

Extending a Generalized Parton Distribution from DGLAP to ERBL (Preliminary results!)

From an Overlap of Light-cone Wave-functions to a Double Distribution

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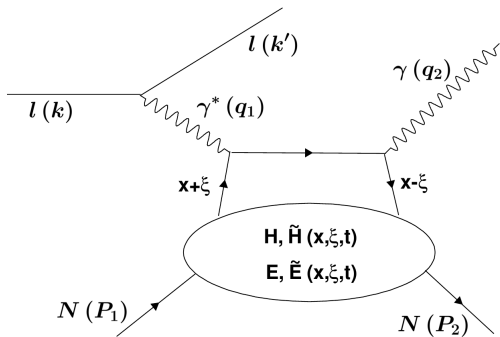
Outline

- 1 Introduction to Generalized Parton Distributions
 - Why Generalized Parton Distributions?
 - Definition and properties
- 2 Overlap and Double Distribution representations of GPDs
 - Overlap of Light-cone wave functions
 - Double Distributions
- 3 From an Overlap of LCWFs to a Double Distribution
 - Inversion of Incomplete Radon Transform
 - Results
- 4 Conclusion

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Experimental access (example: DVCS)



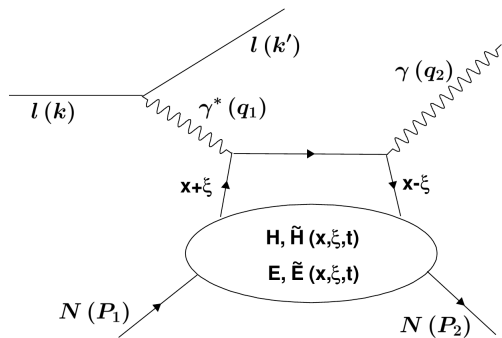
Deeply Virtual Compton Scattering channel of photon electroproduction.

$$\Delta = P_2 - P_1, \quad t = \Delta^2 < 0$$

$$Q^2 = -q_1^2 > 0$$

$$P = \frac{1}{2}(P_1 + P_2), \quad \xi = -\frac{\Delta^+}{2P^+}$$

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Compton Form Factors: [\(Belitsky et al., 2002\)](#)

$$\mathcal{F}(\xi, t, Q^2) = \int_{-1}^1 dx C\left(x, \xi, \alpha_S(\mu_F), \frac{Q}{\mu_F}\right) F(x, \xi, t, \mu_F), \quad (1)$$

where $F \in \{H, E, \tilde{H}, \tilde{E}, \dots\}$ is a Generalized Parton Distribution.

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- Probability density (Fourier transform of GPD): ([Burkardt, 2000](#))

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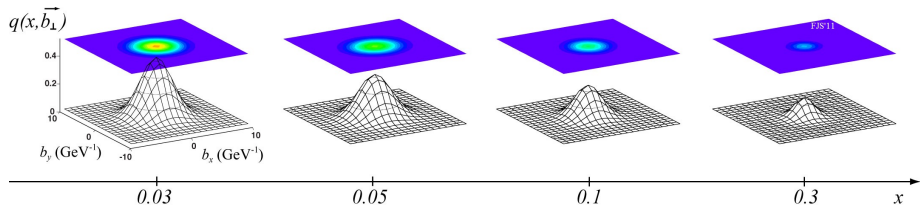


Figure: Hadron tomography.

Nucleon spin

- Ji's decomposition of the nucleon spin: (Ji, 1997)

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- Ji's sum rule:

$$\begin{aligned} J^q &= \frac{1}{2} \int_{-1}^1 x [H^q(x, \xi, 0) + E^q(x, \xi, 0)] dx \\ &= \frac{1}{2} \int_0^1 x [q(x) + \bar{q}(x)] dx + \frac{1}{2} \int_{-1}^1 x E^q(x, 0, 0) dx \\ &= \frac{1}{2} \Delta q + L^q. \end{aligned} \quad (4)$$

Definition of GPDs

- Quark GPD: (Müller et al., 1994; Radyushkin, 1996; Ji, 1997)

$$F^q(x, \xi, t) = \frac{1}{2} \int \frac{dz^-}{2\pi} e^{ixP^+z^-} \langle P_2 | \bar{q}(-z) \gamma^+ q(z) | P_1 \rangle \Big|_{z^+=0, z_\perp=0}, \quad (5)$$

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- Link to PDFs and Form Factors:

$$\int dx H^q(x, \xi, t) = F_1^q(t) \quad , \quad \int dx E^q(x, \xi, t) = F_2^q(t) \quad , \quad (8)$$

$$H^q(x, 0, 0) = \theta(x) q(x) - \theta(-x) \bar{q}(-x). \quad (9)$$

Theoretical constraints on GPDs

Main properties:

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- Positivity: (Pire et al., 1999)

$$H^q(x, \xi, t) \leq \sqrt{q \left(\frac{x - \xi}{1 - \xi} \right) q \left(\frac{x + \xi}{1 + \xi} \right)}. \quad (11)$$

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- ▶ Cauchy-Schwarz theorem in Hilbert space.

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Light-cone wave functions (LCWFs)

(Brodsky and Lepage, 1989)

- A given *hadronic state* is decomposed in a **Fock basis**:

$$|H; P, \lambda\rangle = \sum_{N, \beta} \int [dx]_N [d^2\mathbf{k}_\perp]_N \Psi_{N, \beta}^\lambda(x_1, \mathbf{k}_{\perp 1}, \dots) |N, \beta; k_1, \dots, k_N\rangle, \quad (12)$$

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- For example, for the pion:

$$|\pi\rangle = \sum_{q\bar{q}} \psi_{q\bar{q}}^\pi |q\bar{q}\rangle + \sum_{q\bar{q}g} \psi_{q\bar{q}g}^\pi |q\bar{q}g\rangle + \dots$$

Overlap of LCWFs

(Diehl et al., 2001; Mezrag, 2015)

- The GPD can be then computed as an overlap of LCWFs:

$$H^q(x, \xi, t) = \sum_{N, \beta} \sqrt{1 - \xi}^{2-N} \sqrt{1 + \xi}^{2-N} \sum_{j=1}^N \delta_{S_j, q} \quad (13)$$

$$\int [d\bar{x}]_N [d^2\bar{\mathbf{k}}_\perp]_N \delta(x - \bar{x}_j) \Psi_{N, \beta}^*(\Omega_2) \Psi_{N, \beta}(\Omega_1),$$

in the DGLAP region $\xi < x < 1$ (pion case).

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Double Distributions (DDs)

- DD representation of GPDs:

$$H^q(x, \xi, t) = \int_{\Omega} d\beta d\alpha (F^q(\beta, \alpha, t) + \xi G^q(\beta, \alpha, t)) \delta(x - \beta - \alpha\xi) . \quad (14)$$

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- ▶ Positivity not manifest...

One-component DD (1CDD)

- BMKS gauge: (Belitsky et al., 2001)

$$H(x, \xi, t) = x \int_{\Omega} d\beta d\alpha f_M(\beta, \alpha, t) \delta(x - \beta - \alpha\xi). \quad (16)$$

with

$$\begin{cases} F(\beta, \alpha) = \beta f_M(\beta, \alpha) \\ G(\beta, \alpha) = \alpha f_M(\beta, \alpha) \end{cases}. \quad (17)$$

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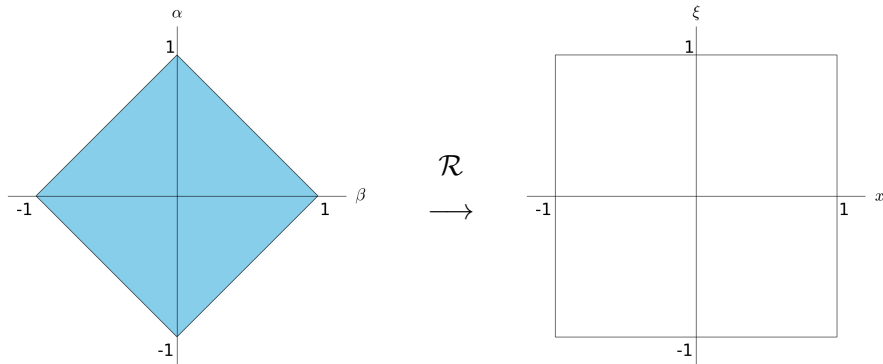
- Pobylitsa gauge: (Pobylitsa, 2004)

$$H(x, \xi, t) = (1 - x) \int_{\Omega} d\beta d\alpha f_P(\beta, \alpha, t) \delta(x - \beta - \alpha\xi). \quad (18)$$

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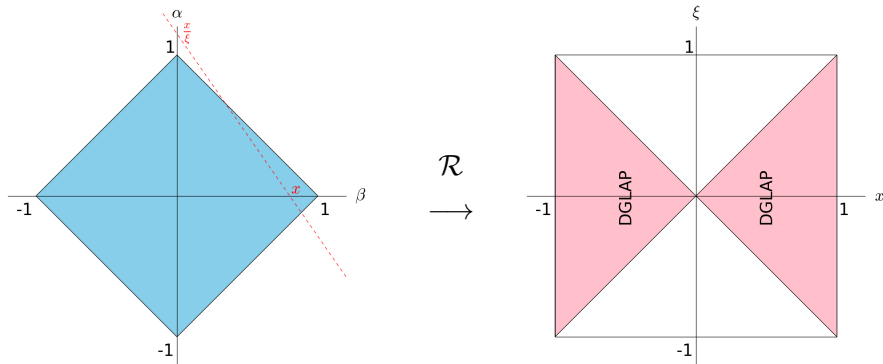
Radon transform



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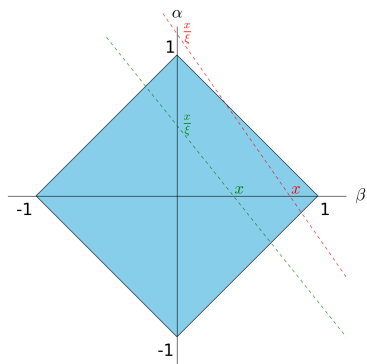
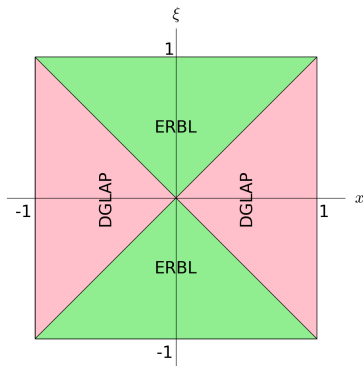


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 \mathcal{R}
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- ▶ ERBL region: $|x| < |\xi|$.

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Problem

Find $f(\beta, \alpha)$ on square $\{|\alpha| + |\beta| \leq 1\}$ such that

$$H(x, \xi)|_{\text{DGLAP}} = \left\{ \begin{array}{c} x \\ 1-x \end{array} \right\} \int d\beta d\alpha f(\beta, \alpha) \delta(x - \beta - \alpha\xi) .$$

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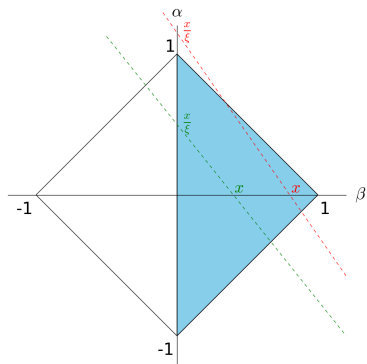
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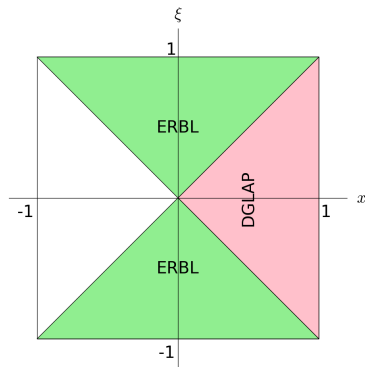
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 - ▶ DD $f(\beta, \alpha)$ **exists** (as a *distribution*) and is **unique** (if it is a *function*).
 - ▶ We can reconstruct the GPD everywhere.

(Moutarde, 2015)

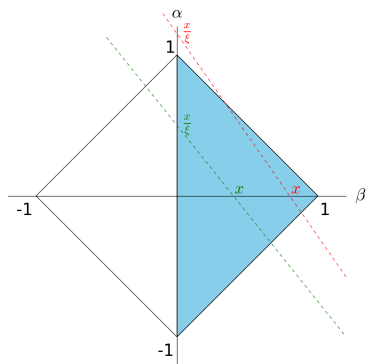
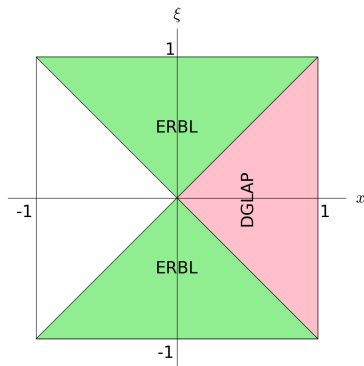
Support properties



\mathcal{R}
→

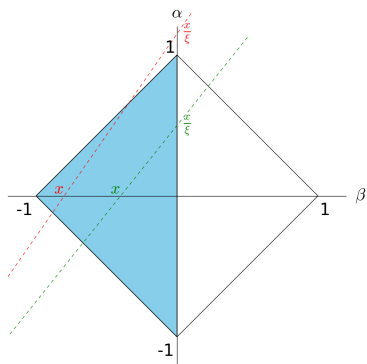
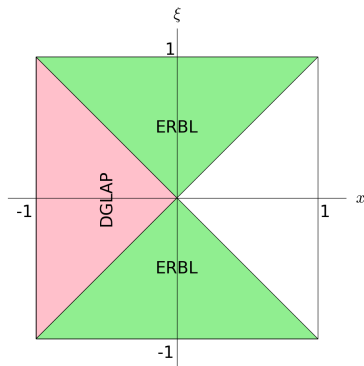


Support properties


 \mathcal{R}
 \longrightarrow


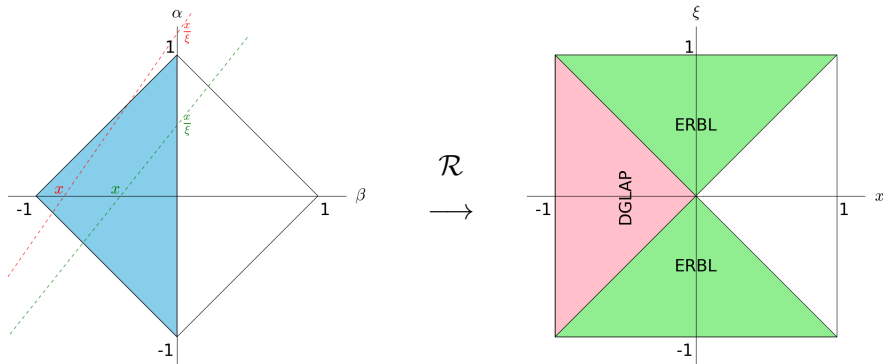
- Valence GPD: $H(x, \xi) = 0$ for $-1 < x < -|\xi| \implies f(\beta, \alpha) = 0$ for $\beta < 0$.

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 \longrightarrow


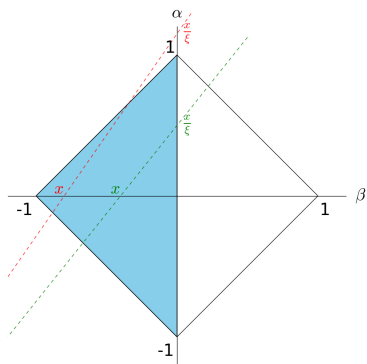
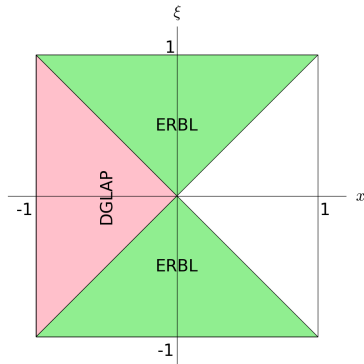
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Support properties



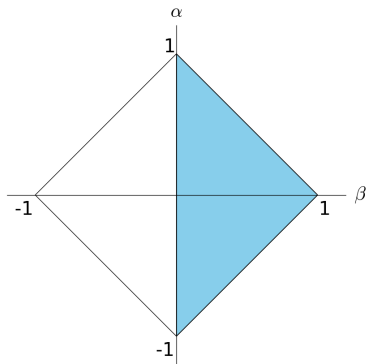
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Support properties

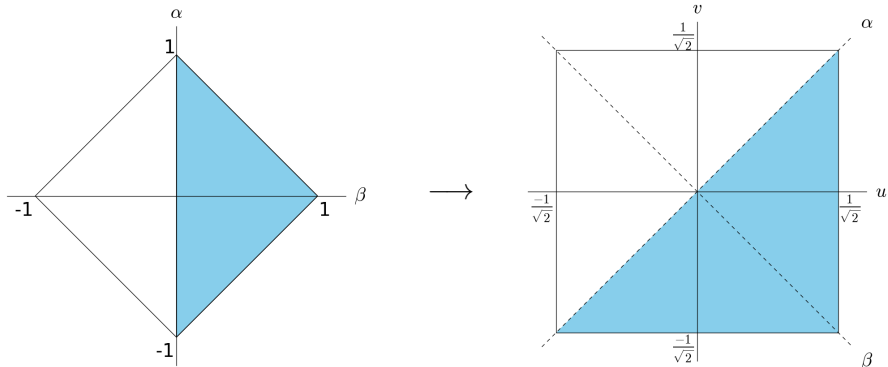

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- Valence GPD: $H(x, \xi) = 0$ for $-1 < x < -|\xi| \implies f(\beta, \alpha) = 0$ for $\beta < 0$.
- Domains $\beta < 0$ and $\beta > 0$ are uncorrelated in the DGLAP region.
- Divide and conquer:
 - ▶ Better numerical stability.
 - ▶ Lesser complexity: $O(N^p + N^p) \ll O((N + N)^p)$.

Domain for the inversion



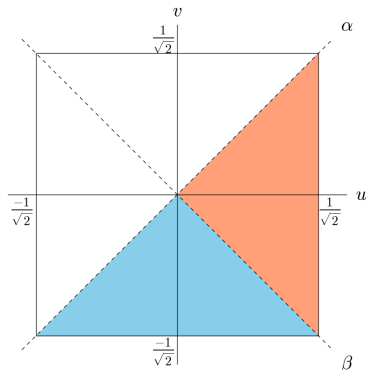
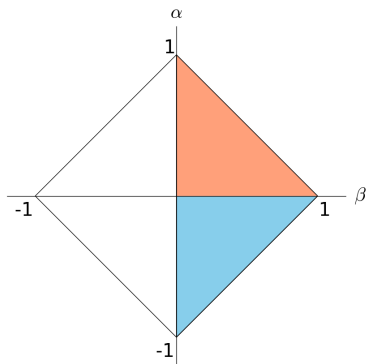
Domain for the inversion



- Rotated square $[-\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}] \times [-\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}]$:

$$\begin{cases} u = \frac{\beta + \alpha}{\sqrt{2}} \\ v = \frac{\alpha - \beta}{\sqrt{2}} \end{cases} \quad (21)$$

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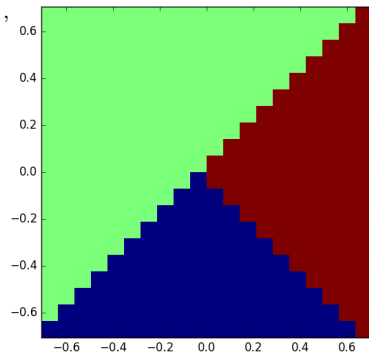
- α -parity of the DD:

$$f(\beta, -\alpha) = f(\beta, \alpha) \quad (22)$$

Discretization

- Discretization of the DD (piece-wise constant):

$$\tilde{f}(u, v) = \sum_{i=0}^{N-1} \sum_{j=0}^{N-1} \tilde{f}_{ij} \mathbf{1}_{[u_i, u_{i+1}]}(u) \mathbf{1}_{[v_j, v_{j+1}]}(v), \quad (23)$$

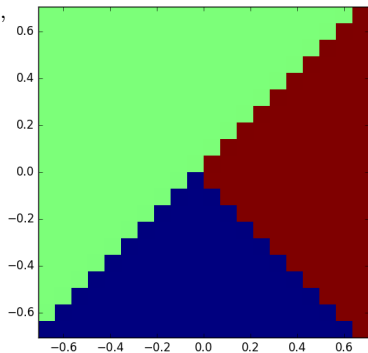


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- Mesh:

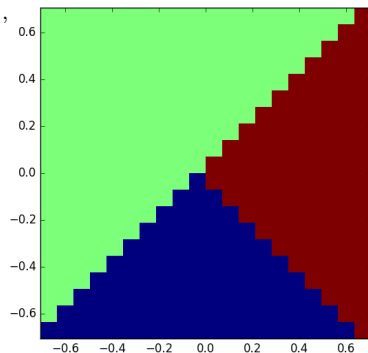


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- Mesh:
 - Cells $(u, v) \rightarrow n$ columns of the matrix.

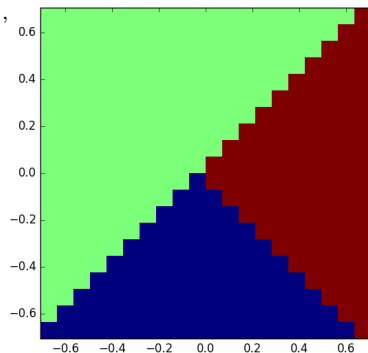


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- Sampling:

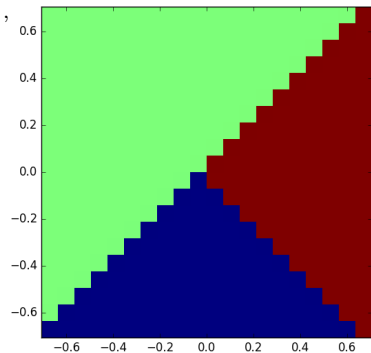


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- Sampling:
 - Random couples $(x, \xi) \rightarrow m \geq n$ lines of the matrix.

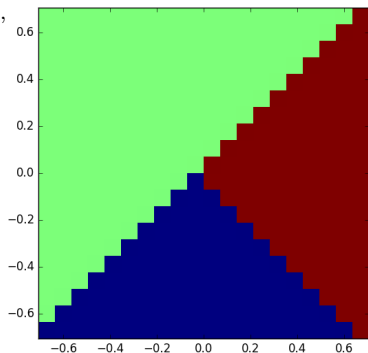


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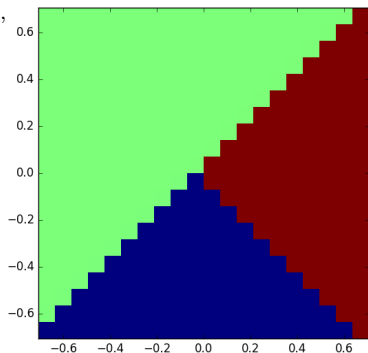


Discretization

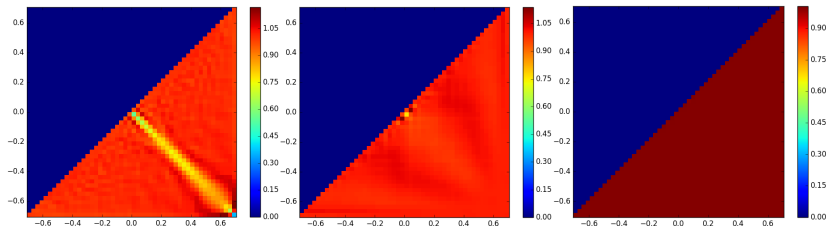
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- Linear problem: $AX = B$ where $B_k = H(x_k, \xi_k)$.
 - ▶ A full-rank: more information but also more noise.



Tests (constant DD)



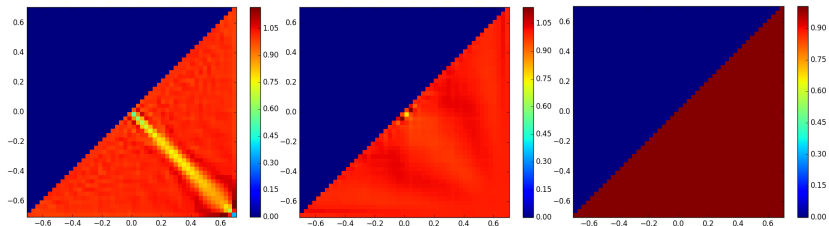
- Test with Constant DD.

$$f(\beta, \alpha) = \begin{cases} 1 & \beta > 0 \\ 0 & \beta < 0 \end{cases}$$

↓

$$H(x, \xi)|_{\text{DGLAP}} = \begin{cases} \frac{2x(1-x)}{1-\xi^2} & |\xi| < x < 1 \\ 0 & -1 < x < -|\xi| \end{cases}$$

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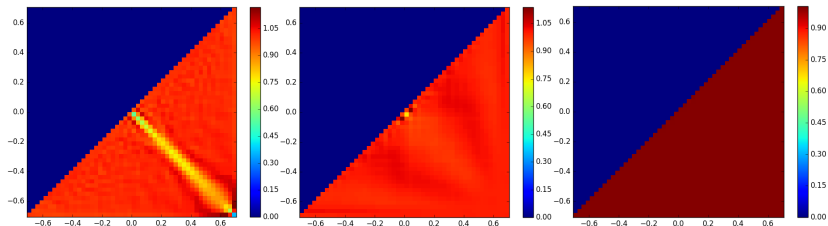
- ▶ Goal: retrieve known DD from DGLAP GPD.

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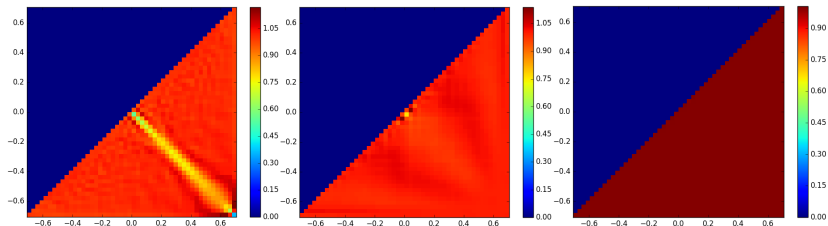
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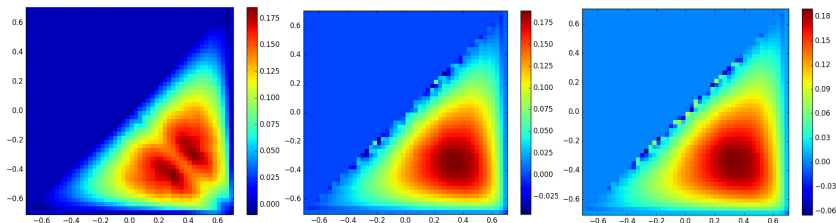
↓

- Consistent problem (discretized DD = theoretical DD):

- ▶ Objective DD retrieved at arbitrary precision: residue decreases to 0 (machine precision).

$$H(x, \xi)|_{\text{DGLAP}} = \begin{cases} \frac{2x(1-x)}{1-\xi^2} & |\xi| < x < 1 \\ 0 & -1 < x < -|\xi| \end{cases}$$

Tests (RDDA)



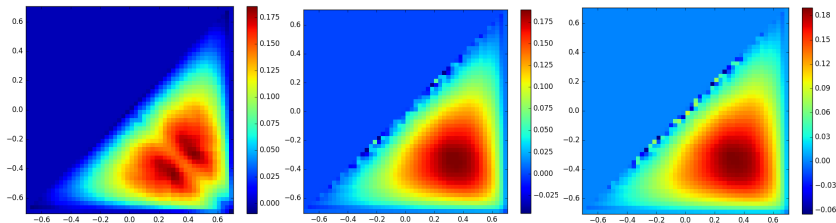
- Test with a DD inspired by Radyushkin ansatz.

$$f(\beta, \alpha) = \begin{cases} \frac{3(\alpha^2 - (1-\beta)^2)\beta}{4(-1+\beta)} & \beta > 0 \\ 0 & \beta < 0 \end{cases}$$

↓

$$H(x, \xi)|_{x>|\xi|} = \frac{(1-x)^3 \left(3\xi + (2x-5)\xi^3 - 3(1-\xi^2)^2 \arctan \xi \right)}{2\xi^3(1-\xi^2)^2}$$

Tests (RDDA)



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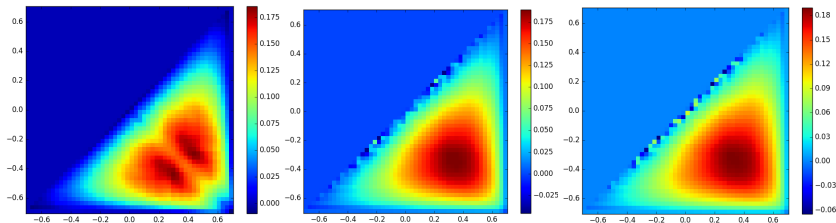
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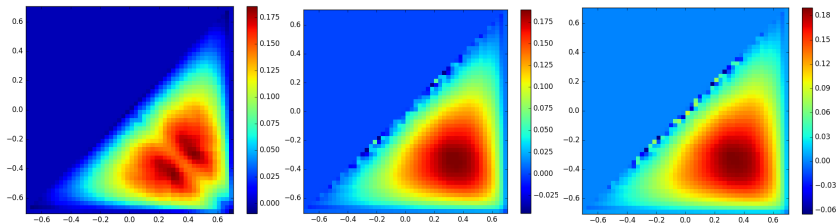
- Smooth function, vanishes at the boundaries.

$$f(\beta, \alpha) = \begin{cases} \frac{3(\alpha^2 - (1-\beta)^2)\beta}{4(-1+\beta)} & \beta > 0 \\ 0 & \beta < 0 \end{cases}$$

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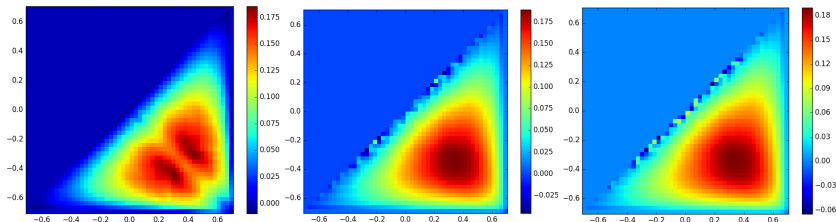
- ▶ Least-squares problem: residue has a finite limit.

$$f(\beta, \alpha) = \begin{cases} \frac{3(\alpha^2 - (1-\beta)^2)\beta}{4(-1+\beta)} & \beta > 0 \\ 0 & \beta < 0 \end{cases}$$

↓

$$H(x, \xi)|_{x>|\xi} = \frac{(1-x)^3 \left(3\xi + (2x-5)\xi^3 - 3(1-\xi^2)^2 \arctan \xi \right)}{2\xi^3(1-\xi^2)^2}$$

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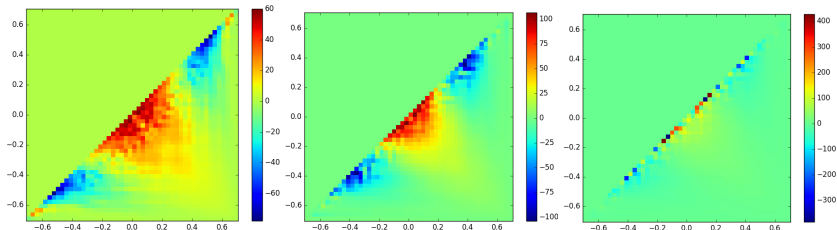
- ▶ Least-squares problem: residue has a finite limit.
- ▶ Compromise between noise on $\beta = 0$ and artifact on $\alpha = 0$.

$$f(\beta, \alpha) = \begin{cases} \frac{3(\alpha^2 - (1-\beta)^2)\beta}{4(-1+\beta)} & \beta > 0 \\ 0 & \beta < 0 \end{cases}$$

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First result



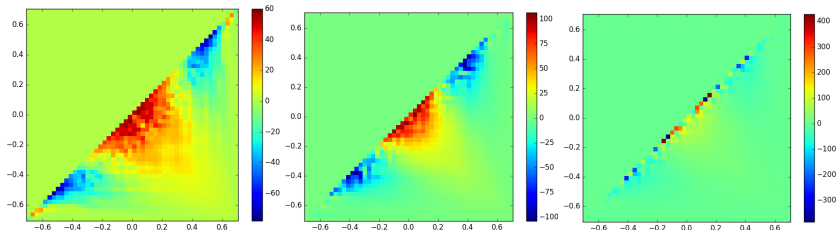
- Real application to an algebraic DSE overlap model.

$$f(\beta, \alpha) = \begin{cases} ? & \beta > 0 \\ 0 & \beta < 0 \end{cases}$$

↓

$$H(x, \xi)|_{x > |\xi|} = 30 \frac{(1-x)^2 (x^2 - \xi^2)}{(1 - \xi^2)^2}$$

First result



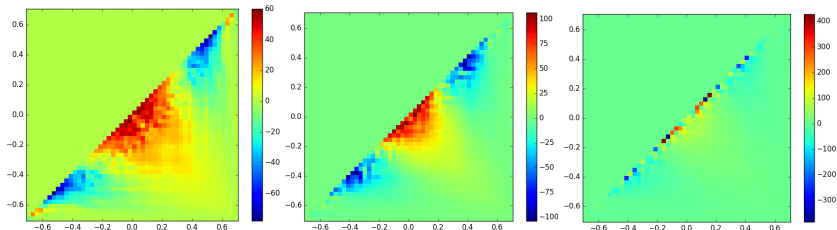
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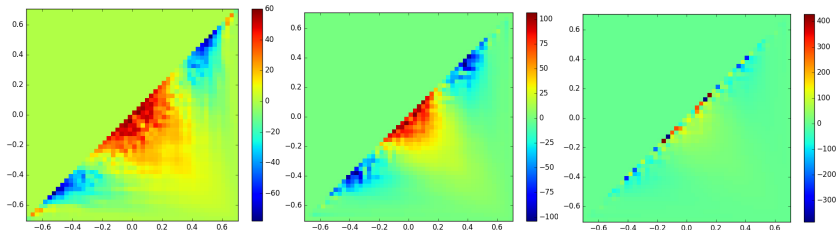
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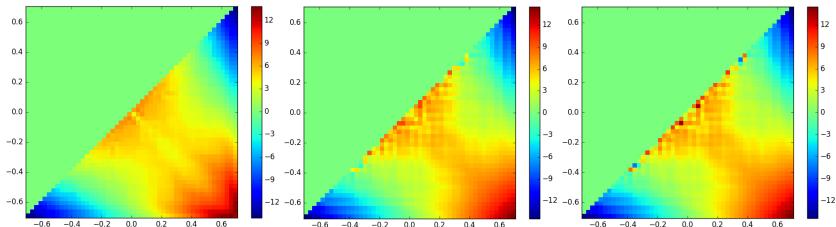
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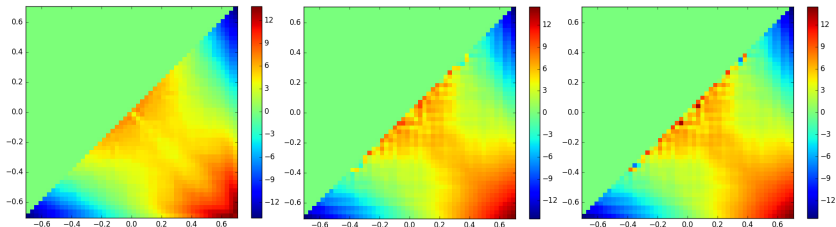
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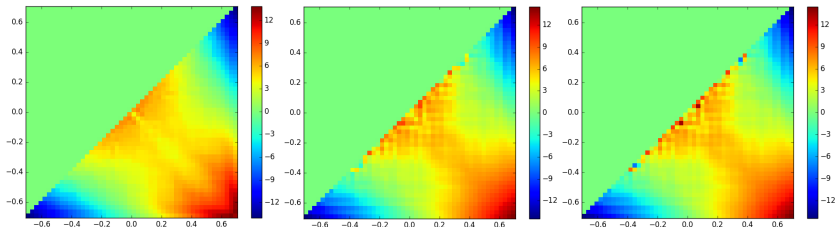
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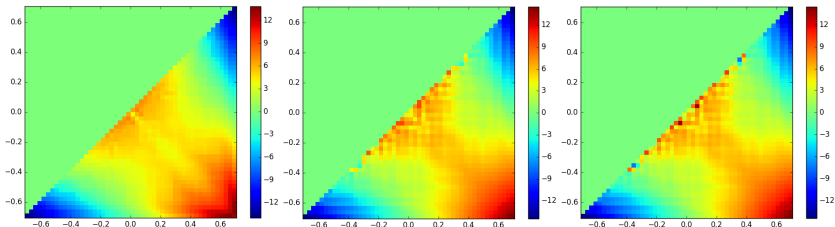
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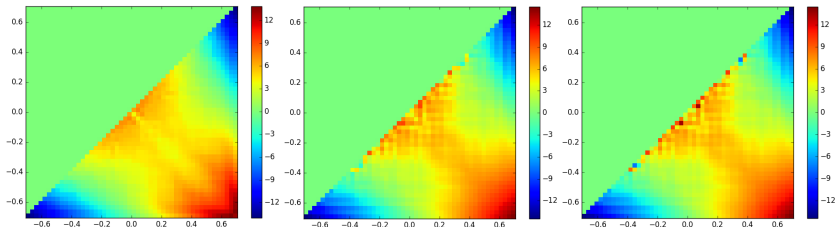
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$$f(\beta, \alpha) = \begin{cases} \frac{30}{4} (1 - 3\alpha^2 - 2\beta + 3\beta^2) & \beta > 0 \\ 0 & \beta < 0 \end{cases}$$

↓

$$H(x, \xi)|_{x>|\xi|} = 30 \frac{(1-x)^2 (x^2 - \xi^2)}{(1-\xi^2)^2}$$

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Quantitative comparison of DDs

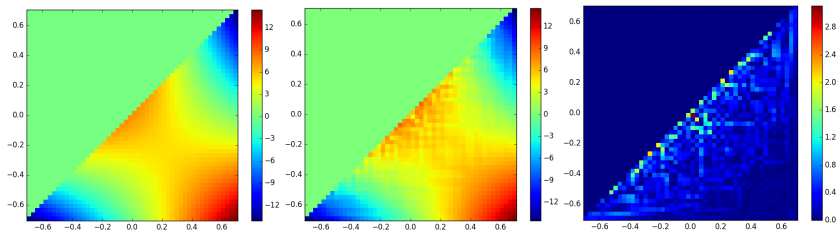


Figure: Quantitative comparison for the Overlap results. Left: Theoretical discretized DD. Middle: Numerical solution at tolerance 10^{-6} . Right: Absolute difference.

$$f(\beta, \alpha) = \begin{cases} \frac{3(\alpha^2 - (1-\beta)^2)\beta}{4(-1+\beta)} & \beta > 0 \\ 0 & \beta < 0 \end{cases}$$

Quantitative comparison of DDs

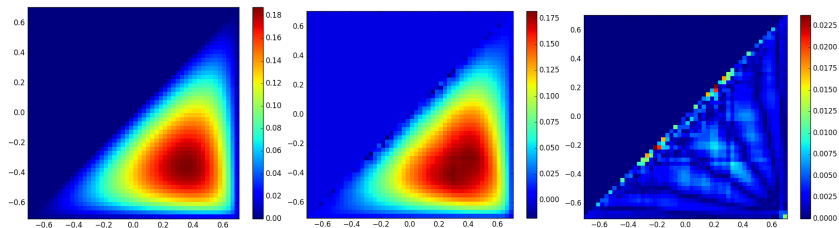


Figure: Quantitative comparison for the RDDA results. Left: Theoretical discretized DD. Middle: Numerical solution at tolerance 10^{-6} . Right: Absolute difference.

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Quantitative comparison of GPDs

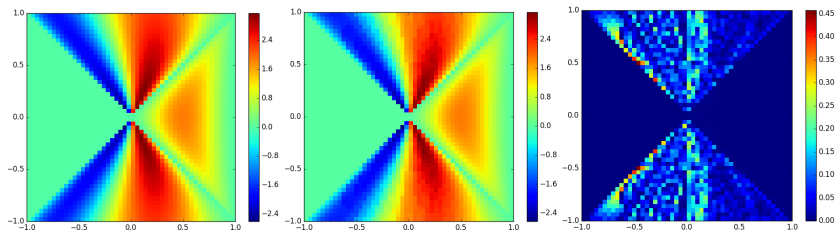


Figure: Quantitative comparison for the Overlap GPD obtained from the numerical DD solution. Left: Theoretical GPD. Middle: Numerical solution at tolerance 10^{-6} . Right: Absolute difference.

$$H(x, \xi)|_{x>|\xi|} = 30 \frac{(1-x)^2 (x^2 - \xi^2)}{(1 - \xi^2)^2}$$

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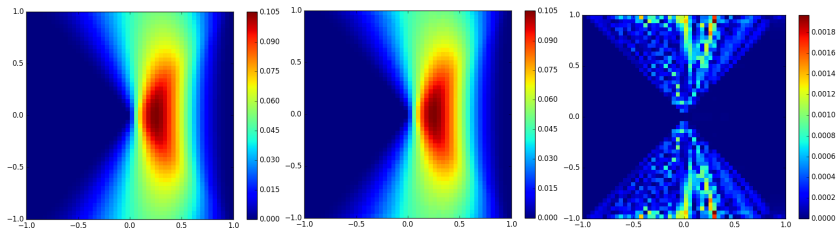


Figure: Quantitative comparison for the RDDA GPD obtained from the numerical DD solution. Left: Theoretical GPD. Middle: Numerical solution at tolerance 10^{-6} . Right: Absolute difference.

$$H(x, \xi)|_{x>|\xi|} = \frac{(1-x)^3 \left(3\xi + (2x-5)\xi^3 - 3(1-\xi^2)^2 \arctan \xi \right)}{2\xi^3 (1-\xi^2)^2}$$

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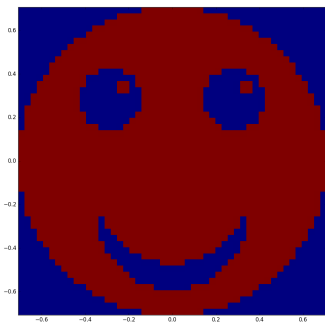
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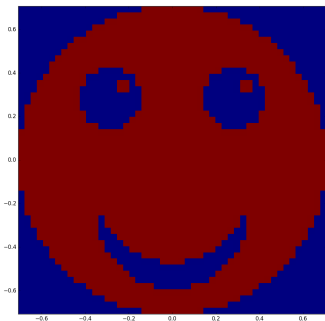
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 - ▶ Any questions?



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DSE Overlap DD

Recall the GPD: ([Mezrag, 2015](#); [Mezrag et al., 2016](#))

$$H(x, \xi) = 30 \frac{(1-x)^2 (x^2 - \xi^2)}{(1-\xi^2)^2}. \quad (24)$$

After trial and error (to reproduce the numerical result in Pobylitsa gauge that looks like a polynomial):

$$f_P(\beta, \alpha) = \frac{30}{4} (1 - 3\alpha^2 - 2\beta + 3\beta^2). \quad (25)$$

And with the mapping from Ref. ([Mueller, 2014](#)) (Eq. (10) of the article), we get the singular indeed DD in BMKS gauge:

$$\begin{aligned} f_M(\beta, \alpha) = & \frac{30}{4} \left(\alpha^2 \left(3 - \frac{3}{(|\alpha| + \beta)^4} \right) + \beta^2 \left(\frac{3}{(|\alpha| + \beta)^3} - 3 \right) \right) \\ & + \beta \left(4 - \frac{4}{(|\alpha| + \beta)^3} \right) + \frac{2}{(|\alpha| + \beta)^2} - 2 \\ & + \frac{30}{4} (1 - \alpha^2) \delta(\beta). \end{aligned} \quad (26)$$