



REGRESSION NETWORKS: PRECISION AND UNCERTAINTY ESTIMATION

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WITH TUTORIALS BY

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MACHINE LEARNING IN PARTICLE THEORY

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LECTURE PLAN

- I: THE PHYSICS PROBLEM
 - PARTON DISTRIBUTIONS AND THEIR DETERMINATION
 - THE PROBLEM OF PDF UNCERTAINTIES
- II: REGRESSION & UNCERTAINTY
 - MONTE CARLO UNCERTAINTIES
 - NEURAL NETWORK REGRESSION
- III: PROPER LEARNING
 - OVERLEARNING AND CROSS-VALIDATION
 - HYPEROPTIMIZATION AND GENERALIZATION
- IV: VALIDATION & TESTING
 - TESTING: DATA REGION AND EXTRAPOLATION REGION
 - THE MEANING OF CORRELATIONS
- V: UNDERSTANDING RESULTS
 - DISTRIBUTION OF RESULTS AND FAITHFULNESS
 - OUTLIERS AND GENERALIZATION

SOME GENERAL REFERENCES

REGRESSION AND STATISTICS

- C. M. Bishop, "Pattern recognition and machine learning", Springer, 2006-2009
- G. James, D. Witten, T. Hastie, R. Tibishrani, "An introduction to statistical learning", Springer, 2021

GENERAL MACHINE LEARNING

- I. Goodfellow, Y. Bengio, A. Courville, "Deep learning", MIT press, 2016
- A. Géron, "Hands-on machine learning with Scikit-Learn & TensorFlow", O'Reilly, 2017

PARTON DISTRIBUTIONS

- S. Forte, "Parton Distributions at the dawn of the LHC", Acta Phys. Pol. **B41** (2010) 2859, arXiv:1011.5247
- S. Forte and S. Carrazza, "Parton distribution functions", in "Artificial intelligence for High Energy Physics", P. Calafiura, D. Rousseau and K. Terao, eds., World Scientific 2022, pag 715, arXiv:2008.12305
- L. Del Debbio, T. Giani, T. Wilson, "Bayesian approach to inverse problems: an application to NNPDF closure testing", Eur. Phys. J. **C82** (2022) 33, arXiv 2111.057887

THE NNPDF CODE https://nnpdf.mi.infn.it/nnpdf-open-source-code/

I: THE PHYSICS PROBLEM

- FACTORIZATION AND PDFs
 - FACTORIZATION: LEPTONIC AND HADRONIC PROCESSES
 - PDFs and their properties
- PDF DETERMINATION
 - FROM DATA TO PDFS
 - PHYSICAL PROCESSES
- UNCERTAINTIES
 - EXPERIMENTAL UNCERTAINTIES AND THEIR CORRELATIONS
 - MISSING HIGHER ORDERS AND THEORY UNCERTAINTIES
- THE PROBLEM OF MODEL UNCERTAINTIES
 - THE PROBLEM OF PDF UNCERTAINTIES
 - PLOTNOMIAL REGRESSION AND ITS PITFALLS



- (R. Roentsch, Les Houches, June 2023)
- PDF ESPRESS THE "PROBABILITY" OF QUARKS OR GLUONS (PARTONS) TO ENTER A COLLISION
- THEIR KNOWLEDGE IS **REQUIRED FOR THE COMPUTATION** OF ANY PROCESS AT THE LHC
- THEIR KNOWLEDGE IS A DOMINANT SOURCE OF UNCERTAINTY

FACTORIZATION

FACTORIZATION IN DEEP-INELASTIC SCATTERING

STRUCTURE FUNCTIONS



$\lambda_l \rightarrow \text{lepton helicity}$	UNPOL.	F_1, F_2	F_3
$\lambda_p \rightarrow \text{proton helicity}$	POL.	g_1	g_4, g_5
1 -			

DEEP-INELASTIC SCATTERING STRUCTURE FUNCTIONS AND PDFS

STRUCTURE FUNCTION=HARD COEFF. (PARTONIC STRUCTURE FUNCTION) \otimes PARTON DISTN. HARD COEFF.: XSECT FOR INCOMING PROTON MOMENTUM $p \Rightarrow$ INCOMING PARTON MOMENTUM $\hat{p} = xp$



 $F_2(x,Q^2) = x \sum_i \int_1^1 \frac{dy}{y} C_i\left(\alpha_s(Q^2), \frac{x}{y}\right) \left[q_i(y,Q^2) + \bar{q}_i(y,Q^2)\right] + C_g\left(\alpha_s(Q^2), \frac{x}{y}\right) g(y,Q^2)$

 q_i quark, \bar{q}_i antiquark, g gluon

• PARTON LUMINOSITY $\mathcal{L}_{ab}(\tau) = \int_{\tau}^{1} \frac{dx}{x} f_{a/h_1}(x) f_{b/h_2}(\tau/x)$

• COEFFICIENT FUNCTION
$$\hat{\sigma}_{q_a q_b \to X} \left(x_1 x_2 s, M_X^2 \right) = \sigma_0 C \left(\frac{M_X^2}{x_1 x_2 s}, \alpha_s(M_H^2) \right)$$

EXAMPLE: THE DRELL-YAN PROCESS AT LEADING ORDER

- Hadronic c.m. energy: $s = (p_1 + p_2)^2$
- Momentum fractions $x_{1,2} = \sqrt{\frac{\hat{s}}{s}} \exp \pm y;$ Lead. Ord. $\hat{s} = M^2$
- Partonic c.m. energy: $\hat{s} = x_1 x_2 s$
- Invariant mass of final state X (dilepton, Higgs,...): $M_W^2 \Rightarrow$ scale of process
- Scaling variable $\tau = \frac{M_X^2}{s}$



$$\Rightarrow M^2 \frac{d\sigma}{dM^2} = \sigma_0 \mathcal{L}(\tau); \quad \sigma_0 = \frac{4}{9} \pi \alpha \frac{1}{s};$$

FACTORIZATION SUMMARY & MELLIN TRANSFORM

- x-SPACE FACTORIZED EXPRESSIONS:
 - LEPTON-HADRON $F(x,Q^2) = \sum_i \int_x^1 \frac{dy}{y} C_i\left(\alpha(Q^2), \frac{x}{y}\right) f_i(y,Q^2) = \left[C_i(\alpha(Q^2)) \otimes f_i(Q^2)\right](x)$
 - HADRON-HADRON $\begin{aligned} &\sigma(x,Q^2) = \sum_{ij} \int_x^1 \frac{dy}{y} \hat{\sigma}_{ij} \left(\alpha(Q^2), \frac{x}{y} \right) \mathcal{L}_{ij} = (y,Q^2) \left[\hat{\sigma}_{ij}(\alpha(Q^2)) \otimes \mathcal{L}_{ij}(Q^2) \right] (x) \\ &\hat{\sigma}_{ij} \left(\alpha(Q^2), x \right) = \sigma_0 C_{ij} \left(\alpha(Q^2), x \right); \\ &\mathcal{L}_{ij}(Q^2)(x) = \int_x^1 \frac{dy}{y} f_i \left(\frac{x}{y} \right) f_j(y,Q^2) = \left[f_i \otimes f_j \right] (x) \end{aligned}$
- MELLIN TRANSFORM: $F(N) = \int_x^1 x^{N-1} f(x) \Leftrightarrow (x) = \int_{-i\infty}^{+i\infty} x^{-N} f(N)$ $h(x) = \int_x^1 \frac{dy}{y} f\left(\frac{x}{y}\right) g(y) = [f \otimes f](x) \Leftrightarrow H(N) = F(N)G(N)$
- *N*-SPACE FACTORIZED EXPRESSIONS:
 - Lepton-hadron $F(N,Q^2) = \sum_i C_i \left(\alpha(Q^2), N \right) f_i(N,Q^2)$
 - HADRON-HADRON $\sigma(x, N) = \sum_{ij} \hat{\sigma}_{ij} \left(\alpha(Q^2), N \right) \mathcal{L}_{ij}(N, Q^2)$ $\mathcal{L}_{ij}(Q^2, N) = f_i(N) f_j(N)$

LARGE/SMALL $x \Leftrightarrow$ LARGE/SMALL N

THE SCALE DEPENDENCE OF PDFS EVOLUTION EQUATIONS

$$\begin{aligned} \frac{d}{dt}q_{NS}(N,Q^2) &= \frac{\alpha_s(t)}{2\pi}\gamma_{qq}^{NS}(N,\alpha_s(t))q_{NS}(N,Q^2), \\ \frac{d}{dt} \left(\begin{array}{c} \Sigma(N,Q^2) \\ g(N,Q^2) \end{array} \right) &= \frac{\alpha_s(t)}{2\pi} \left(\begin{array}{c} \gamma_{qq}^S(N,\alpha_s(t)) & 2n_f\gamma_{qg}^S(N,\alpha_s(t)) \\ \gamma_{gq}^S(N,\alpha_s(t)) & \gamma_{gg}^S(N,\alpha_s(t)) \end{array} \right) \left(\begin{array}{c} \Sigma(N,Q^2) \\ g(N,Q^2) \end{array} \right). \end{aligned}$$

• LOG SCALE
$$t = \ln \frac{Q^2}{\Lambda^2}$$
:

- ANOMALOUS DIMENSIONS VS. SPLITTING FUNCTIONS $\gamma(N, \alpha_s(t)) \equiv \int_0^1 dx \, x^{N-1} P(x, \alpha_s(t))$
- SINGLET $\Sigma(x, Q^2) = \sum_{i=1}^{n_f} (q_i(x, Q^2) + \bar{q}_i(x, Q^2))$ VS. NONSINGLET $q^{NS}(x, Q^2) = q_i(x, Q^2) - q_j(x, Q^2)$ COMBINATIONS OF QUARK PDFS
- **PERTURBATIVE EXPANSION** OF ANOMALOUS DIMENSION $\gamma_i(N, \alpha_s(t)) = \gamma_i^{(0)}(N) + \alpha_s(t)\gamma_i^{(1)}(N) + \ldots \Rightarrow$ LOG RESUMMATION: LO \Leftrightarrow LLQ²; NLO \Leftrightarrow LLQ², ...



- AS Q^2 INCREASES, PDFS DECREASE AT LARGE x & INCREASE AT SMALL x DUE TO RADIATION
- Gluon sector singular at $N=1 \Rightarrow$ gluon grows more at small x
- $\gamma_{qq}(1) = 0 \Rightarrow$ number of quarks conserved

PARTON KINEMATICS vs. HADRON KINEMATICS $\sigma(\tau) = \int_{\tau}^{1} \frac{dy}{y} \sum_{ij} \mathcal{L}_{ij}(y) \hat{\sigma}_{ij}\left(\frac{\tau}{z}\right); \quad \mathcal{L}_{ij}(y) \equiv \int_{y}^{1} \frac{dx_{1}}{x_{1}} q_{i}(y) q_{j}\left(\frac{y}{x_{1}}\right)$

• WHICH PARTON MOMENTUM FRACTIONS CONTRIBUTE TO A GIVEN HADRONIC PROCESS ?

INVERSION OF MELLIN TRANSFORMS $f_n = \int_x^1 x^{n-1} f(x) \Leftrightarrow F(x) = \int_{-i\infty}^{+i\infty} x^{-n} f_n$ integrate to the right of convergence abscissa

- Mellin inversion dominated by saddle point
- POSITION OF SADDLE DEPENDS ON HADRONIC KINEMATICS, CONTROLLED BY PARTON LUMINOSITY DEPENDENCE ON x OF \mathcal{L} POWERLIKE, OF $\hat{\sigma}$ LOGARITHMIC
- PDF PEAKED AT SMALL x ("SEA" \bar{q} vs. "Valence" $q \bar{q}$) \Rightarrow LUMI PEAKS AT SMALL N



PDFs: QUALITATIVE FEATURES



- THE MOMENTUM PROBABILITY DENSITY $xf_i(x)$ IS SHOWN AT TWO DIFFERENT SCALES (LEFT \Rightarrow LOW SCALE; RIGHT \Rightarrow HIGH SCALE)
- As $x \ge 1$ kinematic constraint $f_i(x) = 0$
- "VALENCE" UP AND DOWN: PEAKED AT $x \sim 0.3$; EXPECT $f_x(x) \underset{x \to 1}{\sim} (1-x)_i^{\beta}$
- "SEA" ANTIQUARK AND GLUON GROW AT SMALL x
- "SINGLET" AND GLUON MIX \Rightarrow ALL PDFs look the same as $x \to 0$

PDF DETERMINATION

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

- CC F_1 and F_3 in principle provide four combinations, and NC F_1 two more \Rightarrow All light flavors
- W^{\pm} AND Z PRODUCTION (INCLUDING DOUBLE DIFFERENTIAL: MASS AND RAPIDITY) PROVIDE INDEPENDENT COMBINATIONS
- WHEN PRODUCING ELECTROWEAK FINAL STATES, THE GLUON CAN ONLY BE ACCESSED FROM SCALE DEPENDENCE OR HIGHER ORDERS (DIFFERENTIAL DISTRIBUTIONS) ...EXCEPT IN HIGGS PRODUCTION!
- JET PRODUCTION GIVES A DIRECT HANDLE ON THE GLUON

FLAVOR SEPARATION (DIS & DY) LEADING ORDER PARTON CONTENT

DEEP-INELASTIC SCATTERING



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

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

DRELL-YAN

	$L^{ij}(x_1,$	$(x_2) \equiv q_i(x_1, M^2) \bar{q}_j(x_2, M^2)$
	γ	$\frac{d\sigma}{dM^2 dy}(M^2, y) = \frac{4\pi\alpha^2}{9M^2 s} \sum_i e_i^2 L^{ii}(x_1, x_2)$
June 1	W	$\frac{d\sigma}{dy} = \frac{\pi G_F M_V^2 \sqrt{2}}{3s} \sum_{i,j} V_{ij}^{\text{CKM}} L^{ij}(x_1, x_2)$
	Z	$\frac{d\sigma}{dy} = \frac{\pi G_F M_V^2 \sqrt{2}}{3s} \sum_i \left(V_i^2 + A_i^2 \right) L^{ij}(x_1, x_2)$

 $V_{ij}^{\text{CKM}} \rightarrow \text{CKM}$ matrix (i = u, ct, j = d, sb), $V_{ij}^{\text{CKM}} = 1 + O(\lambda)$; $\lambda = \sin \theta_C \approx 0.22$

FIXED-TARGET DRELL-YAN (TEVATRON) QUARKS AND ANTIQUARK SEPARATION

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

DRELL-YAN p/d ASYMMETRY





CMS (2013)

THE GLUON FROM DIS

SCALE DEPENDENCE OF FLAVOR SINGLET STRUCTURE FUNCTIONS



LARGE x GLUON DIFFICULT TO DETERMINE FROM DEEP-INELASTIC SCATTERING

THE GLUON IN HADRONIC COLLISIONS THE GLUON ONLY INTERACTS THROUGH QCD JETS GLUON



CMS (2018)







• WIDE KINEMATIC REGION AT LHC



UNCERTAINTIES

DATA UNCERTAINTIES: COVARIANCE MATRIX

PREDICTIONS VS. DATA: LOSS

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

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
- $\sigma_{i,\alpha}^{(c)}$: α -TH CORRELATED ADDITIVE SYSTEMATICS FOR *i*-TH DATAPOINT
- $\sigma_{i,\alpha}^{(\mathcal{L})}$: α -th correlated multiplicative systematics for *i*-th datapoint

DATA UNCERTAINTIES: NUISANCE PARAMETERS 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

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

MULTIPLICATIVE UNCERTAINTIES THE D'AGOSTINI BIAS

- NORMALIZATION UNCERTAINTIES IN COVARIANCE MATRIX $(cov)_{ij} = \sigma_{i,n}\sigma_{j,n}D_iD_j$ \Rightarrow MAXIMUM-LIKELIHOOD RESULT BIASED (d'Agostini, 1994)
- EQUIVALENT TO RESCALING DATA BUT NOT UNCERTAINTIES



(CELLO collab., 1987)

• MUST COMPUTE UNCERTAINTY FROM PREVIOUS THEORY PREDICTION RESULT OF PREVIOUS FIT: $(\text{cov})_{ij} = \sigma_{i,n}\sigma_{j,n}T_i^{(0)}T_j^{(0)}$



MISSING HIGHER ORDER (THEORY) UNCERTAINTIES

• PDFs are determined by maximizing the 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 THE THEORY t being correct given data d, which by Bayes is

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

- IF THEORY WAS KNOWN EXACTLY, THEN $P(t) = \delta(t t^{\text{exact}})$
- IN ACTUAL FACT ONLY SOME PERTURBATIVE RESULT t_p is exactly known so $t^{\text{exact}} = t_p + \Delta_p$, where Δ_p includes MHO
- Assuming Δ to be gaussianly distributed, with uncertainty $\sigma_{\rm th}$ and integrating out

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

- THEORETICAL UNCERTAINTY ADDED IN QUADRATURE, PROPAGATES INTO PDF UNCERTAINTY UPON MINIMZATION
- CAN COMPUTE PDF UNCERTAINTY GIVEN MHOU

THEORY COVARIANCE MATRICES

THEORY COVARIANCE MATRIX: processes i, j, scale choice $\{\mu^{(k)}\}$, default $\{\mu_0\}$

$$\sigma_{i,j} = \frac{1}{N} \sum_{k} \left(\sigma_i[\{\mu^{(k)}\}] - \sigma_i[\{\mu_0\}] \right) \left(\sigma_j[\{\mu^{(k)}\}] - \sigma_j[\{\mu_0\}] \right)$$

- 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



MODEL-DEPENDENT REGRESSION

A MODEL-DEPENDENT (HESSIAN) APPROACH

- CHOOSE A FIXED FUNCTIONAL FORM
 - 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
- DETERMINE BEST-FIT VALUES OF PARAMETERS
- DETERMINE ERROR BY PROPAGATION OF ERROR ON PARMS. $\Delta \chi^2 = 1$ ('HESSIAN METHOD');

PARM. SCANS ALSO POSSIBLE ('LAGR. MULTIPLIER METHOD')

THE PROBLEM OF MODEL DEPENDENCE: A DISCOVERY THAT WASN'T

- DISCREPANCY BETWEEN QCD CALCULATION AND CDF JET DATA (1995)
- EVIDENCE FOR QUARK COMPOSITENESS?
- RESULT STRONGLY DEPENDS ON GLUON AT $x \gtrsim 0.1$
- PDF MUST VANISH AT x = 0, BUT (THEN) NO DATA FOR $x \ge 0.05!$



DISCREPANCY REMOVED IF JET DATA USED FOR GLUON DETERMINATION



NEW CTEQ GLUON (1998)

MODEL-DEPENDENT UNCERTAINTIES THE HERA-LHC BENCHMARK PUZZLE

- RESTRICTED AND VERY CONSISTENT DATASET USED
- RESULTS COMPARED TO THEN-BEST RESULT FROM FULL DATASET



BENCHMARK VS DEFAULT GLUON

"...the partons extracted using a very limited data set are completely incompatible, even allowing for the uncertainties, with those obtained from a global fit with an identical treatment of errors...The comparison illustrates the problems in determining the true uncertainty on parton distributions." (R.Thorne, HERALHC, 2005)

REGRESSION THROUGH POLYNOMIALS THE RUNGE PHENOMENON



- INTERPOLATE WITH A POLYNOMIAL A FUNCTION SAMPLED AT EQUALLY SPACED POINTS
- RUNGE/AGNESI FUNCTION: $\frac{1}{1+x^2}$ SAMPLED AT $x_i = \frac{2i}{n} 1$, i = 1, ..., n
- OSCILLATIONS INCREASE AS THE DEGREE OF THE POLYNOMIAL INCREASES
- ALLEVIATED WITH SUITABLE CHOICE OF POLYNOMIALS AND SAMPLING POINTS (CHEBYSHEV NODES)

REGRESSION THROUGH POLYNOMIALS CHEBYSHEV AND LENGTH PENALTY

- OLD IDEA FOR PDF MODELING (PARISI, SOURLAS, 1978): EXPAND PDFS OVER BASIS OF ORTHOGONAL POLYNOMIALS
- GLAZOV, RADESCU, 2009: SYSTEMATIC MONTE CARLO APPROACH
- LENGTH PENALTY STABILIZATION: CONTRIBUTION TO χ^2 proportional to the arclength with weight p
- RESULTS STRONGLY DEPENDENT ON ARBITRARY CHOICE OF p





VALIDATION OF THEORY UNCERTAINTIES NLO SCALE VARIATION VS ACTUAL NNLO CORRECTION



(G. Salam, 2016)

- SCALE VARIATION CANNOT PREDICT OPENING OF NEW CHANNELS OR RESUMMATION
- ISSUE KNOWN IN MOST CASES

