

# COLLINEAR STRUCTURE LECTURE 2

## STEFANO FORTE UNIVERSITÀ DI MILANO & INFN



UNIVERSITÀ DEGLI STUDI DI MILANO DIPARTIMENTO DI FISICA



CFNS EIC school

Stony Brook, June 7 2024

## PDF DETERMINATION $DATA \rightarrow PARTON DISTRIBUTIONS$

#### Experimental data in NNPDF4.0



More than 4000 datapoints!

New processes:

- direct photon
- single top
- dijets
- W+jet
- DIS jet

**ISSUES AND TASKS:** 

- FROM PHYSICAL OBSERVABLES TO PDFS: SOLVE EVOLUTION EQUATIONS, CONVOLUTE WITH PARTON-LEVEL CROSS-SECTIONS
- DISENTANGLING PDFS: CHOOSE A BASIS OF PDFS ( $2N_f$  guarks + 1 gluon) & a set of SUITABLE PHYSICAL PROCESSES TO DETERMINE THEM ALL
- PROBABILITY IN THE SPACE OF FUNCTIONS: CHOOSE A STATISTICAL APPROACH • (MULTIGAUSSIAN, MONTE CARLO, ...)
- **UNCERTAINTY ON FUNCTIONS:** CHOOSE A REGRESSION MODEL

## DISENTANGLING PDFs

- DIS STRUCTURE FUNCTIONS CC  $F_1$  AND  $F_3 \Rightarrow$  FOUR COMBINATIONS, AND NC  $F_1$ TWO MORE  $\Rightarrow$  ALL LIGHT FLAVORS
- $W^{\pm}$  AND Z PRODUCTION (INCLUDING DOUBLE DIFFERENTIAL: MASS AND RAPIDITY)  $\Rightarrow$  INDEPENDENT COMBINATIONS
- TAGGED CHARM  $\Rightarrow$  HEAVY FLAVORS IN INITAL STATE
- DIS+DY  $\Rightarrow$  Gluon from scale dependence or higher orders  $p_T \Rightarrow$  Gluon at LO
- JETS  $\Rightarrow$  GLUON AT LO

#### FLAVOR SEPARATION (DIS & DY) LEADING ORDER PARTON CONTENT

#### DEEP-INELASTIC SCATTERING

	NC	$F_1^{\gamma} = \sum_i e_i^2 \left( q_i + \bar{q}_i \right)$	$\ell$	e	V	A
<b>A</b>	NC	$F_{1}^{Z, \text{ int.}} = \sum_{i} B_{i} (q_{i} + \bar{q}_{i})$	u,c,t	+2/3	$(+1/2 - 4/3\sin^2\theta_W)$	+1/2
	NC	$F_{3}^{Z, \text{ int.}} = \sum_{i} D_{i} (q_{i} + \bar{q}_{i})$	$_{\rm d,s,b}$	-1/3	$(-1/2 + 2/3\sin^2\theta_W)$	-1/2
	CC	$F_1^{W^+} = \bar{u} + d + s + \bar{c}$	ν	Ó	+1/2	+1/2
-	CC	$-F_3^{W^+}/2 = \bar{u} - d - s + \bar{c}$	$_{\mathrm{e},\mu,\tau}$	-1	$(-1/2 + 2\sin^2\theta_W)$	-1/2

 $B_q(Q^2) = -2e_q V_\ell V_q P_Z + (V_\ell^2 + A_\ell^2)(V_q^2 + A_q^2) P_Z^2; D_q(Q^2) = -2e_q A_\ell A_q P_Z + 4V_\ell A_\ell V_q A_q P_Z^2; P_Z = Q^2/(Q^2 + M_Z^2)$ 

$$W^+ \to W^- \Rightarrow u \leftrightarrow d, c \leftrightarrow s; p \to n \Rightarrow u \leftrightarrow d$$

#### DRELL-YAN



#### **ISOSPIN**

**PROTON VS.** NEUTRON  $u^p = d^n$ ;  $d^p = u^n$ ;  $\bar{u}^p = \bar{d}^n$ ;  $\bar{d}^p = \bar{u}^n$ ;  $s^p = s^n$ ;  $\bar{s}^p = \bar{s}^n$ ;  $\bar{c}^p = \bar{c}^n$ 

#### FIXED-TARGET DRELL-YAN (TEVATRON) QUARK ANTIQUARK SEPARATION

#### CHARGE CONJUGATION $\Rightarrow \bar{q}_{\bar{P}} = q_p$

DRELL-YAN p/d ASYMMETRY



#### COLLIDER DRELL-YAN (LHC)



CMS (2013)

## W, Z + TAGGED JET: HEAVIER FLAVORS W + c



- CHARM TAG  $\Rightarrow$  STRANGENESS
- $W^{\pm} \Rightarrow$  STRANGE-ANTISTRANGE SEPARA-TION

Z + c





- CHARM TAG  $\Rightarrow$  CHARM
- LARGE RAPIDITY  $Rightarrow \ x_1 \gg x_2$   $\Rightarrow$  ACCESS SMALL & LARGE x
- INTRINSIC CHARM!

LHCb (2022)

#### GLUON FROM SCALING VIOLATIONS THE GLUON HAS ONLY QCD INTERACTIONS!

SCALE DEPENDENCE OF SINGLET STRUCTURE FUNCTION



 $\Rightarrow$  GLUON AT SMALL *x* ONLY



CMS (2015)



 $v^* < 0.5$ 

CMS (2018)



10<sup>10</sup>

- ONE-JET/DIJET INCLUSIVE USED TO LARGE x GLUON
- WIDE KINEMATIC • REGION AT LHC



• WIDE RAPIDITY RANGE  $\Rightarrow$  MEDIUM AND LARGE x RE-GION

## PDF DETERMINATION SUMMARY

- DEEP-INELASTIC SCATTERING  $\Rightarrow$  CLEAN AND ABUNDANT INFORMATION ON PDFs:
  - HERA  $e^{\pm}p$  CC+NC data  $\Rightarrow$  four independent combinations, wide kinematic region  $\Rightarrow$  light quarks and antiquarks
  - FIXED-TARGET  $\mu p \& \mu d \Rightarrow$  DIRECT HANDLE ON UP-DOWN SEPARATION
  - HERA  $\Rightarrow$  SMALL x GLUON FROM SCALE DEPENDENCE
  - NEUTRINO (ALSO TAGGED c)  $\Rightarrow$  STRANGENESS
- Drell-Yan  $\gamma^*$  on fixed p and d target  $\Rightarrow$  UP-down separation at large x
- LHC W, Z HIGH AND LOW MASS
  - ANTIUP/ANTDOWN FROM W ASYMMETRY
  - FULL FLAVOR SEPARATION IN WIDE KINEMATIC REGION
  - STRANGENESS  $\Leftarrow$  TOTAL CROSS-SECTION AND TAGGED W+c FINAL STATE
  - CHARM  $\Leftarrow$  TAGGED Z + c FINAL STATE
  - GLUON  $\Leftarrow Z$  TRANSVERSE MOMENTUM DISTRIBUTION
- GLUON AT LHC:
  - TOP  $\Rightarrow$  MEDIUM x, FEW DATAPOINTS, HIGH ACCURACY

#### DATA UNCERTAINTIES: COVARIANCE MATRIX APPROACH

PREDICTIONS VS. DATA

$$\chi^{2} = \sum_{i,j}^{N_{\rm pt}} (T_{i} - D_{i}) (\operatorname{cov}^{-1})_{ij} (T_{j} - D_{j})$$

THE COVARIANCE MATRIX

$$\operatorname{cov}_{ij} = \delta_{ij} s_i^2 + \sum_{\alpha=1}^{N_c} \sigma_{i,\alpha}^{(c)} \sigma_{j,\alpha}^{(c)} + \left(\sum_{\alpha=1}^{N_{\mathcal{L}}} \sigma_{i,\alpha}^{(\mathcal{L})} \sigma_{j,\alpha}^{(\mathcal{L})}\right) D_i D_j$$

- $D_i$ : DATA;  $T_i$ : PREDICTION
- $s_i$ : UNCORRELATED STATISTICAL UNCERTAINTY FOR *i*-TH DATAPOINT
- $\sigma_{i,\alpha}^{(c)}$ :  $\alpha$ -th correlated additive systematics for *i*-th datapoint
- $\sigma_{i,\alpha}^{(\mathcal{L})}$ :  $\alpha$ -TH CORRELATED MULTIPLICATIVE SYSTEMATICS FOR *i*-TH DATAPOINT

#### DATA UNCERTAINTIES: NUISANCE PARAMETER APPROACH THE PARAMETERS

$$\chi^2(\{a\},\{\lambda\}) = \sum_{k=1}^{N_{\text{pt}}} \frac{1}{s_k^2} \left( D_k - T_k - \sum_{\alpha=1}^{N_\lambda} \beta_{k,\alpha} \lambda_\alpha \right)^2 + \sum_{\alpha=1}^{N_\lambda} \lambda_\alpha^2$$

 $\beta_{i,\alpha} = \sigma_{i,\alpha}^{(c)} \text{ for } \alpha = 1, \dots, N_c; \ \beta_{i,\alpha} = \sigma_{j,\alpha}^{(\mathcal{L})} D_i \text{ for } \alpha = N_c + 1, \dots, N_{\mathcal{L}}$ BEST-FIT VALUES

$$\lambda_{0\alpha} = \sum_{i=1}^{N_{\text{pt}}} \frac{D_i - T_i}{s_i} \sum_{\delta=1}^{N_{\lambda}} \mathcal{A}_{\alpha\delta}^{-1} \frac{\beta_{i,\delta}}{s_i}$$

#### REDUCED COVARIANCE MATRIX

$$\mathcal{A}_{lphaeta} = \delta_{lphaeta} + \sum_{k=1}^{N_{ ext{pt}}} rac{eta_{k,lpha}eta_{k,eta}}{s_k^2}$$

CONSTRUCTION OF THE COVARIANCE MATRIX: INVERSE

$$(\operatorname{cov})_{ij}^{-1} = \left[\frac{\delta_{ij}}{s_i^2} - \sum_{\alpha,\beta=1}^{N_{\lambda}} \frac{\beta_{i,\alpha}}{s_i^2} \mathcal{A}_{\alpha\beta}^{-1} \frac{\beta_{j,\beta}}{s_j^2}\right],\,$$

THE COVARIANCE MATRIX

$$(\operatorname{cov})_{ij} \equiv s_i^2 \delta_{ij} + \sum_{\alpha=1}^{N_{\lambda}} \beta_{i,\alpha} \beta_{j,\alpha}$$

## A LOOK AT THE EXPERIMENTAL COVARIANCE MATRIX



#### HESSIAN UNCERTAINTIES

• CHOOSE A FIXED FUNCTIONAL FORM  $f(x, Q^2; \vec{p}), p_i, i = 1, ..., N_{\text{par}}$  parameters

- SINCE 1973, PHYSICALLY MOTIVATED ANSATZ  $f_i(x, Q_0^2) = x^{\alpha}(1-x)^{\beta}g_i(x);$  $g_i(x)$  polynomial in x or  $\sqrt{x}$
- MMHT 2015:
  - \* BASIS FUNCTIONS  $g; u_v = u \bar{u}; d_v = d \bar{d}; S = 2(\bar{u} + \bar{d}) + s + \bar{s}; s_+ = s + \bar{s}; \Delta = \bar{d} \bar{u}; s_- = s \bar{s}.$
  - \* FOR ALL BUT  $\Delta s_{-}, g \Rightarrow x f_i(x, Q_0^2) = A x^{\alpha} (1-x)^{\beta} \left(1 + \sum_{i=1}^4 a_i T_i(y(x))\right);$   $T_i$  CHEBYSHEV POLYNOMIALS,  $y = 1 - 2\sqrt{x} \leftrightarrow$  MUST MAP x = [0, 1] INTO y = [-1, 1]; $T_i(-1) = T_i(1) = 1$
  - \* GLUON  $xg(x, Q_0^2) = Ax^{\alpha}(1-x)^{\beta} \left(1 + \sum_{i=1}^2 a_i T_i(y(x))\right) + A'xT\alpha'(1-x)^{\beta'}$
  - \* SEA ASYMMETRY  $x\Delta(x, Q_0^2) = Ax^{\alpha}(1-x)^{\beta}(1+\gamma x+\epsilon x^2)$
  - \* STRANGENESS ASYMMETRY  $x\Delta(x, Q_0^2) = Ax^{\alpha}(1-x)^{\beta}(1-x/x_0)$
  - \* 41 parameters, 4 fixed by sum rules
  - \* 12 parms fixed at best fit, remaining 25 used for covariance matrix  $\Rightarrow$  increased to 30 in MSHT 2019
- EVOLVE TO DESIRED SCALE & COMPUTE PHYSICAL OBSERVABLES
- MINIMUM OF  $\chi^2(\vec{p})$  BEST-FIT VALUES OF PARAMETERS  $p_i^{(0)}$
- COVARIANCE MATRIX IN PARM. SPACE  $\sigma_{ij} = \partial_i \partial_j \chi^2(\vec{p})$

## HESSIAN UNCERTAINTY PROPOAGATION ONE SIGMA PARM. RANGE $\Rightarrow \Delta \chi^2 = 1$ (Error propagation) "PARADOX"

- THE STANDARD DEVIATION OF  $\chi^2$  FOR  $N_{dat}$  DATA  $\sigma_{\chi^2} = \sqrt{2N_{dat}}$ HYPOTESIS-TESTING RANGE: COMPARE  $\Delta \chi^2 = \chi^2 - \langle \chi^2 \rangle$  TO  $\sigma_{\chi^2}^2$ . IF TOO LARGE, SOMETHING WRONG WITH THEORY (OR DATA)
- BUT THE ONE- $\sigma$  RANGE FOR A PARM. OF THE THEORY IS THE CURVE  $\chi^2 \chi^2_{min} = 1$ PARAMETER-FITTING RANGE: UNIT DEVIATION FROM THE PARAMETRIC MINIMUM  $\chi^2_{min}$
- CONSIDER DEVIATIONS  $\Delta_i$  FROM LINEAR FIT y = x + k; DETERMINE INTERCEPT k AS FREE PARAMETER
- IF STANDARD DEVIATION FOR EACH  $\Delta_i$  is  $\sigma_{\Delta}$ , Then average square deviation in units of  $\sigma_{\Delta}$  for  $N_{\rm dat}$  data:  $\sigma_{\chi^2} = N_{\rm dat}$
- BEST-FIT INTERCEPT:  $k = \langle \Delta_i \rangle$
- UNCERTAINTY ON IT:  $\sigma_k = \frac{\sigma_{\Delta}}{N_{\text{dat}}}$
- If  $\Delta k = \sigma_k$ , then  $\Delta \chi^2 = 1$



## TOLERANCE

- IN GLOBAL HESSIAN FITS, UNCERTAINTITES OBTAINED BY  $\Delta\chi^2 = 1$  UNREALISTICALLY SMALL
- UNCERTAINTIES TUNED TO DISTRIBUTION OF DEVIATIONS FROM BEST-FITS FOR INDIVIDUAL EXPERIMENTS



- (MSTW/MMHT) FOR EACH EIGENVECTOR IN PARAMETER SPACE DETERMINE CONFIDENCE LIMIT FOR THE DISTRIBUTION OF BEST-FITS OF EACH EXPERIMENT
- Rescale  $\Delta \chi^2 = T$  interval such that correct confidence intervals are reproduced

## MONTE CARLO UNCERTAINTIES

- DATA+UNCERTAINTIES  $\Rightarrow$  probability  $P(\vec{z})$  (multigaussian);  $z_i, i = 1, \dots, N_{dat}$
- MEAN  $\langle \vec{z} \rangle = \int d^d z X(\vec{z}) P(\vec{z})$ ; COVARIANCE  $\sigma_{ij} = \langle (z_i \langle z_i \rangle)(z_j \langle z_j \rangle) \rangle$
- GENERATE REPLICAS OF ORIGINAL DATA  $ec{z}^{(k)}$ ,  $k=1,\ldots,N_{\mathrm{rep}}$

Experimental data

• MEAN 
$$\langle \vec{z} \rangle = \frac{1}{M_{rep}} \sum_{1}^{N_{rep}} \vec{z}^{(k)}$$
  
replica averages  
vs. central values  
 $100$   
 $100$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $10$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$   
 $100$ 

10 REPLICAS ENOUGH FOR CENTRAL VALS, 100 FOR UNCERTAINTIES, 1000 FOR CORRELNS

10

## MONTE CARLO UNCERTAINTY PROPAGATION



- Determine best-fit PDF replca  $f^i(x,Q_0^)$  for each data replica  $\Rightarrow$  does not have to be min. Of  $\chi^2$
- MC REPRESENTATION OF PROBABILITY DISTRIBUTION IN PDF SPACE

## MONTE CARLO UNCERTAINTIES IMPORTANCE SAMPLING

- PROBABILITY DISTRIBUTION SAMPLED DIRECTLY  $\Rightarrow$  ALL INSTANCES EQUALLY WEIGHTED  $\langle f \rangle = \frac{1}{N} \sum_{I=1}^{N} f_i$
- CONTRAST TO A MODEL DEPENDING ON PARAMETERS  $\theta_i$  WITH KNOWN PROBABILITY  $p(\theta_i)$ :
- $\langle f \rangle = \frac{1}{N} \sum_{i=1}^{N} f(\theta_i) p(\theta_i) = \frac{1}{N} \sum_{i=1}^{N} f(\theta_i^p);$  $\theta_i^p$  sampled with probability  $p(\theta_i)$ 
  - IF  $p(\theta_i)$  SMALL FOR SOME  $\theta_i \Rightarrow$  INEFFICIENCY
  - REDEFINE  $\langle f \rangle = \frac{1}{N} \sum_{I=1}^{N} f(\theta_i) \frac{p(\theta_i)}{q(\theta_i)} q(\theta_i) = \frac{1}{N} \sum_{I=1}^{N} f(\theta_i) \frac{p(\theta_i^q)}{q(\theta_i^q)};$ 
    - $\theta_i^q$  sampled with probability  $q(\theta_i)$
  - OPTIMIZE CHOICE OF  $q( heta_i)$
- EQUAL WEIGHTING  $\Rightarrow$  OPTIMAL CHOICE

## WHY IT IS IMPORTANT

- Space of functions huge 5 bins for 10 pts  $\times$  7 fctns  $\rightarrow$   $5^{70} \sim 10^{49}$  bins
- BUT OBSERVABLES CORRELATED  $\Rightarrow$  DATA TELL US WHICH BINS ARE POPULATED

## $MC \Leftrightarrow HESSIAN$

- TO CONVERT HESSIAN INTO MONTECARLO GENERATE MULTIGAUSSIAN REPLICAS IN PARAMETER SPACE
- ACCURATE WHEN NUMBER OF REPLICAS SIMILAR TO THAT WHICH REPRODUCES DATA





- TO CONVERT MONTE CARLO INTO HESSIAN, SAMPLE REPLICAS  $f_i(x)$  AT A DISCRETE SET OF POINTS & CON-STRUCT THE ENSUING COVARIANCE MATRIX
- EIGENVECTORS OF THE COVARIANCE MATRIX  $\Rightarrow$  A BASIS IN VECTOR SPACE SPANNED BY REPLICAS BY SINGULAR-VALUE DECOMPOSITION
- NUMBER OF DOMINANT EIGENVECTORS  $\sim$  TO NUMBER OF REPLICAS  $\Rightarrow$  ACCURATE REPRESENTATION

## ARE UNCERTAINTIES GAUSSIAN?

- REPLICA HISTOGRAM *i*-TH DATAPOINT  $z_i$  FROM MC  $\Rightarrow$  CONTINUOUS DISTRIBUTION WITH KDE
  - POINT  $\Rightarrow$  KERNEL:  $P(z) = \frac{1}{N_{\text{rep}}} \sum_{k=1}^{N_{\text{rep}}} K(z-z_i);$
  - Gaussian kernel  $K(z z_i) \equiv \frac{1}{h\sqrt{2\pi}} \exp\left(-\frac{(z z_i)^2}{h}\right)$
  - Silverman bandwidth  $h = \sigma_i \left(\frac{4}{3N_{\text{rep}}}\right)^{\frac{1}{5}} \Rightarrow \text{MINIMIZES DIFFERENCE TO GAUSSIAN}$
- DEFINE KULLBACK-LEIBLER DIVERGENCE  $D_{\text{KL}} = \int_{-\infty}^{\infty} P(x) \ln \frac{P(x)}{Q(x)} dx$ BETWEEN A PRIOR P AND ITS REPRESENTATION Q
- COMPUTE  $D_{\rm KL}$  MC prior vs representation & MC prior vs gaussian
- **REPRESENTATIONS** SHOWN: MULTIGAUSSIAN OR MC COMPRESSION (OPTIMAL MC WITH SAME NUMBER OF REPLICAS)



 $D_{KL}$  to Gaussian small!:  $D_{KL} \sim$  percentage difference

#### CAN WE TRUST UNCERTAINTIES? CLOSURE TESTS

- ASSUME UNDERLYING "TRUTH" PDF (SAY A RANDOM PDF REPLICA)
- GENERATE DATA ACCORDING TO STATISTICAL AND CORRELATED SYSTEMATICS (SAY FOR NNPDF4.0 DATASET)
- DETERMINE PDFs & COMPARED TO "TRUTH" BASED ON INDICATORS

THE NATURE OF UNCERTAINTIES

- LEVEL 0:
  - EACH DATAPOINT EQUAL TO THE "TRUTH VALUE"; ZERO UNCERTAINTY
  - FIT  $\rightarrow$  MUST FIND  $\chi^2=0$  (GET BACK "TRUTH")
  - $\chi^2 pprox 0$  both replica to replica and average to truth
  - INTERPOLATION/EXTRAPOLATION UNCERTAINTY
- LEVEL 1:
  - EACH PSEUDO- DATAPOINT IS OBTAINED AS A RANDOM FLUCTUATION WITH GIVEN COVARIANCE MATRIX ABOUT "TRUTH"
    - $\Rightarrow$  "RUN OF THE UNIVERSE"
  - FIT DATA OVER AND OVER AGAIN
  - $\chi^2 pprox 1$  both replica to replica and average to truth
  - FUNCTIONAL UNCERTAINTY
- LEVEL 2:
  - data as in level 1
  - GENERATE DATA REPLICAS OF THESE "DATA"
  - FIT PDF REPLICAS TO DATA REPLICAS
  - $~\chi^2 \approx 2$  replica to replica;  $\chi^2 \approx 1$  average to truth
  - DATA UNCERTAINTY

#### UNCERTAINTIES: TYPE AND SIZE CLOSURE TEST RESULTS (NNPDF4.0)

level  $0~\chi^2$  vs training

- LEVEL 0 (TRUTH DATA)  $\Rightarrow \chi^2 \approx 0$ UNCERTAINTY NONZERO  $\Rightarrow$  INTERPOLATE DISCRETE DATA
- LEVEL 1 (RUNS OF UNIVERSE) ⇒ REPLICAS ALL FITTED TO SAME DATA, UNCERTAINTY NONZERO
   ⇒ DEGENERACY OF BEST-FITS (FUNCTIONAL FORMS)
- Level 0, 1 and 2 uncertainties comparable in size



#### LEVEL 0/1/2 UNCERTAINTIES





GLUON

## "PDF" UNCERTAINTIES SUMMARY

- DATA UNCERTAINTIES  $\Rightarrow$  MULTIGAUSSIAN
- "PDF" UNCERTAINTIES
  - DATA UNCERTAINTY PROPAGATION + MODEL
  - HESSIAN
    - \* ABSOLUTE MINIMUM OF  $\chi^2$  IN PARAMETER SPACE
    - \* MULTIGAUSSIAN
  - MONTECARLO
    - \* IMPORTANCE SAMPLING IN PDF SPACE
    - \* CAN TEST FOR GAUSSIANITY
  - INTERPOLATION, MODEL, DATA  $\Rightarrow$  COMPARABLE SIZE
  - GENERALLY GAUSSIAN

## MISSING HIGHER ORDER (THEORY) UNCERTAINTIES

• MAXIMIZE LIKELIHOOD

$$P = N \exp - \left(\frac{d-t}{2\sigma_{exp}^2}\right)$$

 $d,\,t$  are really vectors and  $1/\sigma^2$  the inverse covariance matrix

• PROBABILITY OF THEORY t GIVEN DATA d; BAYES  $\Rightarrow$ 

 $P(t|d) \propto P(d|t)P(t)$ 

- THEORY KNOWN EXACTLY  $\Rightarrow P(t) = \delta(t t^{\text{exact}})$
- THEORY KNOWN PERTURBATIVELY:  $t_p \Rightarrow t^{\text{exact}} = t_p + \Delta_p$ ;  $\Delta_p \Leftrightarrow \text{MHO}$
- $\Delta$  GAUSSIAN WITH UNCERTAINTY  $\sigma_{\rm th}$ ; INTEGRATE OUT

$$P = N \exp\left[\frac{d - t_p}{2\left(\sigma_{exp}^2 + \sigma_{th}^2\right)}\right]$$

• MHOU + EXP COMBINE IN QUADRATURE

## MISSING HIGHER ORDER (THEORY) UNCERTAINTIES

• FACTORIZED OBSERVABLE (NONSINGLET STRUCTURE FUNCTION):

$$F_2^{\rm NS}(N,Q^2) = xC_{\rm NS}(\alpha_s(Q^2),N) \exp\left[\int_{Q_0^2}^{Q^2} \frac{d\lambda^2}{\lambda^2} \gamma_{\rm NS}\left(\alpha_s(\mu^2),N\right)\right] f^{\rm NS}(Q_0^2)$$

• SOURCES OF MHOU

$$- \gamma_{\rm NS}^{N^k LO}(\alpha_s, N) = \alpha_s \gamma_{\rm NS}^{(0)}(N) + \alpha_s^2 \gamma_{\rm NS}^{(1)}(N) + \alpha_s^{k+1} \cdots + \gamma_{\rm NS}^{(k)}(N) \\ - C_{\rm NS}^{N^k LO}(\alpha_s(Q^2), N) = 1 + \alpha_s C_{\rm NS}^{(1)}(N) + \cdots + \alpha_s^{k+1} C_{\rm NS}^{(k)}(N)$$

#### SCALE VARIATION

• BASIC IDEA:  $\alpha_s(\kappa^2\mu^2) = \alpha_s(\mu^2)[1 + O(\alpha_s)]$ ; at N<sup>k</sup>LO difference  $\Leftrightarrow \beta$ -fctn up to  $\beta_k$ 

$$- \bar{C}(\alpha_s(\kappa_r^2 Q^2, \kappa_r^2) = C\alpha_s(Q^2)[1 + O(\alpha_s)] \Rightarrow \text{FIXES } \bar{C}^{(k)} \text{ IN TERMS OF } C^{(k)})$$
$$- \bar{\gamma}(\alpha_s(\kappa_f^2 Q^2, \kappa_f^2) = \gamma \alpha_s(Q^2)[1 + O(\alpha_s)] \Rightarrow \text{FIXES } \bar{\gamma}^{(k)} \text{ IN TERMS OF } \gamma^{(k)})$$

- $\Delta C = \bar{\gamma}(\alpha_s(\kappa_r^2 Q^2, \kappa_r^2) \gamma(\alpha_s(Q^2) \text{ RENORMALIZATION SCALE } \mu_r = \kappa_r Q \text{ VARN SCALE AT WHICH UV DIVS ARE SUBTRACTED}$
- $\Delta \gamma = \bar{\gamma}(\alpha_s(\kappa_f^2 Q^2, \kappa_f^2) \gamma(\alpha_s(Q^2) \text{ factorization scale } \mu_f = \kappa_f Q \text{ VARN} \text{ scale at which collinear divs are factorized}$ 
  - Change in  $\gamma \Rightarrow$  change in PDF  $f(Q^2) \Rightarrow$  can include  $\Delta \gamma$  as  $\Delta f$
  - Fixed F factorized as  $C\otimes f$   $\Rightarrow$  can include  $\Delta f$  as  $\Delta C$

## MHOU PRESCRIPTIONS

prediction for datapoint *i*, scale choice  $\mu_r^{(k)}, \mu_f^{(k)}$ , default  $\mu_r^0, \mu_f^0$ ;  $\Delta^k(\sigma_i) = \sigma_i[\{\mu^{(k)}\}] - \sigma_i[\{\mu_0\}]$ 

- VARY  $\mu_r$ ,  $\mu_f$  ABOUT  $\mu_0$
- PICK A SET OF POSSIBLE VARIATIONS
  - 3pt  $\mu_r=\mu_f$ ,  $\kappa=2,\,1/2$
  - 9 pt  $\mu_r$ ,  $\mu_f$  varied indep.  $\kappa=2,\,1/2$
  - -~7 PT  $\mu_r,\,\mu_f$  varied indep.  $\kappa=2,\,1/2,$  avoid  $mu_r/\mu_f=4$
- ENVELOPE: TAKE LARGEST AND SMALLEST  $\sigma$  AS UNCERTAINTY BAND
- THEORY COVARIANCE MATRIX:

$$\sigma_{i,j} = \frac{1}{N} \sum_{k} \Delta^{k}(\sigma_{i}) \Delta^{k}(\sigma_{j})$$

- SINGLE PROCESS: k runs over common set of scale choices
- MANY PROCESSES:
  - \* UNCORRELATED RENORMALIZATION: DIFFERENT FOR DIFFERENT HARD PROCESSES
  - \* CORRELATED FACTORIZATION: MHOU OF PERTURBATIVE EVOLUTION UNIVERSAL

## A LOOK AT THE THEORY COVARIANCE MATRIX



#### HEAVY QUARKS: DECOUPLING

- DECOUPLING SCHEME  $\Rightarrow$  HEAVY FLAVOR GRAPHS SUBTRACTED AT ZERO MOMENTUM (Collins, Wilczek, Zee, 1978)
- $N_f = 3$  active flavors in  $\beta$  function & evolution equations
- DECOUPLING VS  $\overline{\mathrm{MS}} \Leftrightarrow$  DIFFERENT RENORMALIZATION & FACTORIZATION SCHEMES

EXAMPLE: PHOTON SELF-ENERGY

$$\Pi^{R}(q^{2}) = \frac{2\alpha}{\pi} \int_{0}^{1} dx x(1-x) \ln \frac{m^{2} - x(1-x)q^{2}}{\mu_{r}^{2}}$$

• 
$$\overline{\text{MS}} \ln \frac{m^2 - x(1-x)q^2}{\mu_r^2} = \ln \frac{q^2}{\mu_r^2} + O\left(\frac{m^2}{q^2}\right) \Rightarrow \text{RUNNING } \alpha$$

• DECOUPLING:  $\ln \frac{m^2 - x(1-x)q^2}{\mu_r^2} = O\left(\frac{q^2}{m^2}\right)$ 

SOLID  $\Rightarrow$  HEAVY; DASHED  $\Rightarrow$  LIGHT

M. Buza et al.: Charm

#### MATCHING

- PDFS,  $\alpha_s$  IN  $N_f = 3$  &  $N_f = 4$ RELATED BY MATCHING CONDITIONS
- DETERMINED BY COMPUTING OPERATOR MATRIX ELEMENTS IN EITHER SCHEME AND EQUATING



**Fig. 2.**  $O(\alpha_s^2)$  contributions to the purely-singlet OME  $A_{q'q}^{PS}$ . Here q and q' are represented by the *dashed* and *solid lines* vertices q' = H these graphs contribute to the heavy-quark OME  $A_{Hq}^{PS}$ 

## HEAVY QUARKS IN DIS

- $\overline{\text{MS}}$  Scheme  $\Rightarrow$  HQ massless parton  $\Rightarrow \ln Q^2/m_h^2$  resummed to all orders by evolution eqns,  $O(m^2/Q^2)$  contributions neglected
- DECOUPLING SCHEME  $\Rightarrow$  HQ in hard xsect  $\Rightarrow$   $O(m^2/Q^2)$  contributions included,  $\ln Q^2/m_h^2$  treated at fixed order

THE BEST OF TWO WORLDS MATCHED SCHEMES ACOT



#### FONLL

COMBINE  $N^i LL$  massless resummed &  $N^j LO$  massive fixed-order  $\Rightarrow$  EXPAND RESUMMED RESULT; REPLACE THE FIRST j orders with their massive COUNTERPARTS

$$F(x,Q^{2}) = F^{(3)}(x,Q^{2}) + F^{(4)}(x,Q^{2}) - F^{\text{overlap}}(x,Q^{2})$$

$$F^{(3)}(x,Q^2) = x \int_x^1 \frac{dy}{y} \sum_{i=g,q,\bar{q}} C_i^{(3)} \left(\frac{x}{y}, \frac{Q^2}{m_h^2}, \alpha_s^{(3)}(Q^2)\right) f_i^{(3)}(y,Q^2)$$
  
$$F^{(4)}(x,Q^2) = x \int_x^1 \frac{dy}{y} \sum_{i=g,q,\bar{q},h,\bar{h}} C_i^{(4)} \left(\frac{x}{y}, \alpha_s^{(4)}(Q^2)\right) f_i^{(4)}(y,Q^2)$$

#### **ADVANTAGES**

- RELIES ON STANDARD FACTORIZATION & DECOUPLING
- THE RESUMMED AND UNRESUMMED ORDERS CAN BE CHOSEN FREELY & INDEPENDENTLY

#### COMPLICATION

• RESUMMED & FIXED-ORDER CALCULATION ARE PERFORMED IN DIFFERENT RENORMALIZATION & FACTORIZATION SCHEMES: 3F (MASSIVE, DECOUPLING) VS. 4F (MASSLESS)

#### SOLUTIONS

- EITHER RE-EXPRESS 3F-SCHEME PDFs &  $\alpha_s$  in terms of the 4F-scheme ones
- OR HAVE SIMULTANEOUSLY 3F & 4F SCHEME  $\alpha_s$  & PDFs

### THE CHARM PDF PERTURBATIVE CHARM

- IN  $N_f = 3$  scheme charm PDF vanishes
- IN  $N_f = 4$  scheme, charm determined by perturbative matching
- STARTING AT NNLO (TWO LOOPS) DOES NOT VANISH AT ANY SCALE



## **INTRINSIC CHARM**

• **DEFINE** CHARM PDF AS OME:

 $\langle p|\bar{c}\gamma^{\mu_1}D^{\mu_2}\dots D^{\mu_n}c|p\rangle = A_c^n p^{\mu_1}\dots p^{\mu_n} - \text{traces}$ 

$$A_c^n = \int_0^1 dx \, x^{n-1} c(x)$$

- Decouple charm mass logs  $\Rightarrow$  choose  $N_f = 3$  scheme
- ALLOW NONVANISHING (SCALE-INDEPENDENT) CHARM PDF
- IN  $N_f = 4$  scheme charm PDF differs from that fixed by matching

NONVANISHING CHARM IN THE  $N_F=3$  (decoupling) scheme  $\Rightarrow$  intrinsic charm

## THEORY UNCERTAINTIES SUMMARY

- THEORY UNCERTAINTIES  $\Rightarrow$  THEORY COVARIANCE MATRIX
- SCALE VARIATION:
  - RENORMALIZATION  $\Rightarrow$  MHOU IN PARTONIC CROSS-SECTION
  - FACTORIZATION  $\Rightarrow$  MHOU IN ANOMALOUS DIMENSION
- HEAVY QUARKS
  - **DECOUPLING** SCHEME  $\Rightarrow$  QUARK MASS EFFECTS INCLUDED
  - $\overline{\mathrm{MS}}$  scheme  $\Rightarrow$  collinear mass logs resummed
  - MATCHING  $\Rightarrow$  BOTH INCLUDED
  - HQ PDF differs from result of matching  $\Rightarrow$  intrinsic HQ

## FOOD FOR THOUGHT

- CAN YOU THINK OF NEW PROCESSES AT EIC FOR PDF DETERMINATION? AND CAN YOU THINK OF A SYNERGY BETWEEN EIC & LHC?
- WHAT MIGHT BE THE REASON WHY TOLERANCE IS NEEDED? AND CAN YOU THINK HOW TO TEST IT?
- CAN YOU THINK OF ALTERNATIVE WAYS OF ESTIMATING MHOUS?
- IF INTRINSIC HQ PDFS ARE NONZERO, HOW DO YOU EXPECT THEIR SIZE TO SCALE WITH THE HQ MASS?