

Status of the Gepard code

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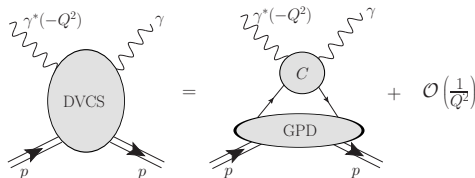


Outline

- ① Introduction to conformal space framework
- ② Software
- ③ Checks and benchmarks

Factorization of DVCS \longrightarrow GPDs

- [Collins et al. '98]



- CFFs are convolution:

$${}^a\mathcal{H}(\xi, t, Q^2) = \int dx T^a(x, \xi, \frac{Q^2}{Q_0^2}) H^a(x, \xi, t, Q_0^2)$$

 $a=q, G$

- $H^a(x, \eta, t, Q_0^2)$ — Generalized parton distribution (GPD)

Modelling GPDs in conformal moment space

- Instead of considering momentum fraction dependence $H(\mathbf{x}, \dots)$
- ... it is convenient to make a transform into complementary space of **conformal moments** j :

$$H_j^q(\eta, \dots) \equiv \frac{\Gamma(3/2)\Gamma(j+1)}{2^{j+1}\Gamma(j+3/2)} \int_{-1}^1 dx \eta^j C_j^{3/2}(x/\eta) H^q(\mathbf{x}, \eta, \dots)$$

- They are analogous to Mellin moments in DIS: $x^j \rightarrow C_j^{3/2}(x)$
- $C_j^{3/2}(x)$ — Gegenbauer polynomials

CFFs as Mellin-Barnes integral

$$\mathcal{H}(\xi, t, Q^2) = \frac{1}{2i} \int_{c-i\infty}^{c+i\infty} dj \xi^{-j-1} \left[i + \tan \left(\frac{\pi j}{2} \right) \right] \\ \times T_j(Q^2/\mu^2, \alpha_s(\mu)) H_j(\xi, t, \mu^2).$$

- Evolution of GPDs:

$$H_j(\eta, t, \mu) = \sum_k E_{jk}(\mu, \mu_0; \eta) H_k(\eta, t, \mu_0),$$

- T_j and E_{jk} known to (N)NLO
- \sum_k is infinite \rightarrow resummation leads to second Mellin-Barnes integral

Advantages of conformal space

- 1 The evolution equations are most simple: There is **no mixing** among moments at LO, and in special (\overline{CS}) scheme not even at NLO
- 2 Powerful analytic methods of **complex j** plane are available (similar to complex angular momentum of Regge theory)
- 3 Stable and fast **computer code** for evolution and fitting
- 4 Moments are equal to matrix elements of **local** operators and are thus directly accessible on the **lattice**

Disadvantages of conformal space

- ❶ Difficult conversion between x -space and j -space. E.g. [Müller and Schäfer '05] toy model:

$$H^{\text{toy}}(x, \eta) = \frac{1}{1-\alpha} \theta(-\eta \leq x) \frac{1}{\eta} \left(\frac{x+\eta}{1+\eta} \right)^{1-\alpha}$$

$$H_j^{\text{toy}}(\eta) \propto \frac{(1+j-\alpha)(2+j-\alpha)}{2(1-\alpha)(2-\alpha)} \frac{\Gamma(3/2)\Gamma(3+j)}{\Gamma(3/2+j)} \\ \times \frac{\eta-1}{2\eta} \left(\frac{\eta}{2}\right)^j {}_3F_2\left(\begin{matrix} -j, 3+j, 2-\alpha \\ 2, 3-\alpha \end{matrix} \middle| \frac{\eta-1}{2\eta}\right)$$

- ❷ Need complex-analytic continuation of functions used for modelling. How to work with numerical models, neural nets?

Gepard* software



Gepard

* Not to be confused with biology software (GENome PAir Rapid Dotter)

Gepard — features

- Gepard implements [Belitsky, Müller et al.] DVCS formulas for all measured DVCS observables
- Implemented GPD/CFF models
 - Goloskokov-Kroll (GK)
 - K.K. and D. Müller (KM)
 - Neural networks
- Fitting procedures
 - MINUIT least-squares fitting
 - Neural networks (PyBrain and TensorFlow (experimental))
- **interactive** work possible (jupyter notebooks)

GeParD — interactive example 1/7

```
In [12]: m = Model.ComptonGepard(p=1, q02=2.)      # initialize NLO KM model for CFFs
         th = Approach.BMK(m)                    # use BMK formulas for DVCS
         th.name = 'NLO preliminary'            # name to be used on plots
```

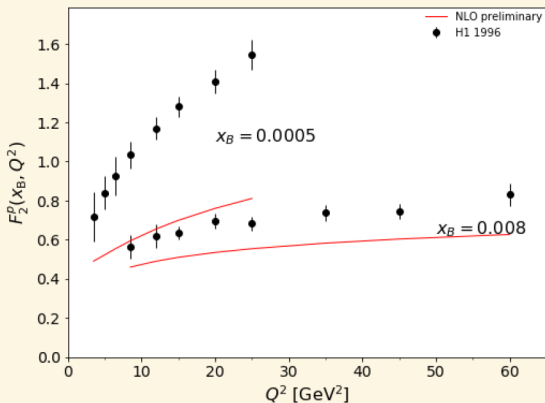
```
In [13]: th.m.parameters['AL0S'] = 1.2          # change some model parameter
```

```
In [14]: utils.describe_data(DISpoints)        # DISpoints = data set
```

npt	x	obs	collab	FTn	id	ref.
8	x	F2	H1	N/A	201	Nucl.Phys.B470(96)3
8	x	F2	H1	N/A	202	Nucl.Phys.B470(96)3
9	x	F2	H1	N/A	203	Nucl.Phys.B470(96)3
9	x	F2	H1	N/A	204	Nucl.Phys.B470(96)3

GeParD — interactive example 2/7

```
In [41]: fig = plots.HERAF2Q2(lines=[th]) # specifically designed plots
```



```
In [18]: th.print_chisq(DISpoints)
```

P(chi-square, d.o.f) = P(3787.21, 85) = 0.0000

GeParD — interactive example 3/7

```
In [19]: %%time
th.model.release_parameters('NS', 'AL0S', 'AL0G')
f = Fitter.FitterMinuit(DISpoints, th)
f.fit()
```

FCN = 72.85271021975446	TOTAL NCALL = 115	NCALLS = 115
EDM = 1.082463486221191e-06	GOAL EDM = 1e-05	UP = 1.0

Valid	Valid Param	Accurate Covar	PosDef	Made PosDef
True	True	True	True	False
Hesse Fail	HasCov	Above EDM		Reach callim
False	True	False		False

±	Name	Value	Hesse Error	Minos Error-	Minos Error+	Limit-	Limit+	Fixed?
0	NS	0.149507	0.00706756					No
1	AL0S	1.06274	0.0197111					No
2	ALPS	0.15		1				Yes

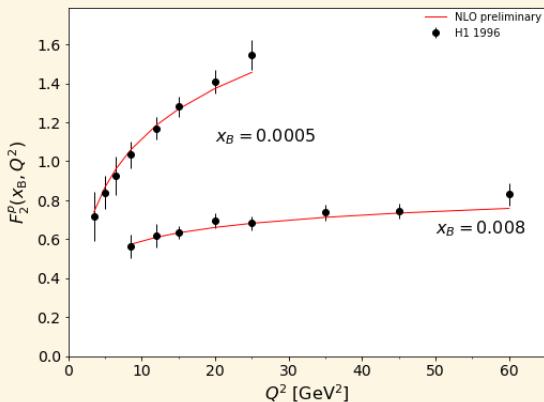
- Two “fitters” are implemented: Minuit and Neural network

GeParD — interactive example 4/7

```
In [20]: th.print_chisq(DISpoints)
```

```
P(chi-square, d.o.f) = P(72.85, 82) = 0.7549
```

```
In [21]: fig = plots.HERAF2Q2(lines=[th])
```



GeParD — interactive example 5/7

In [22]: `utils.describe_data(DVMPpoints)`

```
npt x obs    collab FTn    id ref.
-----
 5 x X      H1      N/A   76 arXiv:0910.5831
20 x X      H1      N/A   79 arXiv:0910.5831
-----
TOTAL = 25
```

Out[22]: 25

In [23]: `th.print_chisq(DVMPpoints)`

P(chi-square, d.o.f) = P(1752.85, 22) = 0.0000

In [24]: `th.model.fix_parameters('NS', 'AL0S', 'AL0G')`
`th.model.release_parameters('M02S', 'M02G', 'SECS', 'SECG', 'THIS', 'THIG')`

In [25]: `%time`
`f = Fitter.FitterMinuit(DVMPpoints, th)`
`f.fit()`

CPU times: user 9min 55s, sys: 582 ms, total: 9min 55s
 Wall time: 25.7 s

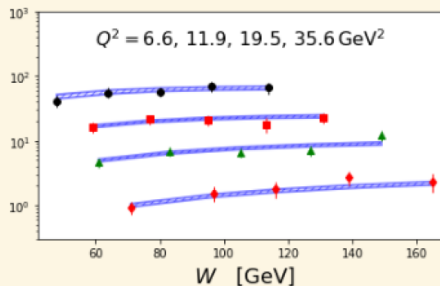
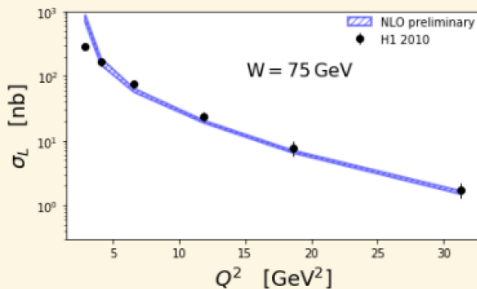
In [30]: `th.print_chisq(DISpoints+DVMPpoints)`

P(chi-square, d.o.f) = P(85.51, 104) = 0.9066

GeParD — interactive example 6/7

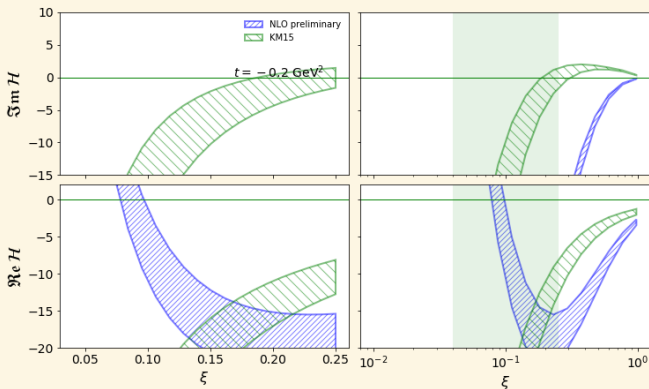
```
fig = plots.DVMP(wdep, bands=[th]).
```

H1 DVMP



GeParD — interactive example 7/7

```
In [33]: fig = plots.CFF(bands=[th, KM15])
```



Code rewrite

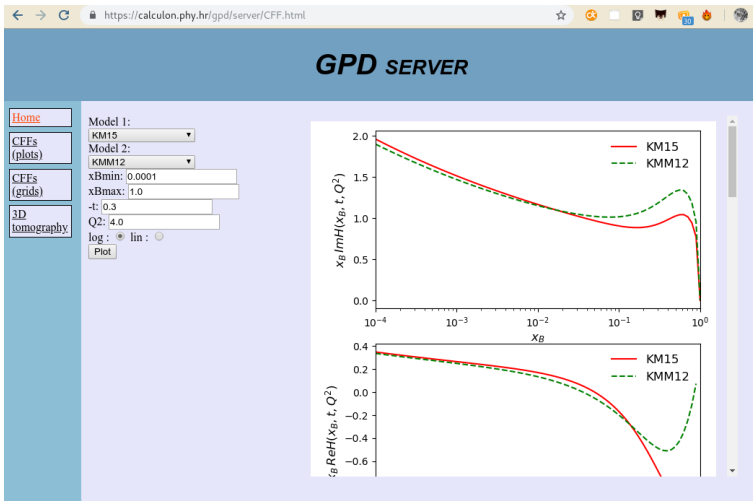
Code is presently in the transition from this:

Language	files	blank	comment	code
Python	46	2289	2865	13061
Fortran 77	120	7822	30240	12190
IPython Notebook	2	0	0	1956
C	5	161	263	454
Mathematica	1	124	15	415
make	4	101	133	296
C/C++ Header	2	40	74	91
DOS Batch	1	0	0	44
Markdown	1	3	0	5
SUM:	182	10540	33590	28512

to this:

Language	files	blank	comment	code
Python	17	1426	1581	5392
SUM:	17	1426	1581	5392

GPD/CFF server



GPD/CFF server

GPD SERVER

Model: KM15

xBmin: 0.01

xBmax: 0.9

-t: 0.3

Q2: 4.0

log: lin

npts: 40

Print

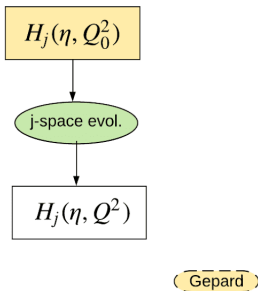
Model: KM15

#	xB	t	Q2	ImH	ReH	ImE	ReE	ImHt	ReHt	ImEt
0.01	-0.3	4.0	116.7022	17.0661	0.0000	1.9001	3.6173	1.0873	0.0000	
0.0112	-0.3	4.0	102.6172	14.5804	0.0000	1.9001	3.5411	1.0715	0.0000	
0.0126	-0.3	4.0	90.2295	12.3970	0.0000	1.9001	3.4661	1.0565	0.0000	
0.0141	-0.3	4.0	79.3347	10.4791	0.0000	1.9001	3.3922	1.0424	0.0000	
0.0159	-0.3	4.0	69.7531	8.7944	0.0000	1.9001	3.3192	1.0290	0.0000	
0.0178	-0.3	4.0	61.3271	7.3148	0.0000	1.9001	3.2473	1.0165	0.0000	
0.02	-0.3	4.0	53.9178	6.0158	0.0000	1.9001	3.1762	1.0048	0.0000	1
0.0224	-0.3	4.0	47.4029	4.8758	0.0000	1.9001	3.1059	0.9940	0.0000	
0.0252	-0.3	4.0	41.6746	3.8757	0.0000	1.9001	3.0363	0.9840	0.0000	
0.0282	-0.3	4.0	36.6384	2.9984	0.0000	1.9001	2.9674	0.9750	0.0000	
0.0317	-0.3	4.0	32.2114	2.2292	0.0000	1.9001	2.8990	0.9668	0.0000	
0.0356	-0.3	4.0	28.3207	1.5553	0.0000	1.9001	2.8311	0.9596	0.0000	
0.0399	-0.3	4.0	24.9018	0.9653	0.0000	1.9001	2.7636	0.9533	0.0000	
0.0448	-0.3	4.0	21.8983	0.4496	0.0000	1.9001	2.6964	0.9480	0.0000	

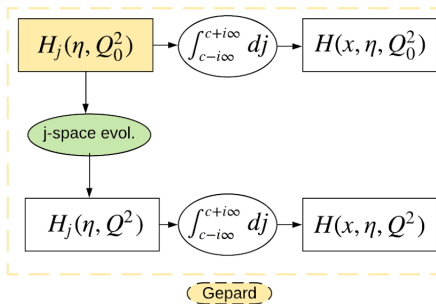
- Gepard code itself is on github.com, but not public yet

Checks and benchmarks

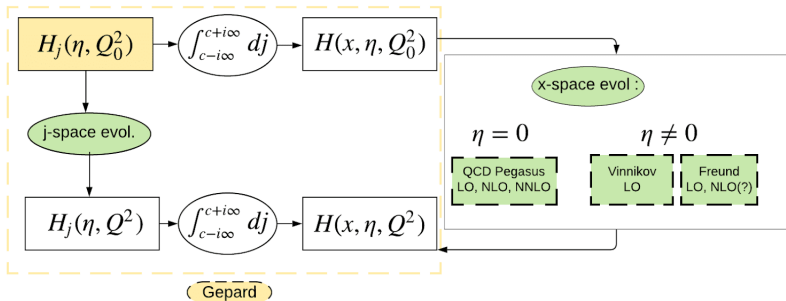
Testing evolution



Testing evolution

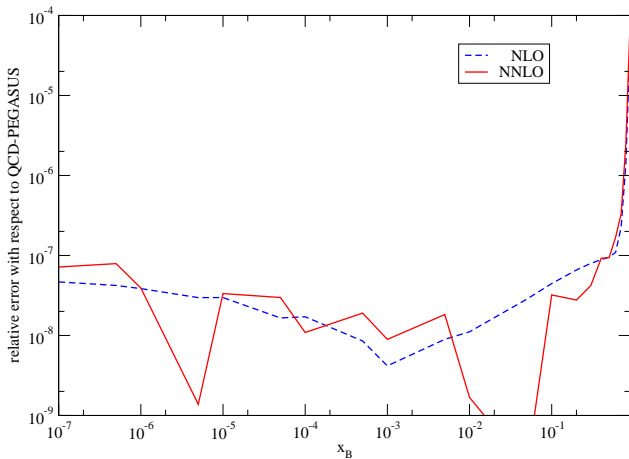


Testing evolution



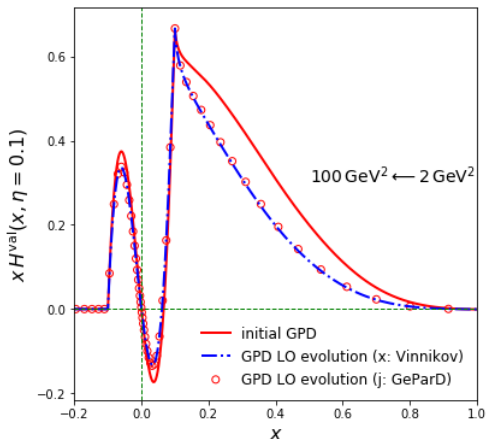
Checking evolution in forward limit

- Comparison to **QCD-Pegasus** PDF evolution software [A. Vogt '04]



Checking LO evolution

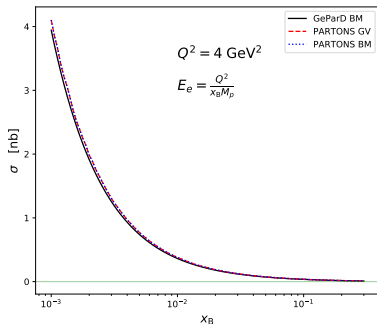
- Comparison to GPD evolution software [Vinnikov '06]



- Maximal relative discrepancy: 2 %.

Gepard vs PARTONS, cross-section

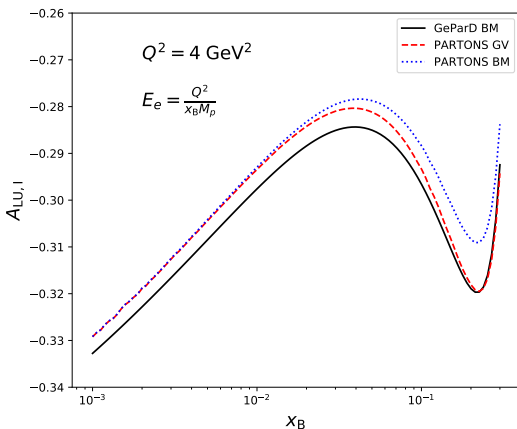
- BM = [Belitsky & Müller], GV = [Guichon & Vanderhaeghen]



- 100 kinematical points:
 - Gepard (python): **24 seconds**
 - PARTONS (C++, XML driven): **9 seconds**

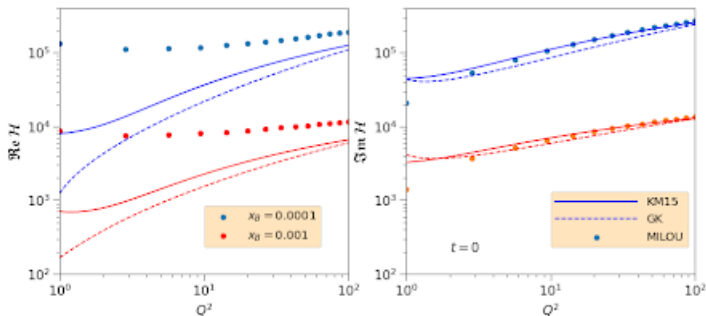
Gepard vs PARTONS, beam spin asymmetry

- BM = [Belitsky & Müller], GV = [Guichon & Vanderhaeghen]



Comparison of popular models

- GK = [Goloskokov & Kroll]
- KM = [K.K. & Müller]
- MILOU = DVCS MC generator, model by [Freund & McDermott]



Outlook

- Implementation of DVMP — **work in progress**
- Going from hybrid to full conformal-space GPD model — **work in progress**
- All the components for the NLO analysis are available — **work in progress**

The End