

Phenomenology of hadron production in e^+e^- collisions



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Outlook

- Theoretical framework for $e^+e^- \rightarrow h X$
- Global fits
- Phenomenological analysis of recent BELLE data
 - z -dependence and choice of collinear FFs
 - Model for P_T -dependence
 - Hypotheses for g_K (work in progress)

Theoretical framework for $e^+e^- \rightarrow h X$

M. Boglione, A. Simonelli JHEP 02 (2021) 076

$$\begin{aligned}
 \frac{d\sigma^{\text{NLO, NLL}}}{dz_h dT dP_T^2} &= \\
 &= -\sigma_B \pi N_C \frac{\alpha_S(Q)}{4\pi} C_F \frac{3 + 8 \log(1 - T)}{1 - T} \exp \left\{ -\frac{\alpha_S(Q)}{4\pi} 3C_F (\log(1 - T))^2 \right\} \times \\
 &\times \sum_f e_f^2 \int \frac{d^2 \vec{b}_T}{(2\pi)^2} e^{i \frac{\vec{P}_T}{z_h} \cdot \vec{b}_T} \tilde{D}_{1, H/f}^{\text{NLL}}(z_h, b_T, Q, (1 - T) Q^2) \left[1 + \mathcal{O} \left(\frac{M_H^2}{Q^2} \right) \right] \\
 \\
 \tilde{D}_{1, H/f}(z, b_T; \mu, \zeta) &= \underbrace{\frac{1}{z^2} \sum_k \int_z^1 \frac{d\rho}{\rho} d_{H/k}(z/\rho, \mu_b) [\rho^2 \mathcal{C}_{k/f}(\rho, \alpha_S(\mu_b))]}_{\text{TMD at reference scale}} \times \\
 &\times \underbrace{\exp \left\{ \frac{1}{4} \tilde{K}(b_T^*; \mu_b) \log \frac{\zeta}{\mu_b^2} + \int_{\mu_b}^{\mu} \frac{d\mu'}{\mu'} \left[\gamma_D(\alpha_S(\mu'), 1) - \frac{1}{4} \gamma_K(\alpha_S(\mu')) \log \frac{\zeta}{\mu'^2} \right] \right\}}_{\text{Perturbative Sudakov Factor}} \\
 &\times \underbrace{(M_D)_{j, H}(z, b_T) \exp \left\{ -\frac{1}{4} g_K(b_T) \log \frac{z_h^2 \zeta}{M_H^2} \right\}}_{\text{Non-Perturbative content}}.
 \end{aligned}$$

Theoretical framework for $e^+e^- \rightarrow h X$

M. Boglione, A. Simonelli, *Eur. Phys. J. C* 81 (2021)

$$D = D^* \sqrt{M_S}$$



Usual definition
TMD FF

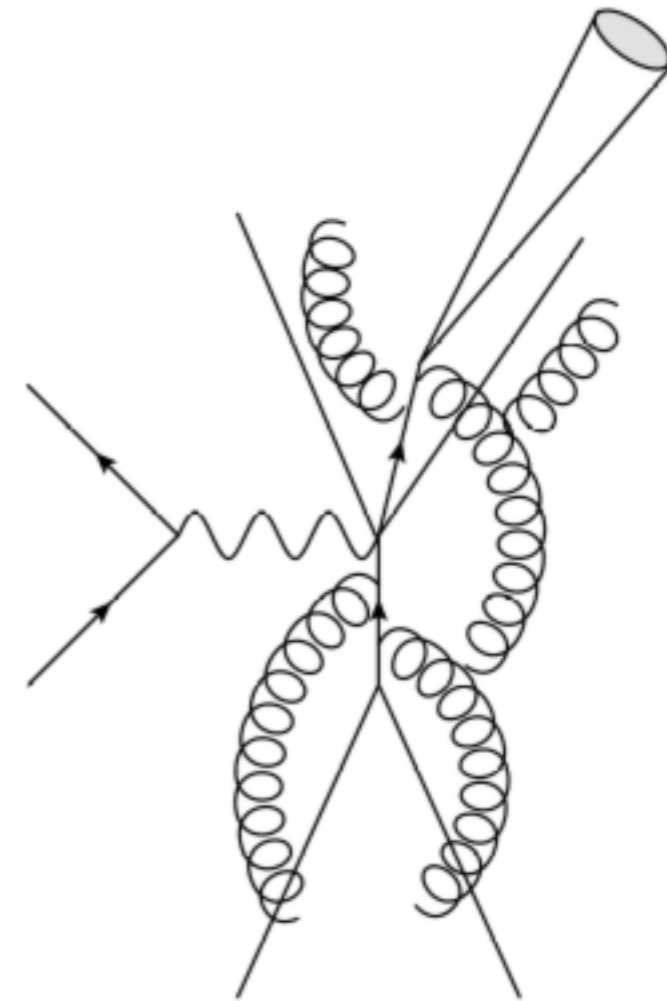


For this process
TMD FF



Soft non-
perturbative
Function

$$e^+e^- \rightarrow hX$$



$$\frac{d\sigma}{dP_T} = d\hat{\sigma} \otimes D^*(P_T)$$

Theoretical framework for $e^+e^- \rightarrow h X$

M. Boglione, A. Simonelli, Eur. Phys. J. C 81 (2021)

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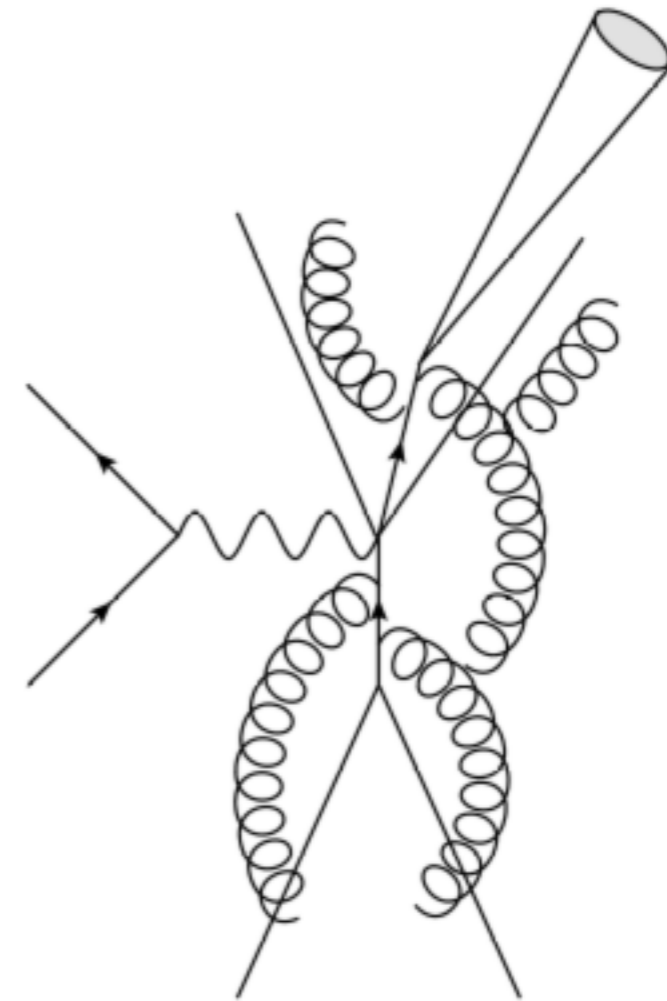


Same
constraints
to collinear FF

$$g_K(b_T) = \tilde{K}(b_T^*; \mu) - \tilde{K}(b_T; \mu)$$

Same function for non-perturbative evolution

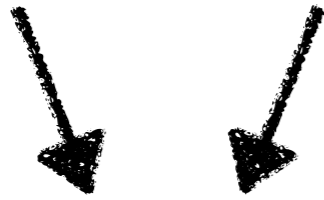
$$e^+ e^- \rightarrow h X$$



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Theoretical framework for $e^+e^- \rightarrow h X$

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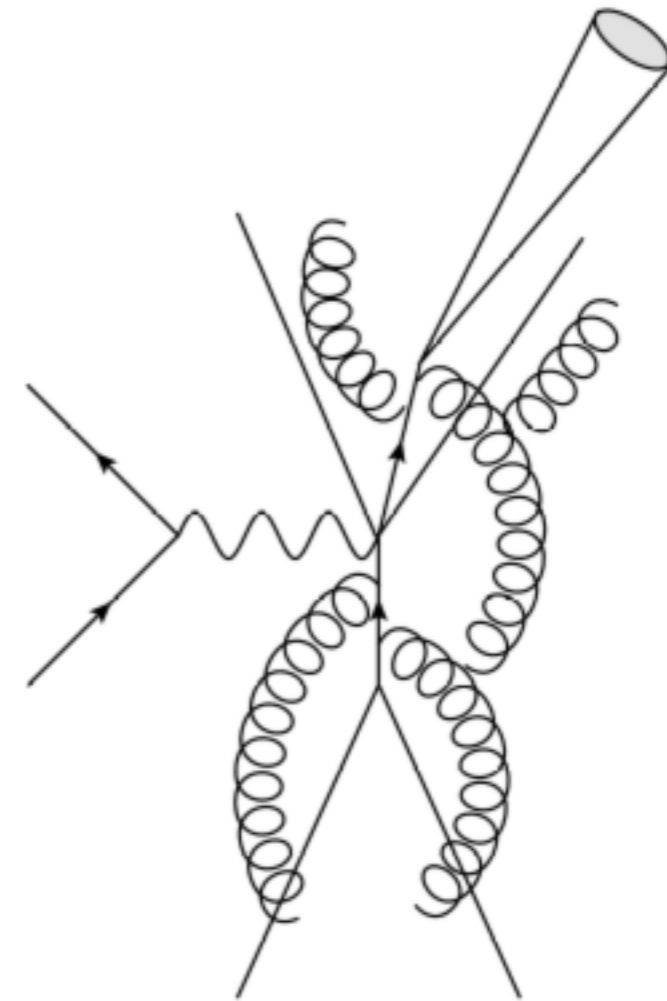


What is the effect of the collinear FFs (and PDFs in general) ?

Large- b_T behaviour of g_K ?

M. Boglione, A. Simonelli, Eur. Phys. J. C 81 (2021)

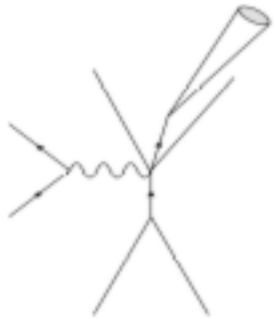
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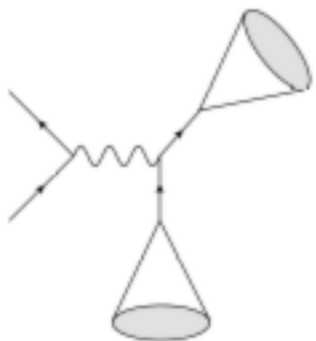
Global Fits

Possible roadmap



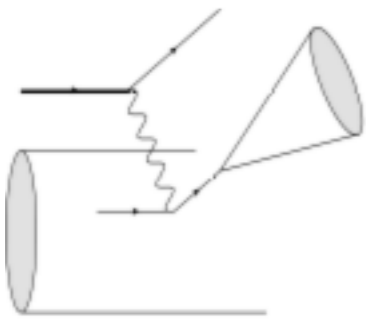
1.

Extraction of the unpolarized TMD FF, D^* , for charged pions from BELLE data (using factorization definition)



2.

Two non-perturbative functions:
 D^* , known from step 1
Soft Model M_S



3. SIDIS

Three non-perturbative functions in the cross section
 D^* , known from step 1.
Soft Model M_S , known from step 2.
Extraction of the TMD PDF, F^* (in the factorization definition, $F^* \neq F$).

(Global) Fits

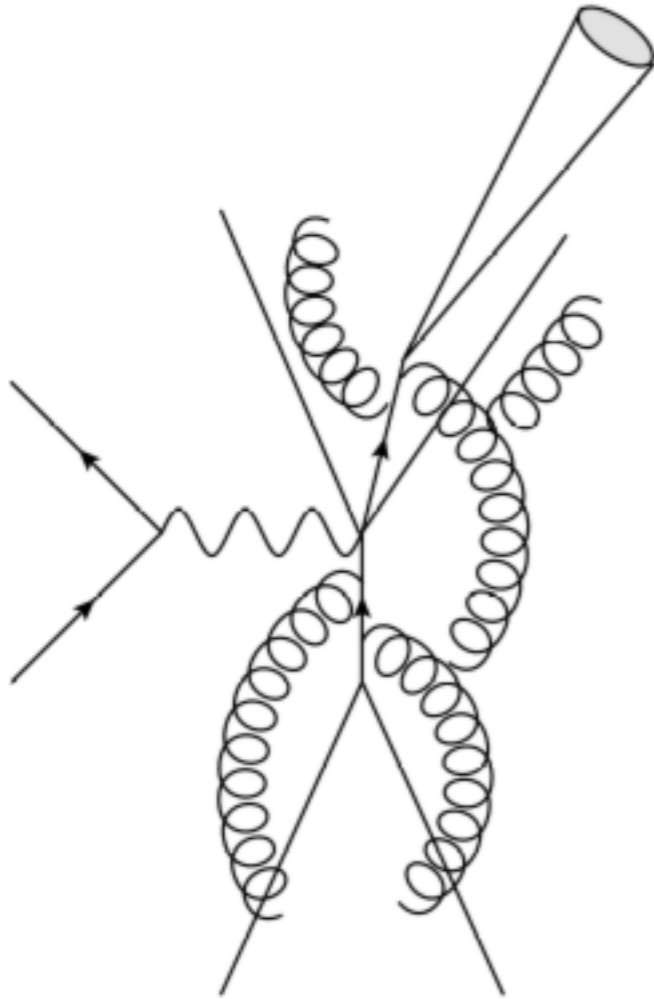
Some important aspects to consider:

- Which collinear functions are more appropriate?
- Which regions in bT are being mapped by extractions.
- Constraints of bT -behaviour for TMDs.
- Physical pictures/theoretical arguments /models (not parametrizations)
- Non perturbative evolution (gK) should be consistent with SIDIS, DY, $e+e^-$ two-hadron production.

Phenomenological analysis of recent BELLE data

Data overview

$$e^+e^- \rightarrow hX \quad (\text{Charged pions})$$



Binned in P_T , z_h and T (thrust)

$$0.06 < P_T < 2.5 \text{ GeV}$$

$$0.125 < z_h < 0.975$$

$$0.6 < T < 0.975$$

zh-dependence and choice of collinear FFs

- We compare results obtained with NNFFnlo and JAM20nlo

$$\begin{aligned}
 \tilde{D}_{1,H/f}(z, b_T; \mu, \zeta) &= \underbrace{\frac{1}{z^2} \sum_k \int_z^1 \frac{d\rho}{\rho} d_{H/k}(z/\rho, \mu_b) [\rho^2 \mathcal{C}_{k/f}(\rho, \alpha_S(\mu_b))]}_{\text{TMD at reference scale}} \times \\
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zh-dependence and choice of collinear FFs

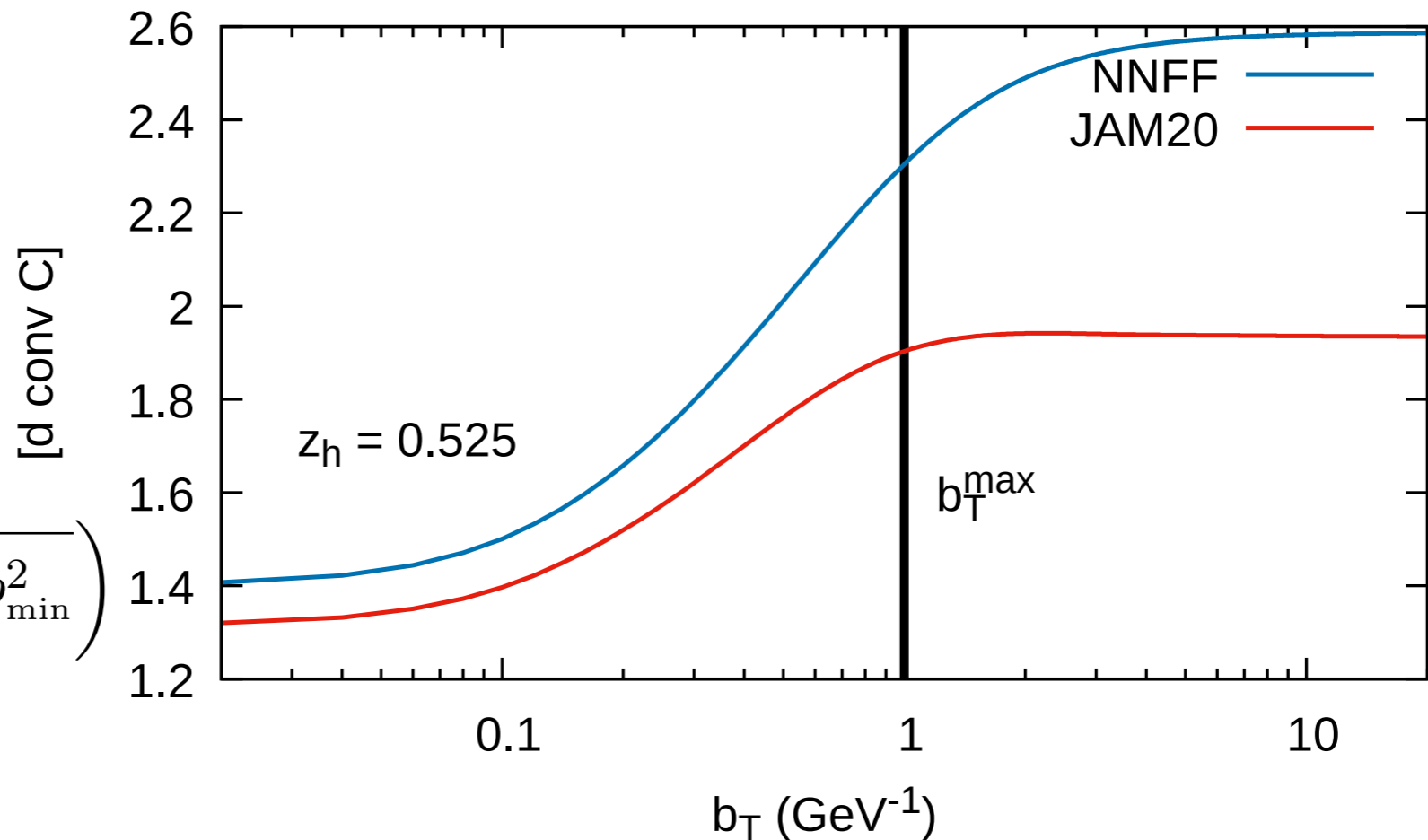
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$$\frac{1}{z^2} \sum_k \int_z^1 \frac{d\rho}{\rho} d_{H/k}(z/\rho, \mu_b) [\rho^2 C_{k/f}(\rho, \alpha_S(\mu_b))]$$

$$\mu_b = \frac{2e^{-\gamma_E}}{b_T^*(b_T)}$$

$$\vec{b}_T^*(b_T) = \frac{\vec{b}_T}{\sqrt{1 + b_T^2/b_{\max}^2}}$$

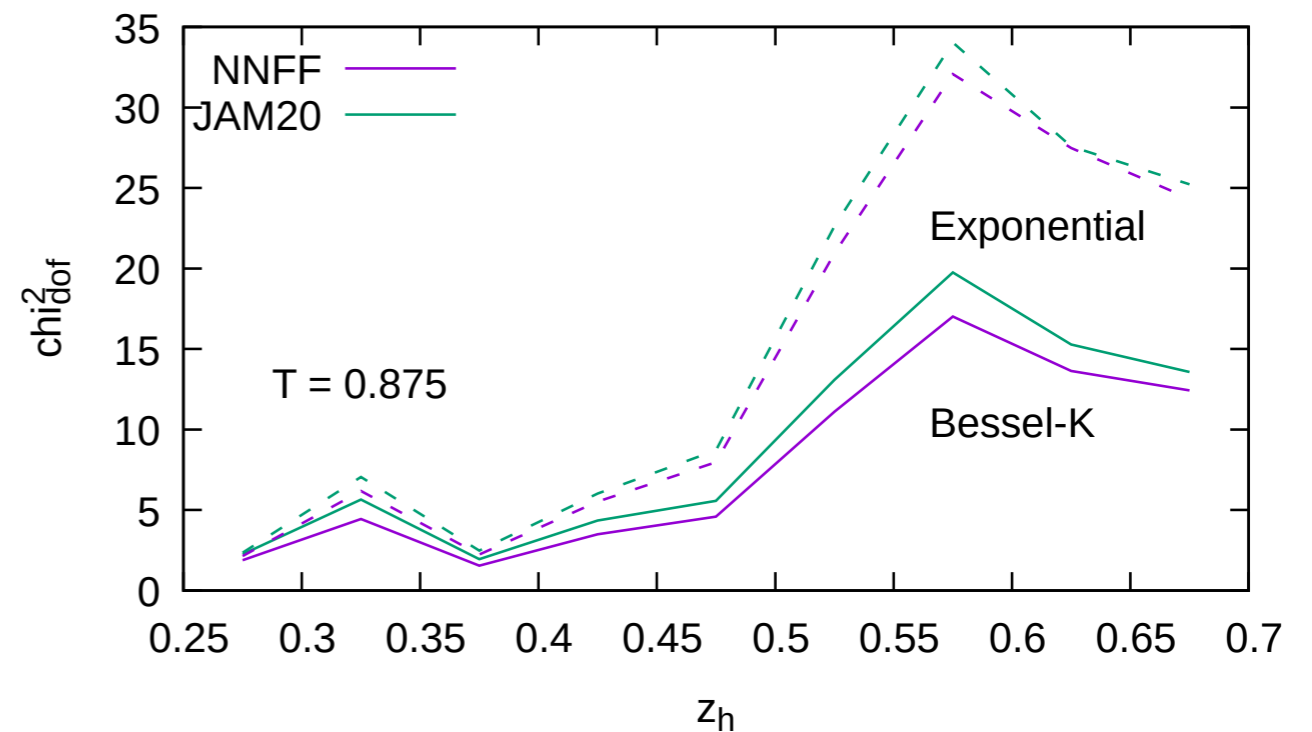
$$\vec{b}_T^*(b_c(b_T)) = \vec{b}_T^* \left(\sqrt{b_T^2 + b_{\min}^2} \right)$$



zh-dependence and choice of collinear FFs

- We compare results obtained with NNFFnlo and JAM20nlo

Nomenclature	M_D -model	parameters
z_h -independent models		
1) Exponential-q	$e^{-(M_0 b_T)^q}$	M_0, q
2) Bessel-K	$\frac{2^{2-p} (b_T M_0)^{p-1}}{\Gamma(p-1)} K_{p-1}(b_T M_0)$	M_0, p



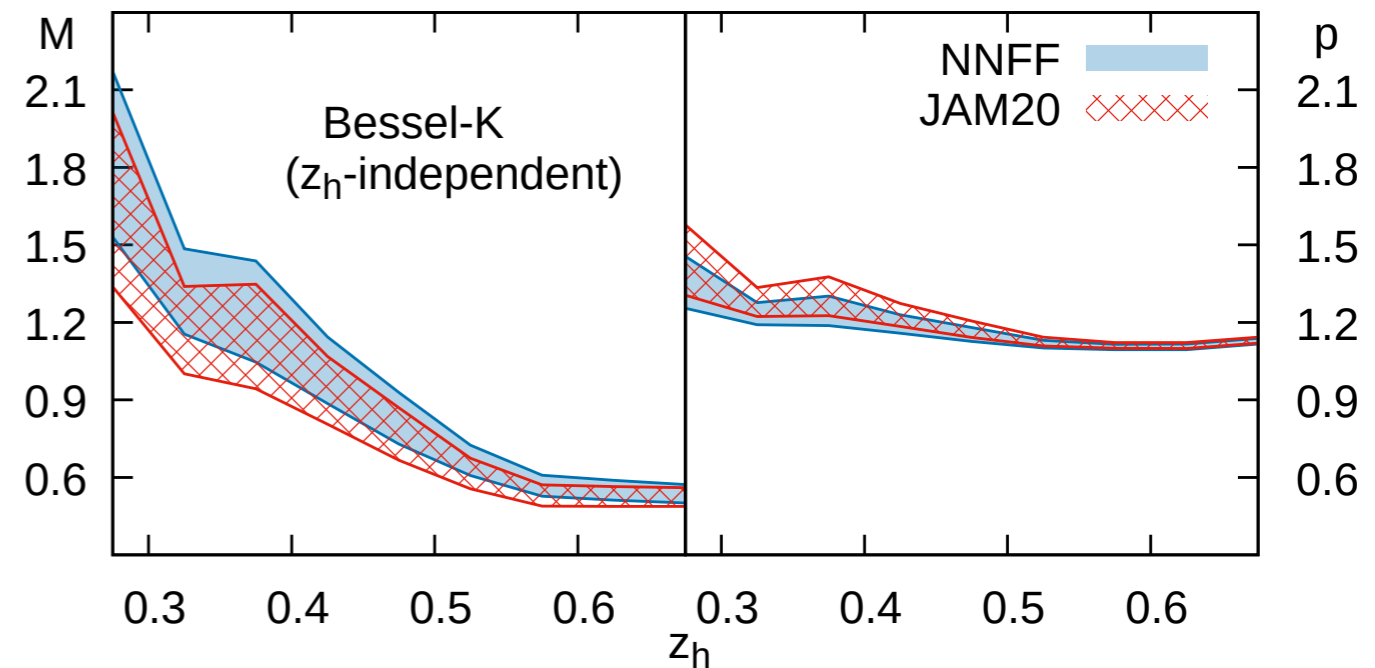
**Proxy models: performed fits
at fixed $T=0.875$.
One INDEPENDENT fit for each
 z_h -bin in the range
 $0.25 < z_h < 0.7$**

**NNFF perform better.
(non-conclusive due to large values
of χ^2/dof)**

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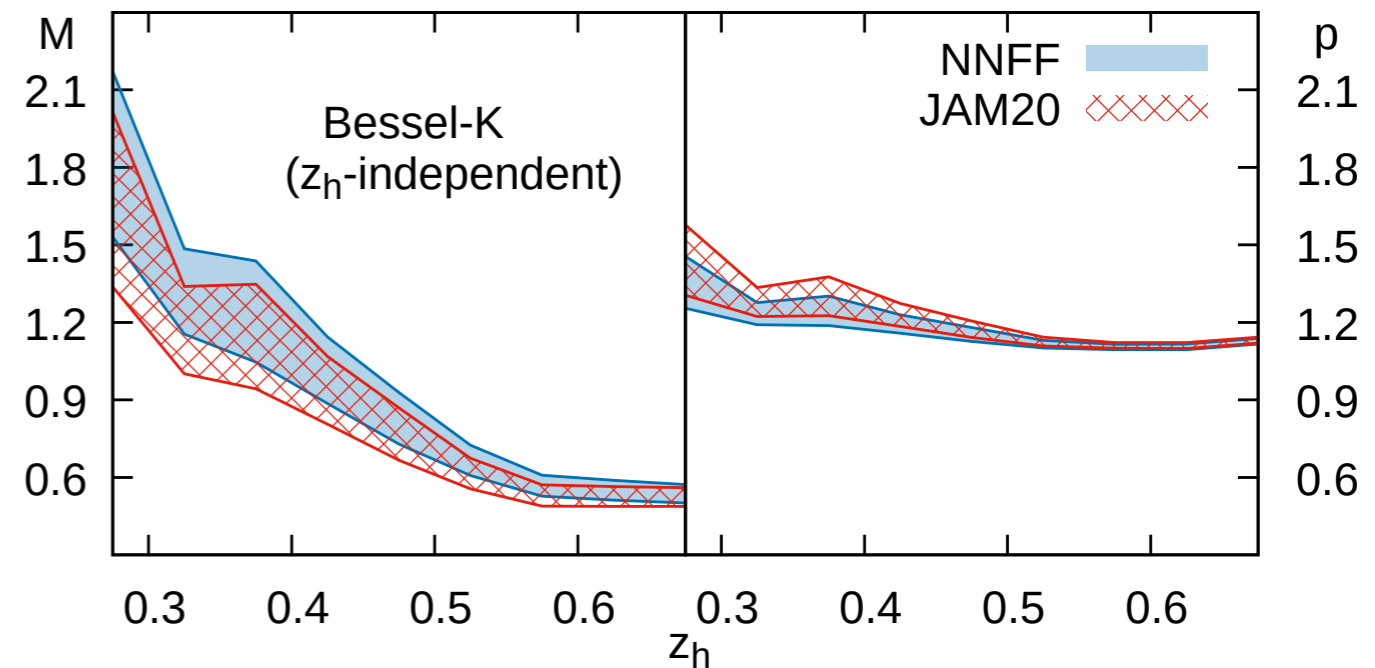
**One INDEPENDENT fit for each
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**Stronger zh-dependence in *dimensionfull*
parameter**

zh-dependence and choice of collinear FFs

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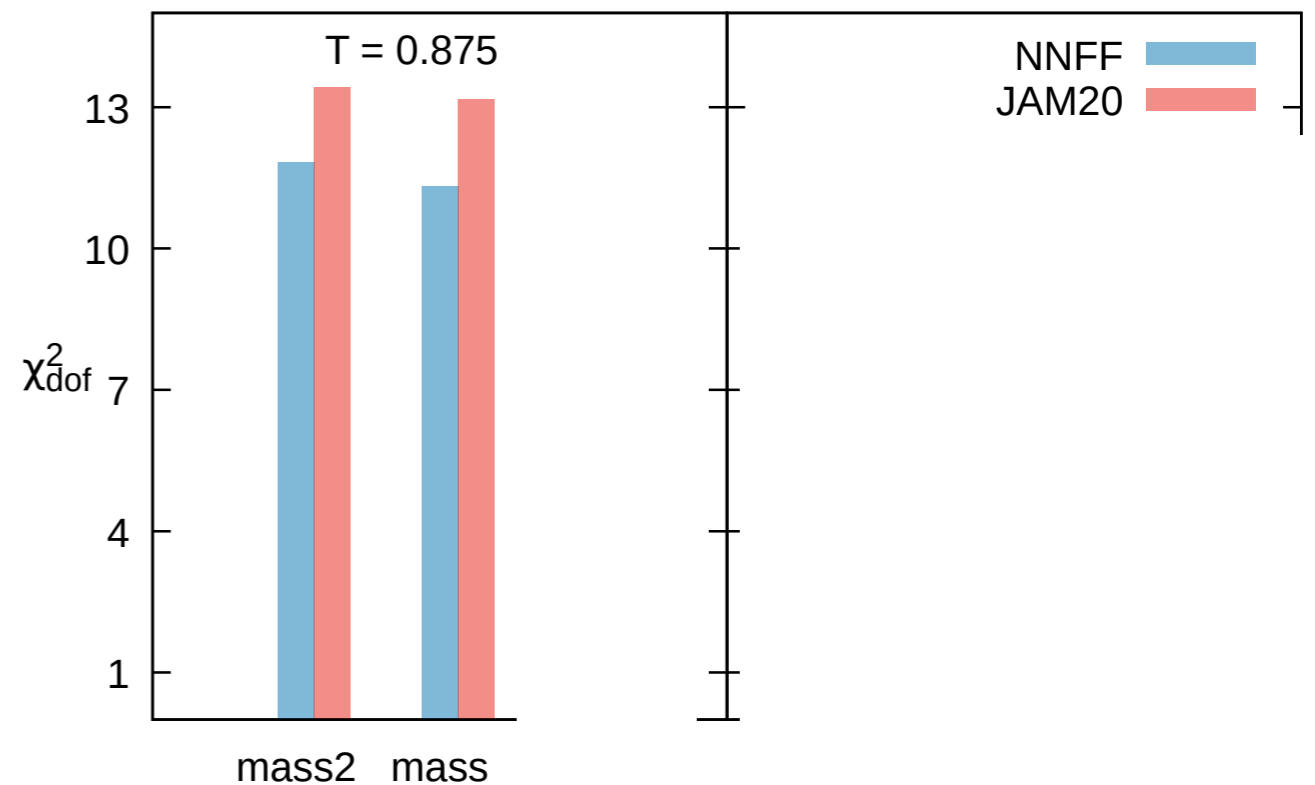
Stronger z_h -dependence in *dimensionfull* parameter

Next Step, try fitting
 z_h -bins simultaneously
 (fixed $T=0.875$)

zh-dependence and choice of collinear FFs

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z_h -dependent models		
3) Bessel-K- M_1^z	$M_0 \rightarrow M_1 (1 - \eta_1 \log(z_h))$	M_1, η_1, p
4) Bessel-K- M_2^z	$M_0 \rightarrow M_2 \left(1 + \frac{\eta_2}{z_h^2}\right)$	M_2, η_2, p

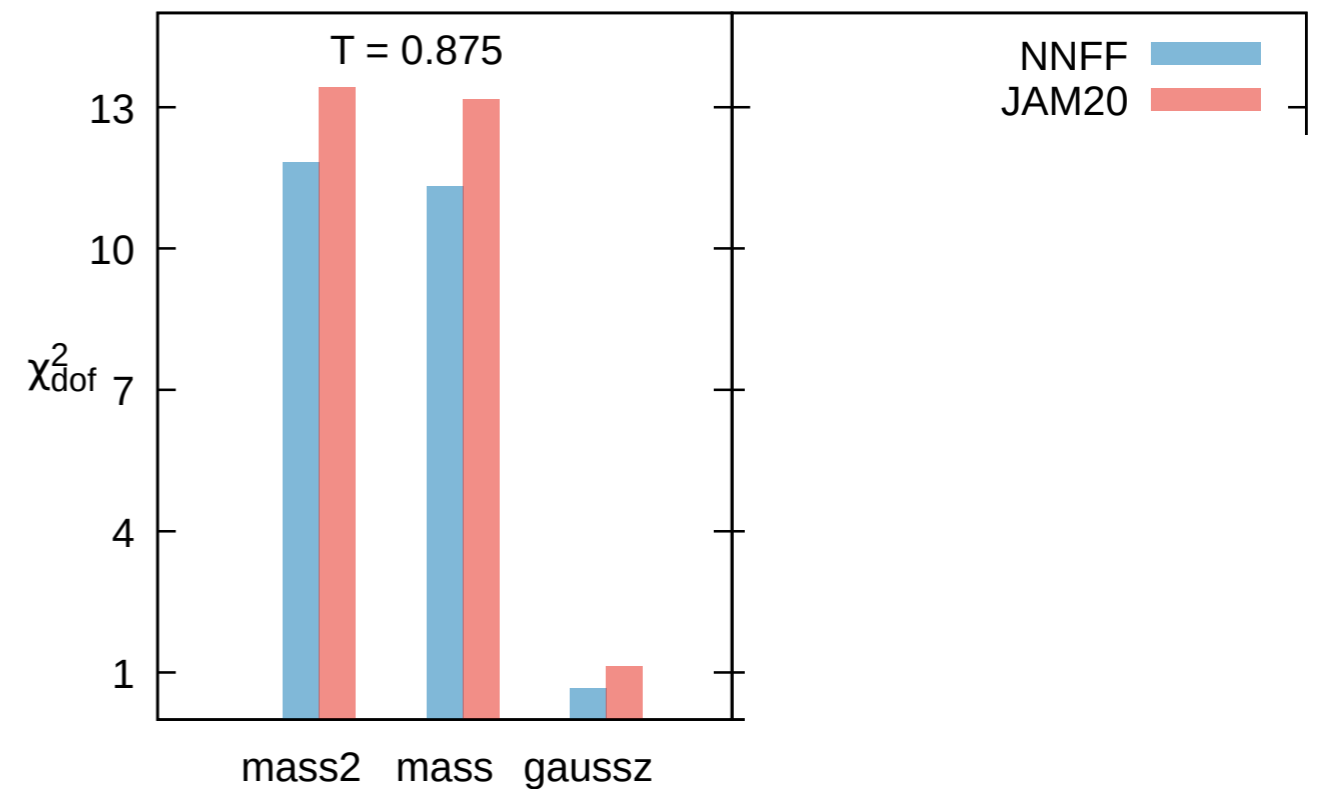


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zh-dependence and choice of collinear FFs

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5) Bessel-K- M_g^z	$e^{(M_g b_T)^2 \log(z_h)} \times \text{Bessel-K}$	M_g, M_0, p



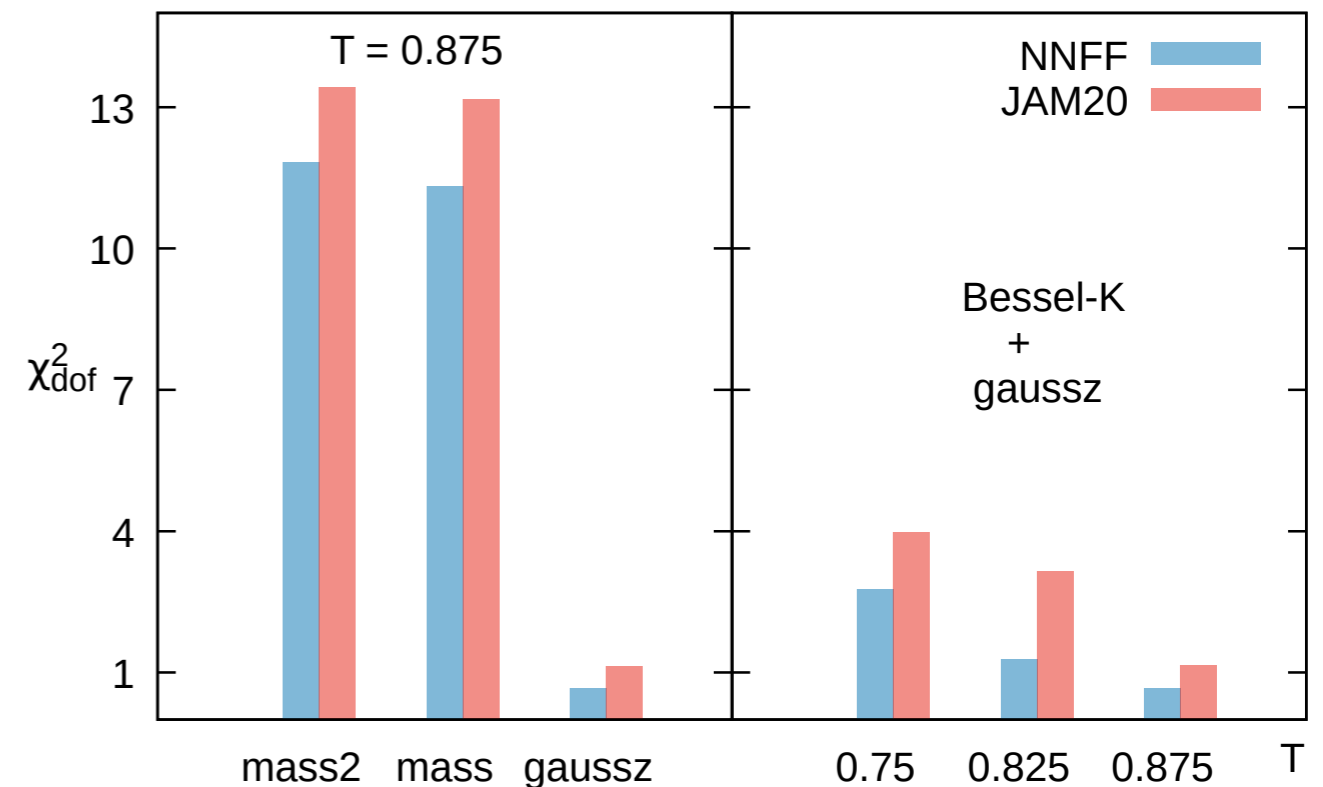
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+1 dimensionfull parameter

zh-dependence and choice of collinear FFs

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**Next Step, try fitting
zh-bins simultaneously
(fixed T=0.750, 0.825, 0.875)**

Systematically NNFFs outperform

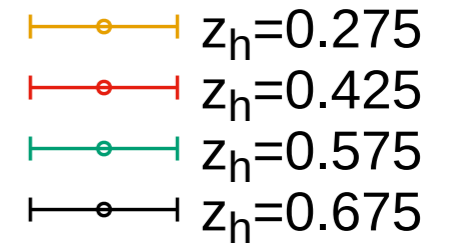
zh-dependence and choice of collinear FFs

- Where did the improvement come from?

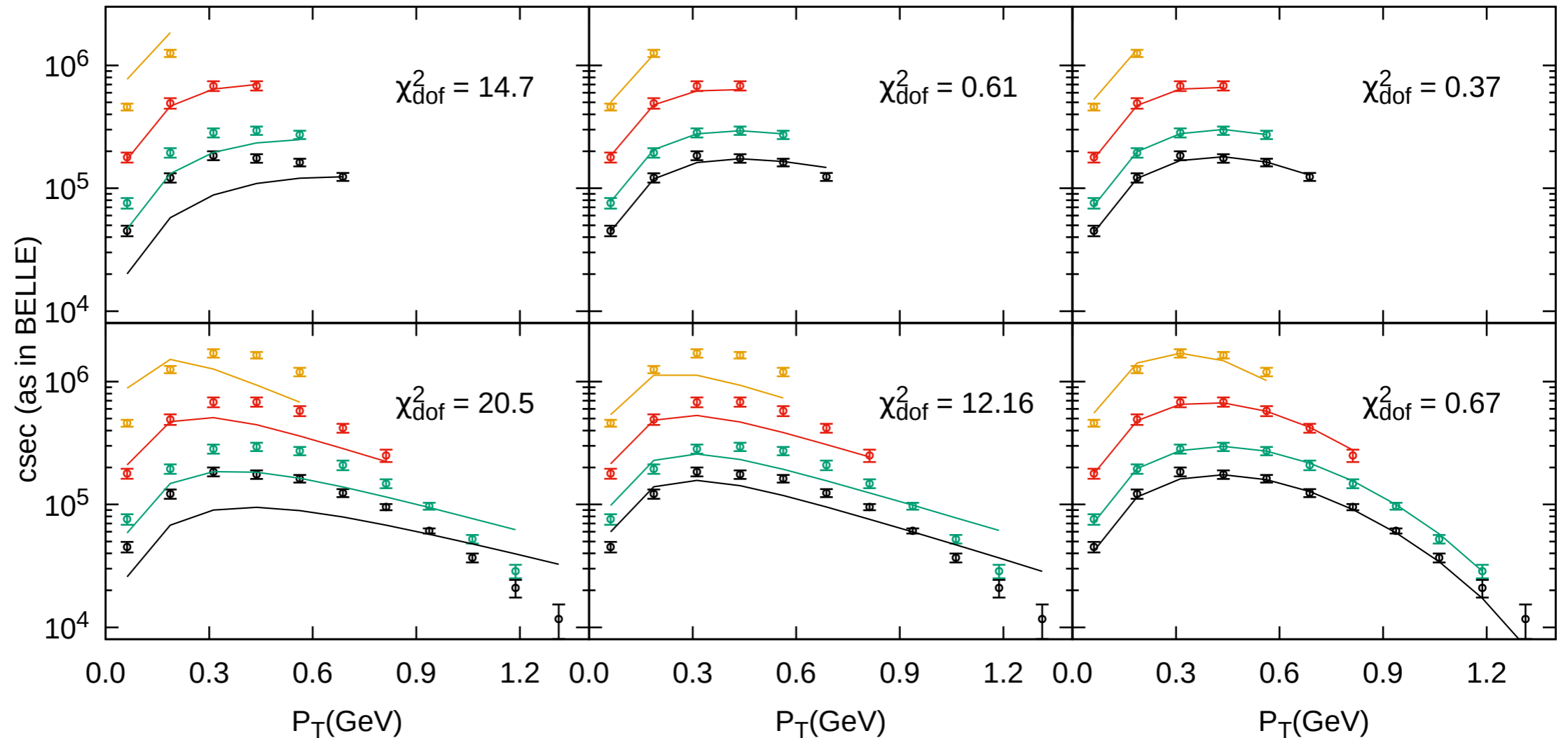
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zh-dependence and choice of collinear FFs

- Where did the improvement come from?



$q_T/Q < 0.1$



$q_T/Q < 0.2$

2) Bessel-K

3) Bessel-K- M_1^z

5) Bessel-K- M_g^z

Model for PT-dependence

$$M_D = \frac{2^{2-p}(b_T M_0)^{p-1}}{\Gamma(p-1)} K_{p-1}(b_T M_0) \times F(b_T, z_h)$$

$$M_z = M_0(1 - \eta \log(z_h))$$

Nomenclature	F -model	parameters
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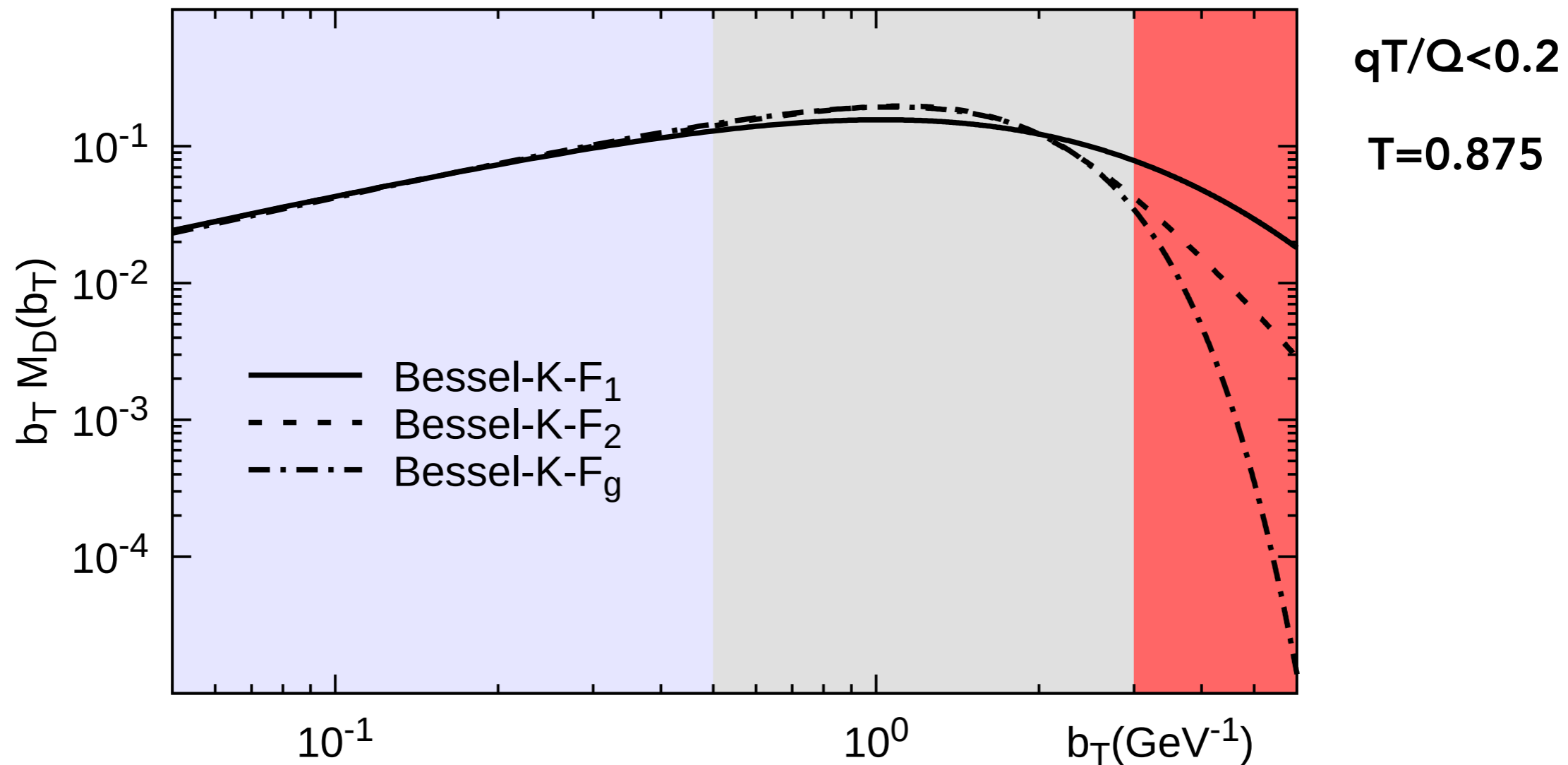
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$\chi^2_{\text{d.o.f.}}$ (fixed- T fits)			
$T = 0.750 \quad 0.825 \quad 0.875$			
Bessel-K- F_1	3.06	1.24	0.65
Bessel-K- F_2	3.02	1.26	0.97
Bessel-K- F_g	2.82	1.29	0.68

**Fit all zh-bins in the range
0.25 < zh < 0.7
qT/Q < 0.2**

Hard to discriminate bT-models

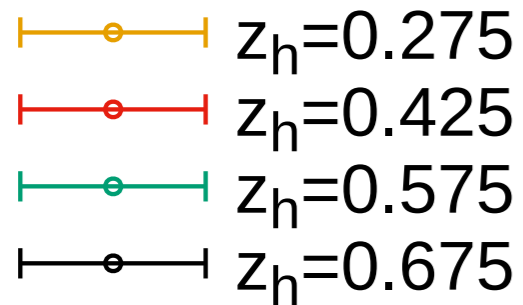
Model for PT-dependence



\uparrow
MD \rightarrow 1 for
 $b_T \rightarrow 0$

\uparrow
Largest proved
distance
(somewhere here)

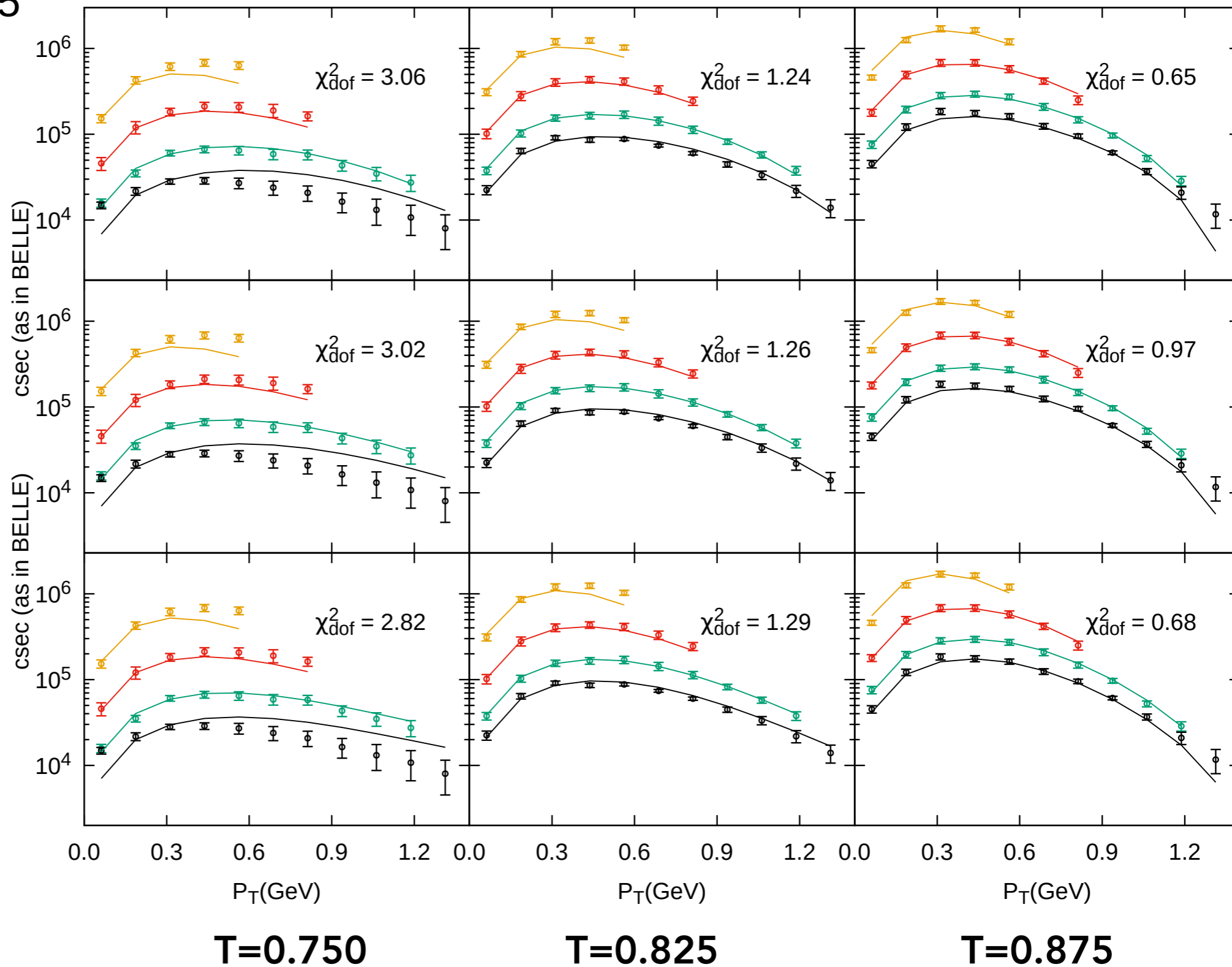
\uparrow
Different
asymptotic
behaviour



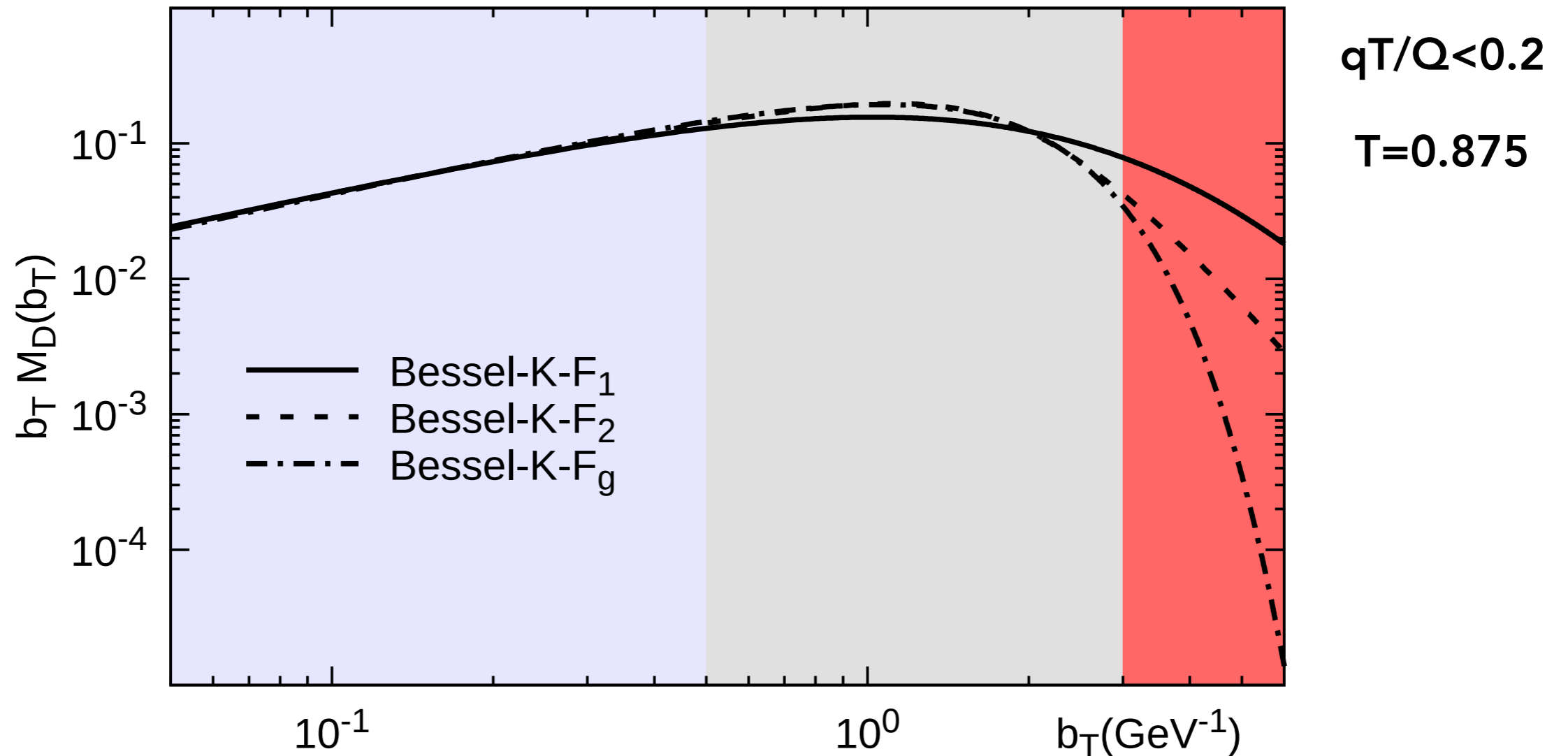
1) Bessel-K- F_1

2) Bessel-K- F_2

3) Bessel-K- F_g



Model for PT-dependence



All three models match the data, different asymptotic behaviour
Which choice?

Model for PT-dependence

P. Schweitzer, M. Strikman, and C. Weiss, JHEP 1301, 163 (2013)

J. Collins and T. Rogers, Phys. Rev. D91 (2015) 074020, [1412.3820].

$$\text{TMDs} \longrightarrow \frac{1}{b_T^\alpha} e^{-mb_T} \quad \text{as} \quad b_T \rightarrow \infty.$$

We consider $\log(M_D) \sim -C b_T + o(b_T)$

Non-
perturbative
function



Mass parameter

Model for PT-dependence

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Focus on these

We consider $\log(M_D) \sim -C b_T + o(b_T)$

Non-
perturbative
function



Mass parameter



Hypotheses for gK (work in progress)

We consider $\log(M_D) \sim -C b_T + o(b_T)$

$$\text{TMDFF} \sim M_D \exp\left(-\frac{g_K(b_T)}{4} \log\left(\frac{\zeta}{\zeta_0}\right)\right)$$

$$\log(\text{TMDFF}) \sim -C b_T - \frac{g_K(b_T)}{4} \log\left(\frac{\zeta}{\zeta_0}\right) + o(b_T)$$

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$$\log(\text{TMDFF}) \sim -C b_T - \frac{g_K(b_T)}{4} \log\left(\frac{\zeta}{\zeta_0}\right) + o(b_T)$$

If $g_K(b_T) = O(b_T)$, the argument of exponential may flip sign depending on ζ_0 and ζ (relevant for global fits)

Hypotheses for gK (work in progress)

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$$\log(\text{TMDFF}) \sim -C b_T - \underbrace{\frac{g_K(b_T)}{4} \log\left(\frac{\zeta}{\zeta_0}\right)}_{\text{bracket}} + o(b_T)$$

If $g_K(b_T) = o(b_T)$ limit of TMDFF always zero for $b_T \rightarrow \infty$

This avoids a fast decaying behaviour for $g_K(b_T)$ as $b_T \rightarrow \infty$

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We consider $\log(M_D) \sim -C b_T + o(b_T)$

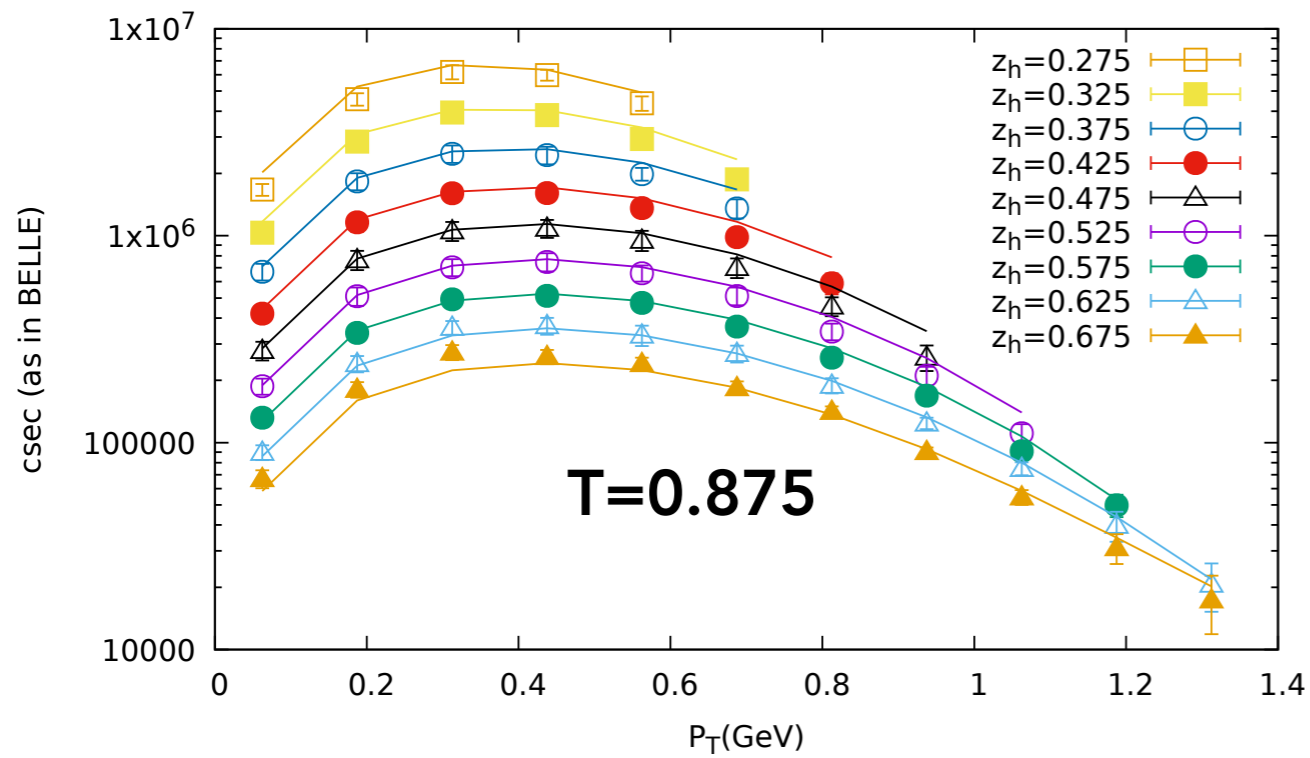
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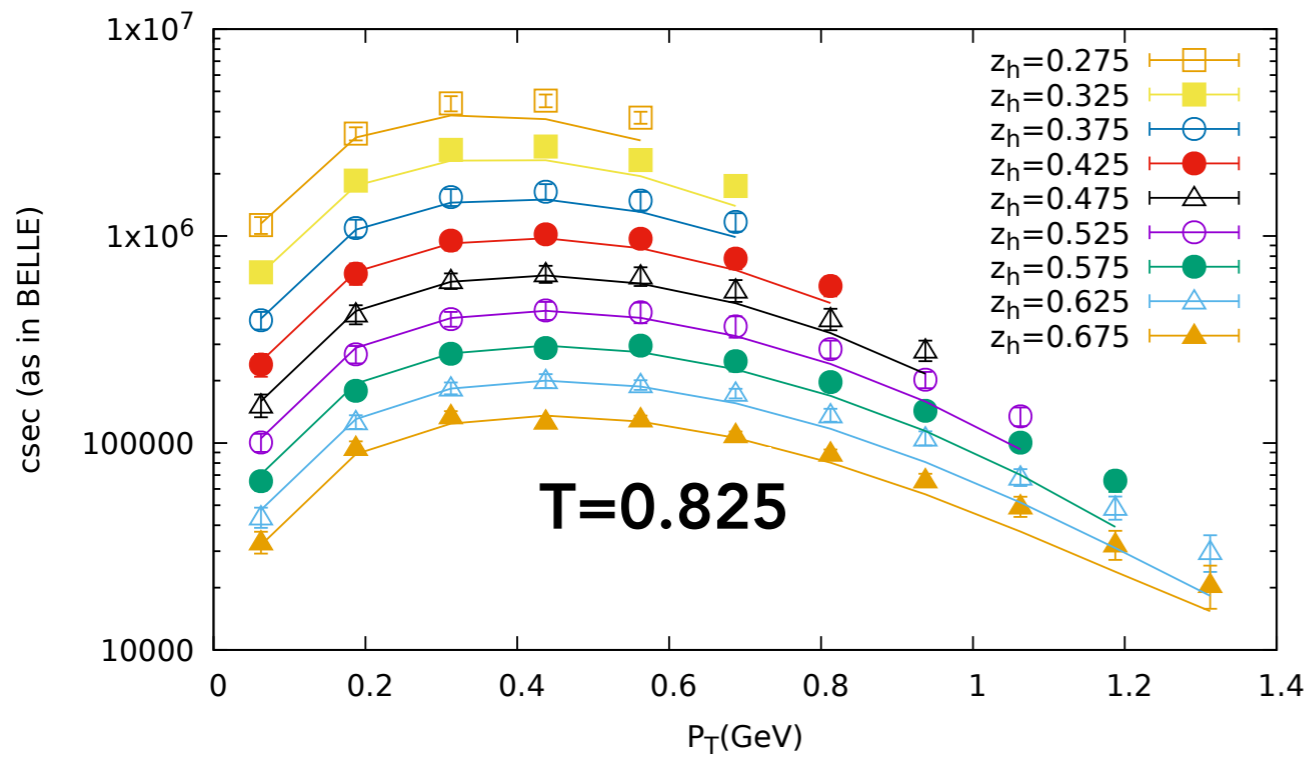
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This avoids a fast decaying behaviour for $g_K(b_T)$ as $b_T \rightarrow \infty$

For instance $g_K = \log(1 + (M_k b_T)^{p_k})$



$q_T/Q < 0.2$



$\chi^2/\text{dof} = 2.1$

Final remarks

- NNFF seem to work well to extract TMDFF in Belle data
- Gaussian behaviour in bT describes the data but NOT because of its asymptotic behaviour
- Cannot constrain phenomenologically the asymptotic behaviour of the TMDFF
- Instead we use the hypothesis that its logarithm should be dominantly linear.
- We attempted to describe multidimensional data without modifying such leading behaviour of the TMDFF by parameterising g_k with a logarithmic asymptotic behaviour.
- Results are promising but must improve limited range of T

Thanks

