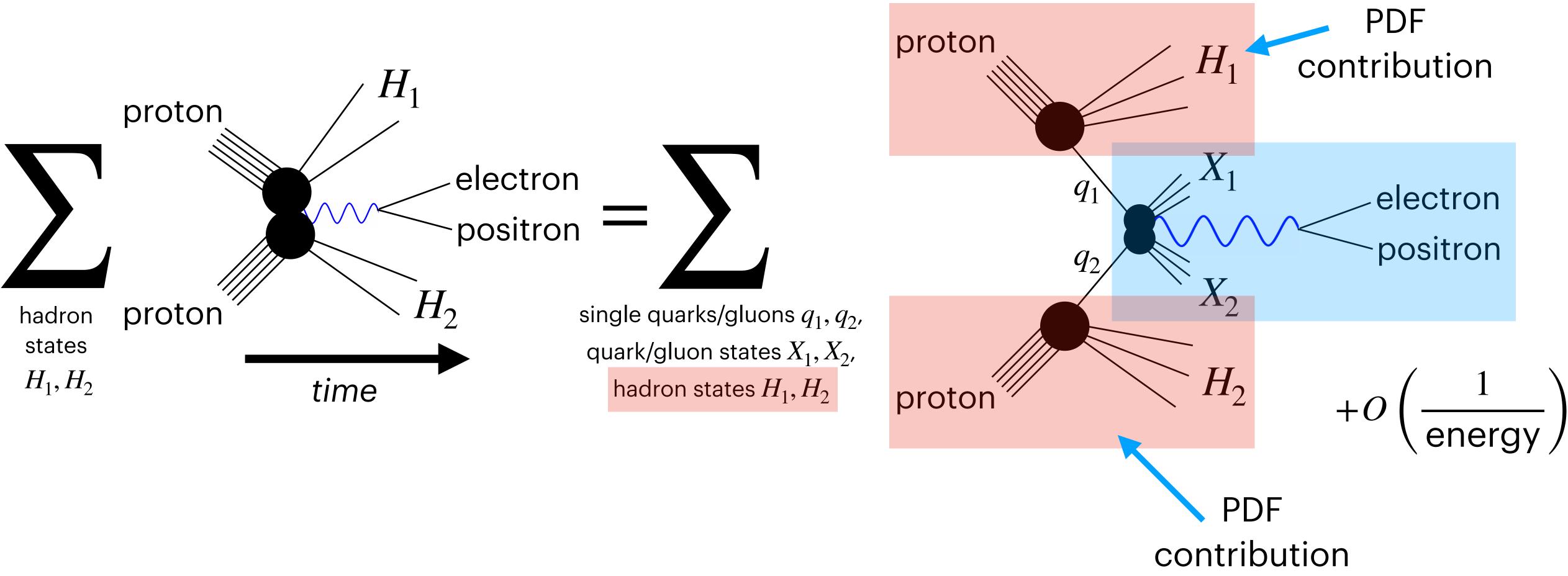
electron-positron pair, plus any hadrons.



• Importantly, PDFs are **universal.** The **same** parton distributions can **also** be used in the **Drell-Yan process**: the collision of two protons to make an

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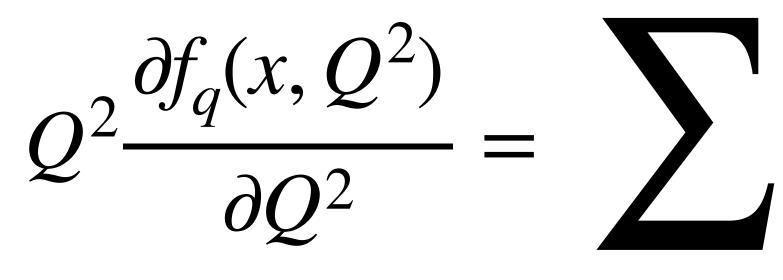
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# **Scaling of PDFs**

- IR divergences.
- Symanzik equation for the PDFs called the **DGLAP equation**:



quarks/gluons q

and can be determined perturbatively.

• Whilst the PDFs are non-perturbative, we can still say something about their  $Q^2$ -dependence, which enters the PDFs when we **absorb collinear** 

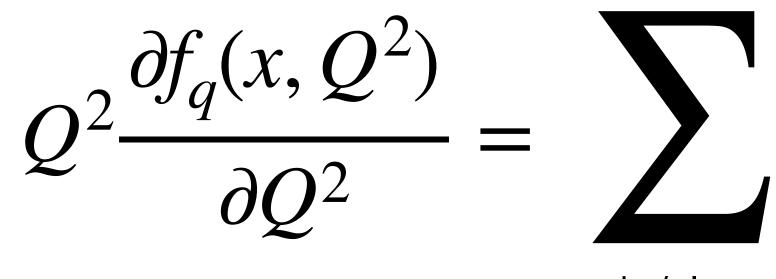
Just as in standard UV renormalisation theory, this leads to a Callan-

$$\int_{x}^{1} \frac{dy}{y} P_{qq'}\left(\frac{x}{y}\right) f_{q'}(x, Q^2)$$

• The functions (technically distributions)  $P_{qq'}$  are called **splitting functions** 



# **Scaling of PDFs**



quarks/gluons  $q' \;\; \mathcal{X}$ 

- This means if we know the PDFs at some **initial energy scale**  $Q_0$ , we can compute them at some energy scale  $Q > Q_0$  by solving DGLAP.
- In particular, only the x-dependence of the PDFs is truly **unknown**.
- We can obtain this x-dependence by fits to collider data, as we shall now describe...

 $Q^{2} \frac{\partial f_{q}(x,Q^{2})}{\partial O^{2}} = \sum \int \frac{dy}{v} P_{qq'}\left(\frac{x}{v}\right) f_{q'}(x,Q^{2})$ 

# **Summary of PDFs**

of hadron being collided, not on the process.

integro-differential equations called the DGLAP equations.

 The PDFs have unknown x-dependence, which must be obtained through fits to experimental data.

The non-perturbative structure of hadrons can be parametrised by parton distribution functions  $f_q(x, Q^2)$ , which depend only on the type

• The PDFs have **known**  $Q^2$ -dependence, described by a linear system of



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- of possible PDFs is **infinite-dimensional**. What do we do?
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- Example functional form:

large and small x behaviour motivated by **Regge theory** 

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 $f(x, Q_0^2) = Ax^{\alpha}(1 - x)^{\beta} (1 + ax^{1/2} + bx + cx^{3/2})$ polynomial in  $\sqrt{x}$ 

measures the **goodness of fit** of our model:

$$\chi^2 = (data - theory)^T c$$

# • The best-fit parameters are found by **minimising the** $\chi^2$ -statistic, which

#### $covariance^{-1}(data - theory)$

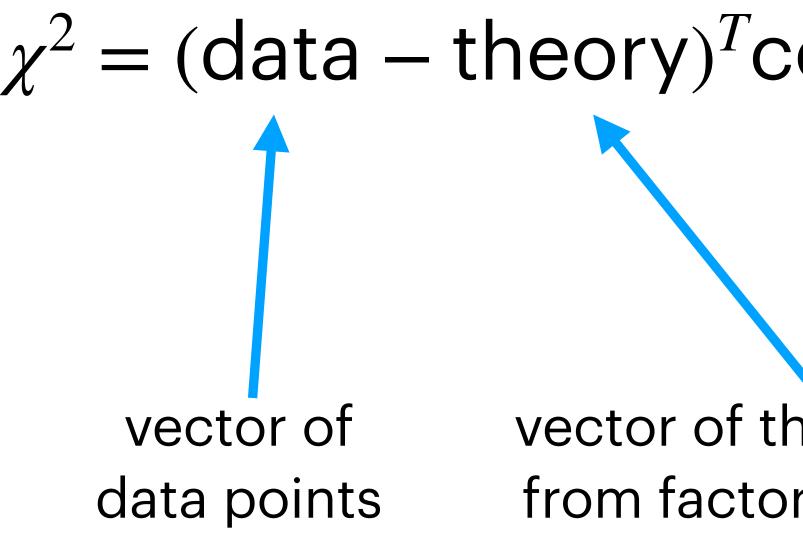
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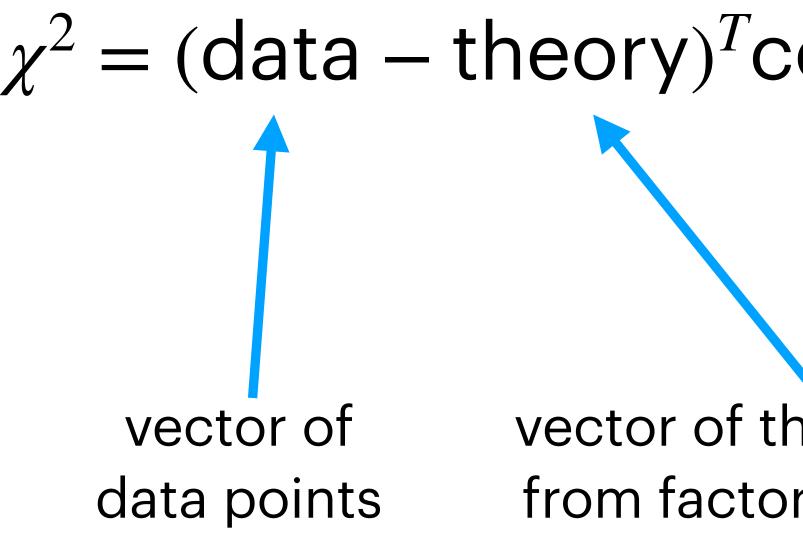


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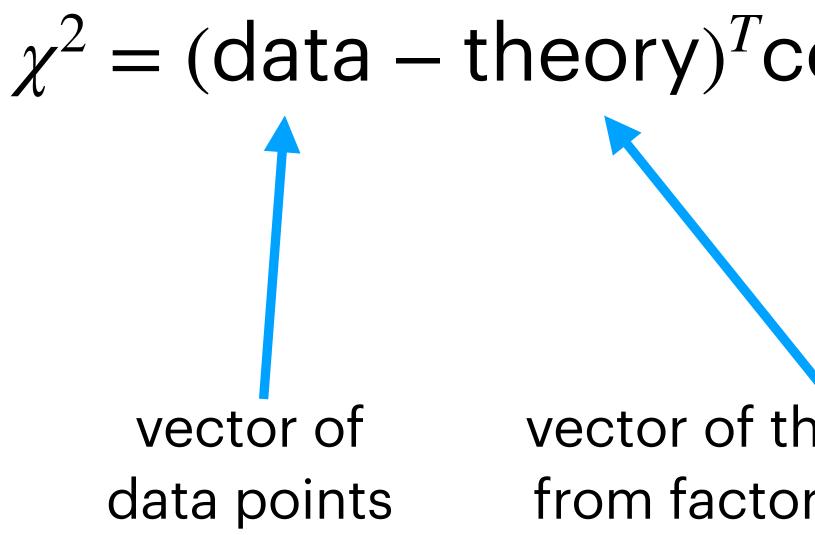
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experimental\* covariance matrix

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**uncertain**, we don't require such precise agreement.

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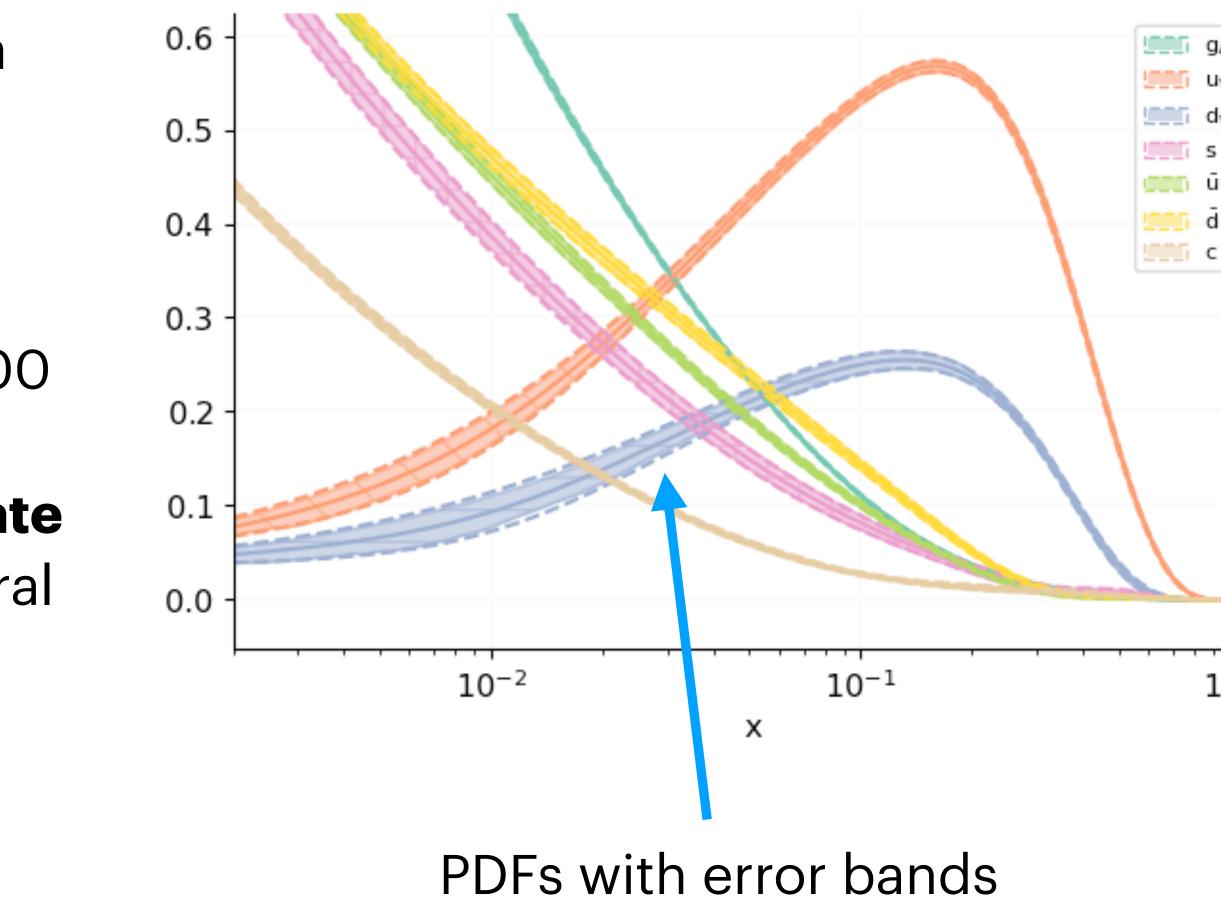
General idea: we want theory to be close to data, but if the data is more

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### The choice of functional form

The choice of functional form that we have suggested so far is:

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• This seems a bit arbitrary though! To try to remove as much **bias** as **neural network** instead:

$$f(x, Q_0^2) = Ax^{\alpha}(1 - x)^{\beta} \mathsf{NN}(x, \omega)$$

has network parameters  $\omega$ .

possible, another possible choice is to parametrise the PDFs using a

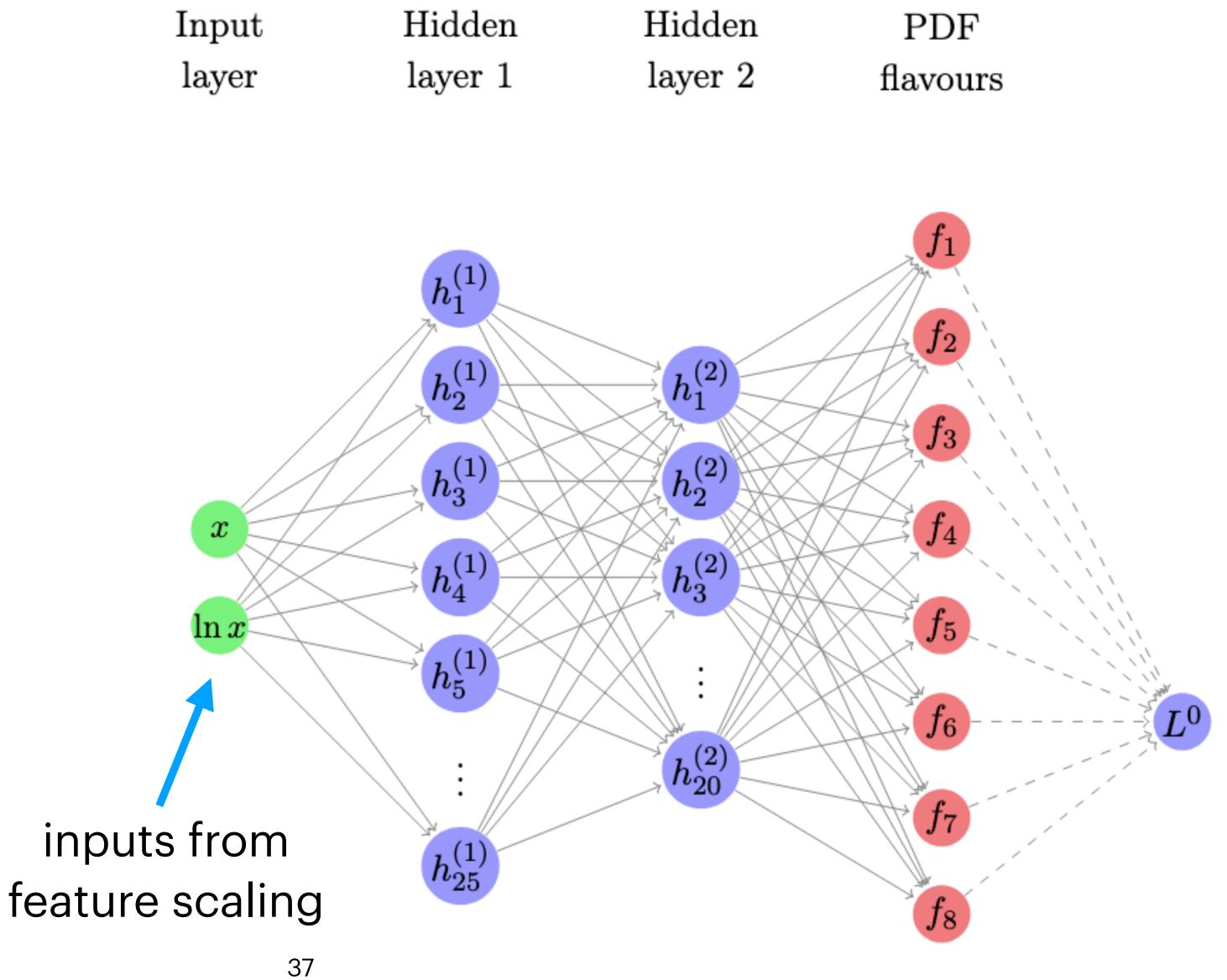
• Here,  $NN(x, \omega)$  is a **neural network** which takes in x as an argument, and

# The choice of functional form

$$f(x, Q_0^2) = Ax^{\alpha}(1 - x)^{\beta} \mathsf{NN}(x, \omega)$$

- The neural network parametrisation is used by the **NNPDF** collaboration, whose fitting code is **publicly** available.
- See 2109.02653 and 2109.02671 for details.

Input layer

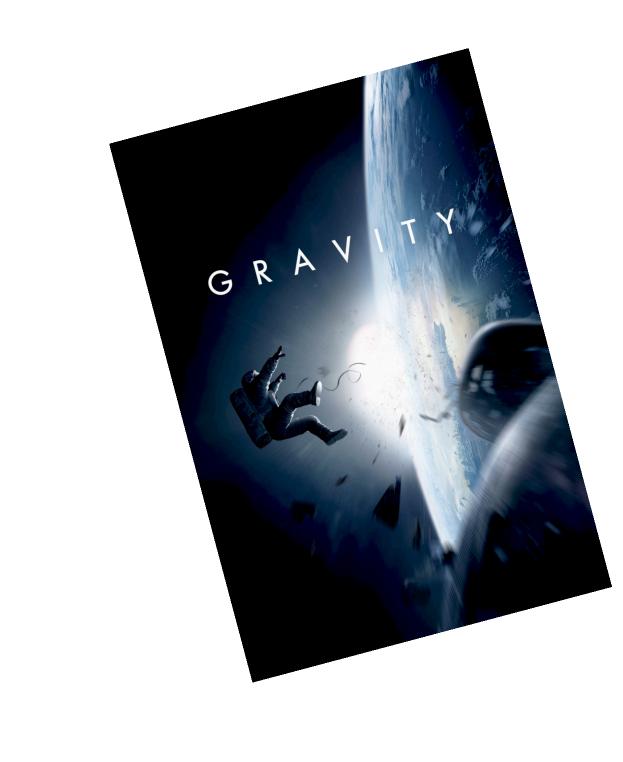


# **3. - Joint PDF-SMEFT fits**

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  - many more...

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• For example, to include dark matter in the Standard Model, we might hypothesise new particles and add them in. The Standard Model Lagrangian density is augmented to:



# $\mathscr{L}_{new} = \mathscr{L}_{SM} + \mathscr{L}_{dark}$ matter

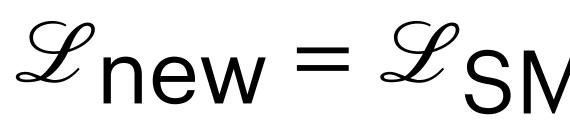
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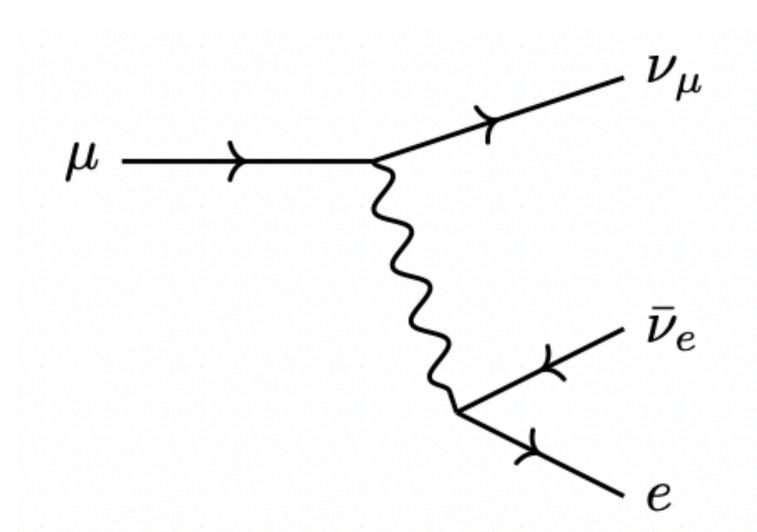
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- However, there are **thousands** of possibilities, so just guessing particles seems a bit like stabbing in the dark!
- Some models are more motivated than others, but it would be nice to have a more general approach... 49

 Fortunately, the language of effect this problem.

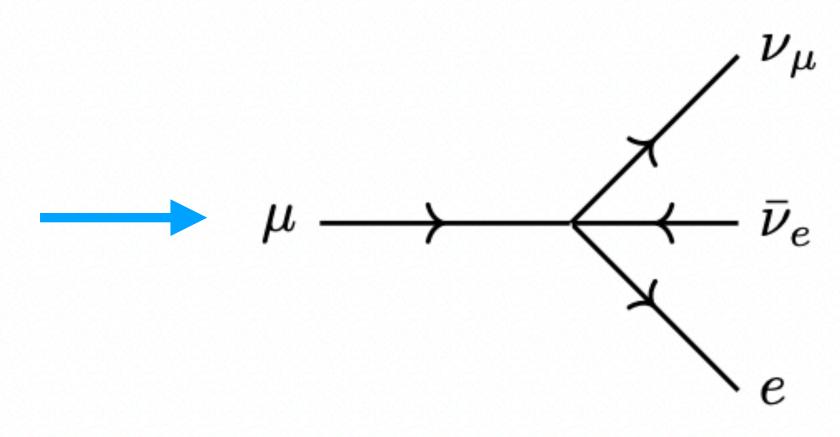
### • Fortunately, the language of effective field theory exists to help us tackle

- this problem.
- Idea: at low energies we can integrate out heavy particles from a theory, giving effective non-renormalisable interactions:



Integrating out particles can also yield shifts in SM couplings.

### • Fortunately, the language of **effective field theory** exists to help us tackle



respecting the SM symmetries):



### • Since **any**\* heavy particle manifests at low energies as non-renormalisable interactions, if we are hunting for extensions of the SM, we can simply add on all non-renormalisable operators built from the SM fields (and

 $\mathscr{L}_{\mathsf{SMFFT}} = \mathscr{L}_{\mathsf{SM}} + \mathscr{L}_5 + \mathscr{L}_6 + \cdots$ 

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## $S_{SM} + \mathscr{L}_5 + \mathscr{L}_6 + \cdots$

 We can organise the additional non-renormalisable operators by their mass dimension, with higher-dimensional operators being suppressed by **powers of**  $1/\Lambda$ , where  $\Lambda$  is a characteristic scale of the New Physics.



## $\mathscr{L}_{\mathsf{SMEFT}} = \mathscr{L}_{\mathsf{SM}} + \mathscr{L}_5 + \mathscr{L}_6 + \cdots$

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- However, the number of operators decreases significantly if we assume additional symmetries, e.g. no baryon number violation. There are only 59 operators if we assume flavour universality.
- The main sectors studied so far are: top, Higgs and electroweak physics.

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- Finally, note that various fitting groups just fit the SMEFT couplings, for example the SMEFiT collaboration, and the FitMaker collaboration.
- In particular, SMEFiT and FitMaker both assume a SM PDF input. This could be problematic because the PDFs were fitted assuming no New Physics...

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- Fix SMEFT parameters (usually to zero),  $c = \overline{c}$ :

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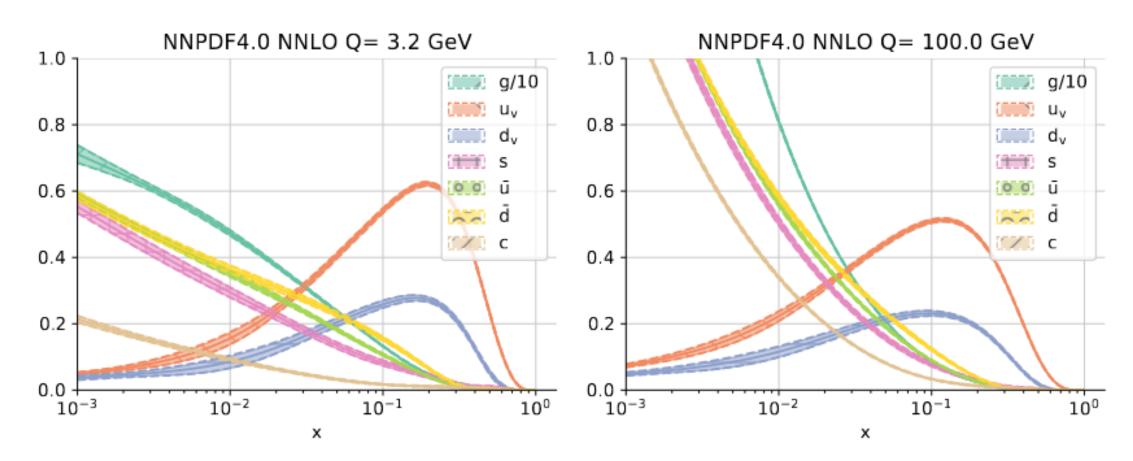
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- E.g. NNPDF4.0 fit, Ball et al., 2109.02653.

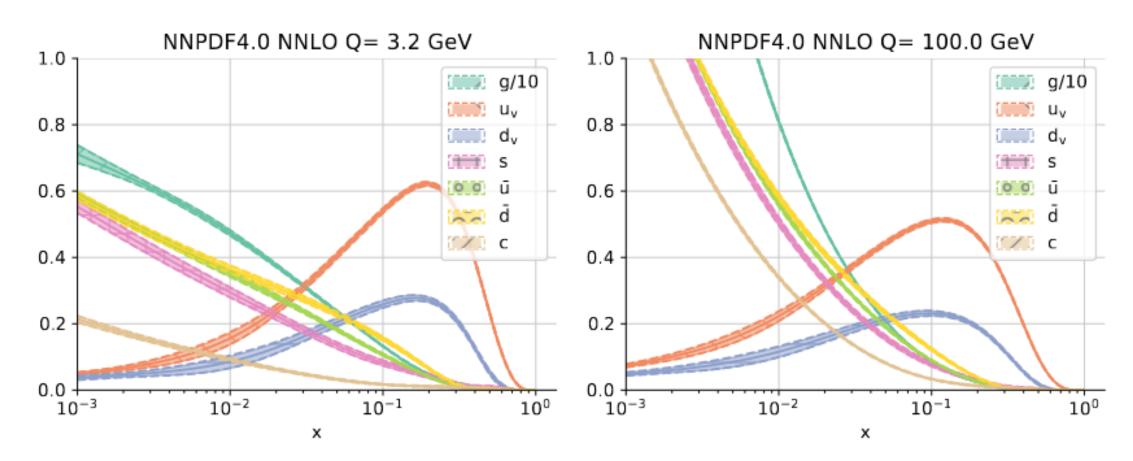




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- E.g. NNPDF4.0 fit, Ball et al., 2109.02653.





### **SMEFT** parameter fits

• Fix PDF parameters  $\theta = \overline{\theta}$ :

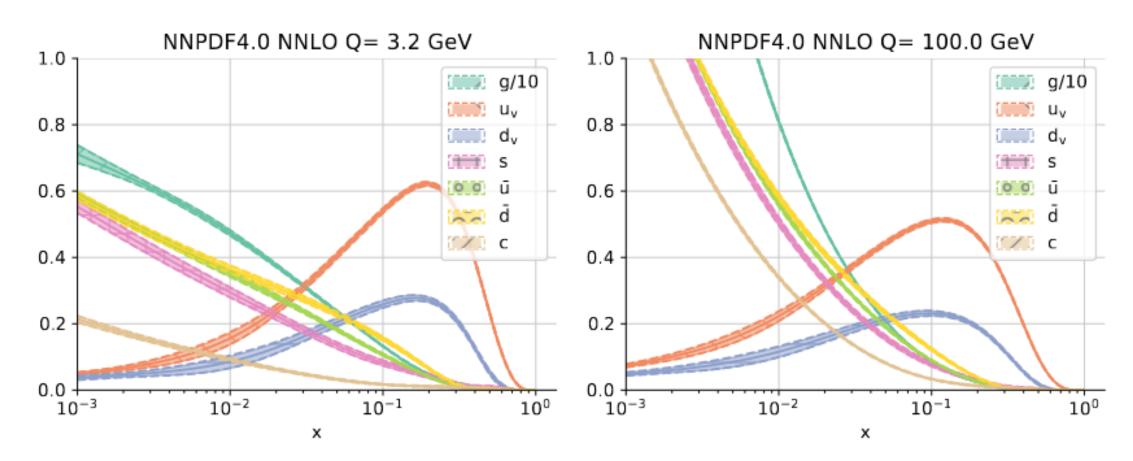
### $\sigma(c,\overline{\theta}) = \hat{\sigma}(c) \otimes \mathsf{PDF}(\overline{\theta})$

Optimal SMEFT parameters  $c^*$  then have an **implicit dependence** on PDF choice:  $c^* = c^*(\overline{\theta}).$ 

- In more detail ( $\otimes$  is shorthand for the **Mellin convolution**)... **PDF** parameter fits
- Fix SMEFT parameters (usually to zero),  $c = \overline{c}$ :

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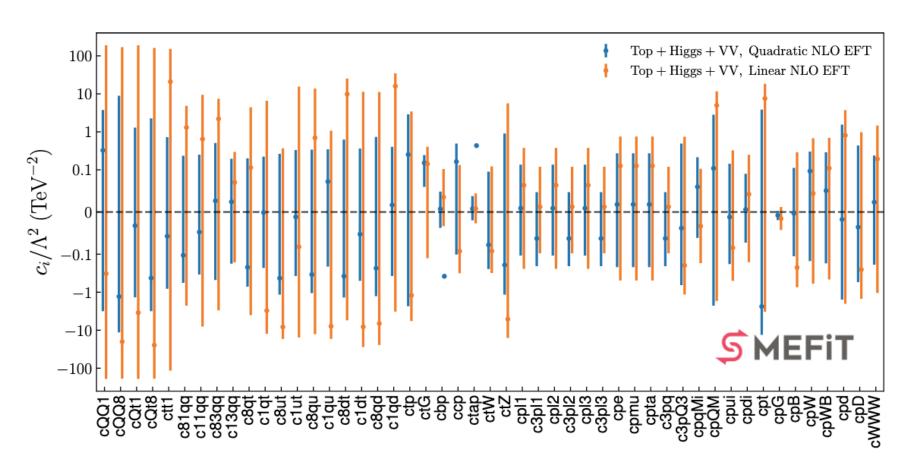


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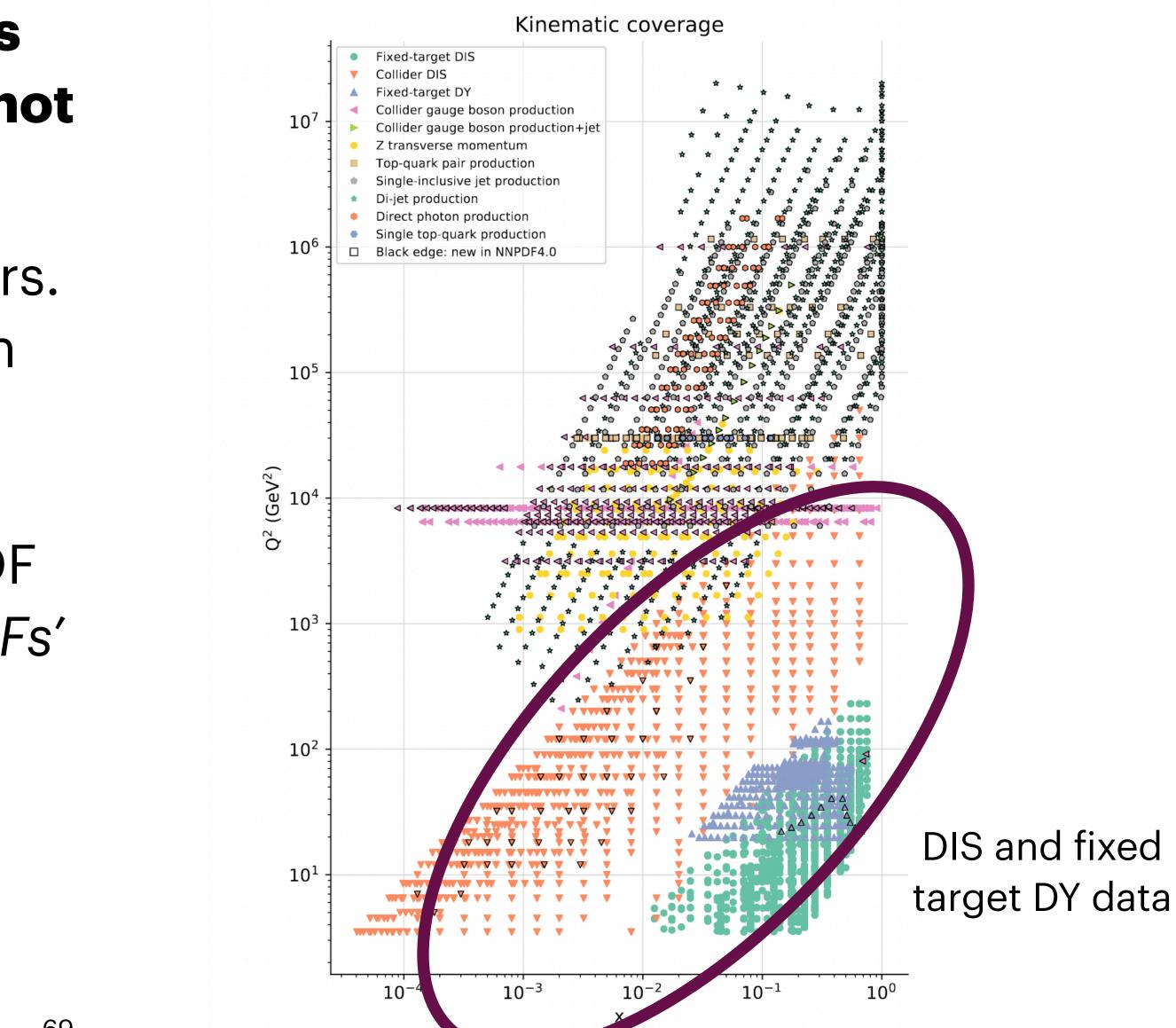
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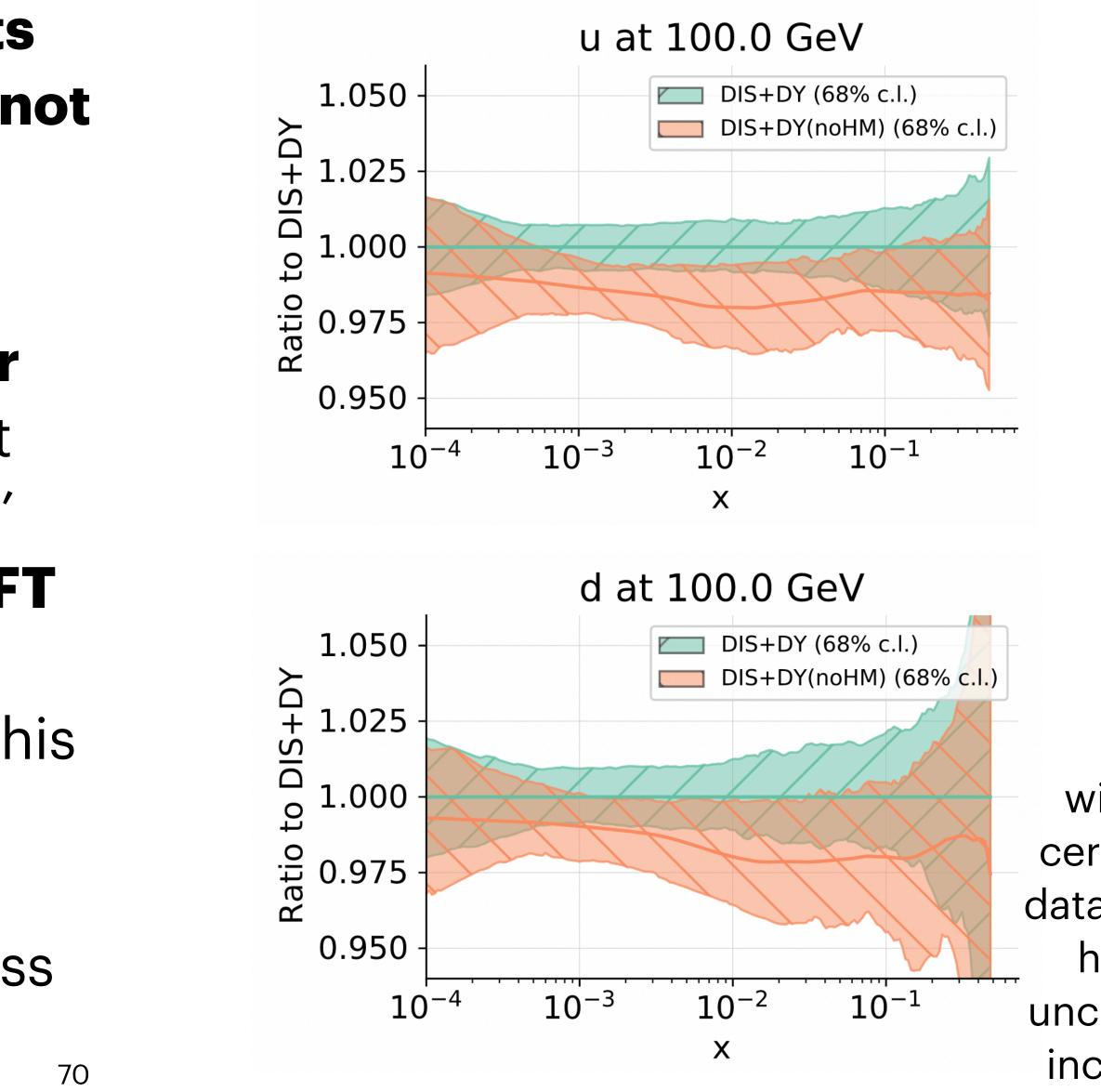
We could even miss New Physics, or see New Physics that isn't really

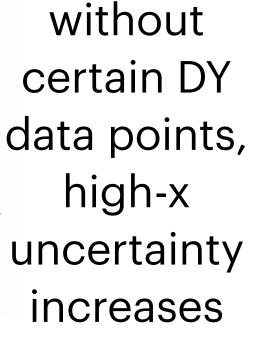
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  - It depends on the SMEFT operators. Some operators (e.g. four-fermion operators) will **contaminate DIS** and DY data, which comprise the majority of the data going into PDF fits. So often 'uncontaminated PDFs' don't exist!
  - Right: kinematic coverage of NNPDF4.0 by dataset.



- Question 1: Can't I just use PDF sets which are fitted using data that is not affected by SMEFT operators?
  - Furthermore, if we include more data in a PDF fit, we obtain **better** quality fits. Therefore, we expect that using 'uncontaminated PDFs' will result in poorer quality SMEFT fits; we won't be using the 'best' quality' PDFs that are available - this is shown explicitly in Greljo et al., 2104.02723, where PDF sets including and excluding high-mass DY data are compared.



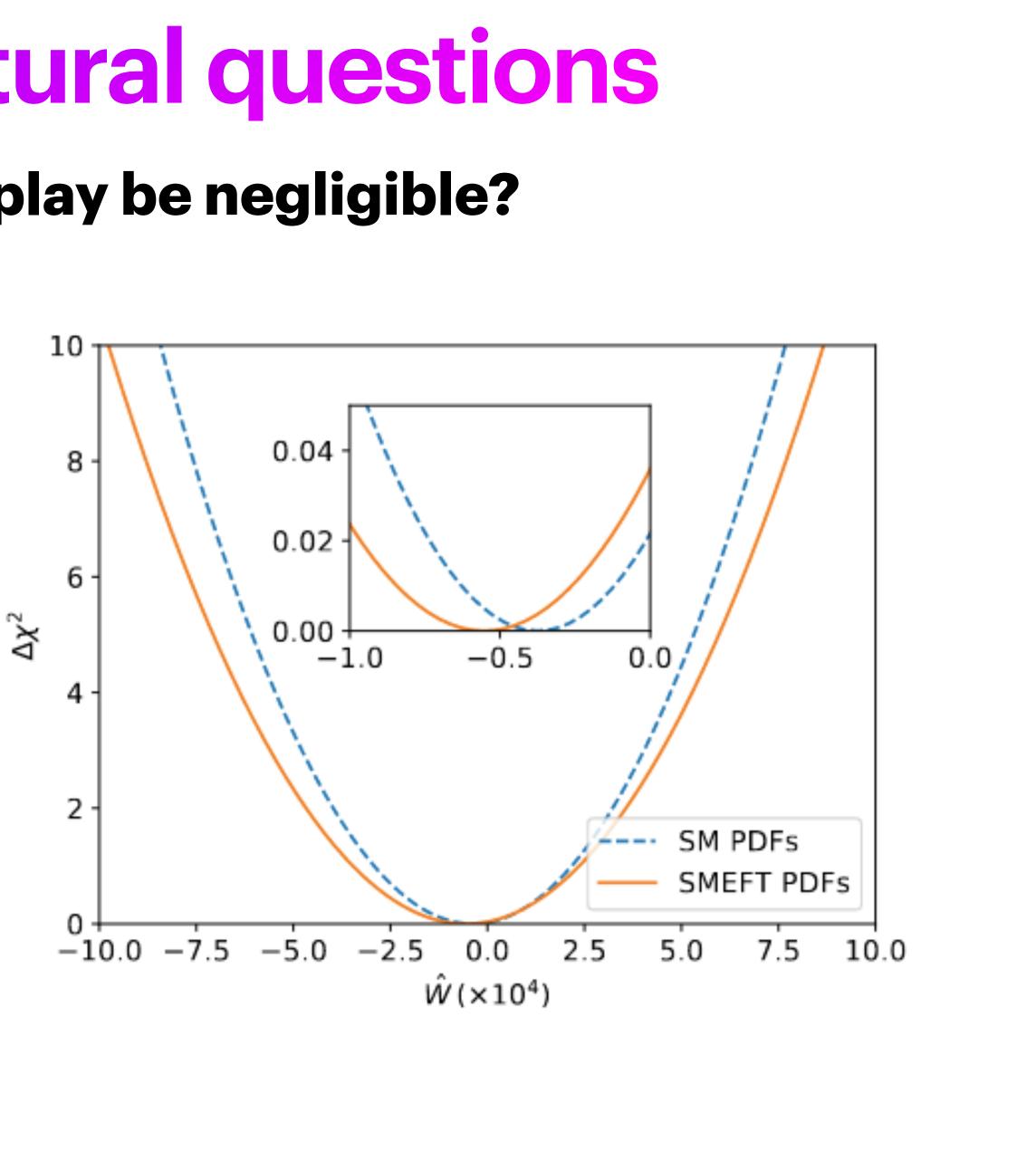


• Question 2: Won't the PDF-SMEFT interplay be negligible?

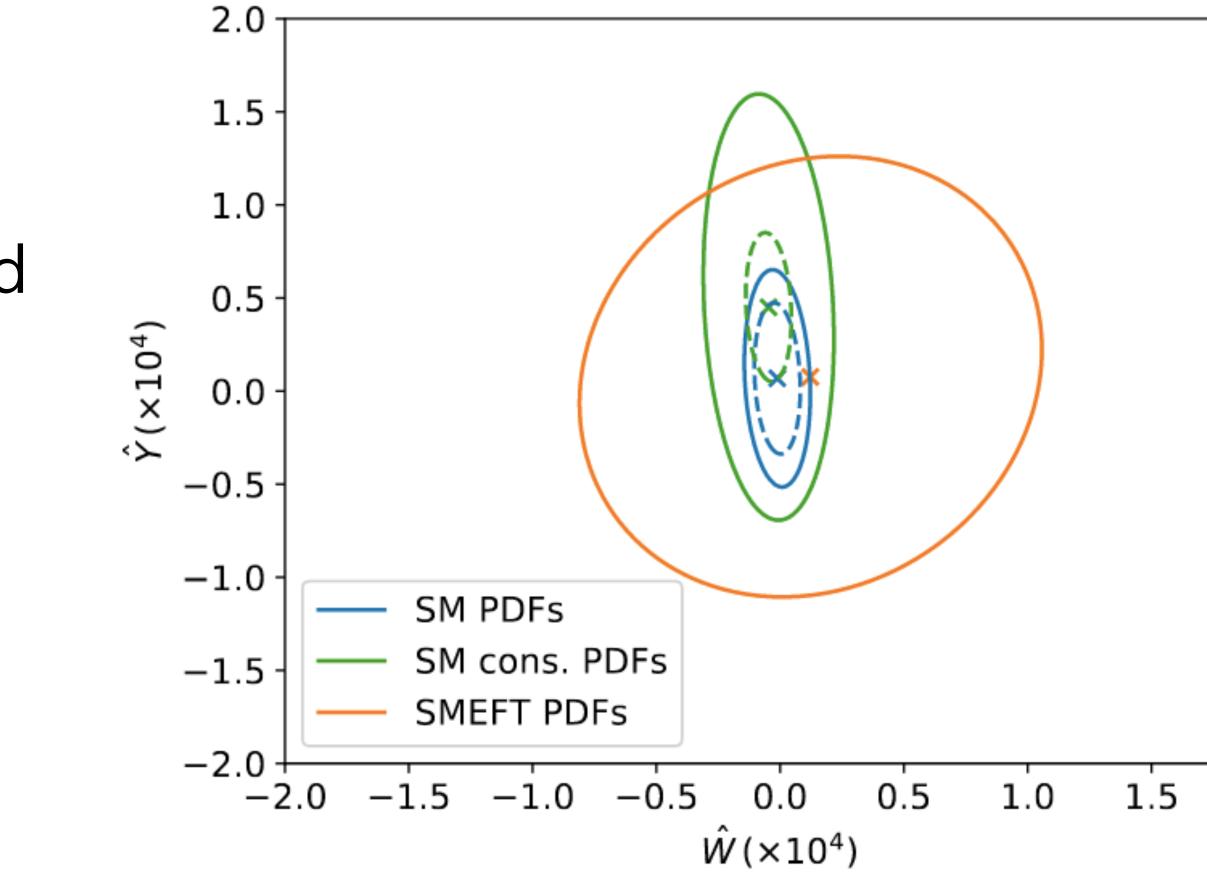
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  - Similarly, it was shown in the PBSP team's earlier study, Greljo et al., 2104.02723, that interplay is mild between the W, Y operators and PDFs using current DIS and DY data.

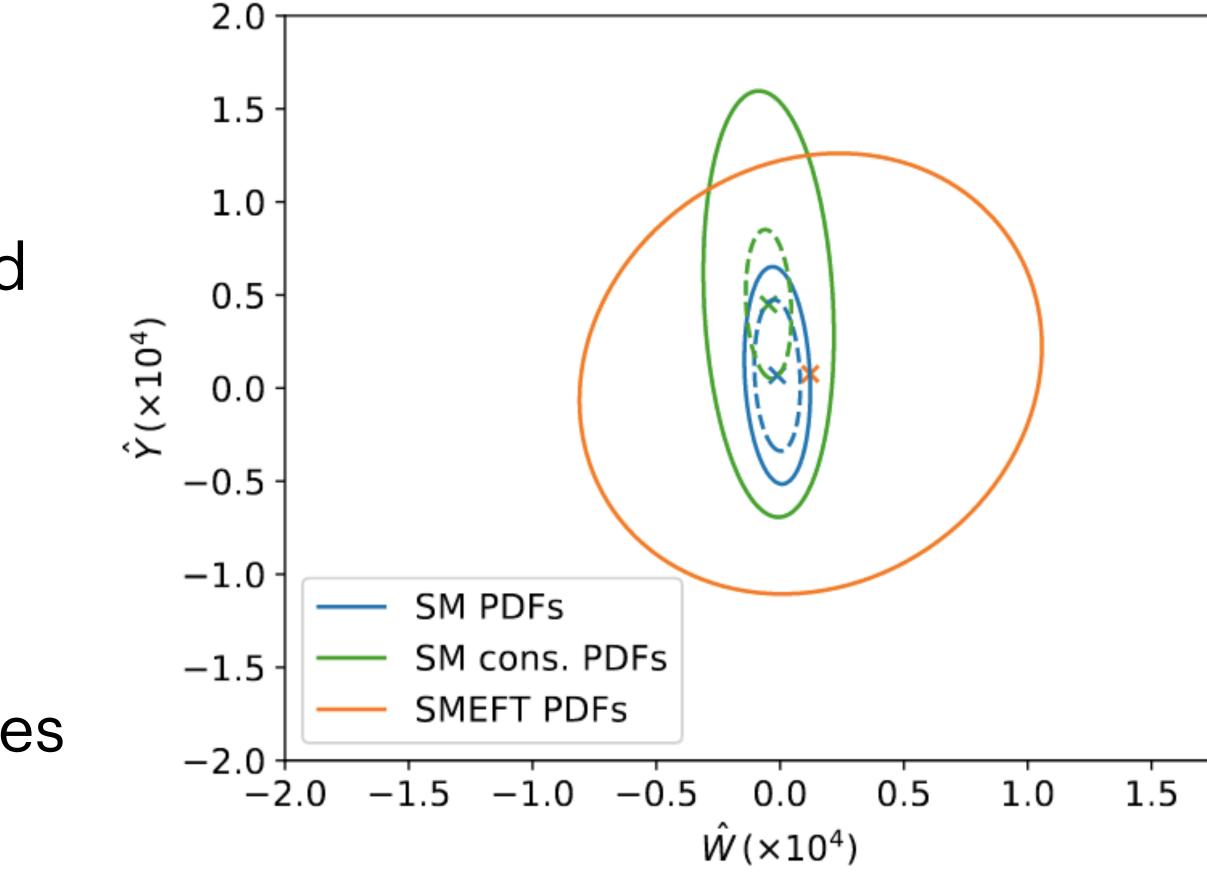


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  - However, it was also shown in Greljo et al., 2104.02723, that interplay is **very significant** between the  $\hat{W}$ ,  $\hat{Y}$  operators and PDFs using projected high**luminosity DY data**.
  - We see that using fixed PDFs results in a **significant** underestimation of uncertainties on the WCs - we might wrongly conclude **New Physics**!





# 4. - The SIMUnet methodology for joint PDF-SMEFT fits

we now need an **efficient methodology** to perform the fits.

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#### **1. 'Scan' methodology**

- Select a grid of benchmark SMEFT points.
- Perform PDF fits at each benchmark point.
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• Model the  $\chi^2$ -surface as a neural network, with inputs given by PDF parameters

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### See 2201.06586 and 2211.01094

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See 2201.06586 and 2211.01094

### **3. SIMUnet methodology**

- Extend the NNPDF replica networks with a new layer with edges corresponding to the WCs.
- Train the network as per an NNPDF fit, but also learning the WCs.

#### See 2201.07240

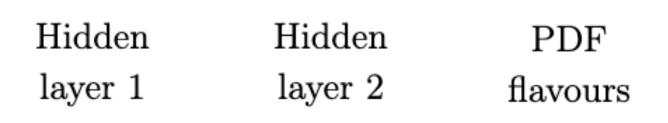


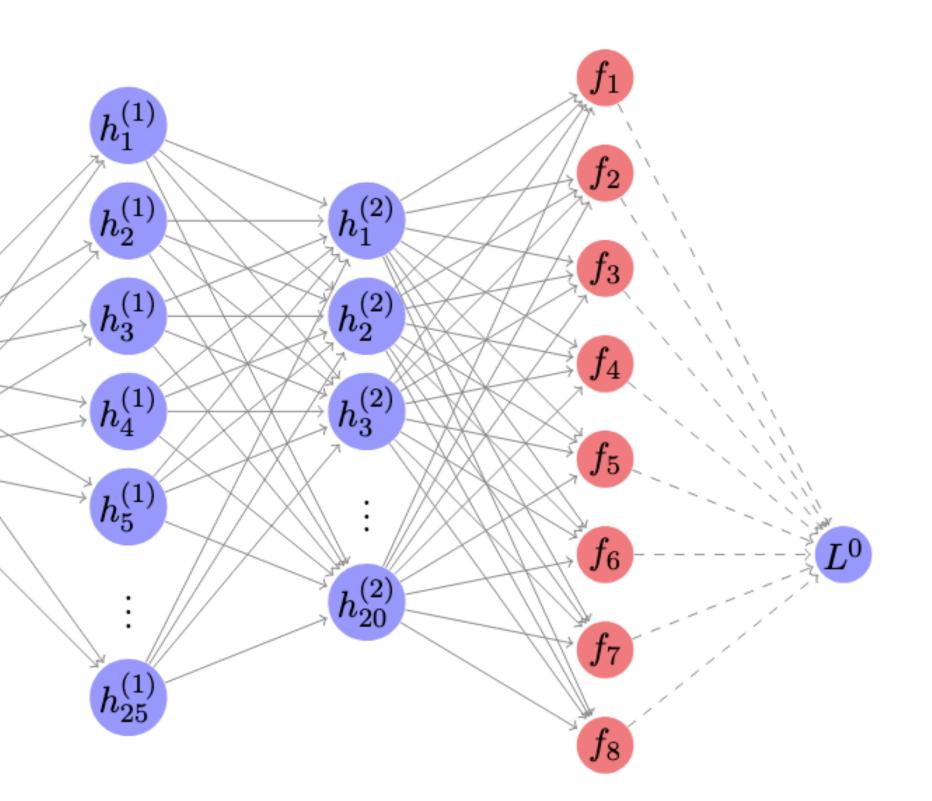
 The SIMUnet methodology extends the existing NNPDF neural network with an additional convolution layer.

Input layer

 $\ln x$ 





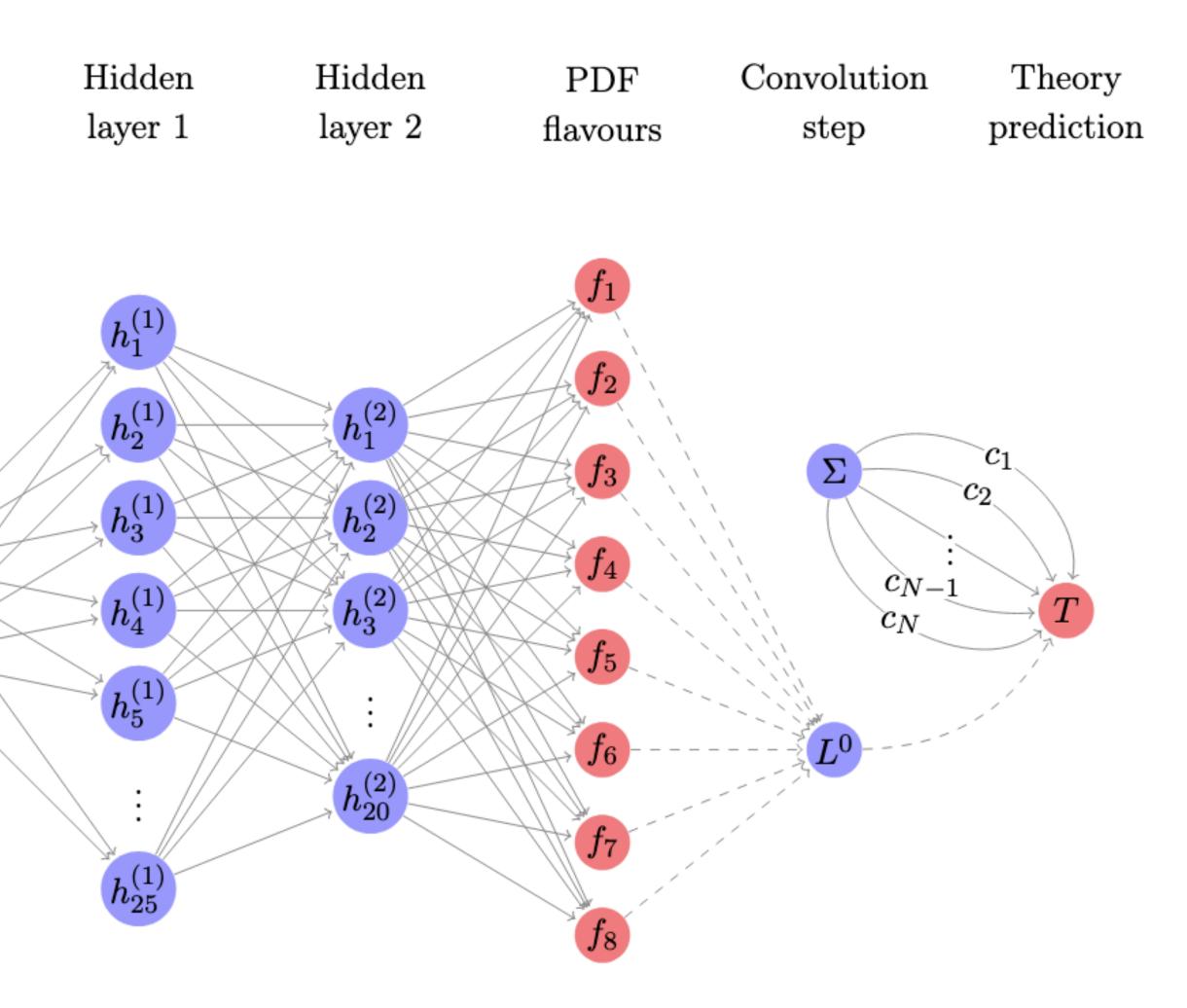


- The SIMUnet methodology extends the existing NNPDF **neural network** with an additional convolution layer.
- The SMEFT couplings are added as weights of neural network edges, and are trained alongside the PDFs.

Input layer

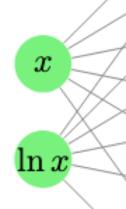
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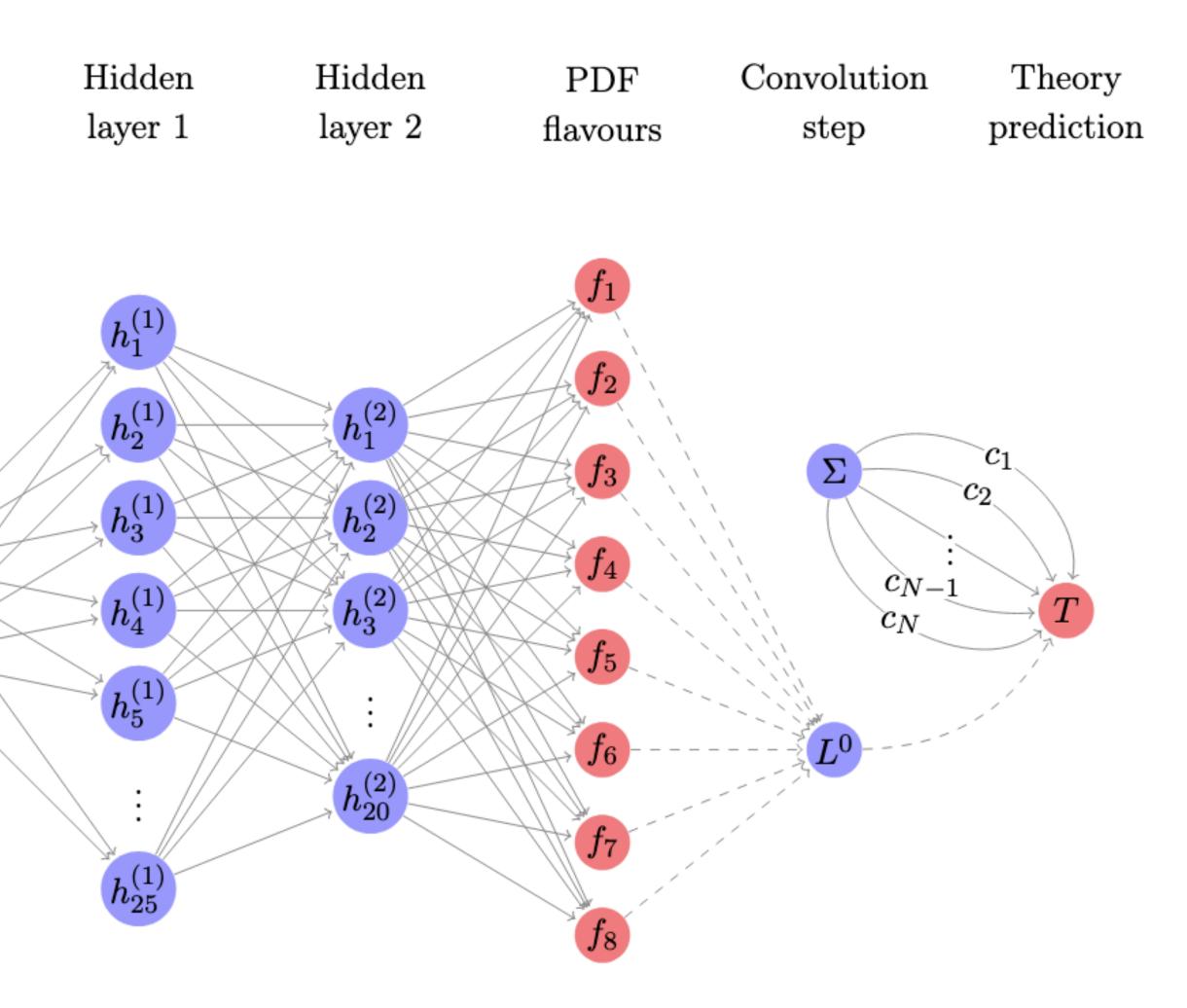


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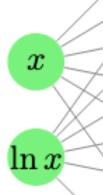




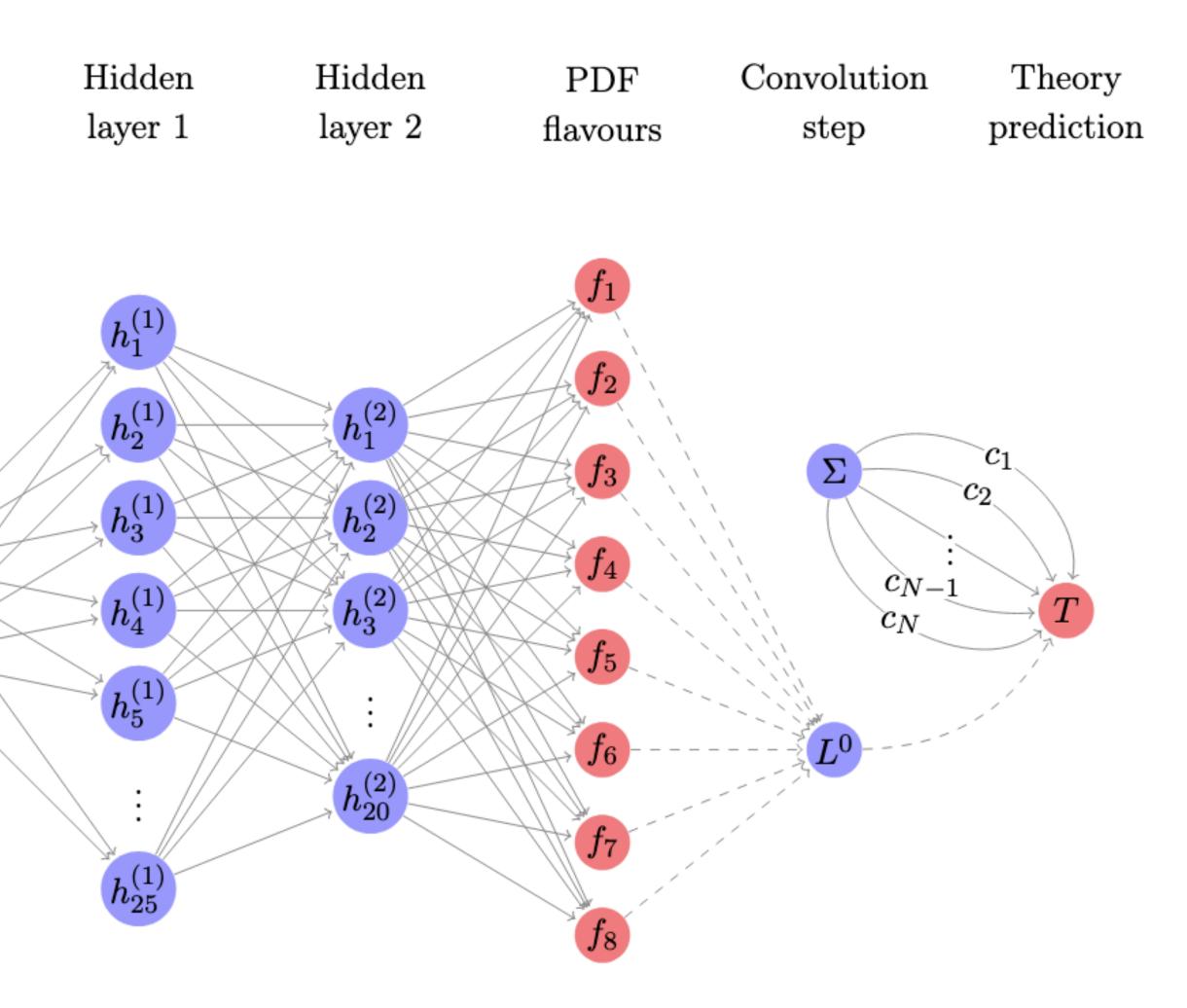
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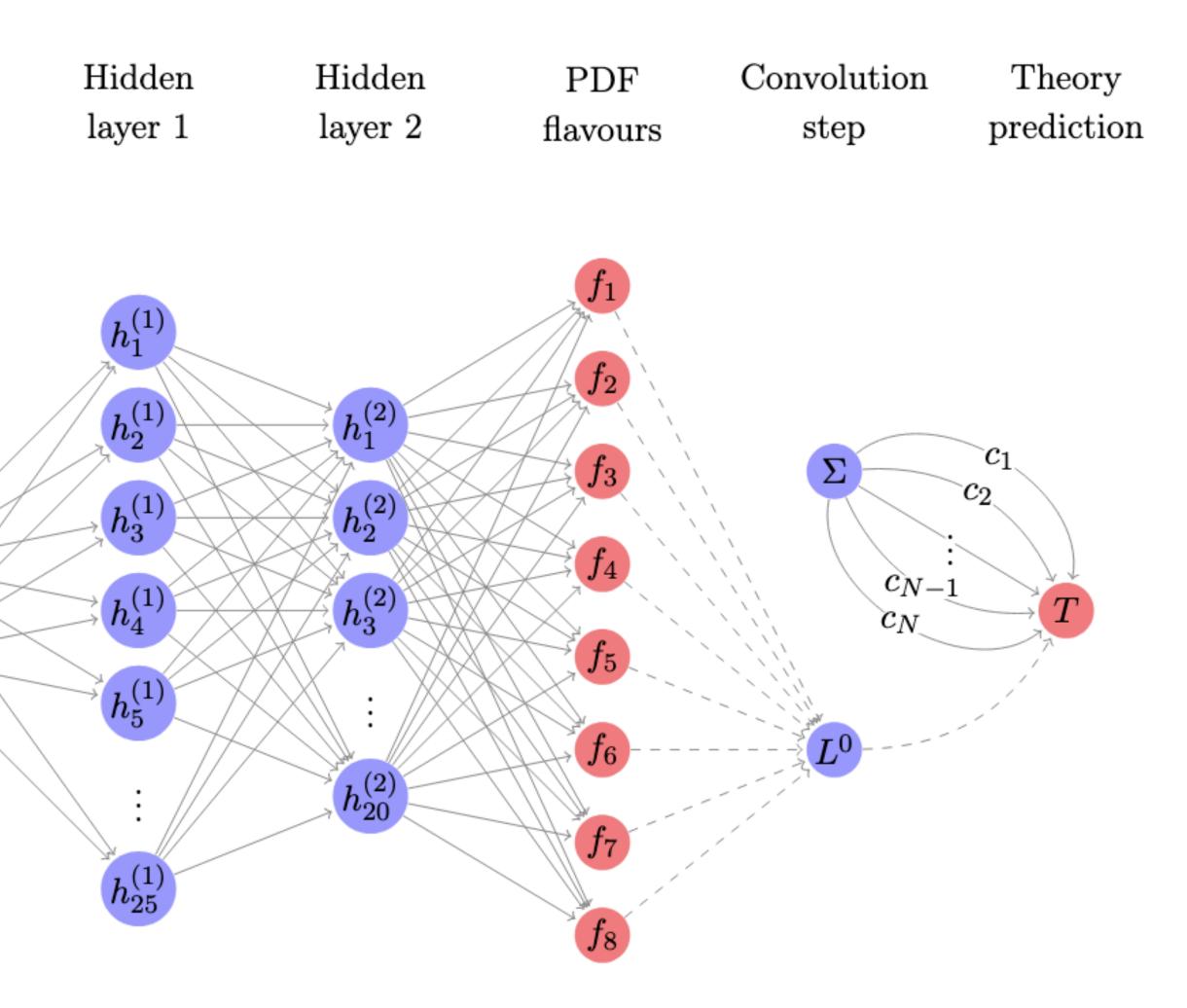
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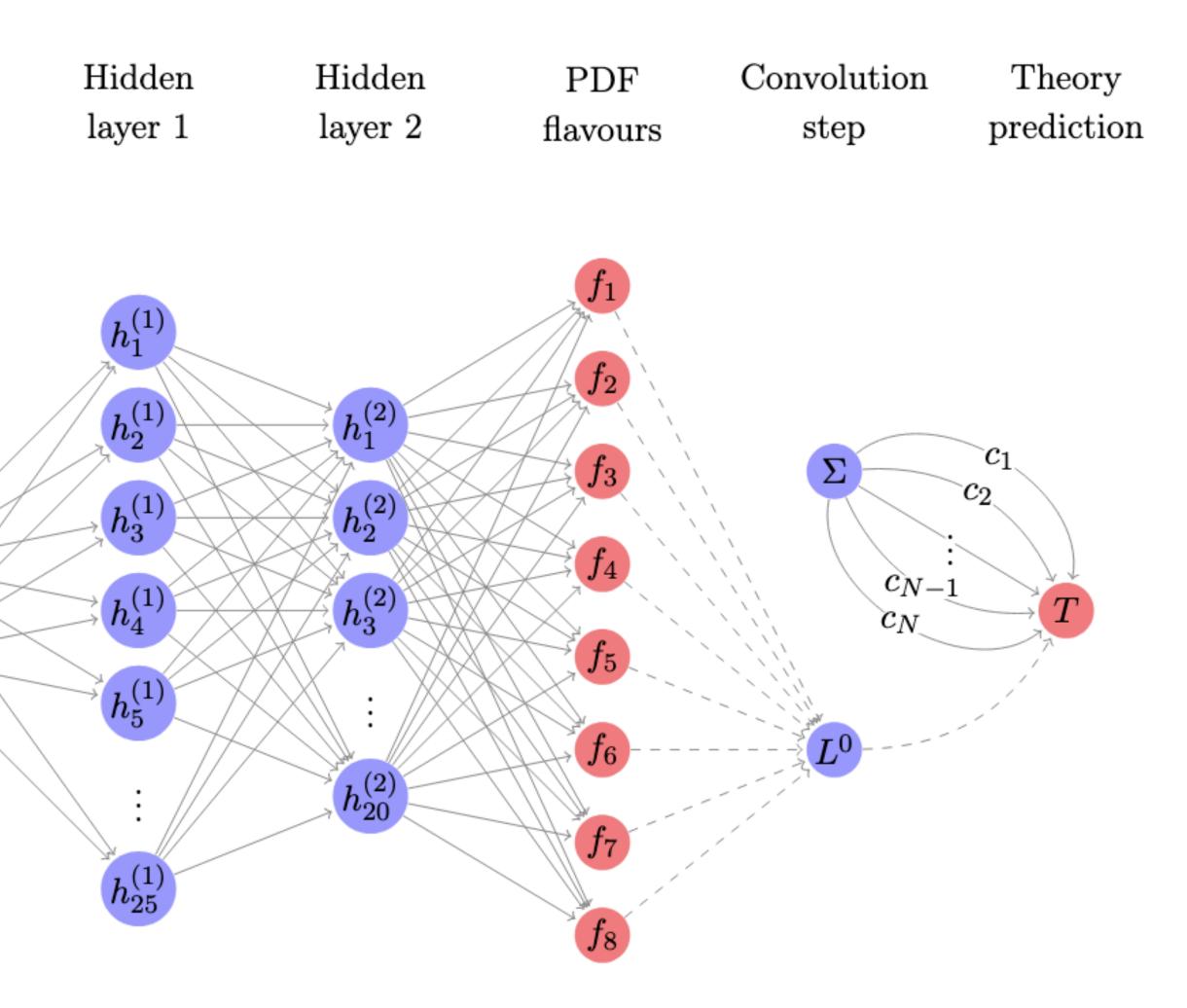
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Input layer

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- Can easily include **PDF**independent observables.
- Can perform **fixed PDF** fits by freezing the PDF part of the network.





## 5. - The top quark legacy of the LHC Run II for PDF and SMEFT analyses

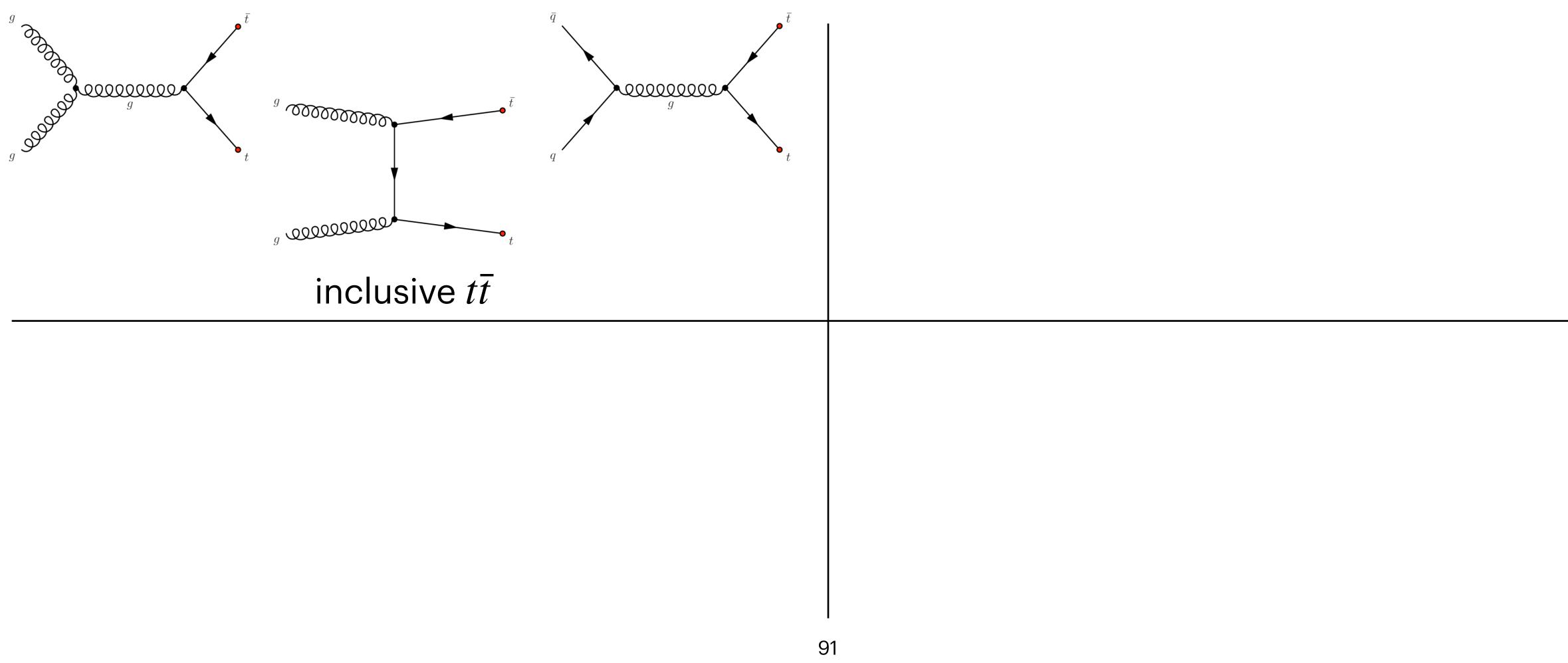
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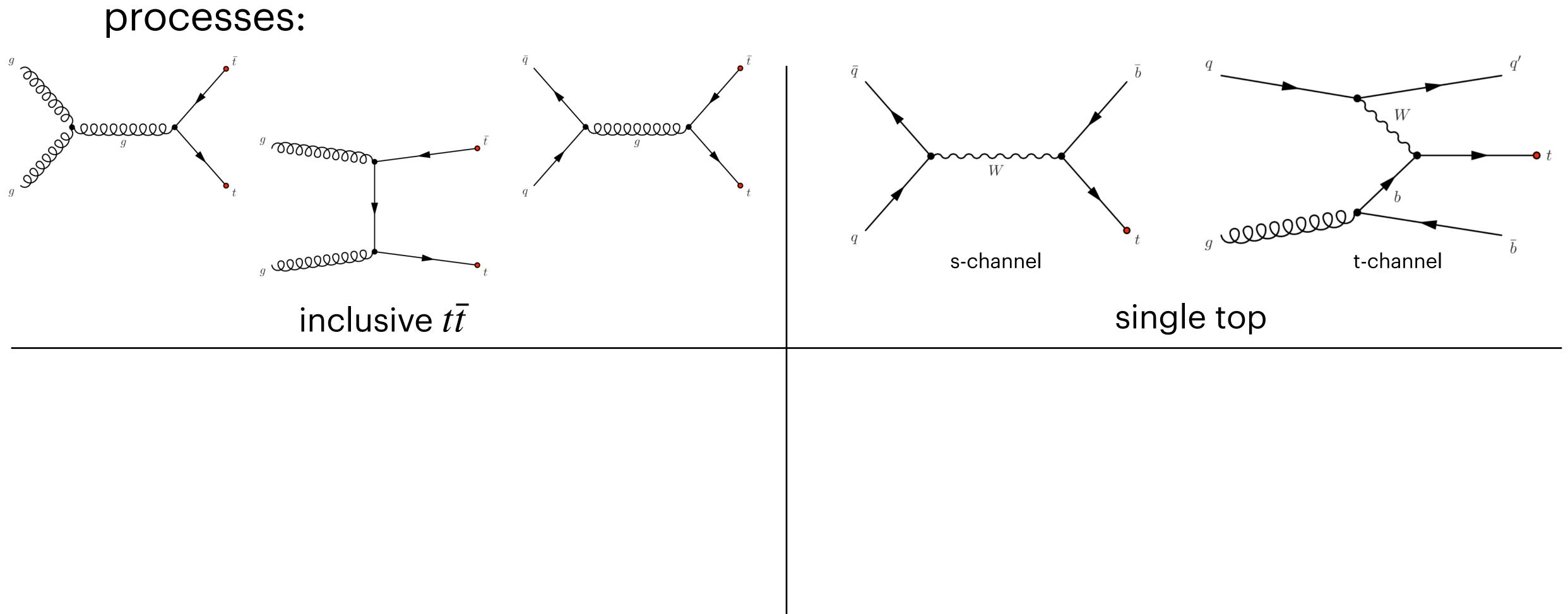
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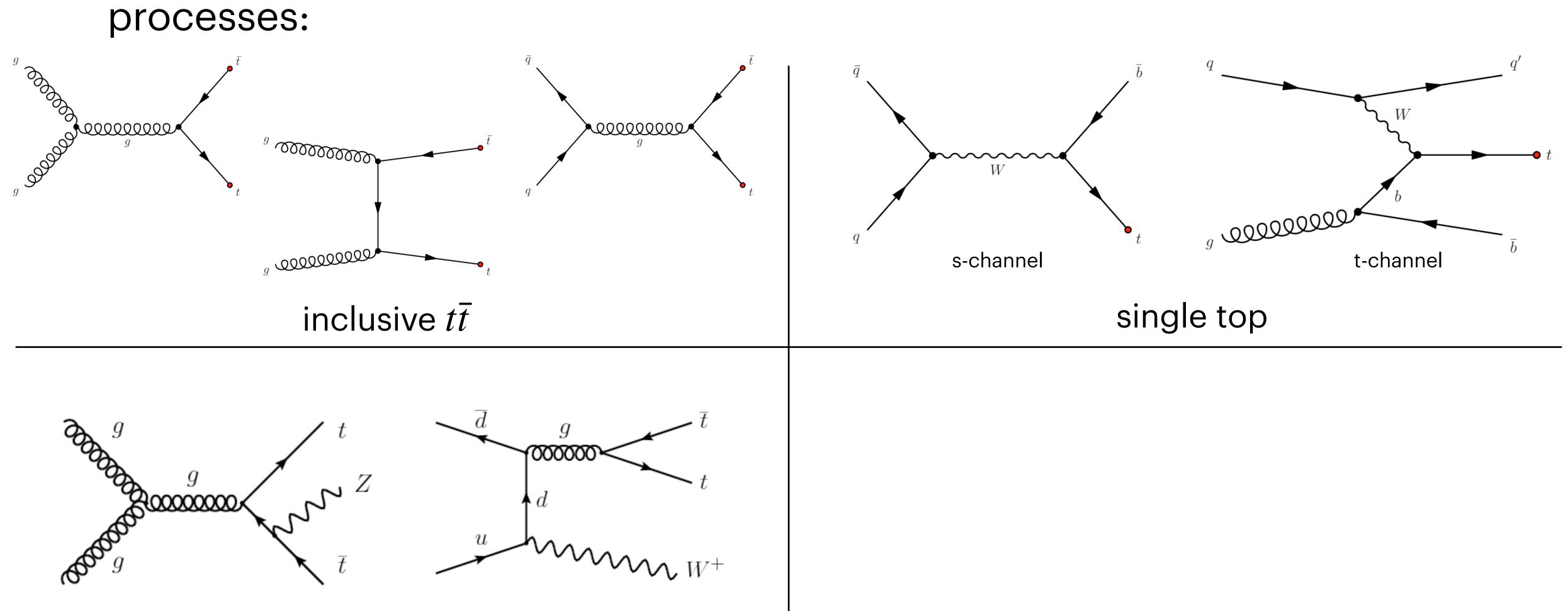
#### processes:

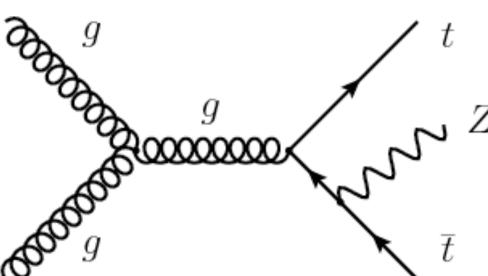


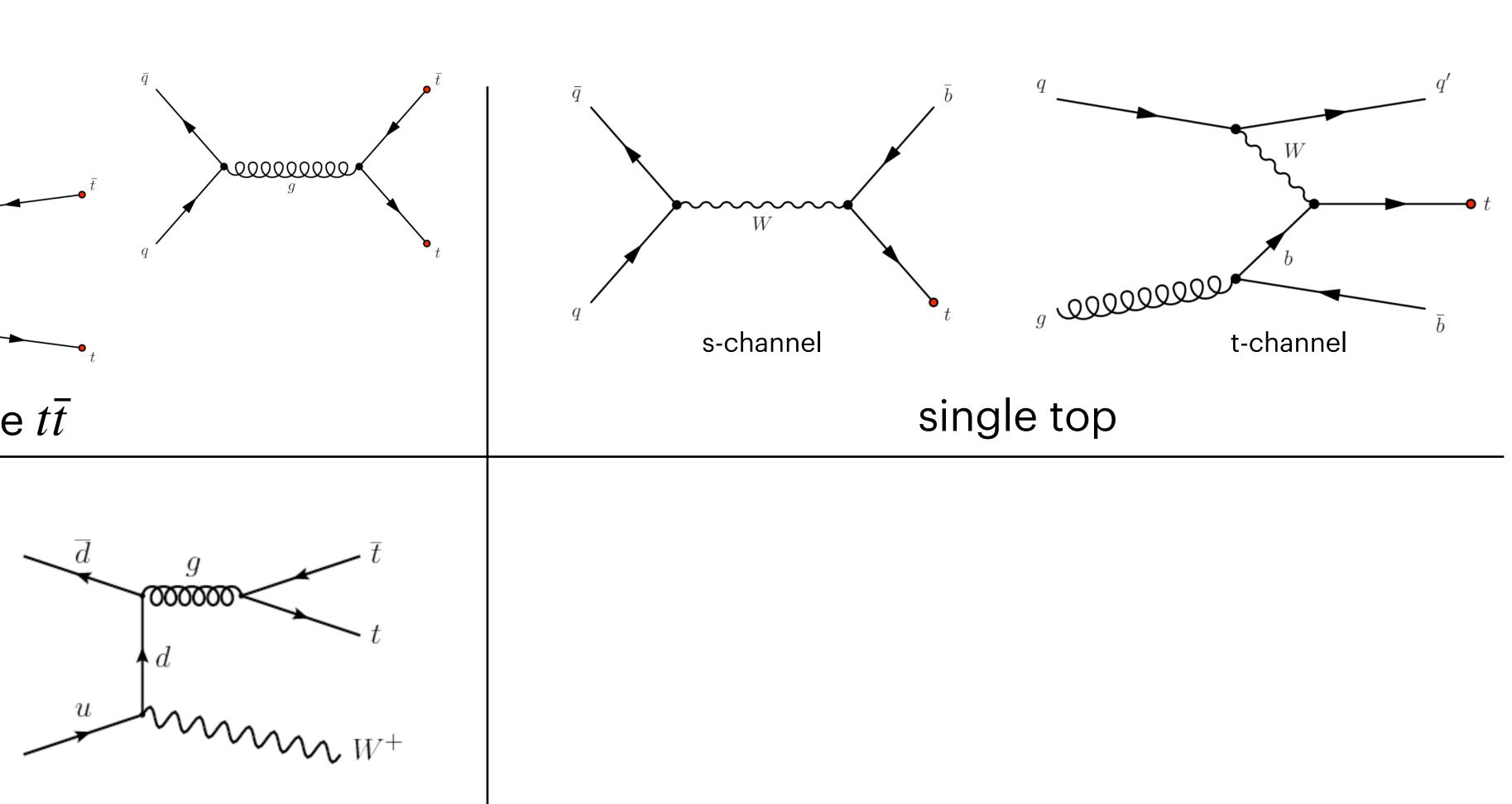
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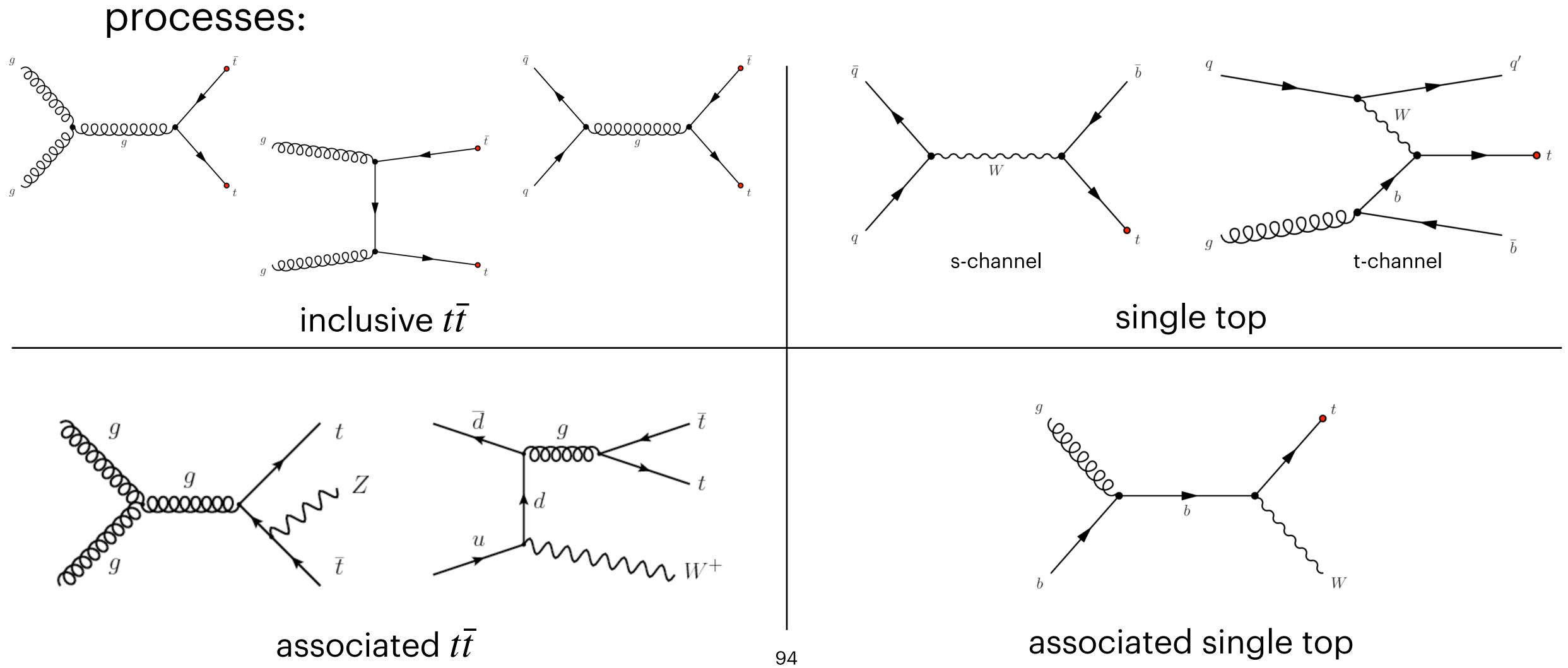


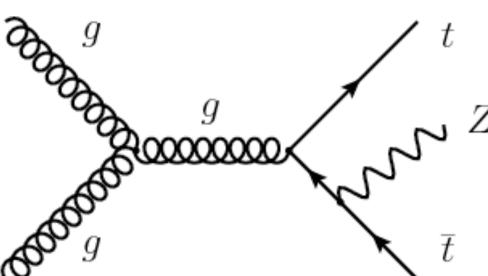


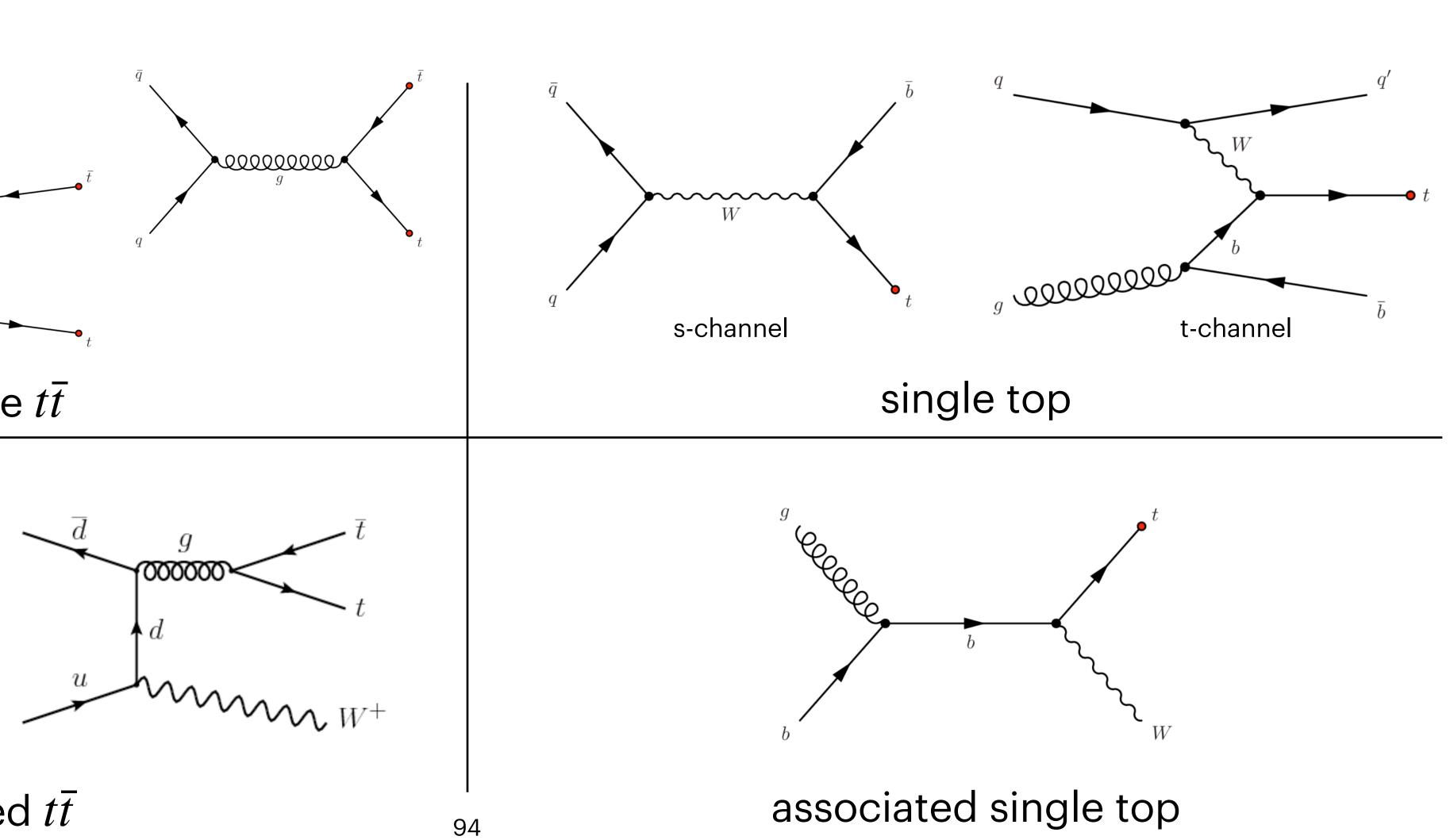


associated  $t\bar{t}$ 

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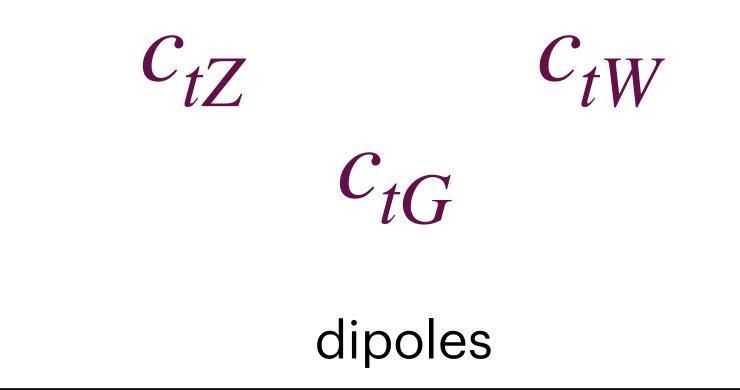




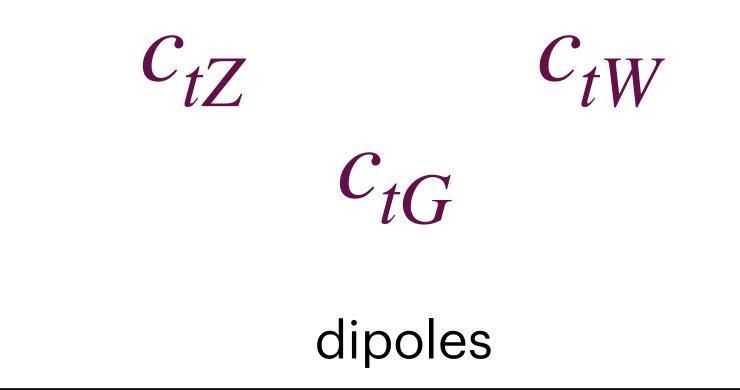


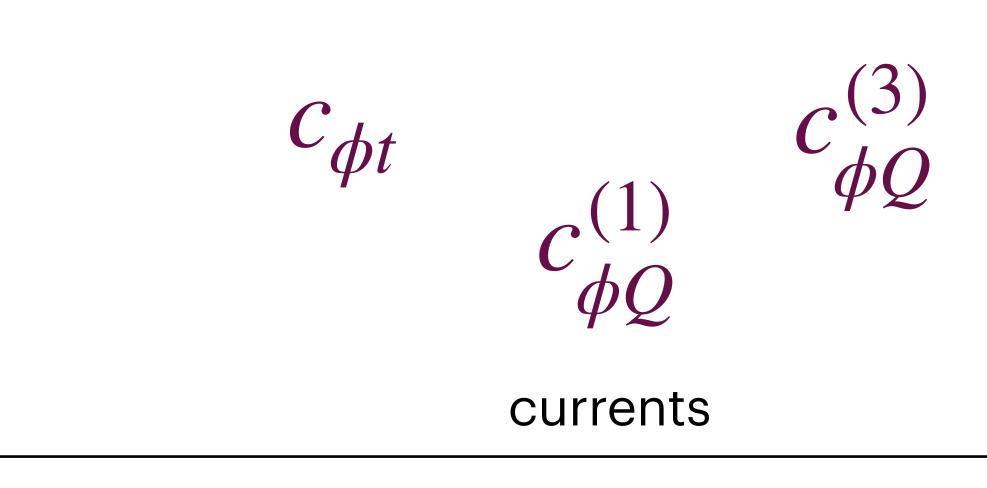
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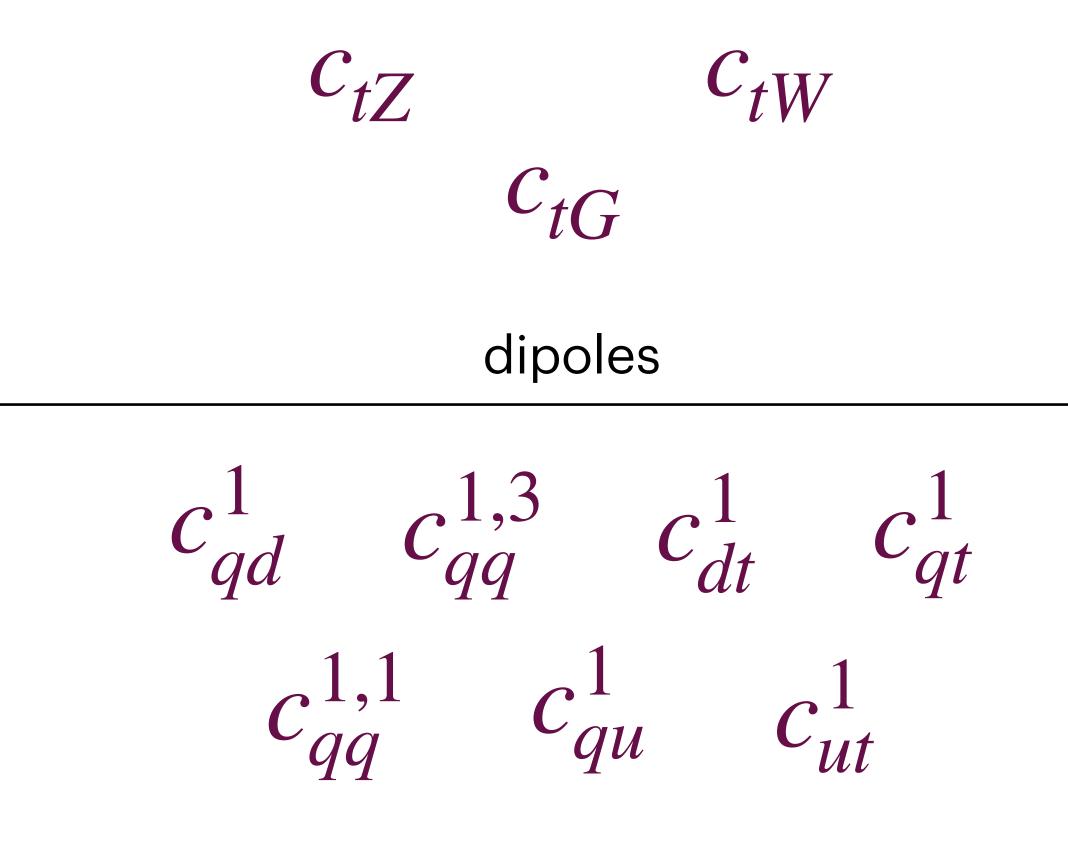


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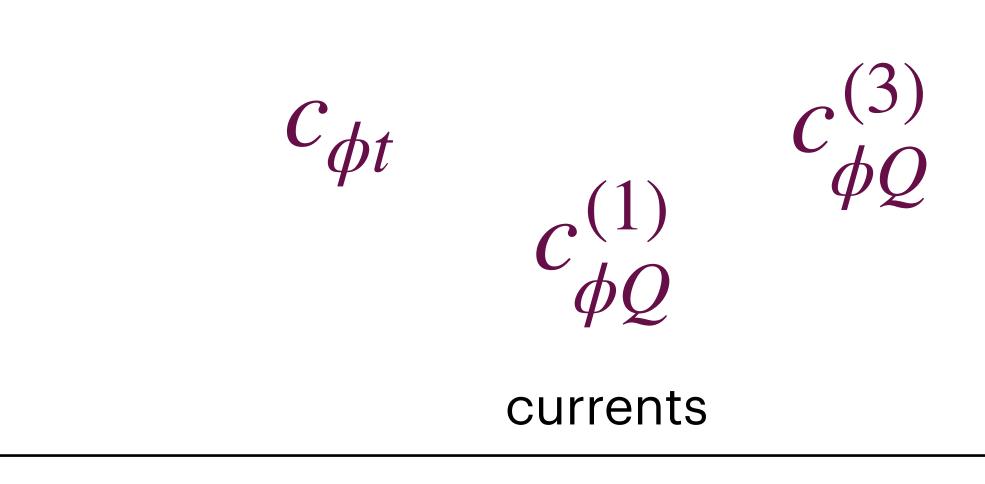




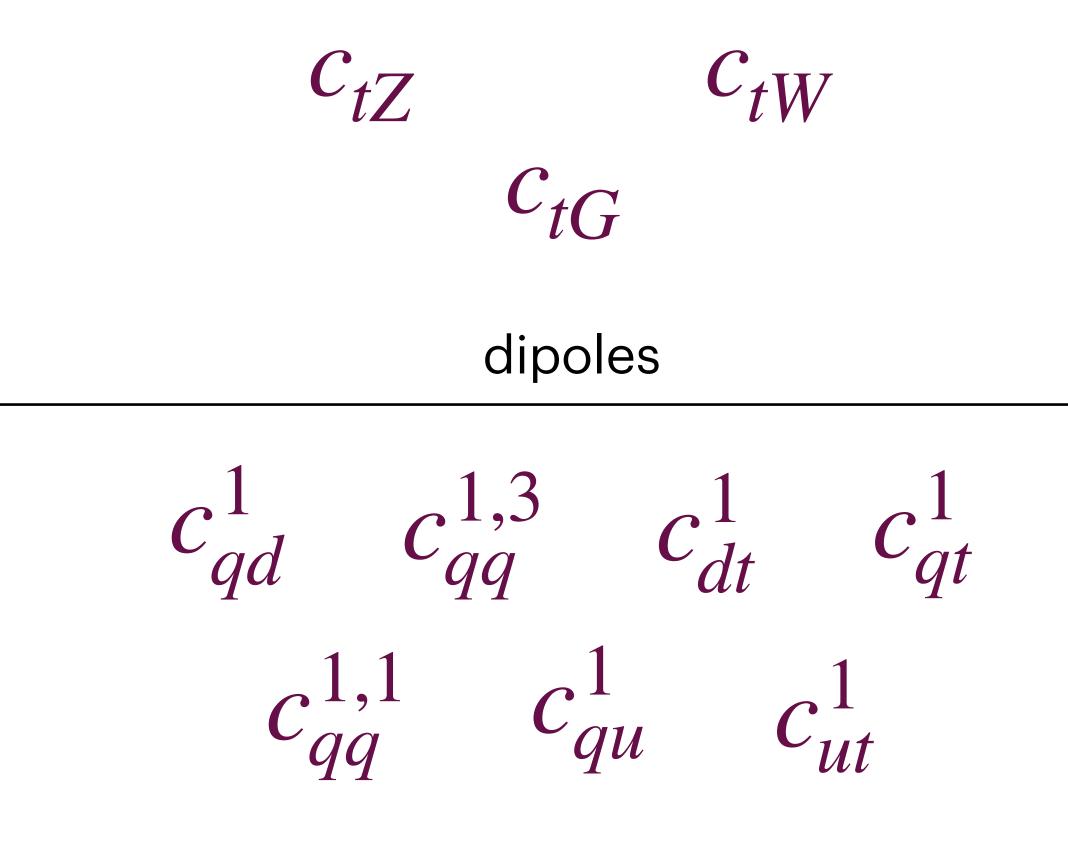
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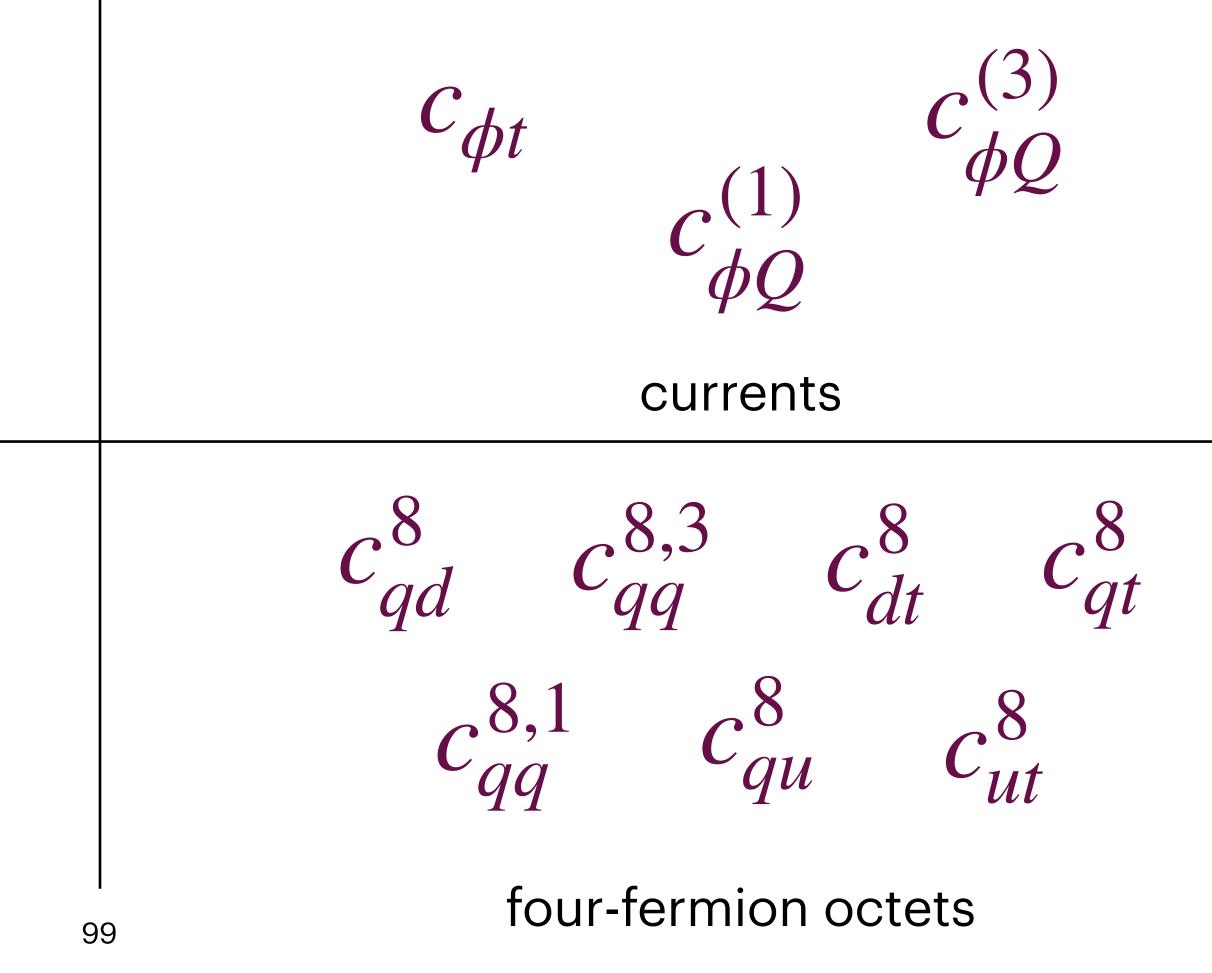
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### Key questions for the rest of the talk:

# 1. How do WC bounds compare between fixed PDF EFT-fits and simultaneous fits?

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fits and simultaneous PDF-EFT fits?

# 2. How do PDFs compare between SM PDF

 Using the SIMUnet methodology, we have performed simultaneous determinations of PDFs and top-sector WCs using the **most** comprehensive and up-to-date LHC top dataset possible.

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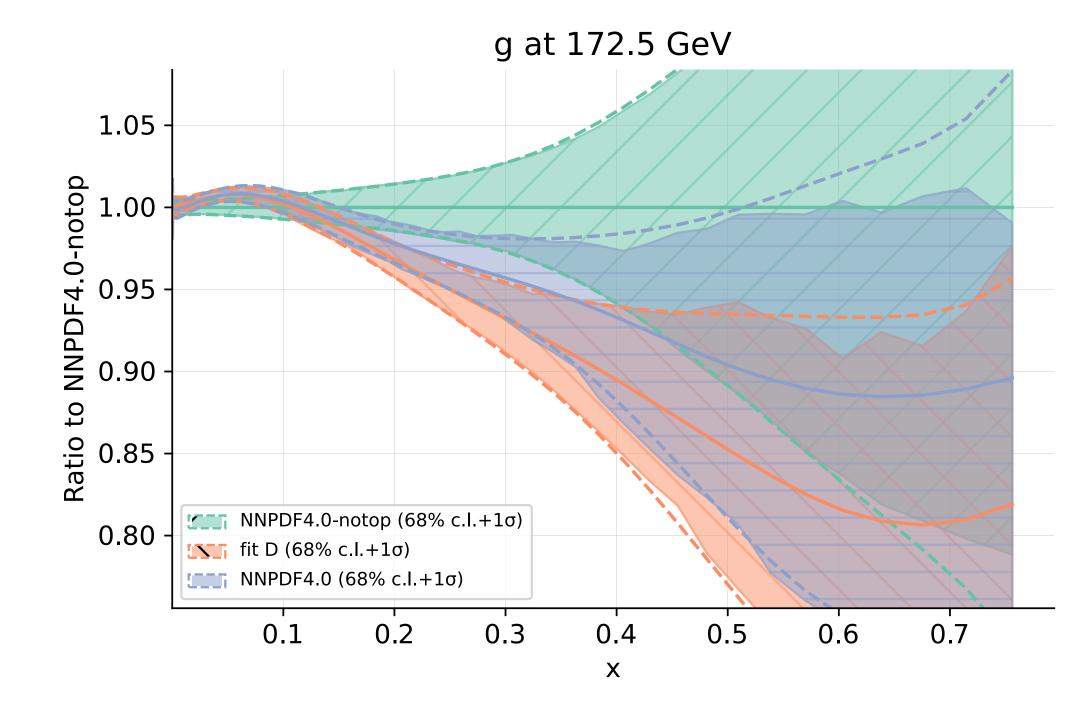
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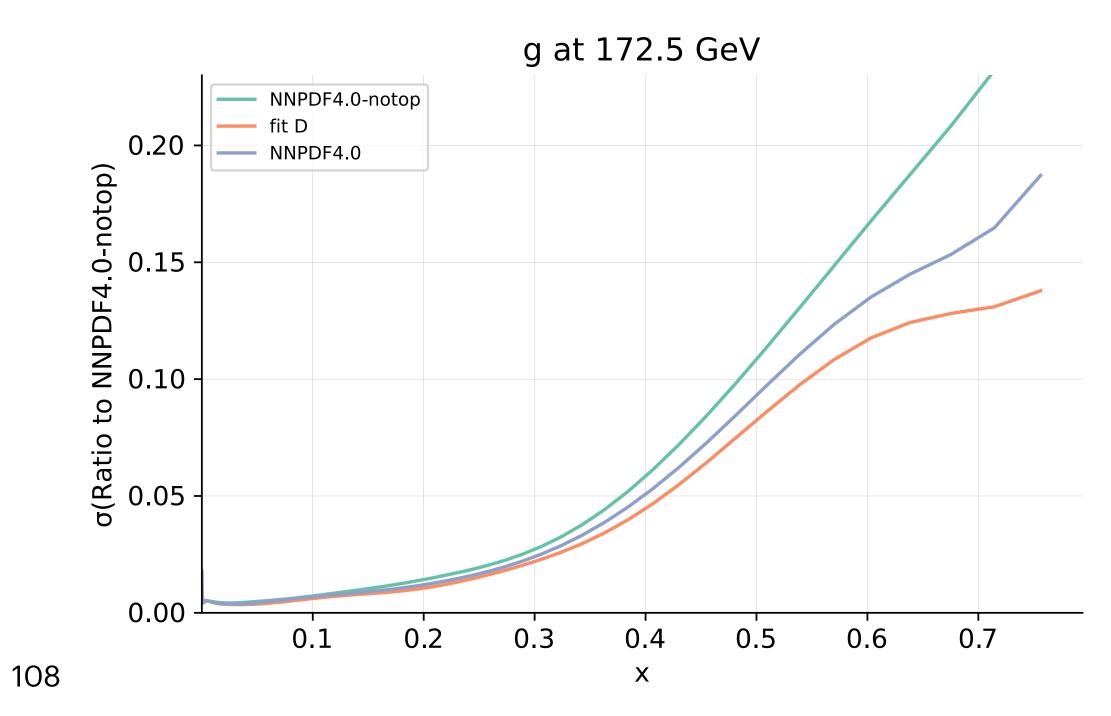
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- We work with theory predictions accurate to NNLO in QCD in the SM, and include NLO QCD in the SMEFT. Some fits are linear in the SMEFT, some are quadratic - a point we will return to.

### **PDFs in the SM - impact of inclusive** $t\bar{t}$ and single-top

- First, we consider the impact of our dataset on PDFs in the SM.
- Begin by considering the updates to the **inclusive**  $t\bar{t}$  and **single-top** dataset relative to NNPDF4.0. If we perform a SM PDF fit using only our new inclusive  $t\bar{t}$  and single-top data, we see a more pronounced effect on the large-x gluon relative to NNPDF4.0. The uncertainty is also further reduced.

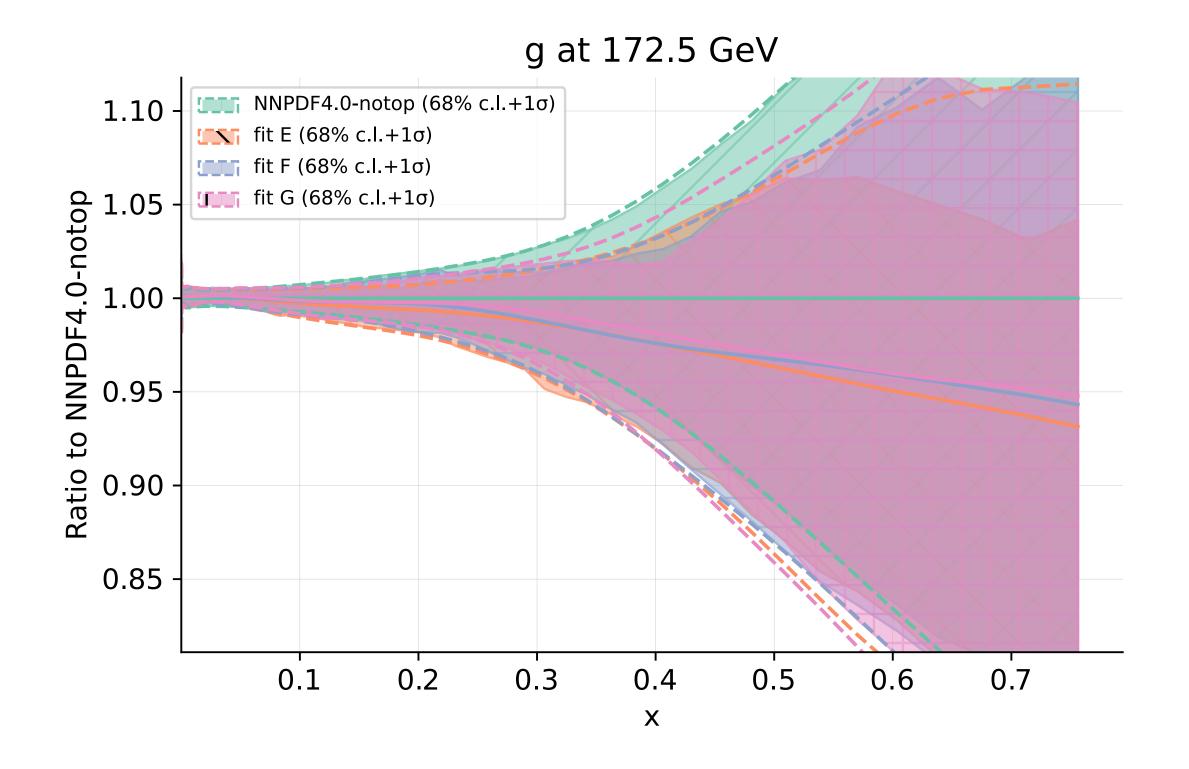




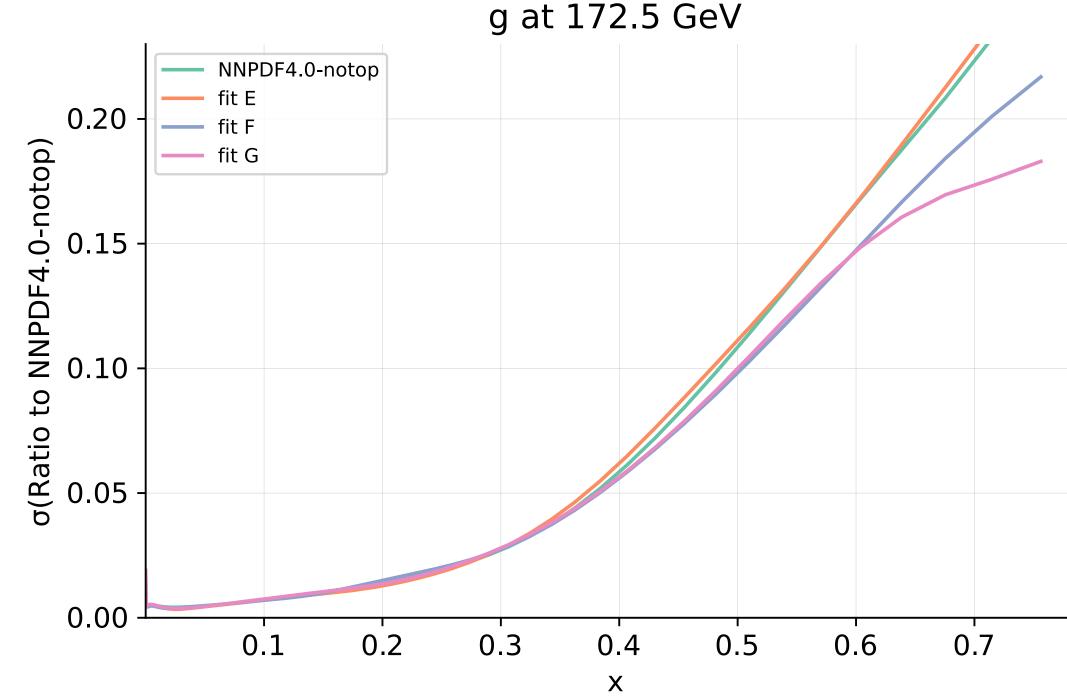


#### **PDFs in the SM - impact of associated top**

reducing it at large-x, and fractionally reducing uncertainty.

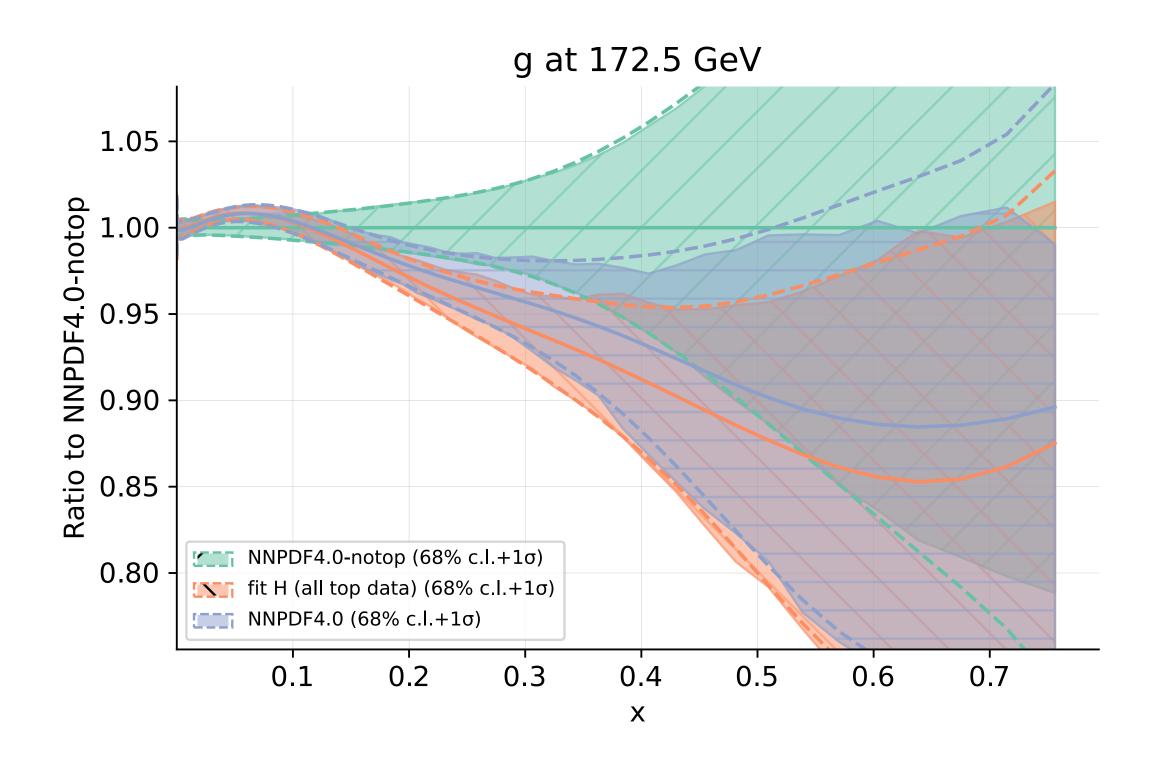


• Next, for the first time we consider the impact of **associated top data** in a PDF fit. There is only a very mild effect on the central value of the gluon,



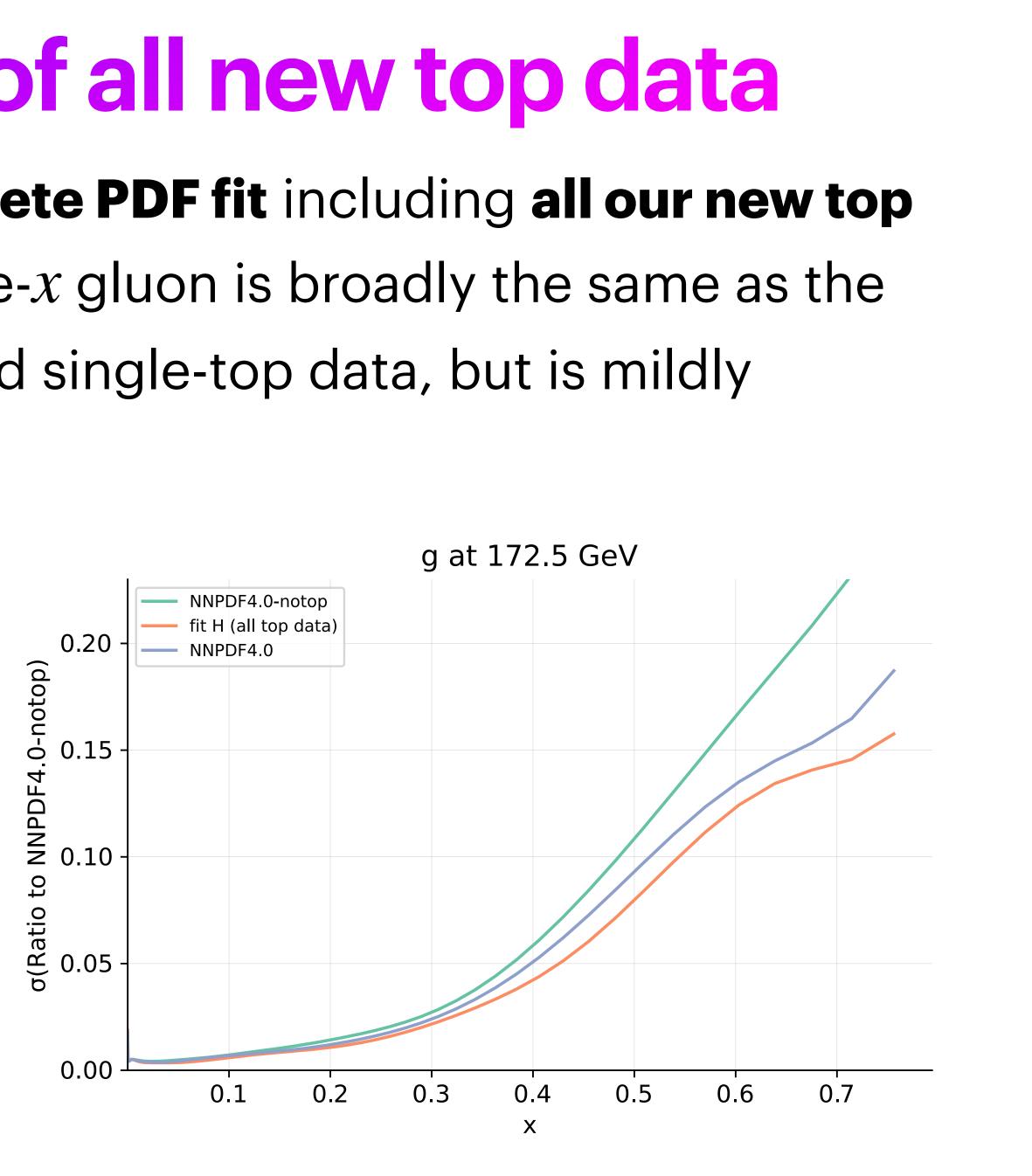
### PDFs in the SM - impact of all new top data

tempered by the associated top data.



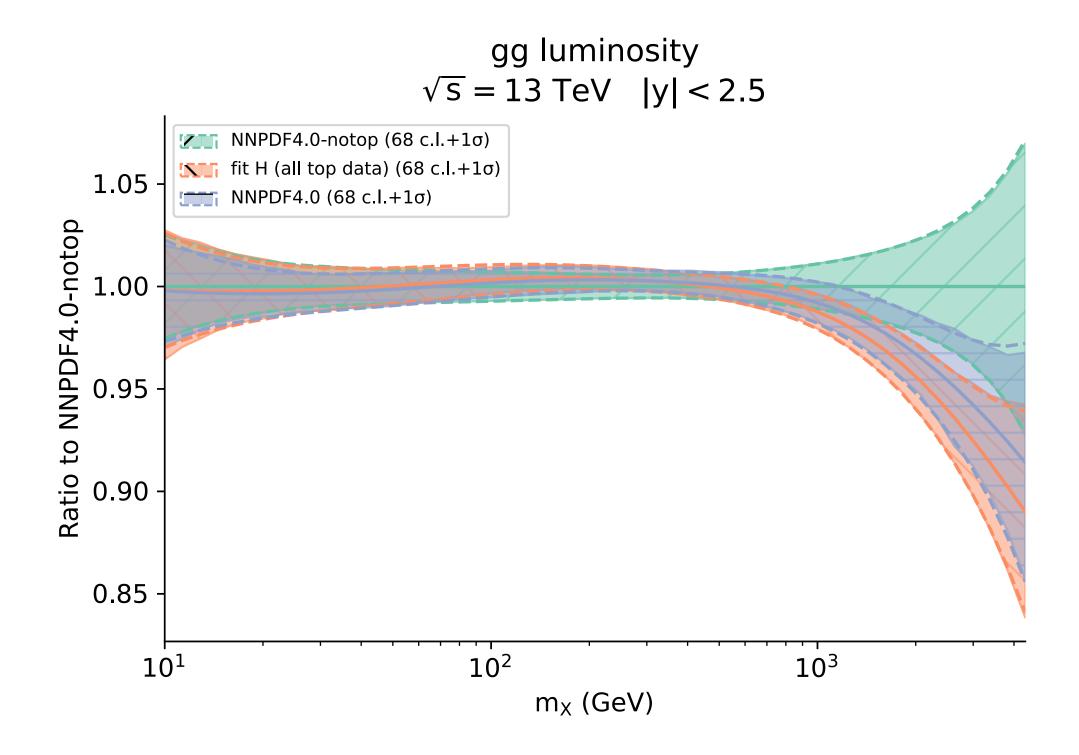
#### • Finally, we present the results of a complete PDF fit including all our new top

**data**. As expected, the effect on the large-*x* gluon is broadly the same as the effect of just including the inclusive  $t\bar{t}$  and single-top data, but is mildly

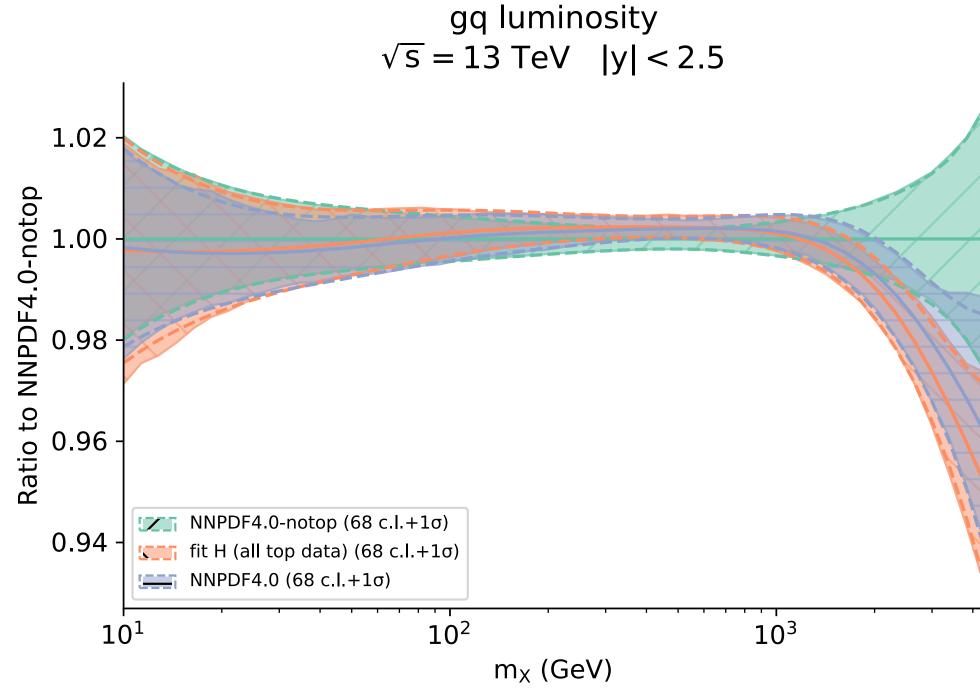


#### PDFs in the SM - impact of all new top data

• A similar trend holds for the **PDF luminosities**, with our new updated fit to NNPDF4.0 at very large invariant mass.

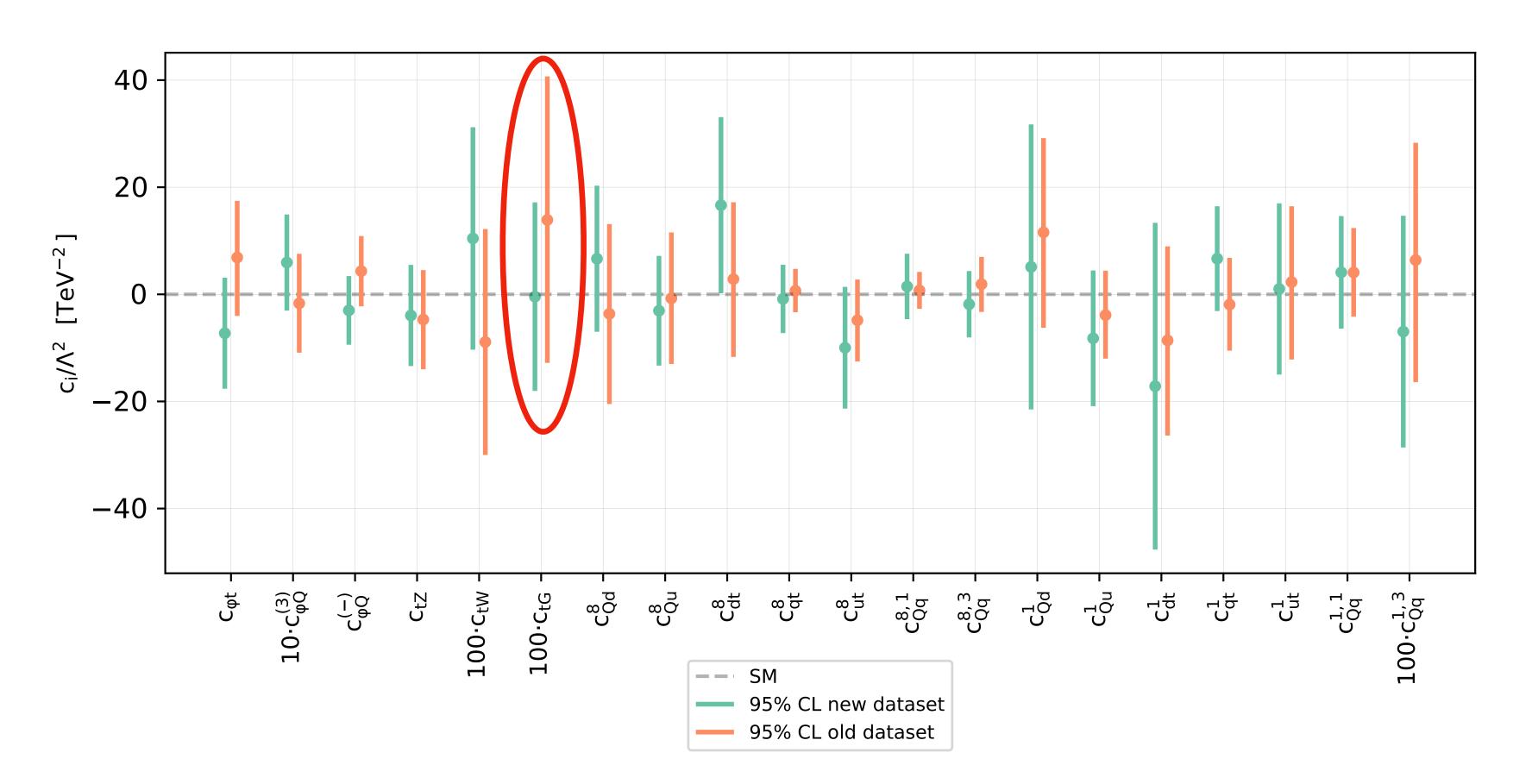


# compatible with NNPDF4.0, but with the central luminosity reduced relative



#### **SMEFT-only fits: linear SMEFT**

- relative to previous SMEFT-fits, namely **SMEFiT**.
- At the **linear level** in the SMEFT, best improvement is seen in  $C_{tG}$ , whose bound undergoes a 35% tightening this is traced to more precise total tt measurements.

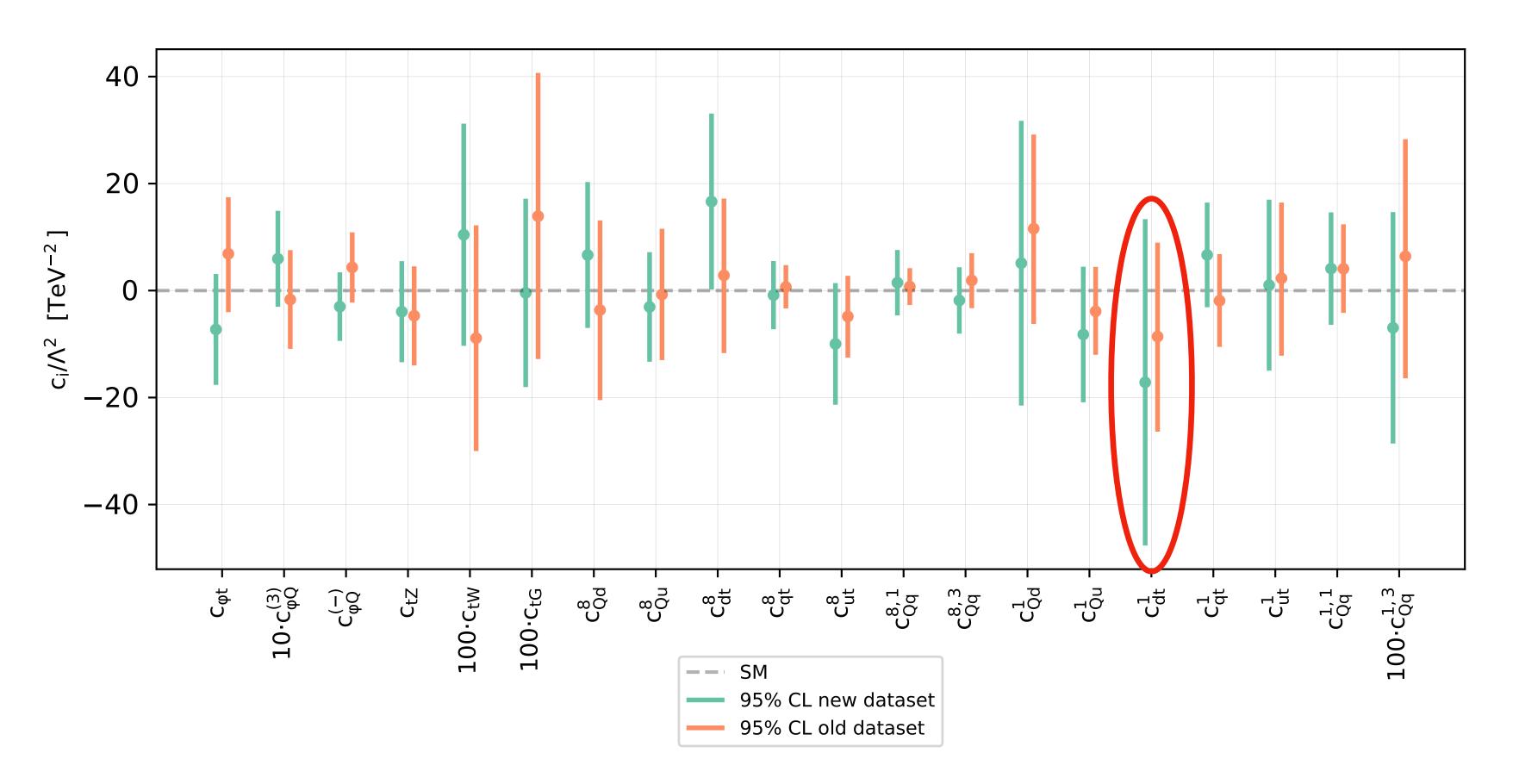




# • We have also performed SMEFT-only fits to see the impact of our new dataset

### **SMEFT-only fits: linear SMEFT**

- broadening of the constraint.
- Some coefficients have **broader bounds** than previously obtained, in particular some of the four-fermion operators.
- However, bounds are very weak here anyway, and likely challenge EFT validity.

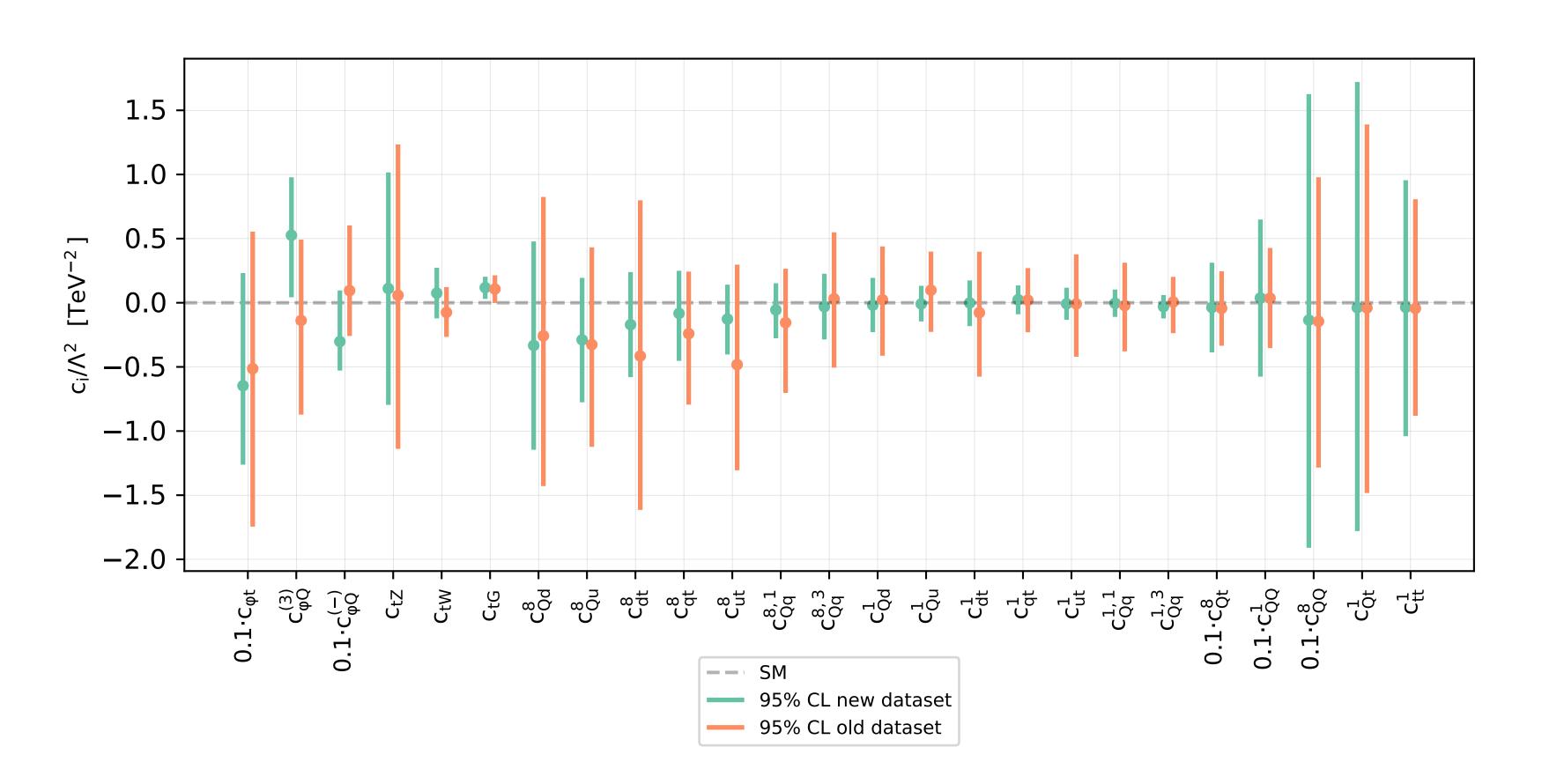




• Some other coefficients undergo a **shift in the central value**, but no tightening or

### **SMEFT-only fits: quadratic SMEFT**

- Results are much more promising when quadratic SMEFT effects are
- Only the five **four**heavy operators experience broadening relative to the old dataset. This could point to some inconsistency in the *ttttt* and *ttbb* data, but with such large uncertainties, it is difficult to be precise.



# included. A **significant tightening** of bounds is seen for most operators.

#### **PDF-SMEFT** correlation

• We can try to get intuition for the result of the joint PDF-SMEFT fit by considering the **PDF-SMEFT correlation** in the SMEFT-only fits.

#### **PDF-SMEFT** correlation

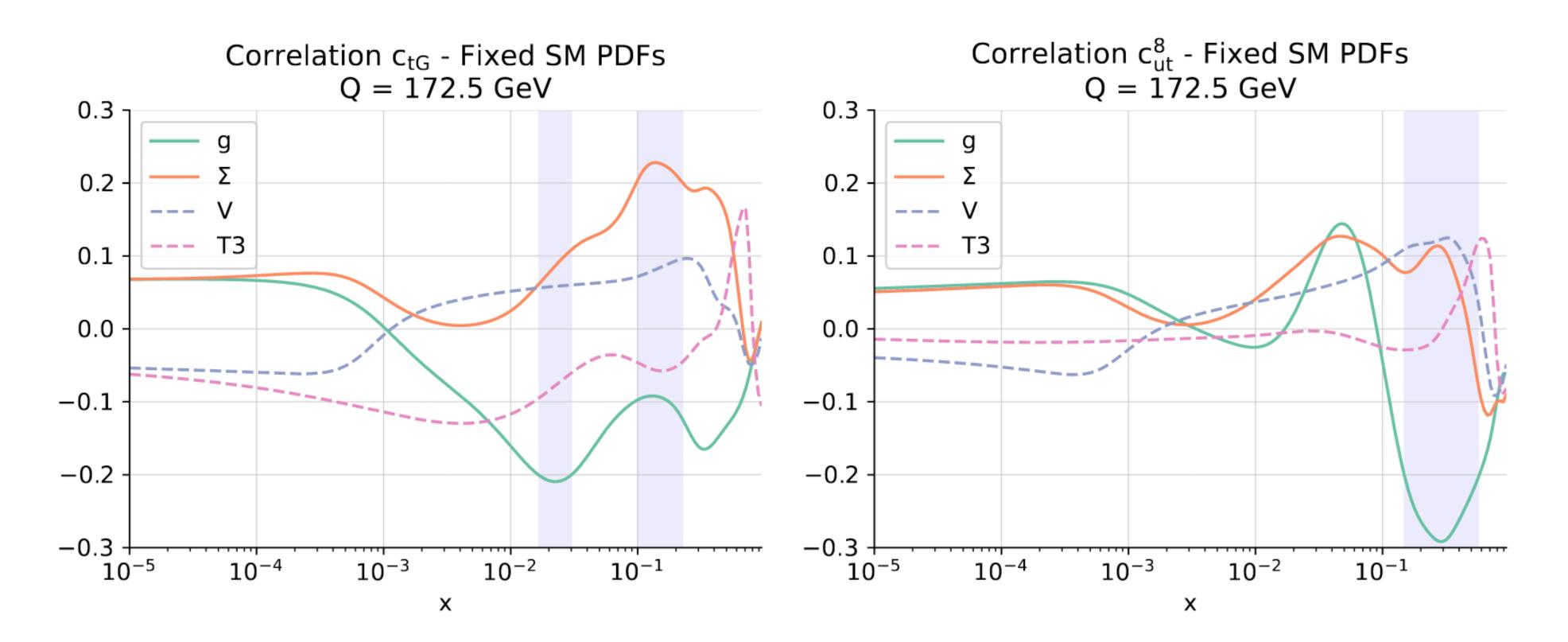
• We can try to get intuition for the result of the **joint PDF-SMEFT** fit by considering the **PDF-SMEFT correlation** in the SMEFT-only fits.

• This is defined for each Wilson coefficient and each PDF flavour by:

$$\rho\left(c, f(x, Q^2)\right) = \frac{\left\langle c^{(k)} f^{(k)}(x, Q^2) \right\rangle_k - \left\langle c^{(k)} \right\rangle_k \left\langle f^{(k)}(x, Q^2) \right\rangle_k}{\sqrt{\left\langle \left(c^{(k)}\right)^2 \right\rangle_k - \left\langle c^{(k)} \right\rangle_k^2} \sqrt{\left\langle \left(f^{(k)}(x, Q^2)\right)^2 \right\rangle_k - \left\langle f^{(k)}(x, Q^2) \right\rangle_k^2}}$$

#### **PDF-SMEFT** correlation

suggesting that the interplay will also be **relatively mild**.

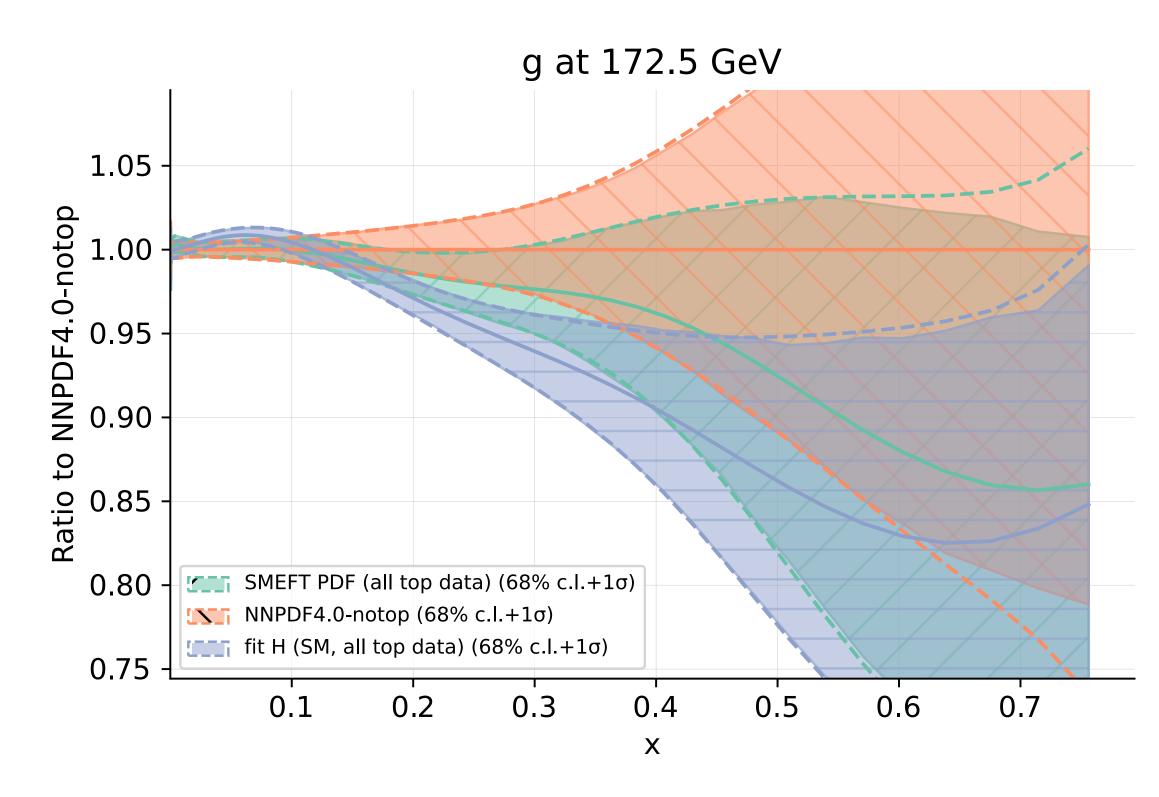


• We see the **strongest correlation** between the Wilson coefficients and the gluon PDF at high-x, as to be expected. The correlation is still **mild** though,

Now, let's do the joint fit...



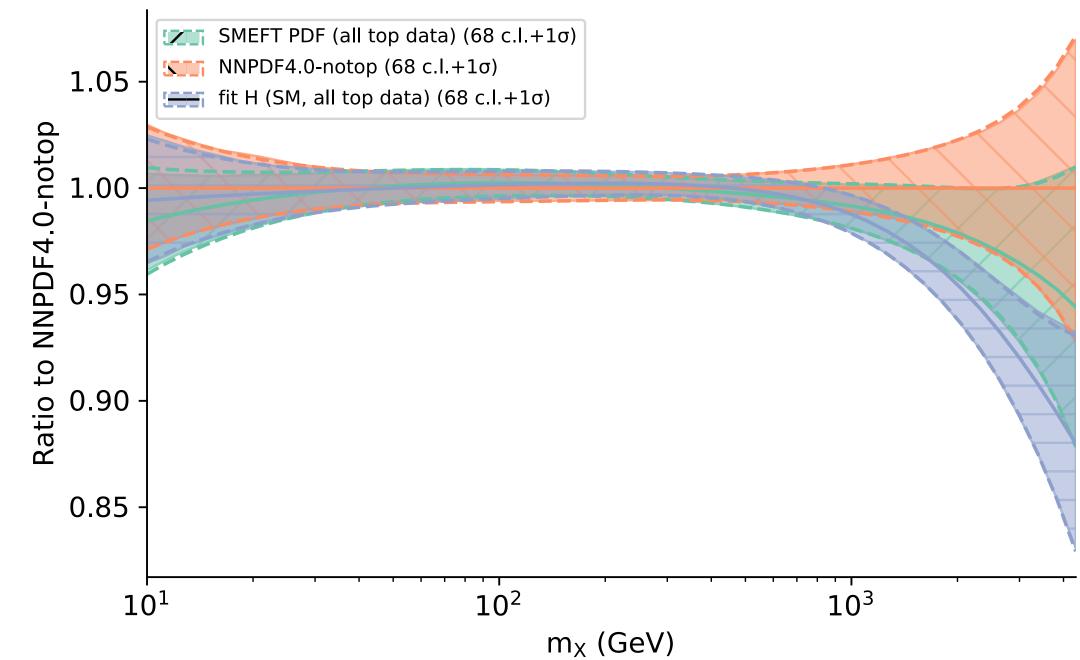
#### **Joint PDF-SMEFT fits: linear SMEFT**



• Finally, we present the key result of the work: a simultaneous determination of PDFs and SMEFT Wilson coefficients. We start assuming linear SMEFT.

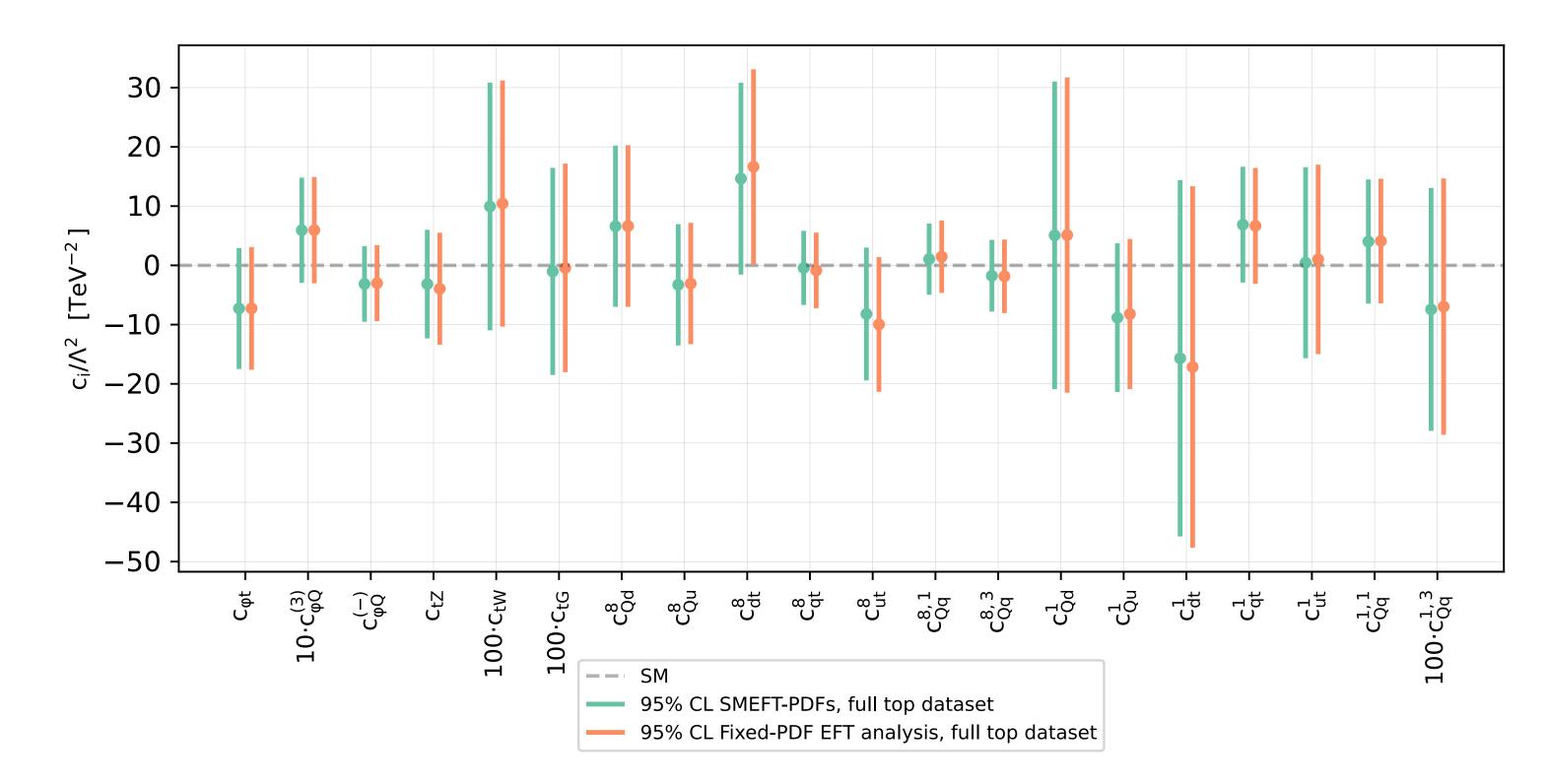
• In terms of the gluon PDFs and luminosities, we find that a simultaneous determination reduces the pull of the top data from the non-top baseline.

> gg luminosity  $\sqrt{s} = 13 \text{ TeV}$



#### **Joint PDF-SMEFT fits: linear SMEFT**

• On the other hand, we find that the bounds on the Wilson coefficients are **very stable** between a simultaneous PDF-SMEFT fit and a SMEFT-only fit.



effects are currently subdominant.

This indicates that within a linear EFT interpretation of the top data, the PDF

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• An upcoming publication will describe the issue in more detail; for now, here's the

$$t(c) = t^{SM} + t^{lin}c + t^{quad}c^2$$

ullet

$$c_p(d_p) = \operatorname{arg\,min}_c \left( \frac{(t(c) - d_p)^2}{\sigma^2} \right)$$

For simplicity, consider a single data point d with experimental variance  $\sigma^2$ , which we attempt to describe using the **quadratic** theory, involving a single theory parameter c:

The Monte-Carlo replica method propagates the uncertainty from the data to the theory parameter by fitting to **pseudodata**. We sample lots of pseudodata replicas from a normal distribution based on the data,  $d_p \sim N(d, \sigma^2)$ , and define the corresponding **parameter** 

**replicas** to be a random function of the pseudodata given by minimising the  $\chi^2$ -statistic:

parameter replicas analytically; it is given by:

$$P_{c^{(i)}}(c) \propto \delta\left(c + \frac{t^{\text{lin}}}{2t^{\text{quad}}}\right) \int_{-\infty}^{t_{\text{min}}} dx \ \exp\left(-\frac{1}{2\sigma^2}(x-d)^2\right) + \frac{2}{|2ct^{\text{quad}} + t^{\text{lin}}|} \exp\left(-\frac{1}{2\sigma^2}(d-t(c))^2\right)$$

Here, t<sub>min</sub> is the minimum value of the theory (which is a parabola).

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  - why...?

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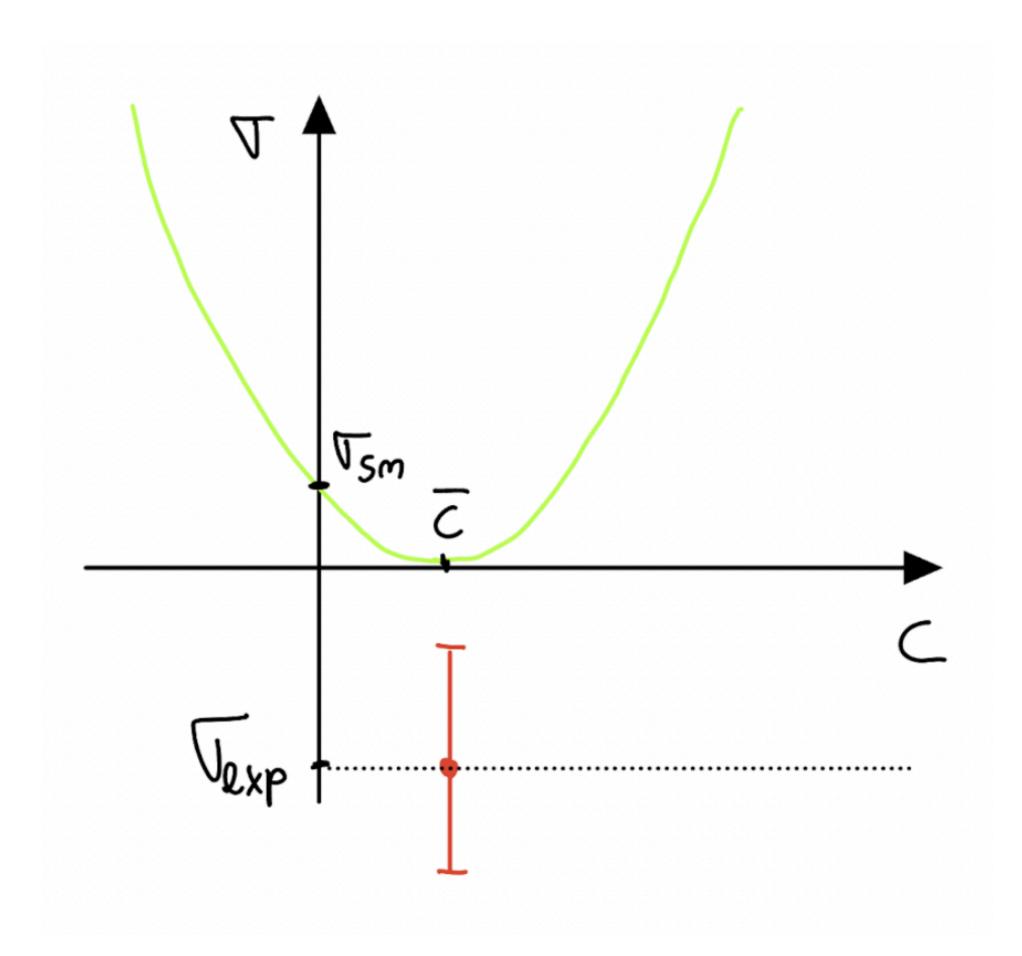
- Part of the distribution looks like a **scaled version** of what we would expect

- There is also a **delta function spike** in the distribution - interesting to ask:

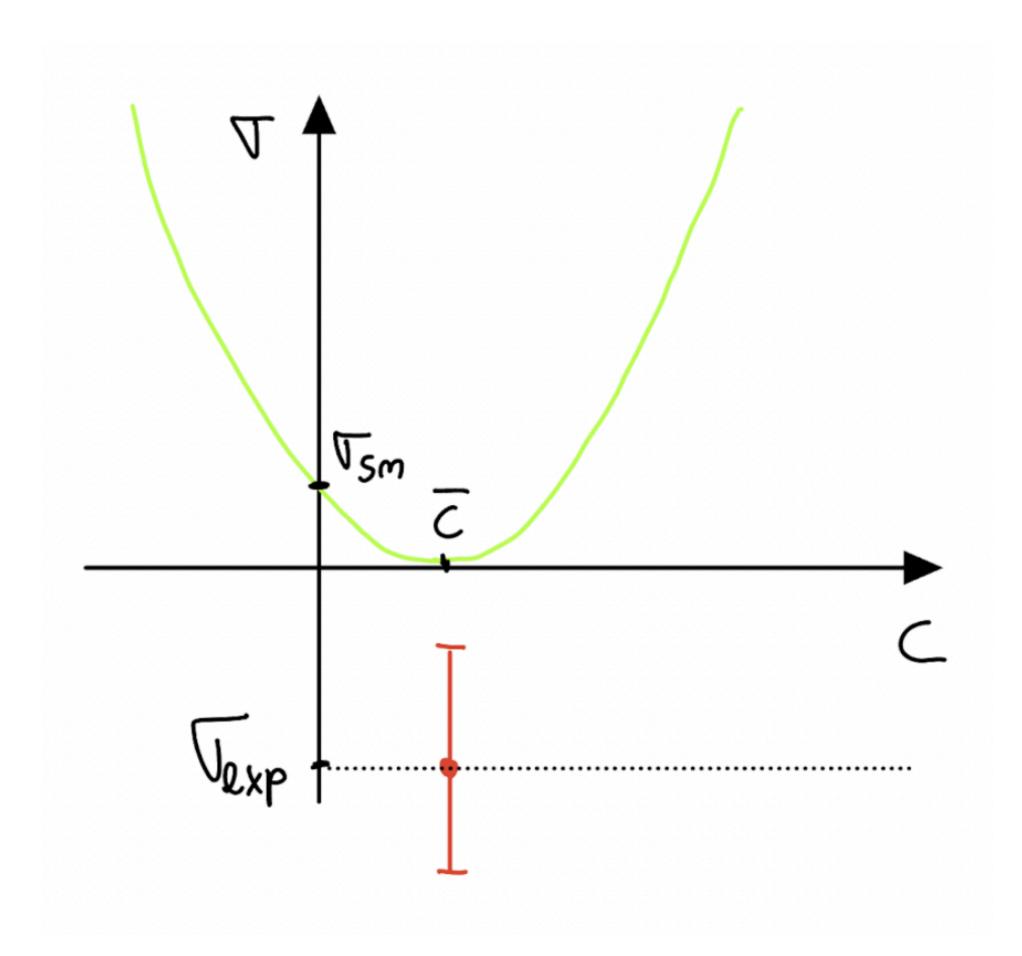


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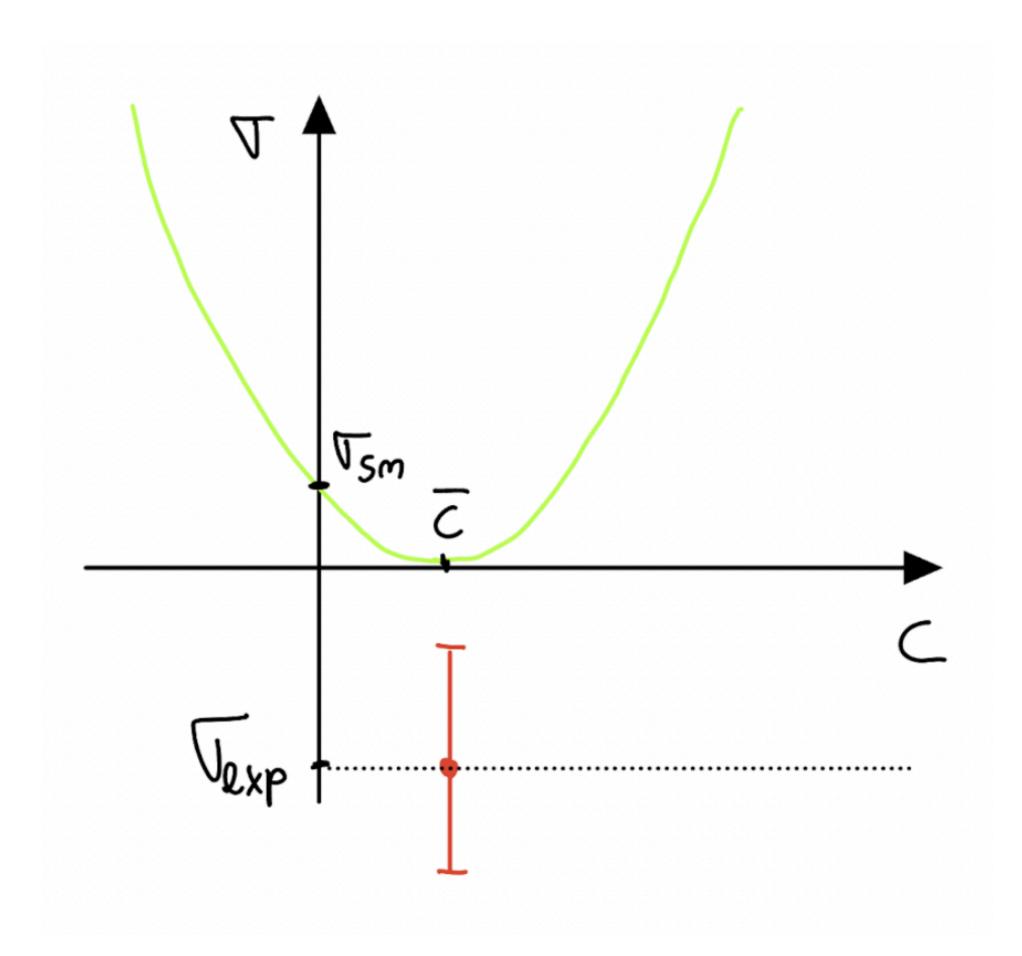


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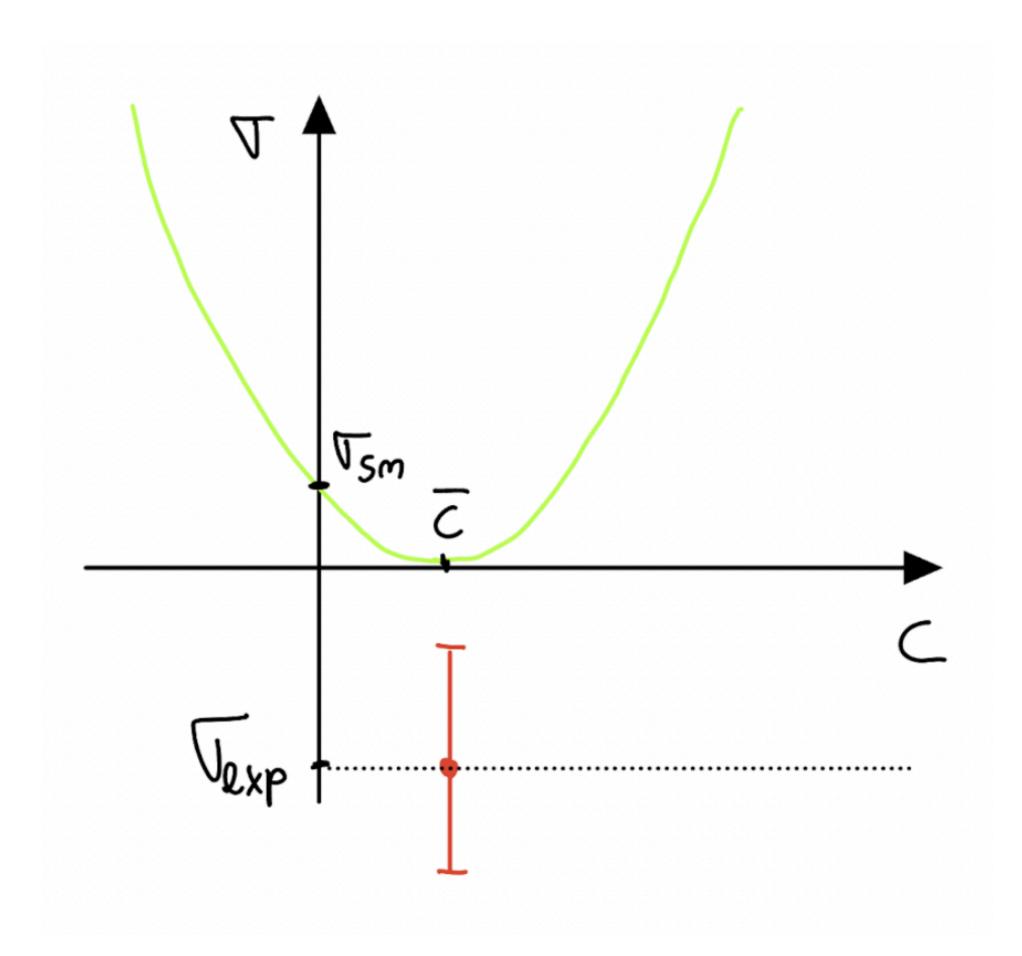
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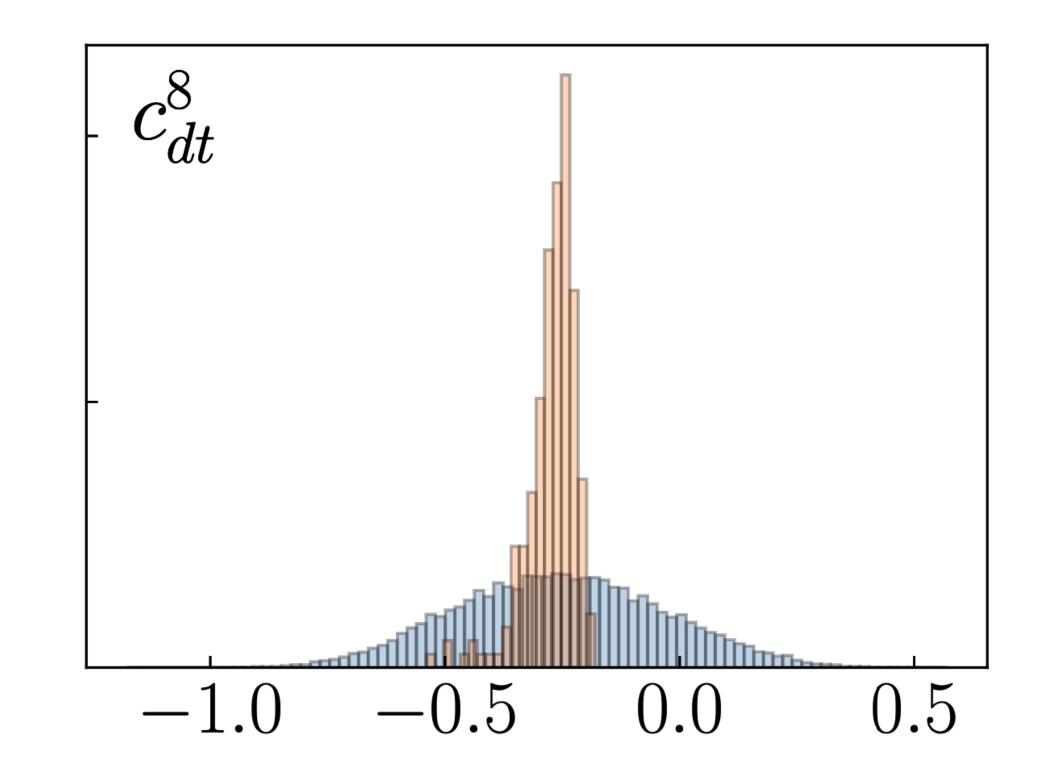
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- This occurs if the experimental data falls **below the minimum** of the theory, or **above but close** to the minimum.
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- This gives rise to the spike in the distribution at  $c = -t^{\text{lin}}/2t^{\text{quad}}$

- method (orange) and a Bayesian method with uniform prior (blue).
- We see that Monte-Carlo massively underestimates uncertainties.



• These problems extend to our top fit... for example in a realistic quadratic fit of one operator  $c_{dt}^8$ , we get the following comparison between the Monte-Carlo

#### Key questions for the future:

#### Can the MC replica method be modified to agree with Bayesian methods?

To what extent do existing fits (in the SMEFT world, PDF world, and beyond) that use the MC replica method underestimate uncertainties?





# Conclusions

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important in future analyses (especially as we enter Run III).

2104.02723 showing the need for simultaneous extraction; (ii) a

# Simultaneous determination of PDFs and BSM parameters, will be very

• Members of the **PBSP team** have already produced three works in the direction of simultaneous PDF-SMEFT fits: (i) a phenomenological study methodology (SimuNET, 2201.07240) capable of fast simultaneous fitting; (iii) a comprehensive simultaneous extraction of PDFs and SMEFT couplings from the full LHC Run II top dataset, 2303.06159.

# Thanks for listening! Questions?