

# 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
  - 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

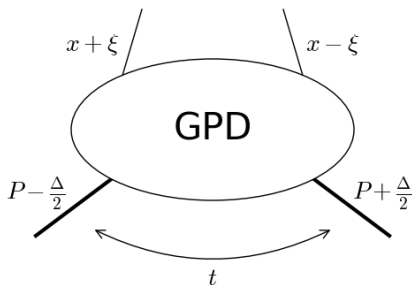
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# Definition of GPDs

- Quark GPD (twist-2, spin-0 hadron): (Müller et al., 1994; Radyushkin, 1996; Ji, 1997)

$$H^q(x, \xi, t) = \frac{1}{2} \int \frac{dz^-}{2\pi} e^{i x P^+ z^-} \left\langle P + \frac{\Delta}{2} \left| \bar{q}(-z) \gamma^+ q(z) \right| P - \frac{\Delta}{2} \right\rangle \Big|_{z^+=0, z_\perp=0} \quad (1)$$



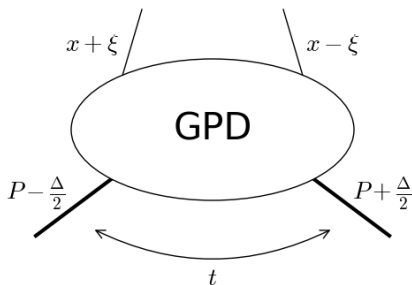
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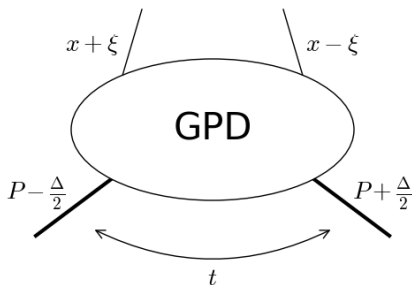
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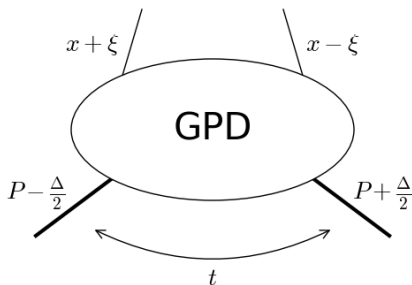
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- More GPDs for spin- $\frac{1}{2}$  hadrons.

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- Impact parameter space GPD (at  $\xi = 0$ ): (Burkardt, 2000)

$$q(x, b_\perp) = \int \frac{d^2\vec{\Delta}_\perp}{(2\pi)^2} e^{-ib_\perp \cdot \vec{\Delta}_\perp} H^q(x, 0, -\Delta_\perp^2). \quad (2)$$

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Main properties:

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- Link to PDFs and Form Factors:

$$\int dx H^q(x, \xi, t) = F_1^q(t) \quad , \quad \int dx E^q(x, \xi, t) = F_2^q(t) \quad , \quad (5)$$

$$H^q(x, 0, 0) = \theta(x) q(x) - \theta(-x) \bar{q}(-x). \quad (6)$$

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# Overlap of Light-cone wave functions

(Brodsky and Lepage, 1989)

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

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$$H^q(x, \xi, t) = \sum_{N, \beta} \sqrt{1 - \xi}^{2-N} \sqrt{1 + \xi}^{2-N} \int [d\bar{x}]_N [d^2\bar{\mathbf{k}}_\perp]_N \delta(x - \bar{x}_q) \Psi_{N, \beta}^*(\Omega_2) \Psi_{N, \beta}(\Omega_1), \quad (8)$$

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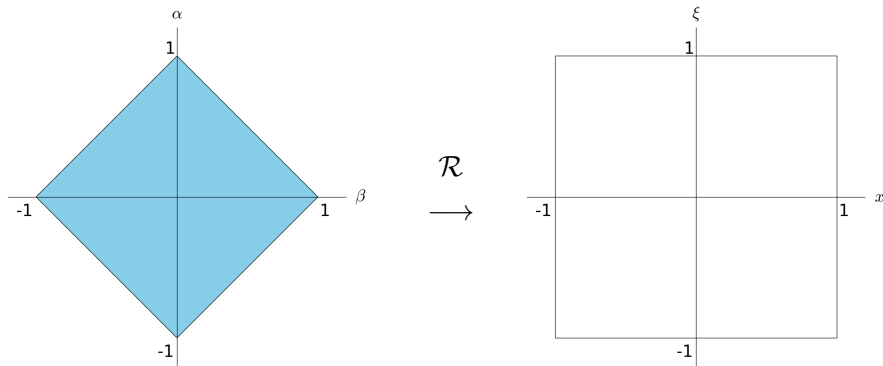
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- Pobylitsa gauge (One Component DD): ([Pobylitsa, 2003](#))

$$H(x, \xi, t) = (1 - x) \int_{\Omega} d\beta d\alpha f_P(\beta, \alpha, t) \delta(x - \beta - \alpha\xi), \quad (11)$$

with

$$\begin{cases} F(\beta, \alpha) &= (1 - \beta) f_P(\beta, \alpha) \\ G(\beta, \alpha) &= -\alpha f_P(\beta, \alpha) \end{cases}. \quad (12)$$

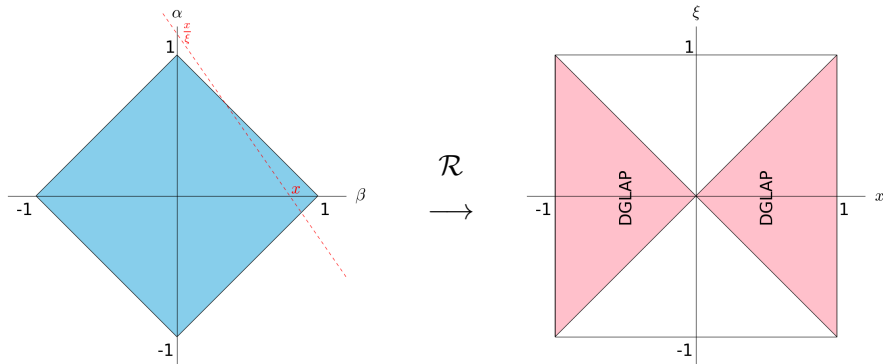
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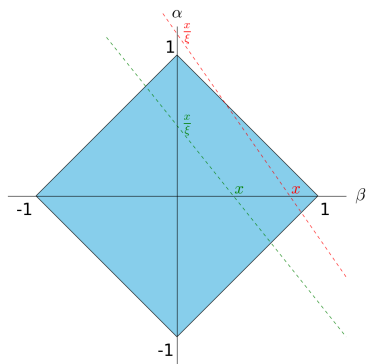
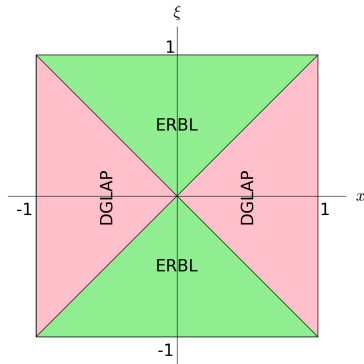


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  - ▶ DD  $f(\beta, \alpha)$  **exists** (as a *distribution*) and is **unique** (if it is a *function*).

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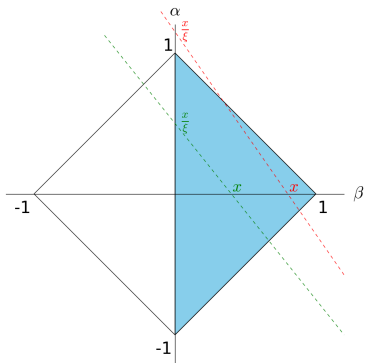
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Find  $f(\beta, \alpha)$  on square  $\{|\alpha| + |\beta| \leq 1\}$  such that

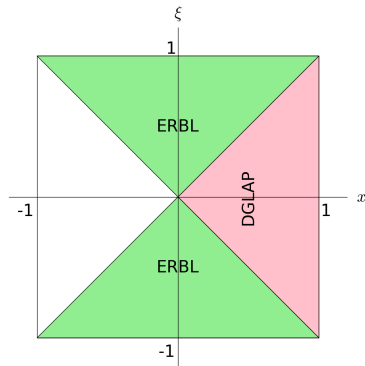
$$H(x, \xi)|_{\text{DGLAP}} = (1-x) \int d\beta d\alpha f(\beta, \alpha) \delta(x - \beta - \alpha\xi).$$

- If model fulfills Lorentz invariance: ([Moutarde, 2015](#))
  - ▶ DD  $f(\beta, \alpha)$  **exists** (as a *distribution*) and is **unique** (if it is a *function*).
  - ▶ We can reconstruct the GPD everywhere.

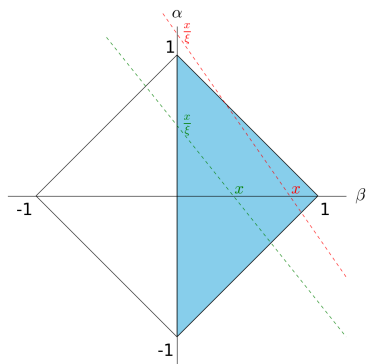
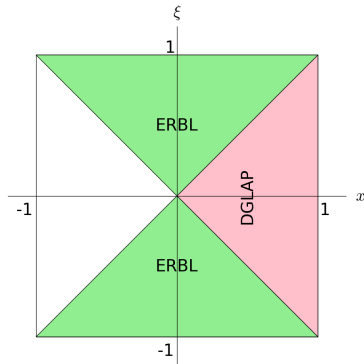
# Support properties



$\mathcal{R}$   
→



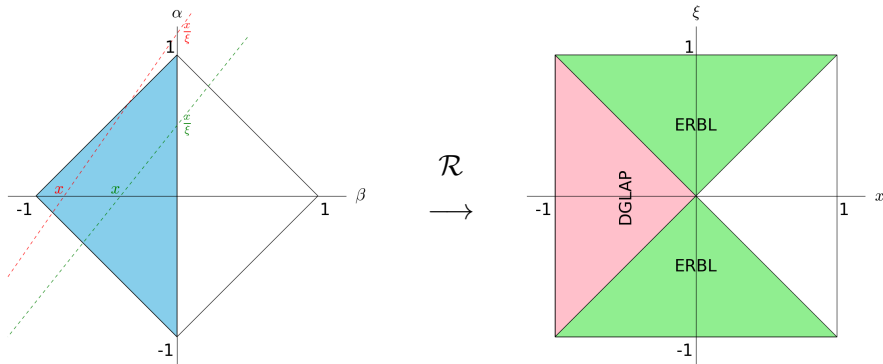
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 $\mathcal{R}$   
 $\longrightarrow$ 


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

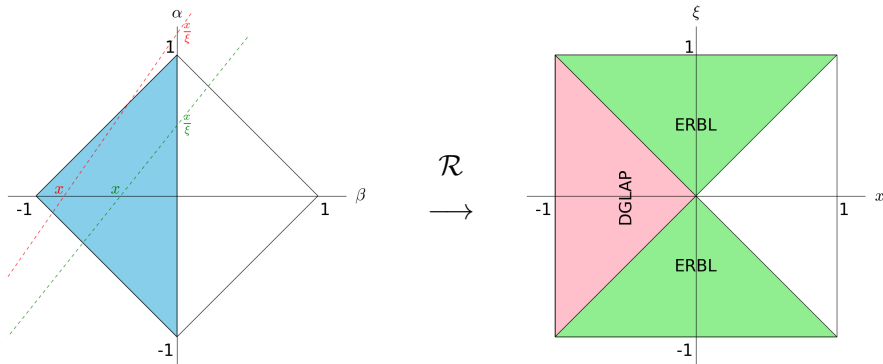


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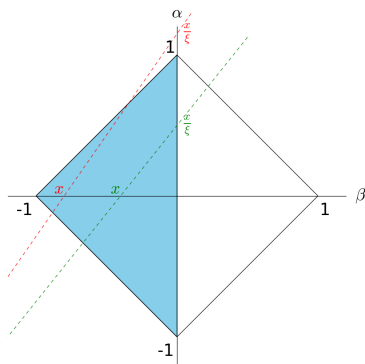
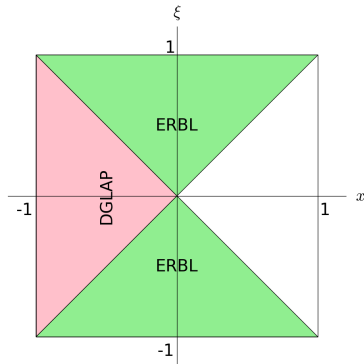
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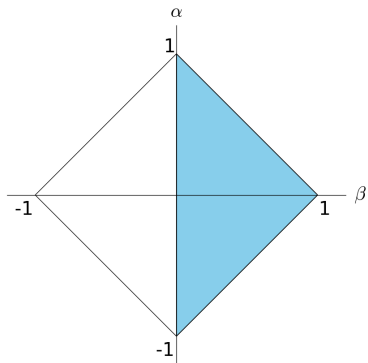
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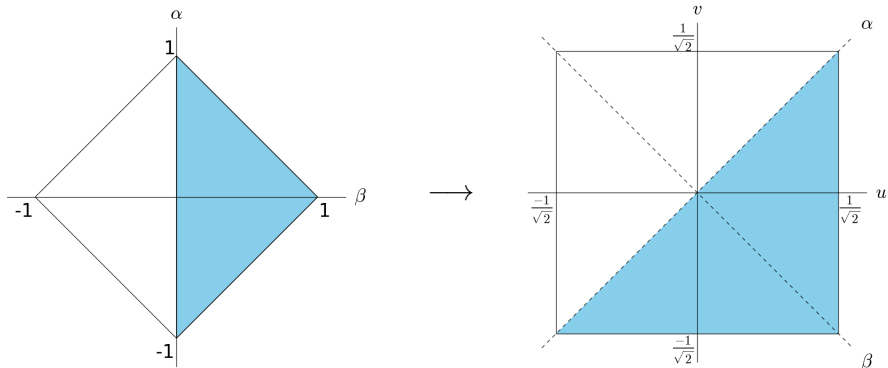

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- Domains  $\beta < 0$  and  $\beta > 0$  are uncorrelated in the DGLAP region.
- Divide and conquer:
  - ▶ Better numerical stability.
  - ▶ Lesser complexity:  $O(N^P + N^P) \ll O((N + N)^P)$ .

# Domain for the inversion



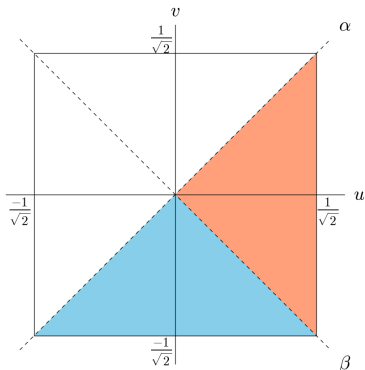
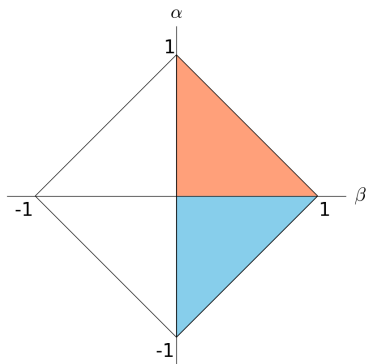
# Domain for the inversion



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

$$\begin{cases} u = \frac{\beta + \alpha}{\sqrt{2}} \\ v = \frac{\alpha - \beta}{\sqrt{2}} \end{cases}, \quad (14)$$

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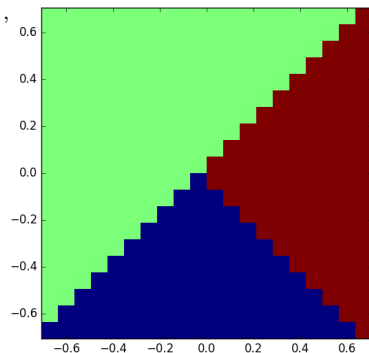
- $\alpha$ -parity of the DD:

$$f(\beta, -\alpha) = f(\beta, \alpha) \quad (15)$$

# Discretization

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

$$\tilde{f}(u, v) = \sum_{i=0}^{N-1} \sum_{j=0}^{N-1} \tilde{f}_{ij} \mathbf{1}_{[u_i, u_{i+1}]}(u) \mathbf{1}_{[v_j, v_{j+1}]}(v), \quad (16)$$

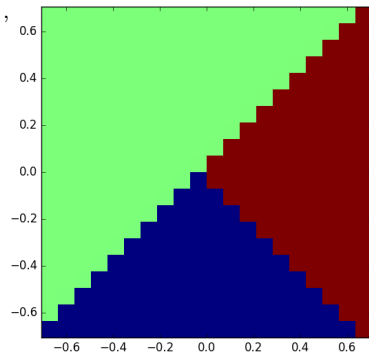


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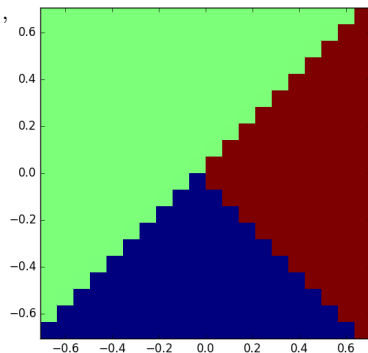


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  - Cells  $(u, v) \rightarrow n$  columns of the matrix.

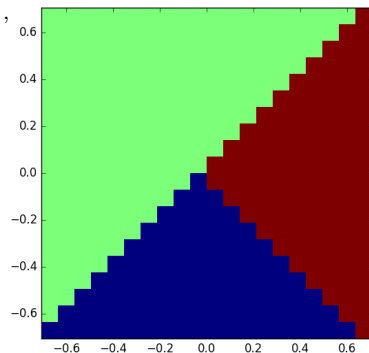


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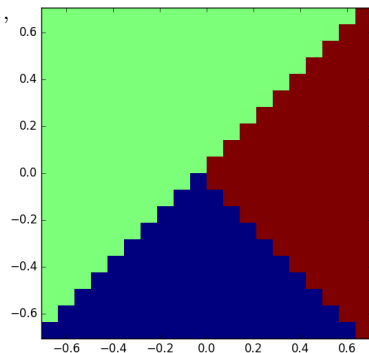


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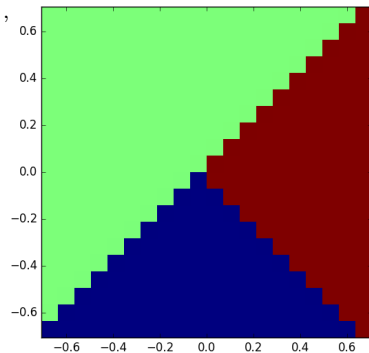


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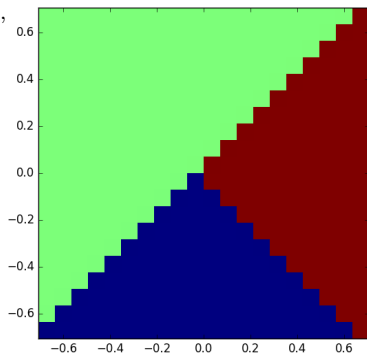


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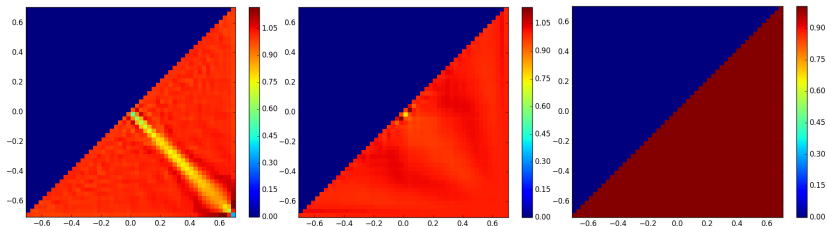
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  - A full-rank: more information but also more noise.



# Test (constant DD)



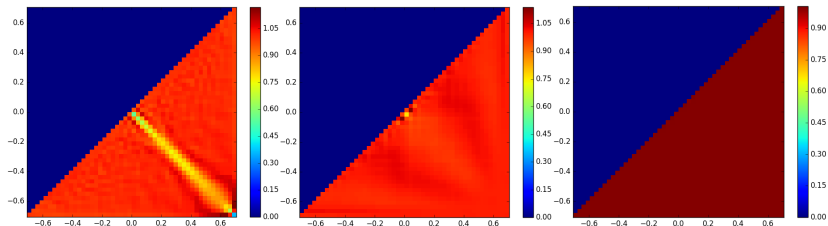
- Test with Constant DD.

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

↓

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

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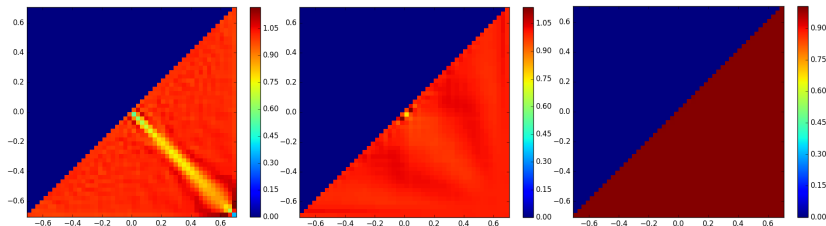
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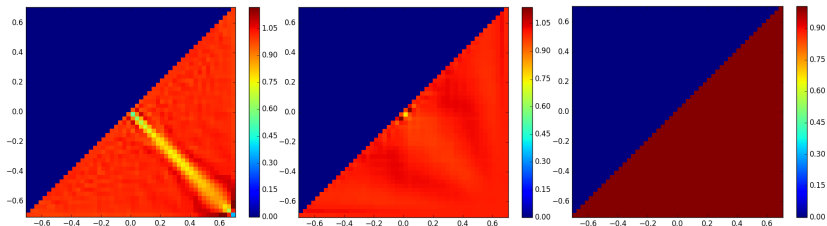
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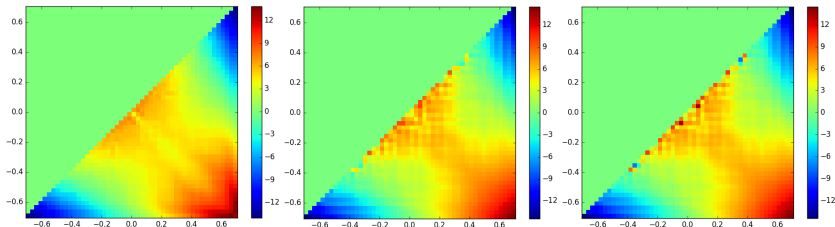
↓

- Consistent problem (discretized DD = theoretical DD):

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

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# First result



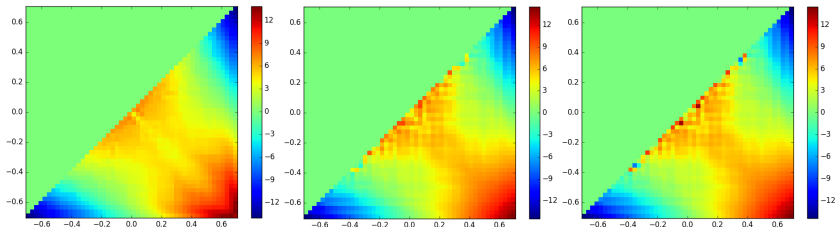
- Real application to a DSE toy model.

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

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

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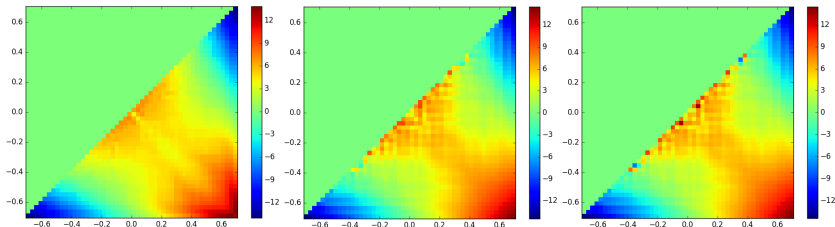
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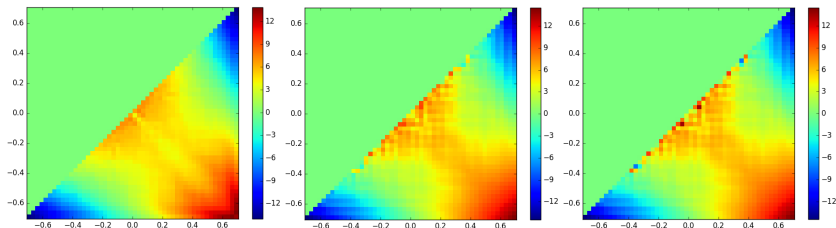
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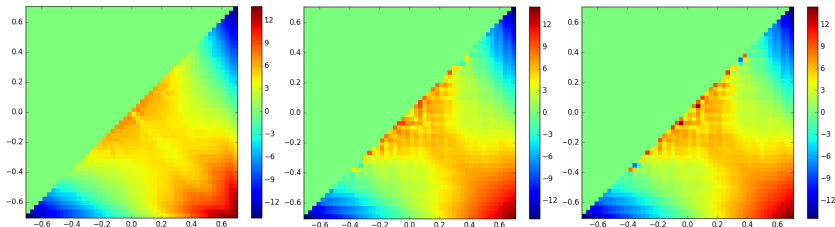
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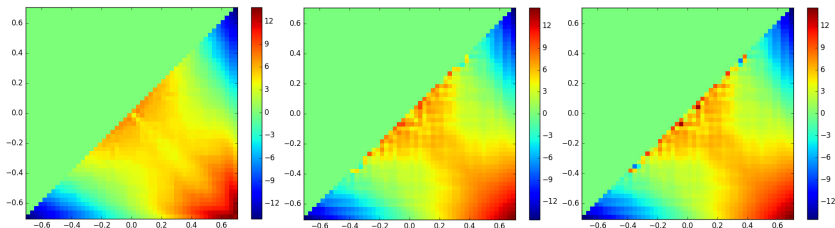
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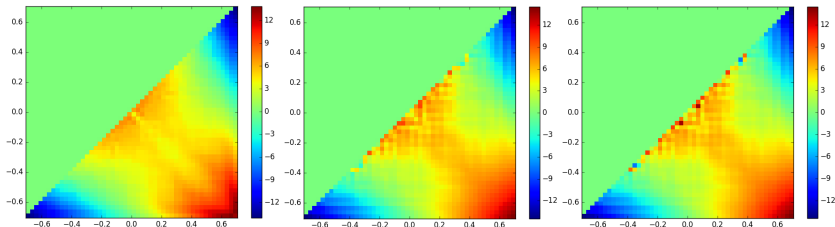
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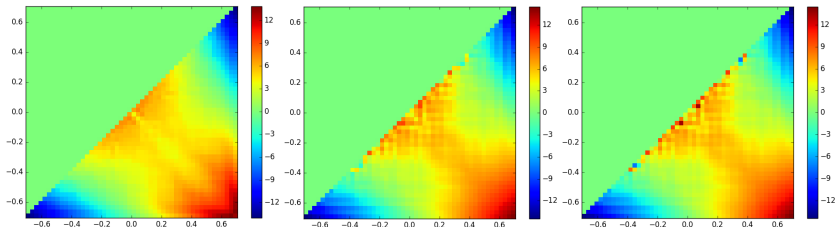
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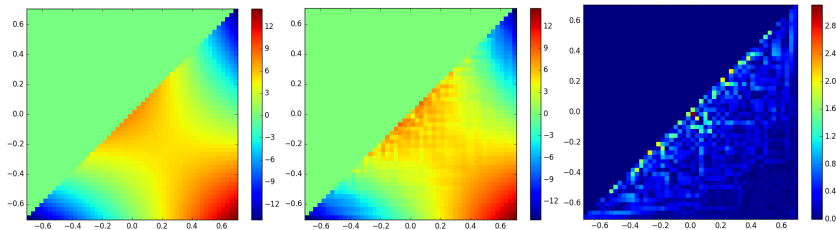
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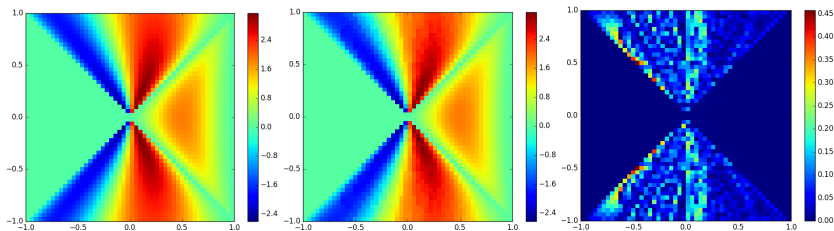
# Quantitative comparison of DDs



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

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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.

$$H(x, \xi) = \begin{cases} 30 \frac{(1-x)^2 (x^2 - \xi^2)}{(1-\xi^2)^2} & x > |\xi| \\ \frac{15(x-1)(x^2 - \xi^2)(\xi^2 + 2|\xi|x + x)}{2|\xi|^3(|\xi| + 1)^2} & |x| < |\xi| \end{cases}$$

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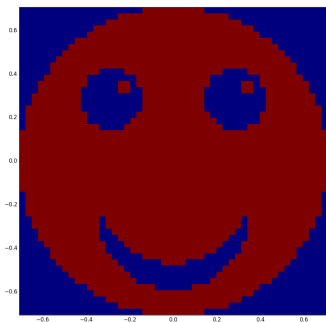
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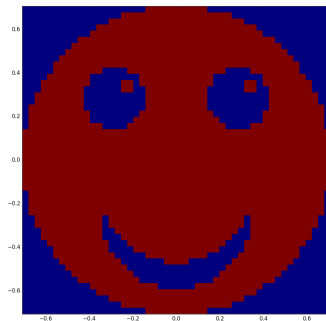
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- Important points:
  - ▶ Compromise with respect to noise and convergence.
  - ▶ Pobylitsa gauge is promising.
- In the future:
  - ▶ Different methods: Basis functions, Bayesian methods, etc.
  - ▶ Handling of errors.
- Thank you!



# Summary

- Extension of the DSE overlap toy model.
- Systematic procedure for GPD modeling from first principles:
  - ▶ LCWFs  $\xrightarrow{\text{Overlap}}$  GPD in DGLAP  $\xrightarrow{\text{Inverse Radon Transform}}$  DD  $\xrightarrow{\text{RT}}$  GPD.
  - ▶ Both polynomiality and positivity!
- Important points:
  - ▶ Compromise with respect to noise and convergence.
  - ▶ Pobylitsa gauge is promising.
- In the future:
  - ▶ Different methods: Basis functions, Bayesian methods, etc.
  - ▶ Handling of errors.
- Thank you!
  - ▶ Any questions?



# Bibliography I

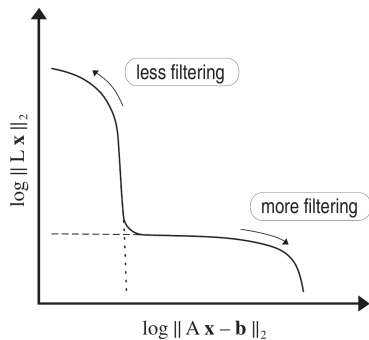
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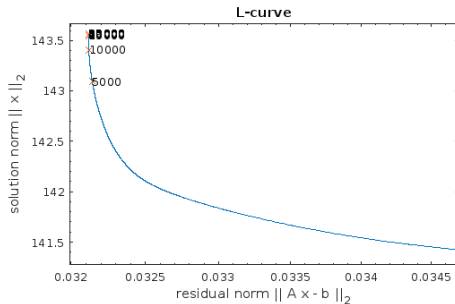
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# Discrete ill-posed problem



Theoretical “L-curve”: curve parameterized by the regularization factor.

(fig. taken from Ref. [\(Hansen, 2007\)](#))



L-curve with the iteration number as regularization factor.