Studies of the \mathbf{k}_{\perp} -dependence in uPDFs

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Summary: During the studentship Gaussian (normal) parametrizations of the uPDFs' k_{\perp} dependence are compared to measurements of inclusive dijet events at Hera. It is found that for a fixed width of 1 GeV a vanishing or small mean in the distribution is favored.

1 Introduction

Unintegrated parton density functions (uPDFs) are part of the CCFM[1, 2, 3, 4] description of high energy scattering processes of hadrons. An introduction to this theory can for example be found in [5, 6], and in the references therein. Also notations not explicitly defined here may be found in these references.

Events generated with Gaussian parametrizations of the uPDFs' k_{\perp} dependence are in the following compared to measured inclusive dijet events in e^+p collisions at the Hera collider ($E_{e^+} = 27.5 \,\text{GeV}, E_P = 820 \,\text{GeV}$) published by the H1 collaboration[7].

2 Simulation steps

The simulation of dijet events given an uPDF starting parametrization can be divided in the following computational task:

- 1. Calculate the CCFM evolution kernel using the Monte Carlo (MC) program SMALLX[8]. Store the kernel as a grid in the file test-grid.dat. (Input parameters: Figure 8)
- 2. Use the kernel to generate a grid for $\mathcal{A}(x, \bar{q}, k_{\perp})$. (CALCGLUE[9]) Results are stored in the file ccfm-test.dat
- 3. Generate hadron level events with CASCADE 2.2.0[6], populate histograms with the HZTOOL-4.3[10] subroutine hz03160.
- 4. Compare MC to data distributions via the χ^2 statistic defined below to assess the data's sensitivity to the parametrization.

Definition of the χ^2 statistic used:

$$\chi^{2} = \sum_{n=0}^{N} \frac{\left(Y_{\text{data}}^{(n)} - Y_{\text{MC}}^{(n)}\right)^{2}}{\left(\delta Y_{\text{data}}^{(n)}\right)^{2} + \left(\delta Y_{\text{MC}}^{(n)}\right)^{2}}$$

(1)

With

 $\begin{array}{lll} Y^{(n)}_{\rm data/MC} & : & {\rm Content} \mbox{ of bin } n \mbox{ of the data or MC histogram} \\ \delta Y^{(n)}_{\rm data} & : & {\rm Total \ error \ of \ bin } n \mbox{ of the data \ histogram} \\ \delta Y^{(n)}_{\rm MC} & : & {\rm Total \ error \ of \ bin \ } n \mbox{ of the MC \ histogram} \\ N & : & {\rm Number \ of \ bins} \end{array}$

3 Inspected dijet distributions

While various distributions of dijet events are published in [7], the focus here is on two distributions which are potentially sensitive to the gluon k_{\perp} distribution:

- 1. The inclusive dijet cross section multiplied by $\langle x \rangle$, averaged over x and Q^2 in dependence of Δ (Figure 2 in [7]).
- 2. The fraction S(120) of events where the two hardest jets open an azimuthal angle less than 120° (Figure 9 in [7]) in dependence on x. Explicitly S is defined as

$$S(\alpha) = \frac{\int_0^{\alpha} N_{\text{dijet}}(\Delta\phi^*, x, Q^2) \mathrm{d}\Delta\phi^*}{\int_0^{180^{\circ}} N_{\text{dijet}}(\Delta\phi^*, x, Q^2) \mathrm{d}\Delta\phi^*}.$$
(2)

4 uPDF starting parametrization—Gaussian for k_{\perp}

The uPDF starting parametrization is split into an x dependent part, implemented in CALCGLUE, and a k_{\perp} starting distribution, implemented in SMALLX. The following distributions are used here:

• The x starting distribution is chosen according to a recent fit to F_2 data[11],

$$A_{0}(x) = Nx^{-B}(1-x)^{C}(1-Dx) \quad \text{with} \quad \begin{cases} N = 0.417 \\ B = 0.125 \\ C = 4.0 \\ D = -9.2 \end{cases}$$
(3)

(See Figure 9 for the modifications applied to calcglu.F.)

• The probability density for generating the starting k_{\perp} values in SMALLX is a Gaussian distribution in k_{\perp} , *i.e.*

$$\mathcal{P}(k_{\perp}) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(k_{\perp}-\mu)^2}{2\sigma^2}}.$$
(4)

(See Figure 6 for the modifications applied to sminfn_gluon.f.)

For the two parameters μ and σ the following one–dimensional χ^2 scans are performed.

4.1 χ^2 scan: μ varied, σ fixed

For μ in the range 0.12 - 12.0 GeV and $\sigma = 1 \text{ GeV}$, per parameter point 10.000.000 events are generated with SMALLX and 5.000.000 with CASCADE.

 χ^2/N is shown in Figure 1 for both distributions described in section 3. For the parameter points highlighted in Figure 1, the distributions of [7] enlisted in section 3 are reproduced in Figure 4.



(a) χ^2 scan for the distributions of Figure 2 of [7]



Figure 1: χ^2 scans: Generated with 10.000.000 events in SMALLX and 5.000.000 events in CASCADE. The colored dots mark the Monte Carlo distributions in Figures 4.

Obviously, the best description of the data examined is given by small values of μ .

4.2 χ^2 scan: μ fixed, σ varied

For $\mu = 0 \text{ GeV}$ and σ in the range 0.25 - 4 GeV, 10.000.000 events are generated with SMALLX and 5.000.000 with CASCADE per parameter point.

 χ^2/N is shown in Figure 2 for both distributions described in section 3. The distributions of [7] are reproduced in Figure 5 for the parameter points highlighted in Figure 2.

As can be seen in Figure 5, the program chain failed to produce realistic events for $\mu = 0 \text{ GeV}$ and small means ($\sigma = 0.25 \text{ GeV}$). This is probably due to the collinear cutoff[6] Q_0 , which was chosen here to be $Q_0 = 1.3 \text{ GeV}$, see the listing in Figure 8.

The distribution of $S(120^{\circ})$ in Figure 5a seems to favor $\sigma \leq \mathcal{O}(1 \text{ GeV})$, as even the MC distribution for $\sigma = 1.25 \text{ GeV}$ overshoots the data. However, this enhancement as well as the rather flat distribution for $0.75 \text{ GeV} \leq \sigma \leq 1.5 \text{ GeV}$ in Figure 2 can probably be explained with the large cutoff Q_0 .



Figure 2: χ^2 scans: Generated with 10.000.000 events in SMALLX and 5.000.000 events in CASCADE. The colored dots mark the MC distributions in Figures 5.

5 uPDF starting parametrization—Gaussian for \mathbf{k}_{\perp}^2

As a variation, a Gaussian distribution in k_{\perp}^2 is examined, i.e.

$$\mathcal{P}(k_{\perp}^2) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(k_{\perp}^2 - \mu)^2}{2\sigma^2}}.$$
(5)

(See Figure 7 for the modifications applied to sminfn_gluon.f.)

A χ^2 scan has been performed for $\sqrt{\mu}$ in 0.01–3.5 GeV and $\sqrt{\sigma} = 1$ GeV with 10.000.000 events in SMALLX and 5.000.000 events in CASCADE. The numeric results are depicted in Figure 3 for both distributions of section 3.

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References

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(a) χ^2 scan for the distributions of Figure 2 of [7]

(b) χ^2 scan for the distributions of Figure 9 in [7]

Figure 3: χ^2 scans: Generated with 10.000.000 events in SMALLX and 5.000.000 events in CASCADE.

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Figure 4: Figures corresponding to Figures 2 and 9 of [7]. Black: data, colored: MC samples from the scan (see Figure 1). The ranges of the axes are chosen in accordance with the original figures (up to the factor of 10^{-3} in the x axis of (b)).



Figure 5: Figures corresponding to Figures 2 and 9 of [7]. Black: data, colored: MC samples from the scan (see Figure 2). The ranges of the axes are chosen in accordance with the original figures (up to the factor of 10^{-3} in the x axis of (b)).

```
43,44c43,44
<
<
        Double Precision sigm, xmean
- -
>
        Double Precision sigmin, xmeanin, hskt, hsphi
        common /gaussalb/sigmin,xmeanin
>
84,90c84,87
             P(1,0) = SMrgau(1,0.0d0,Qx)
<
<
             P(2,0) = SMrgau(2,0.0d0,Qx)
   new test for Albert
<
  С
             sigm = 3.0/sqrt(2.)
< c
              xmean = 3.
< c
              P(1,0) = SMrgau(1,xmean,sigm)
< c
< c
              P(2,0) = SMrgau(2, xmean, sigm)
- - -
             hskt= SMrgau(1, xmeanin, sigmin)
>
             hsphi=2*acos(-1.)*SMRGEN(I)
>
>
             P(1,0)= hskt*sin(hsphi)
>
             P(2,0) = hskt*cos(hsphi)
```

Figure 6: Modifications applied to sminfn_gluon.f[8] in .diff format—Gaussian k_{\perp}

```
43,44c43,44
<
<
        Double Precision sigm, xmean
>
        Double Precision sigmin, xmeanin, hskt, hskt2, hsphi
>
        common /gaussalb/sigmin,xmeanin
84,85d83
<
             P(1,0) = SMrgau(1,0.0d0,Qx)
             P(2,0) = SMrgau(2,0.0d0,Qx)
<
89,90c87,91
< c
              P(1,0) = SMrgau(1, xmean, sigm)
<
 С
              P(2,0) = SMrgau(2, xmean, sigm)
_ _
             hskt2= SMrgau(1, xmeanin**2, sigmin**2) !absolute value?
>
             hskt=sqrt(abs(hskt2))
>
             hsphi=2*acos(-1.)*SMRGEN(I)
>
>
             P(1,0)= hskt*sin(hsphi)
>
             P(2,0) = hskt*cos(hsphi)
```

Figure 7: Modifications applied to sminfn_gluon.f[8] in .diff format—Gaussian k_{\perp}^2 .

Ipgg = 1 ns = 1 Qg = 1.3 Qs = 1.3 Xnorm = 1. oneLoop = 0 Iglu = 1



```
659,662c659,662
         A1 = 0.49
<
         A2 = 0
<
<
         A3 = 4.9
<
         A4 = 0.
- -
>
           A1 = 0.417
>
           A2 = 0.125
>
           A3 = 4.0
           A4 = -9.2
>
672c672,673
<
            else
_ _ _
>
           else
>
              test = A1 * 1./x0**A2 * (1.-x0)**A3 * (1.-A4*x0)*xpqs(0)/x0
674d674
            test =A1*5.*(1./x0**A2)*(1-x0)**A3*(1.-A4*x0)*xpqs(0)/x0
<
```

Figure 9: Modifications applied to calcglu.F[9] in .diff format