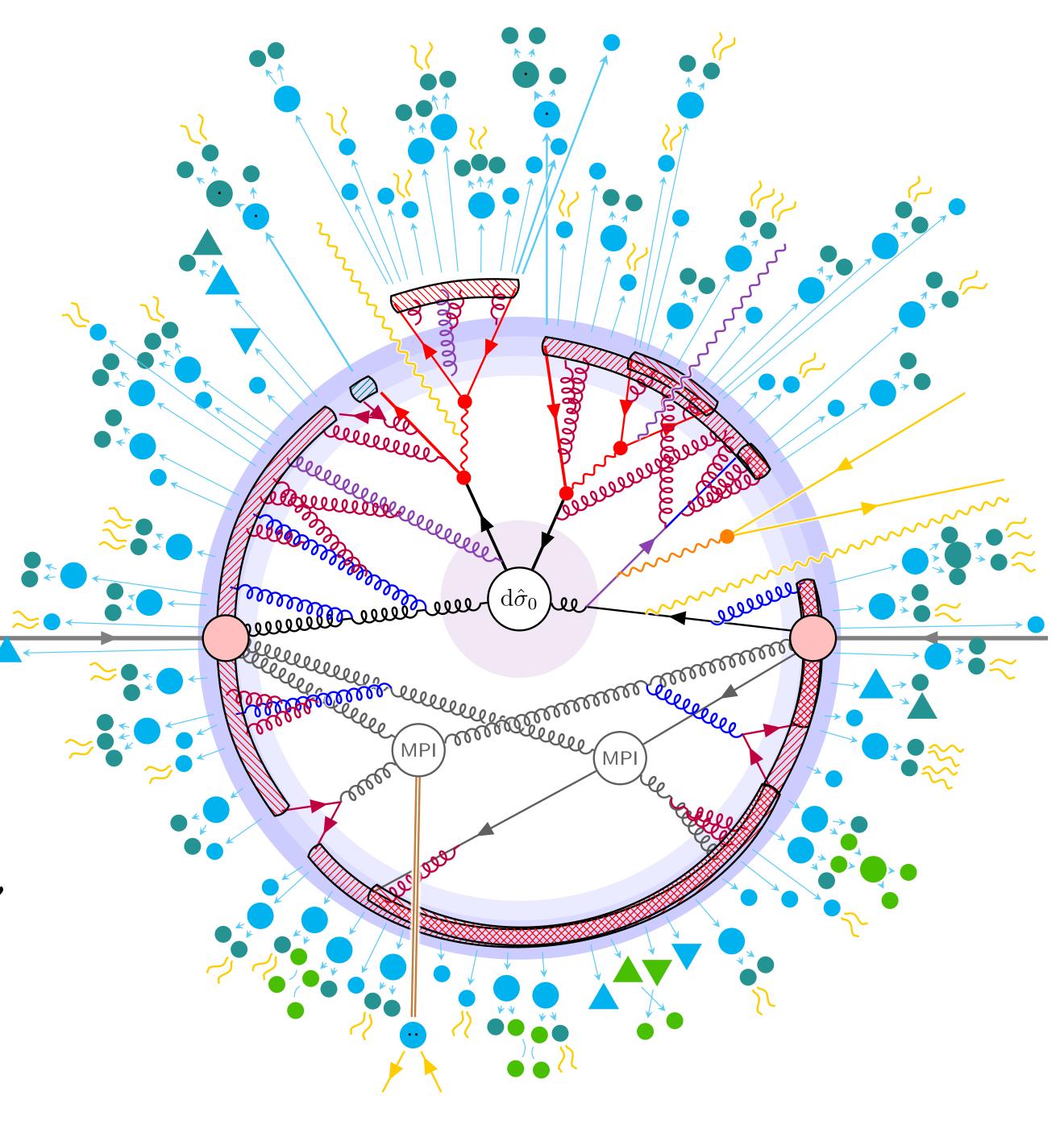
Updates towards Hyperoptimizing on PDF Distributions





Tanjona R. Rabemananjara w/ J. M. Cruz-Martinez, J. Rojo, R. Stegeman, and ASDI eScience Engineer

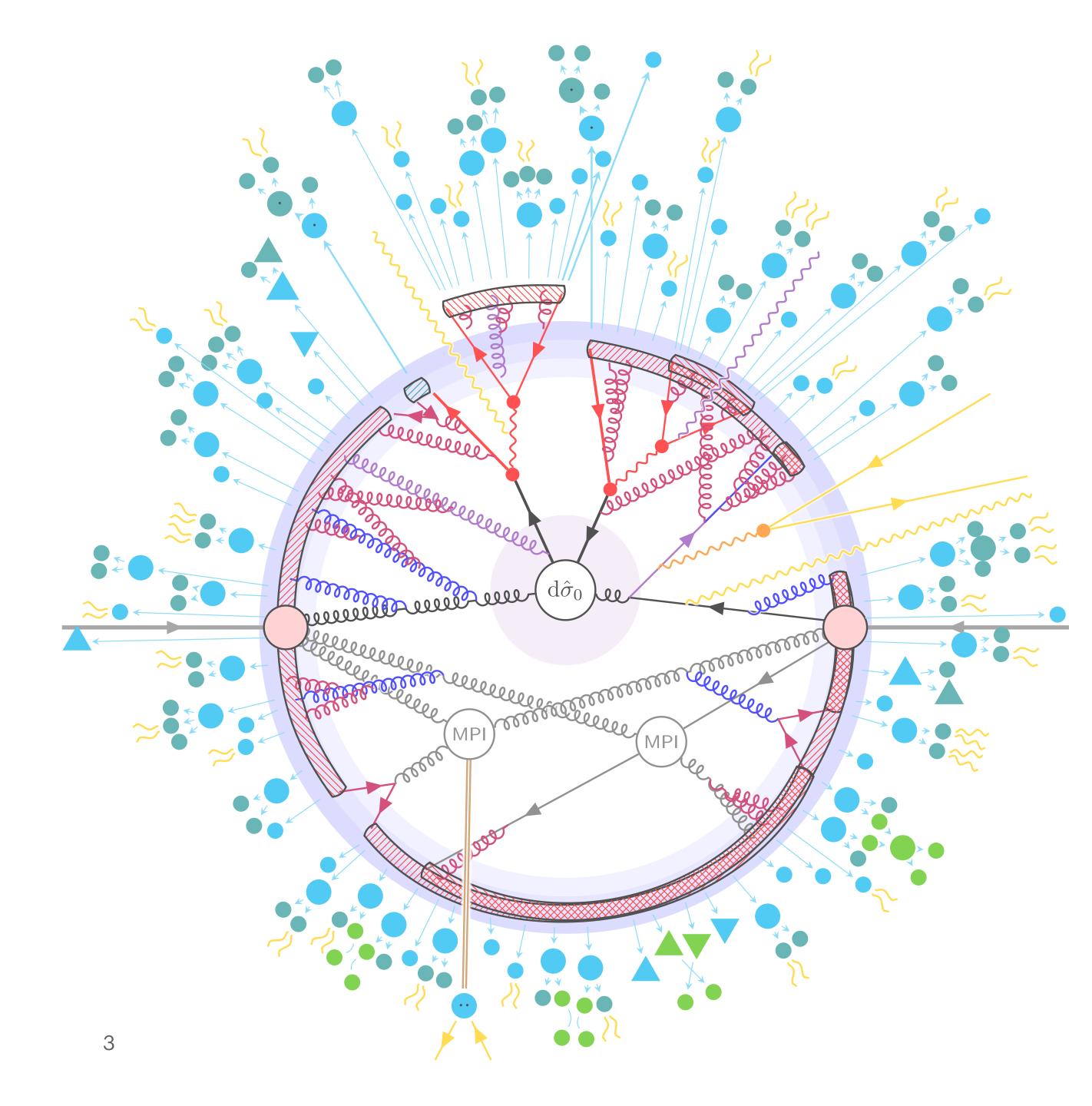


Hyperoptimizing PDF Distributions:

- 1. On the χ^2 of the mean
- 2. On the "Roy-ness"
- 3. On the bias/variance ratio in CT
- 4. On $|\chi^2 1|$ for non-fitted folds
- 5. Combinations of the above
- 6. More ideas ...?

SHORT OUTLINE OF THE TALK:

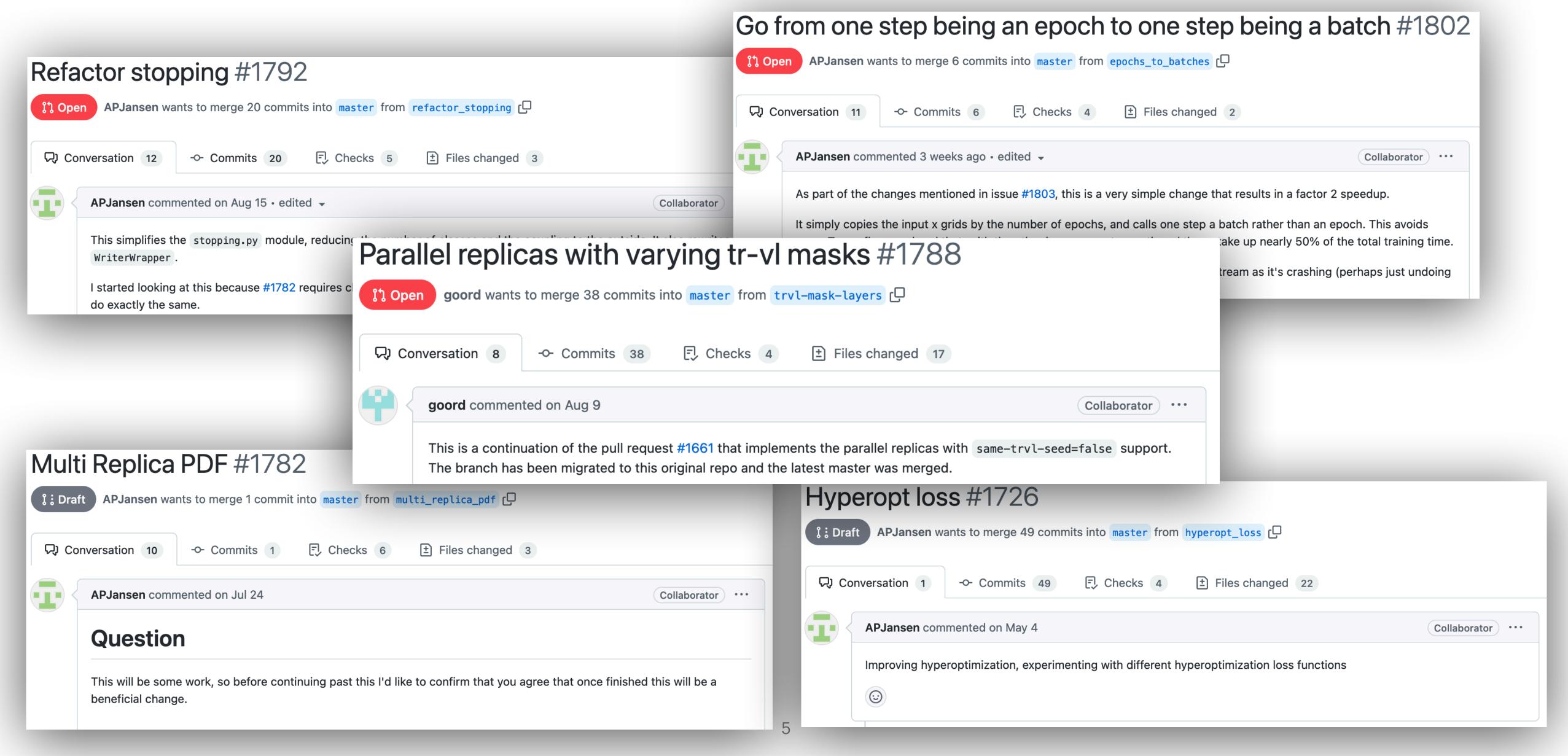
- 1. Technical Developments
- 2. The Curse of Hyperoptimizing on Distributions
- 3. Taming the Issues
- 4. Preliminary Results
- 5. Short-term Plan & Outlook



Multiple Refactoring (Several PRs Merged Already)



Many more Improvements still Ongoing (crucial for next steps!)



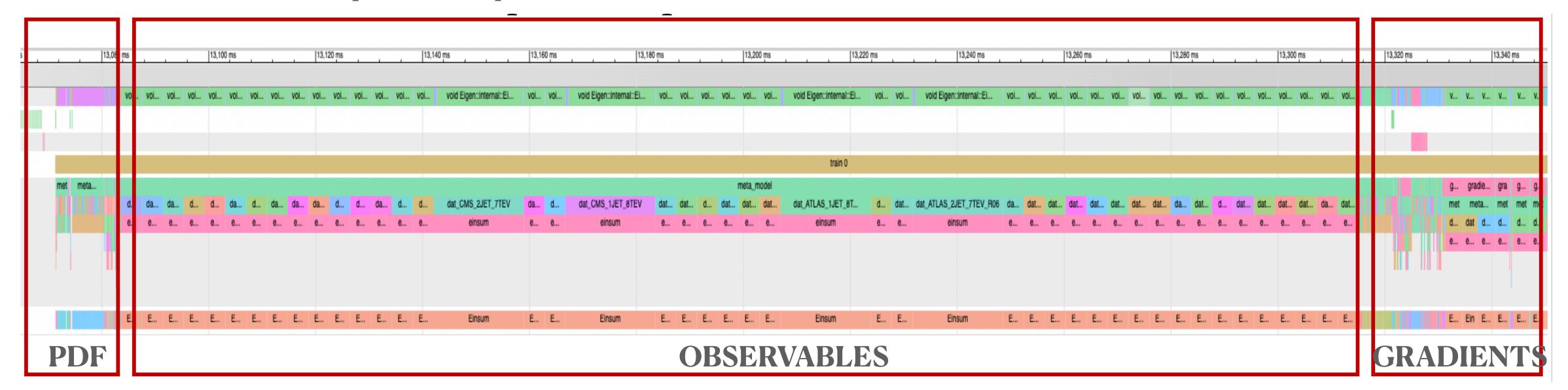
The Curse of Hyperoptimizing Distributions

Hyperopt with full distribution in a Global Fit:

\tilde{I} —	1	1	$n_{\text{fold}} N_{\text{rep}}$
<i>L</i> –	$\overline{n_{ m fold}}$	$\overline{N_{\rm rep}}$	$\sum_{k=1}^{\infty} \sum_{r=1}^{\infty} \chi_{k,r}$

Nb. Replicas	Nb. GPUs	Nb. Trials	Time (s)	SUB/Trial	Projected Full Hyperopt (s)
50	1	2	693	2.5k	360k
100	1	2	965	3k	680k

Performance of 100 replicas setup with Tensorboard:



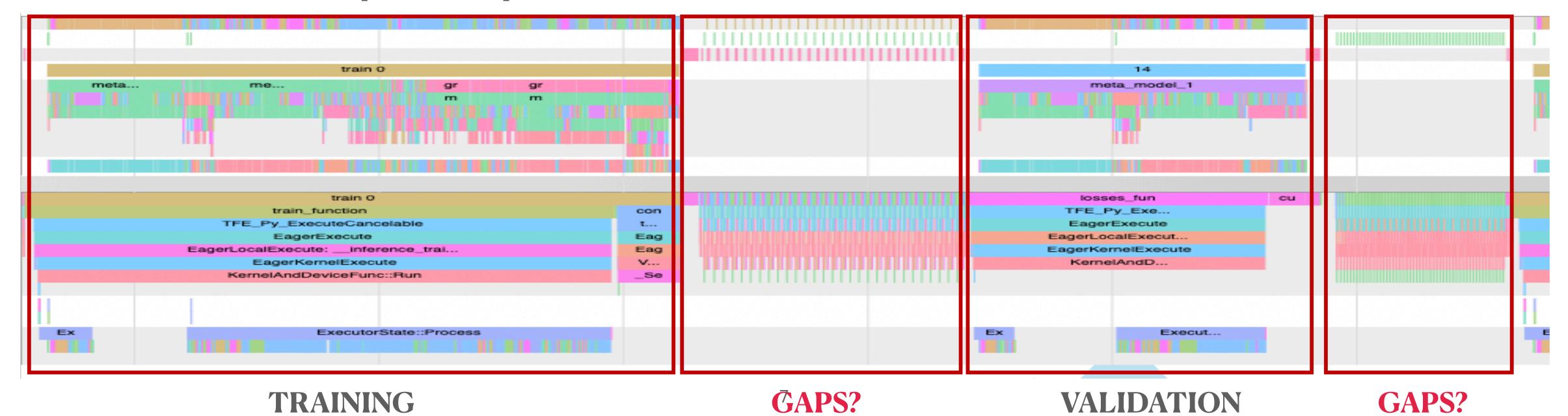
The Curse of Hyperoptimizing Distributions

Hyperopt with full distribution in a Global Fit:

$$\tilde{L} = \frac{1}{n_{\text{fold}}} \frac{1}{N_{\text{rep}}} \sum_{k=1}^{n_{\text{fold}}} \sum_{r=1}^{N_{\text{rep}}} \chi_{k,r}^2$$

Nb. Replicas	Nb. GPUs	Nb. Trials	Time (s)	SUB/Trial	Projected Full Hyperopt (s)
50	1	2	693	2.5k	360k
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Performance of 100 replicas setup with Tensorboard:

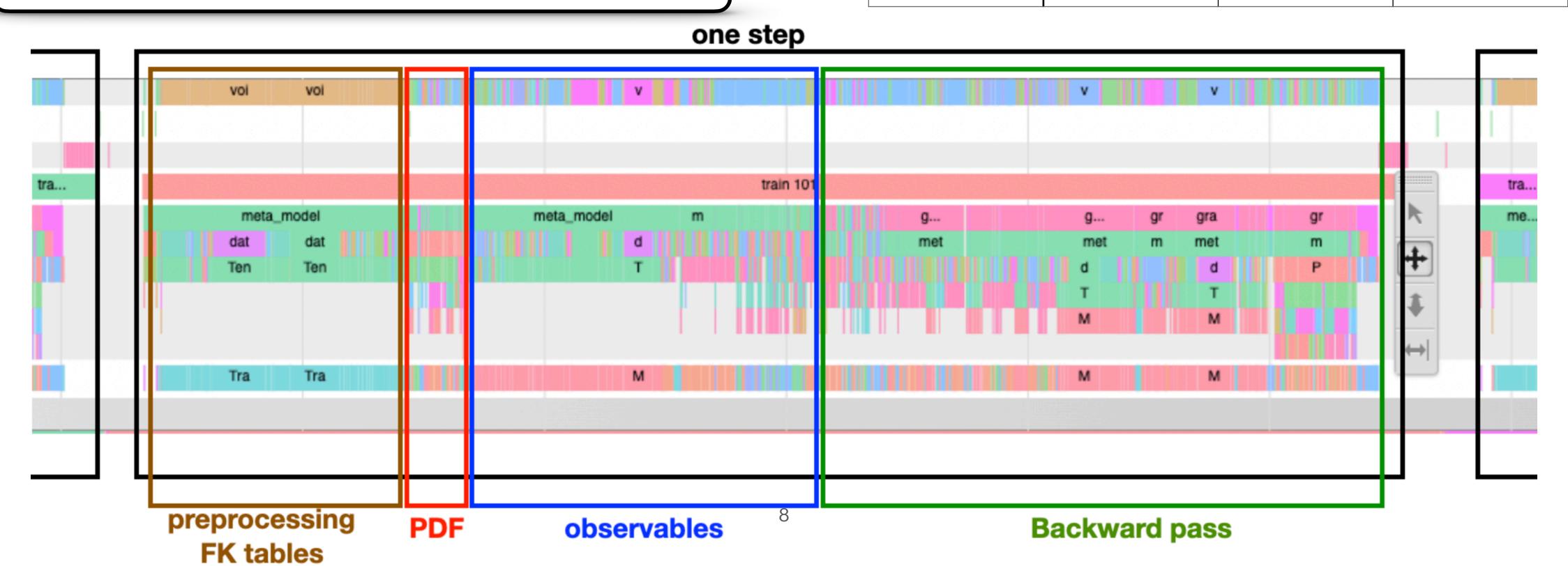


Clever Solutions

- 1. Re-order indexes for Tensor Multiplications
- 2. Fixed Shapes of Masks (Crucial for DY FK tables)
- 3. Using a **Single PDF Model** for all the replicas (similar to what is currently done for flavours)
- 4. Improve computation of the Validation

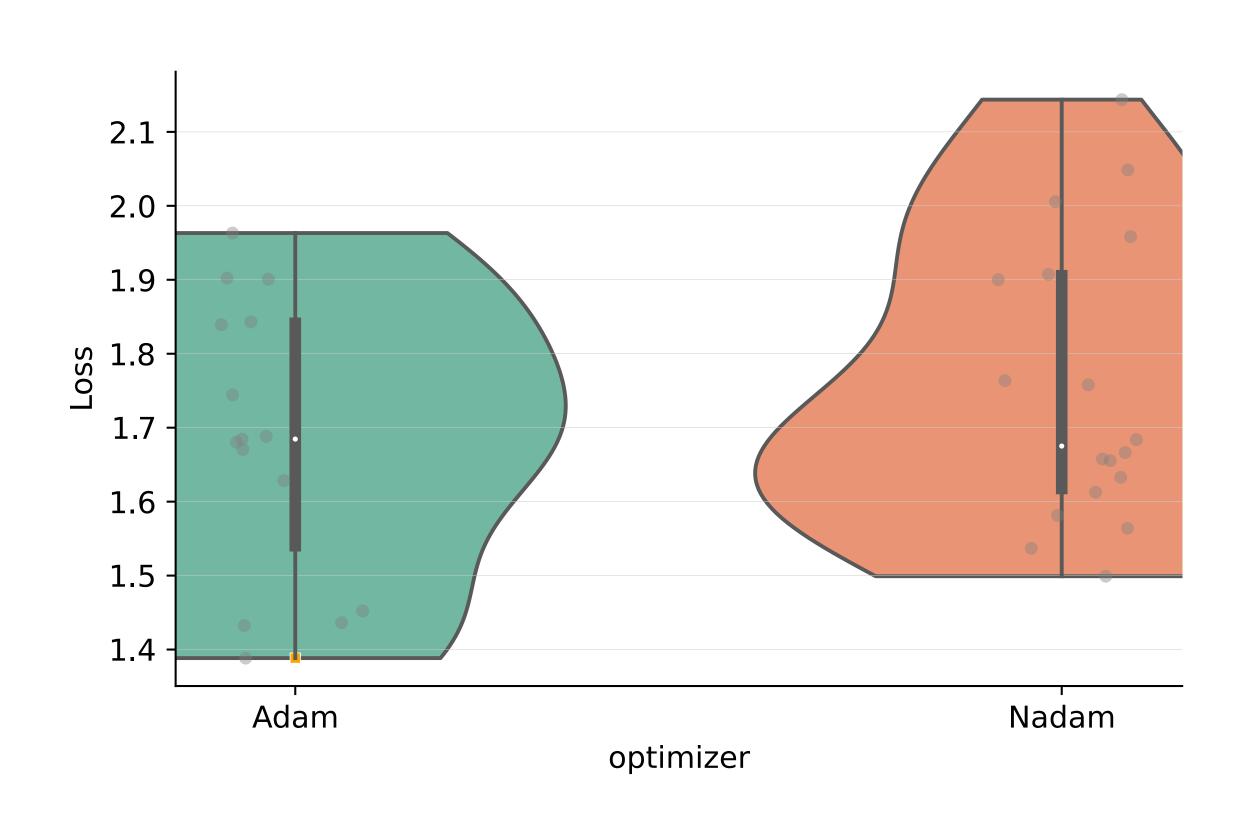
Full HyperOpt w/ 100 replicas & 5 Folds

	Time (mn)	Hrs/Trial	SBU/Trial
No. Improv.	28	47	6k
1 ⊕ 2	11	19	2k
⊕ 3	2.5	4	530
⊕ 4	1.3	1.6	220



New Hyperparameters Configurations

Parameter	NNPDF4.0	
Architecture 2-14-42-8		2-25-20-8
Activation function		hyperbolic tangent
Initializer	${\tt glorot_normal}$	
Optimizer	Adam	Nadam
Clipnorm	2.8×10^{-6}	6.0×10^{-6}
Learning rate	1.2×10^{-3}	2.6×10^{-3}
Maximum # epoch	17×10^3	
Stopping patience	10% of max epochs	
Initial positivity $\Lambda^{(}$	185	
Initial integrability	10	

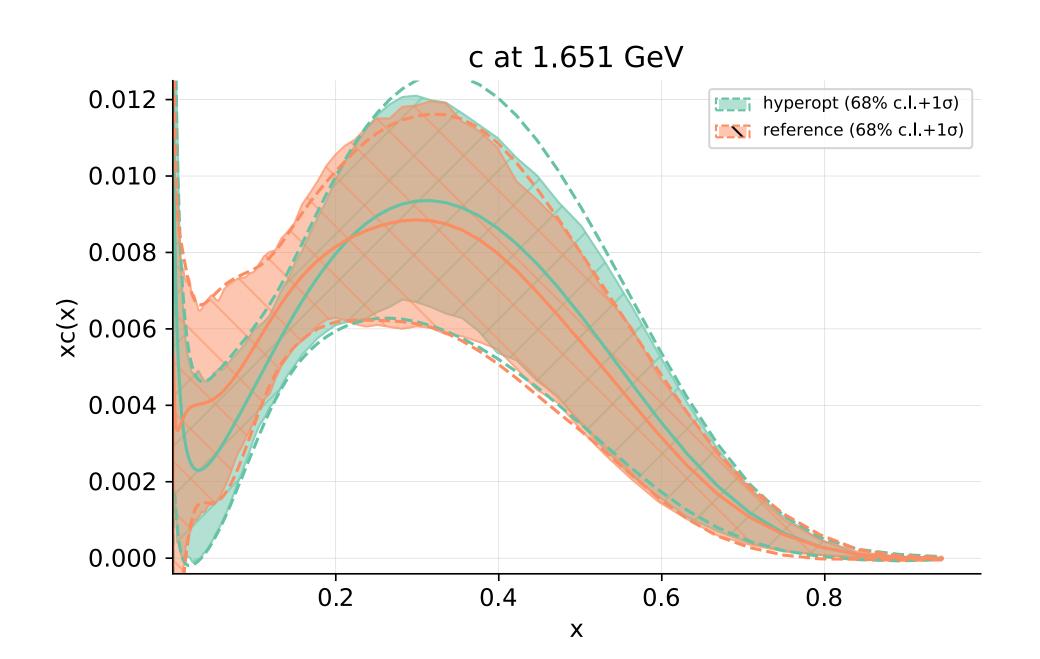


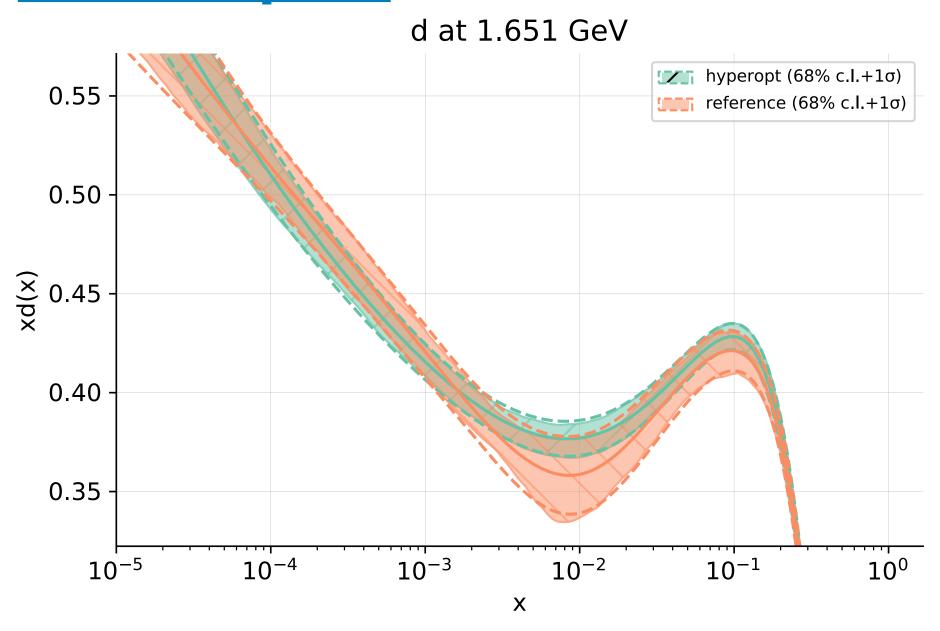
Initial integrability $\Lambda^{(int)}$ 10

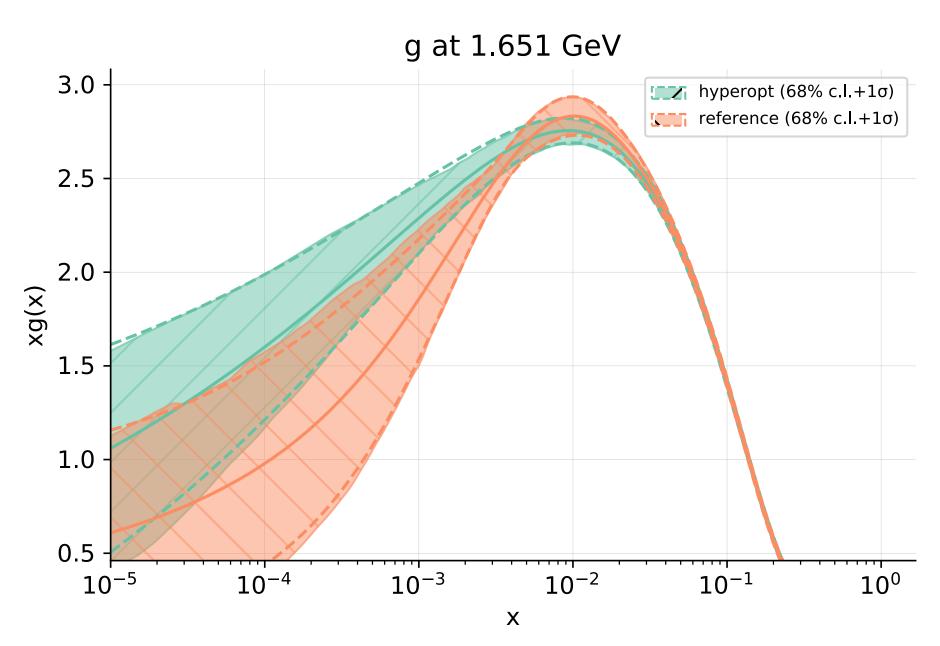
PDF Fit Results (Full Report)

10

	hyperopt	reference
χ ²	1.19543	1.17051
< E _{trn} >	2.281±0.061	2.256±0.059
< E _{val} >	2.37±0.10	2.36±0.10
< TL>	15500±2300	13100±2500
< x ² >	1.212±0.013	1.194±0.015
φ	0.1289±0.0051	0.1549±0.0048

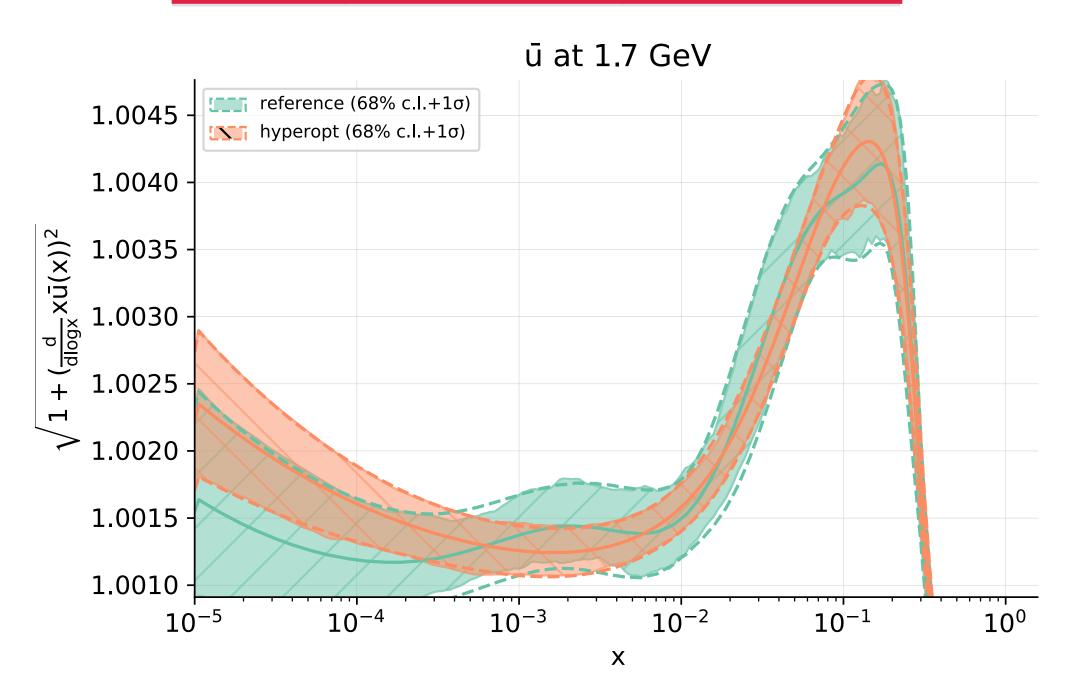


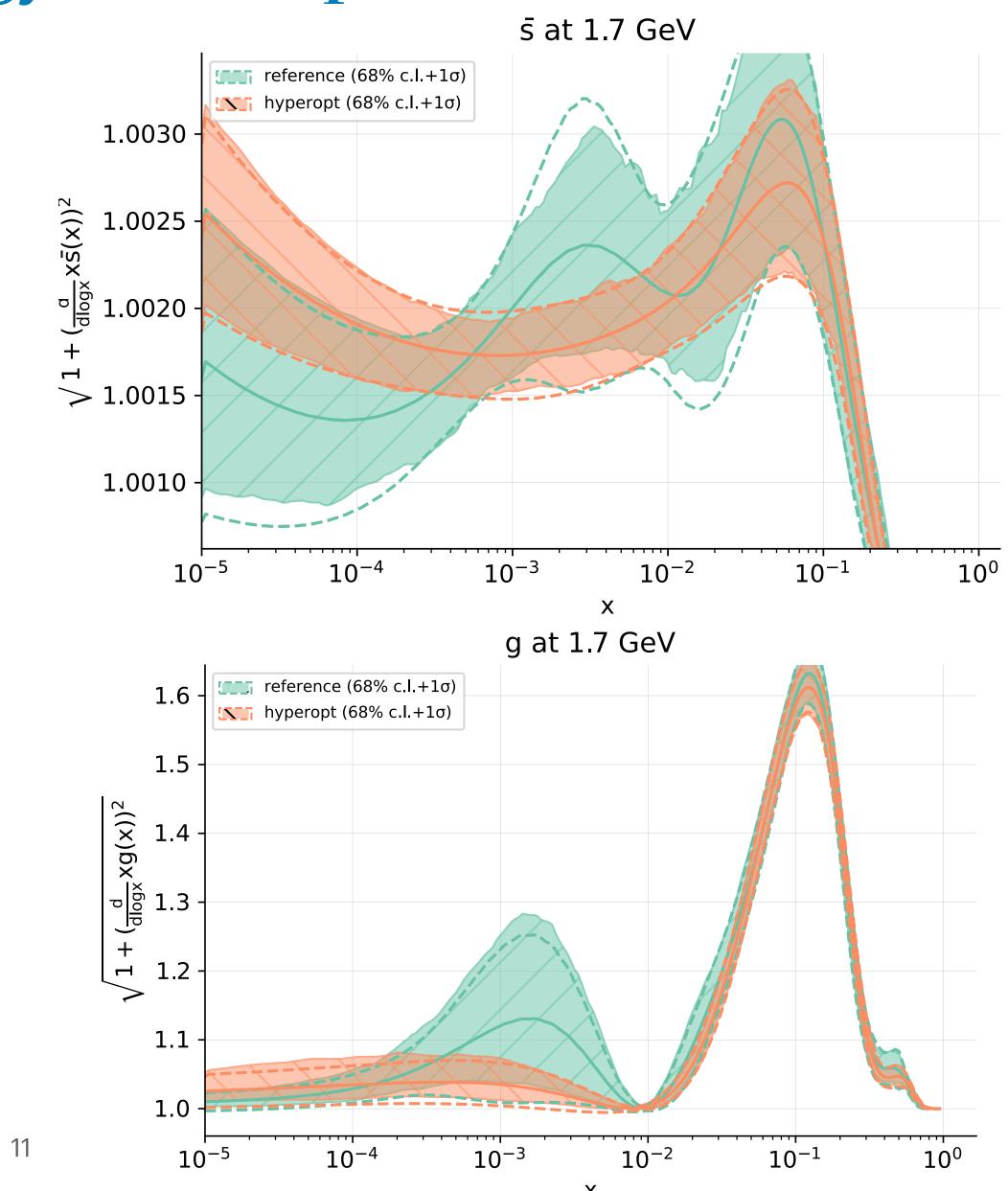




Kinetic Energy (Full Report)

	hyperopt	reference
χ ²	1.19543	1.17051
< <i>E</i> _{trn} >	2.281±0.061	2.256±0.059
< E _{val} >	2.37±0.10	2.36±0.10
< TL>	15500±2300	13100±2500
< x ² >	1.212±0.013	1.194±0.015
φ	0.1289±0.0051	0.1549±0.0048





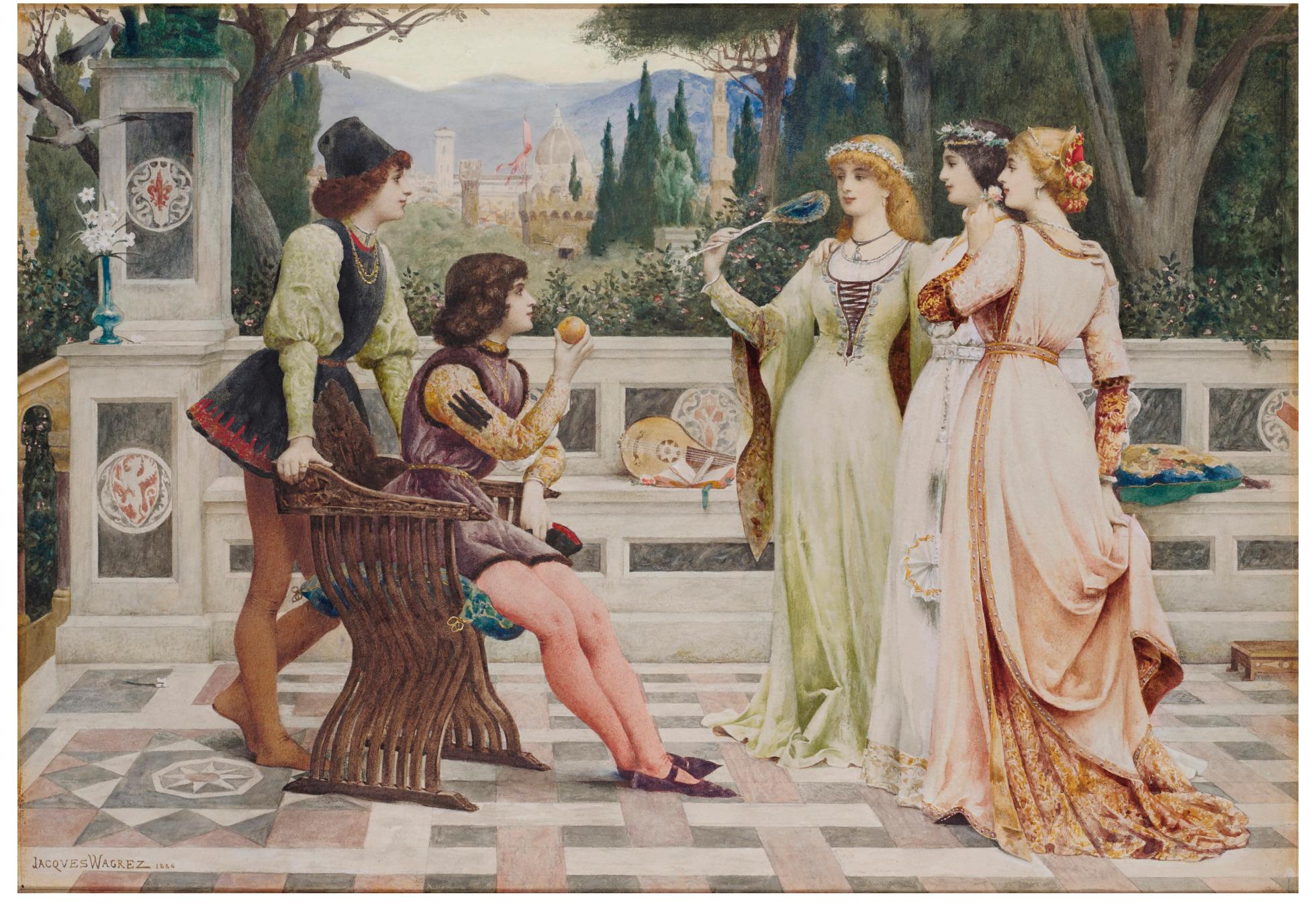
Short-Term Plan & Outlook

- O Finish implementation of all technical improvements
- O Closure Tests / Future Tests / etc.
- O New ideas/propositions?

Steps remaining

Unfortunately I'll have very little time in the next month to work on this (holidays and other projects). Below I'll list the steps necessary to integrate it, and where help could be useful.

- 1. Get the trvl-mask-layers branch to pass the tests.
- Some help here would be useful, @Radonirinaunimi would take a look
- 2. The changes in point 1 above are in a branch gpu-go-brrr (;P), off of trvl-mask-layers. This can be merged into it.
- 3. Once merged they should be tested and reviewed.
- What would be super useful here is to have a list of runcards to test, and also the actual results from master. Since this is already much faster, it's a lot easier if we can only run them on this branch, and have something to compare against already.
- 4. The rewriting from epochs to batches is independent of all the other changes. If anyone wants to pick that up that'd be great. I started it in !\ Go from one step being an epoch to one step being a batch #1802
- 5. The rewriting of the 3 "models" into one and instead 3 losses (By which btw I don't necessarily mean that we put that part in an actual keras. Loss or something, not sure if that's efficient or not, just to not repeat the computations), I think this is also relatively independent of the rest. If anyone wants to do this that'd be great, it doesn't have the highest payout/effort ratio of all these, so I can also do it myself after the last point.
- 6. The multi replica PDF, this is the most work, and the most specialized, so I think it's best if I focus on this. It's not very far I think, I need to look at \\ \frac{1}{2}\) Refactor stopping #1792 again, then I hope write another PR that concatenates the PDFs as soon as possible and makes the remaining logistics around it compatible but hopefully remain numerically identical, and finally drop in the actual joint model. This then needs to be tested again, but since I'll be able to reuse the results from master at step 3 it should be quick.



The Judgement of Paris by Jacques-Clément Wagrez