

Semi-leptonic $B \rightarrow \pi$, $B_s \rightarrow K$ decays & Bayesian inference for z -expansion fits

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Lattice meets Continuum, Siegen, DE

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- 1 Motivation and the status quo
- 2 Challenges on the lattice
- 3 An example: $B_s \rightarrow K\ell\nu$ [JTT, RBC/UKQCD PRD 107 (2023) 114512]
- 4 z-expansions and Bayesian inference
- 5 Improvements and Summary

Motivation and the status quo

Why $|V_{ub}|$

C _[Cabibbo '63] KM _[Kobayashi, Maskawa '73]

- Status: [PDG'24]

	<i>d</i>	<i>s</i>	<i>b</i>
<i>u</i>	0.97367(32)	0.22431(85)	0.00367(15) [†]
<i>c</i>	0.221(4)	0.975(6)	0.0398(6) [†]
<i>t</i>	0.0086(2)	0.0415(9)	1.010(27)

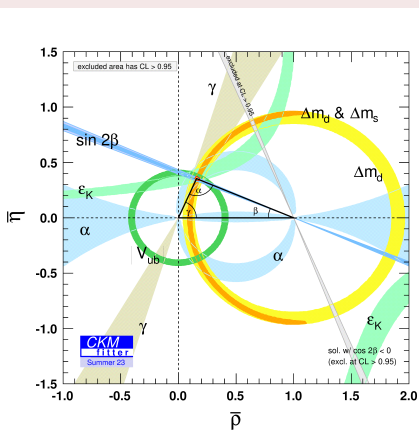
$${}^\dagger V_{ub}^{\text{inc}} = 0.00413(26), \quad V_{cb}^{\text{inc}} = 0.0422(5)$$

- Uncertainties [%]

$$\begin{pmatrix} 0.03 & 0.4 & 4.1^\dagger \\ 1.8 & 0.6 & 1.5^\dagger \\ 2.3 & 2.2 & 2.7 \end{pmatrix}$$

- V_{ub} exc+inc: 5.2% ($\sqrt{\chi^2} = 1.4$)

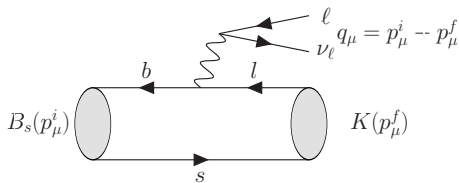
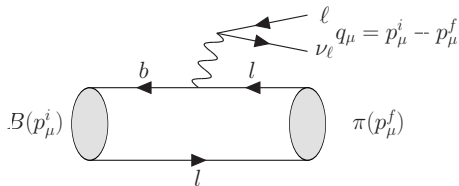
Unitarity Triangle [CKMfitter'05 + web updates]



$|V_{ub}|$ least well known

$B \rightarrow \pi$ vs $B_s \rightarrow K$

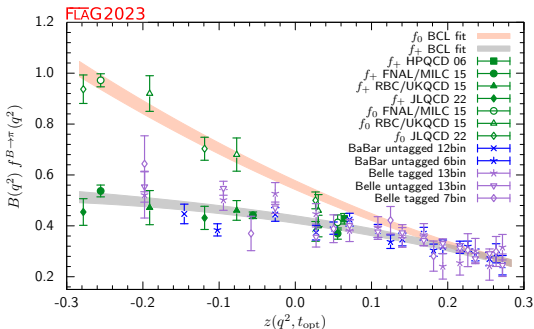
$$\frac{d\Gamma(B_{(s)} \rightarrow Pl\nu_\ell)}{dq^2} = |V_{ub}|^2 \mathcal{K} \left[\left(1 + \frac{m_\ell^2}{2q^2} \right) |f_+(q^2)|^2 + \mathcal{K}_2 m_\ell^2 |f_0(q^2)|^2 \right]$$



- Only differ in spectator quark:
⇒ Very similar computation on the lattice
- Complementary channels to extract $|V_{ub}|$
- Most experimental data obtained for $\ell \in \{e, \mu\}$, so $m_\ell \sim 0$
⇒ $f_+(q^2)$ “more important”

Experimental situation for $b \rightarrow ul\nu$

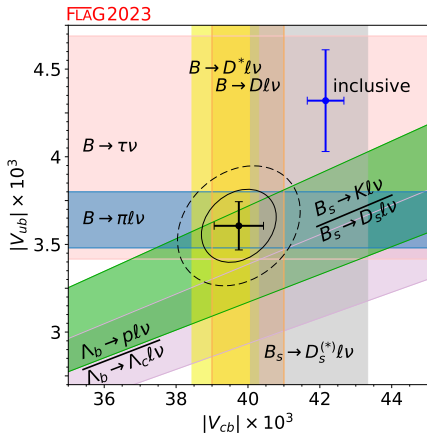
$$B \rightarrow \pi l \nu$$



Belle & BaBar multiple bins (\uparrow)
 + Future: Belle II, LHCb

- $B_s \rightarrow K l \nu$
 LHCb 2 bins for
 $B_s \rightarrow K l \nu$ (normalised
 by $B_s \rightarrow D_s$)
 LHCb working on more
 bins
- $\Lambda_b \rightarrow p l \nu$
 LHCb $\Lambda_b \rightarrow p l \nu$
 (normalised by $\Lambda_b \rightarrow \Lambda_c$)
- ...

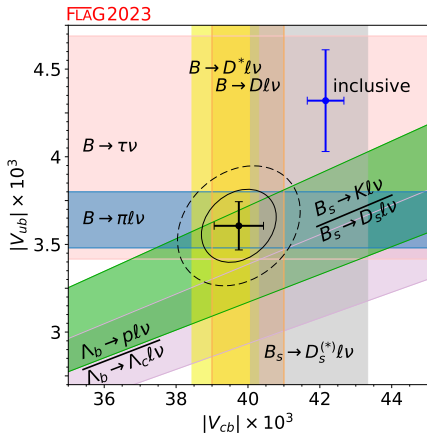
What is the status of V_{ub} ? Let's look at FLAG!



Consistency between different determinations ✓ (or is it?!)

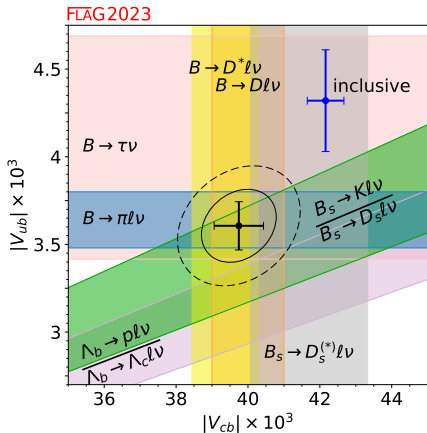
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$B \rightarrow \tau \nu$: good agreement! ✓



Consistency between different determinations ✓ (or is it?!)

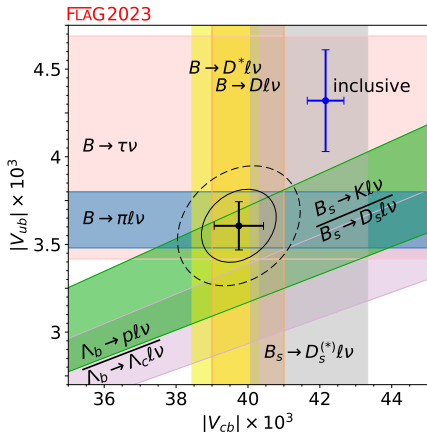
What is the status of V_{ub} ? Let's look at FLAG!



$B \rightarrow \tau \nu$: good agreement! ✓
 $\Lambda_b \rightarrow p$: Only a single result ✓(?)

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What is the status of V_{ub} ? Let's look at FLAG!



$B \rightarrow \tau \nu$: good agreement! ✓
 $\Lambda_b \rightarrow p$: Only a single result ✓(?)
 $B \rightarrow \pi \ell \nu$: $\rho \sim 2 \times 10^{-5}$ ✗

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2111.09849

$B \rightarrow \pi$ ($N_f = 2 + 1$)

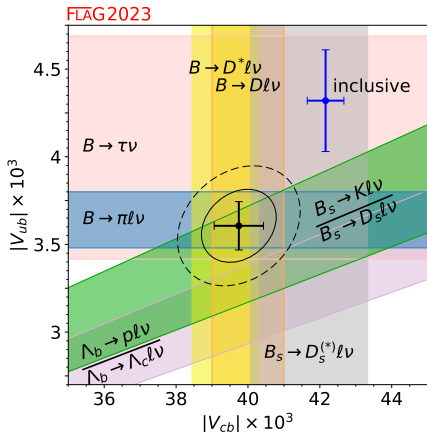
	Central Values	Correlation Matrix				
a_0^+	0.423 (21)	1	-0.00466	-0.0749	0.402	0.0920
a_1^+	-0.507 (93)	-0.00466	1	0.498	-0.0556	0.659
a_2^+	-0.75 (34)	-0.0749	0.498	1	-0.152	0.677
a_0^0	0.561 (24)	0.402	-0.0556	-0.152	1	-0.548
a_1^0	-1.42 (11)	0.0920	0.659	0.677	-0.548	1

Table 46: Coefficients and correlation matrix for the $N^+ = N^0 = 3$ z-expansion fit of the $B \rightarrow \pi$ form factors f_+ and f_0 . The coefficient a_2^0 is fixed by the $f_+(q^2 = 0) = f_0(q^2 = 0)$ constraint. The chi-square per degree of freedom is $\chi^2/\text{dof} = 43.6/12$ and the errors on the z-parameters have been rescaled by $\sqrt{\chi^2/\text{dof}} = 1.9$. The lattice calculations that enter this fit are taken from FNAL/MILC 15 [58], RBC/UKQCD 15 [59] and JLQCD 22 [60]. The parameterizations are defined in Eqs. (533) and (534).

Consistency between different determinations ✓(or is it?!)

∃ sys errors \Rightarrow ρ only indicative

What is the status of V_{ub} ? Let's look at FLAG!



- $B \rightarrow \tau \nu$: good agreement! ✓
- $\Lambda_b \rightarrow p$: Only a single result ✓(?)
- $B \rightarrow \pi l \nu$: $p \sim 2 \times 10^{-5}$ ✗
- $B_s \rightarrow K l \nu$: $p \sim 7 \times 10^{-6}$ ✗

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$B_s \rightarrow K$ ($N_f = 2 + 1$)

	Central Values	Correlation Matrix						
a_0^+	0.370(21)	1.	0.2781	-0.3169	-0.3576	0.6130	0.3421	0.2826
a_1^+	-0.68(10)	0.2781	1.	0.3672	0.1117	0.4733	0.8487	0.8141
a_2^+	0.55(48)	-0.3169	0.3672	1.	0.8195	0.3323	0.6614	0.6838
a_3^+	2.11(83)	-0.3576	0.1117	0.8195	1.	0.2350	0.4482	0.4877
a_0^0	0.234(10)	0.6130	0.4733	0.3323	0.2350	1.	0.6544	0.5189
a_1^0	0.135(86)	0.3421	0.8487	0.6614	0.4482	0.6544	1.	0.9440
a_2^0	0.20(35)	0.2826	0.8141	0.6838	0.4877	0.5189	0.9440	1.

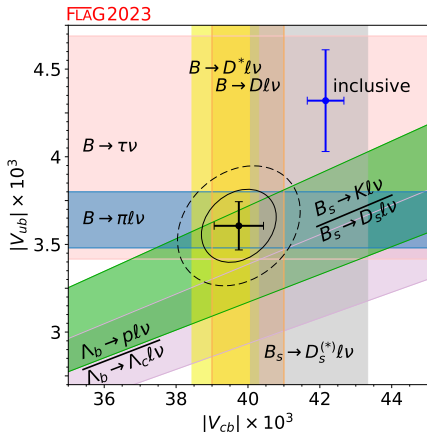
Table 48: Coefficients and correlation matrix for the $N^+ = N^0 = 4$ z-expansion of the $B_s \rightarrow K$ form factors f_+ and f_0 . The coefficient a_0^0 is fixed by the $f_+(q^2 = 0) = f_0(q^2 = 0)$ constrain. The chi-square per-degree of freedom is $\chi^2/\text{dof} = 3.82$ and the errors on the z -parameters have been rescaled by $\sqrt{\chi^2/\text{dof}} = 1.95$.

(I counted 7 fit parameters and 19 datapoints \Rightarrow 12 dof's)

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What is the status of V_{ub} ? Let's look at FLAG!



- $B \rightarrow \tau \nu$: good agreement! ✓
- $\Lambda_b \rightarrow p$: Only a single result ✓(?)
- $B \rightarrow \pi l \nu$: $p \sim 2 \times 10^{-5}$ ✗
- $B_s \rightarrow K l \nu$: $p \sim 7 \times 10^{-6}$ ✗
- $|V_{ub}| (B \rightarrow \pi)$: $p \sim 3 \times 10^{-5}$ ✗

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$B \rightarrow \pi l \nu$ ($N_f = 2 + 1$)

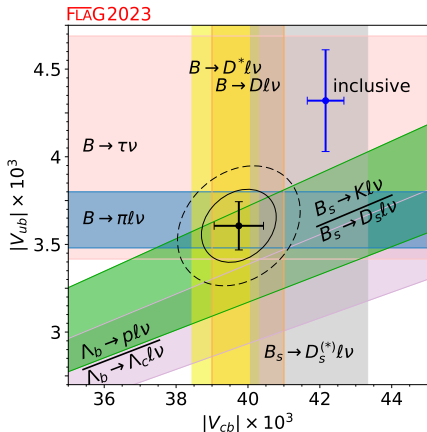
	Central Values	Correlation Matrix					
$ V_{ub} \times 10^3$	3.64 (16)	1	-0.812	-0.108	0.128	-0.326	-0.151
a_0^+	0.425 (15)	-0.812	1	-0.188	-0.309	0.409	0.00926
a_1^+	-0.441 (39)	-0.108	-0.188	1	-0.498	-0.0343	0.150
a_2^+	-0.52 (13)	0.128	-0.309	-0.498	1	-0.190	0.128
a_0^0	0.560 (17)	-0.326	0.409	-0.0343	-0.190	1	-0.772
a_1^0	-1.346 (53)	-0.151	0.00926	0.150	0.128	-0.772	1

Table 57: $|V_{ub}|$, coefficients for the $N^+ = N^0 = N^T = 3$ z-expansion of the $B \rightarrow \pi$ form factors f_+ and f_0 , and their correlation matrix. The chi-square per degree of freedom is $\chi^2/\text{dof} = 116.7/62 = 1.88$ and the errors on the fit parameters have been rescaled by $\sqrt{\chi^2/\text{dof}} = 1.37$. The lattice calculations that enter this fit are taken from FNAL/MILC [58], RBC/UKQCD [59] and JLQCD [60]. The experimental inputs are taken from BaBar [161, 162] and Belle [163, 164].

Consistency between different determinations ✓(or is it?!)

∃ sys errors $\Rightarrow p$ only indicative

What is the status of V_{ub} ? Let's look at FLAG!



- $B \rightarrow \tau \nu$: good agreement! ✓
- $\Lambda_b \rightarrow p$: Only a single result ✓(?)
- $B \rightarrow \pi \ell \nu$: $p \sim 2 \times 10^{-5}$ ✗
- $B_s \rightarrow K \ell \nu$: $p \sim 7 \times 10^{-6}$ ✗
- $|V_{ub}| (B \rightarrow \pi)$: $p \sim 3 \times 10^{-5}$ ✗

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$B \rightarrow \pi \ell \nu$ ($N_f = 2 + 1$)

$ V_{ub} \times 10^3$	Central Values	Correlation Matrix					
		1	-0.812	-0.108	0.128	-0.326	-0.151
a_0^+	3.64 (16)	1	-0.812	-0.108	0.128	-0.326	-0.151
a_0^+	0.425 (15)	-0.812	1	-0.188	-0.309	0.409	0.00926
a_1^+	-0.441 (39)	-0.108	-0.188	1	-0.498	-0.0343	0.150
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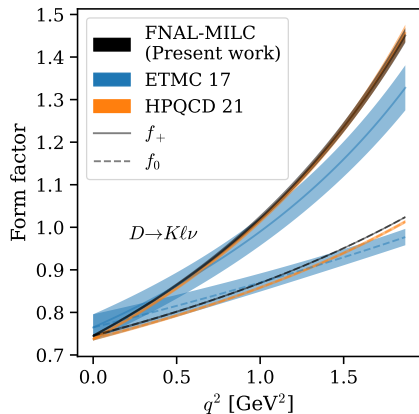
∃ sys errors $\Rightarrow p$ only indicative

We need to scrutinise this!

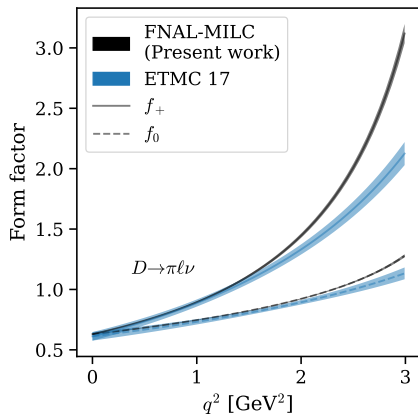
What about charm decays?

Not unique to b -decays: Similar tensions for $c \rightarrow s$ and $c \rightarrow d$

[FNAL/MILC'22]



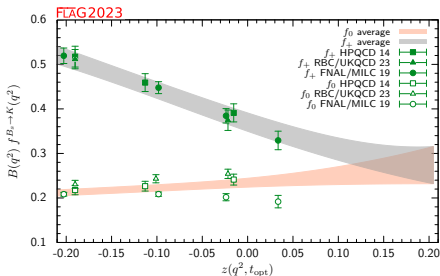
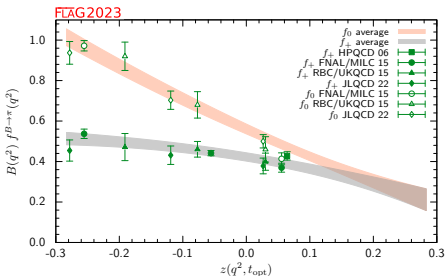
[FNAL/MILC'22]



Puzzling: Tensions at large q^2 where data should be most precise.

⇒ We need to resolve these discrepancies

FLAG's summary of $B \rightarrow \pi$ and $B_s \rightarrow K$



- f_+ looks fine, f_0 shows some tensions
- Most experimental data obtained for $\ell \in \{e, \mu\}$, so $m_\ell \sim 0$ and recall:

$$\frac{d\Gamma(B_s \rightarrow Pl\nu_\ell)}{dq^2} = |V_{ub}|^2 \mathcal{K} \left[\left(1 + \frac{m_\ell^2}{2q^2} \right) |f_+(q^2)|^2 + \mathcal{K}_2 m_\ell^2 |f_0(q^2)|^2 \right]$$

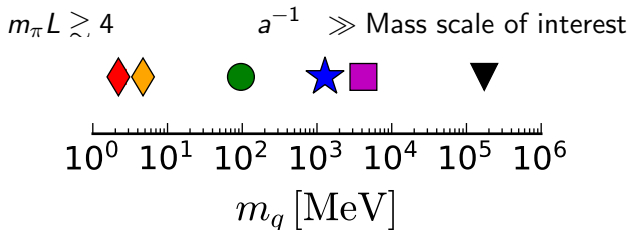
Does that mean V_{ub} should be fine? **X**

- kinematic extrapolation (z-expansion) stabilised by kinematic constraint $f_0(0) = f_+(0)$, so f_0 does impact CKM determinations!

Challenges on the lattice

Multiple scale problem on the lattice: back of the envelope

Control effects of IR (finite volume) and UV (discretisation) regulators:



For $m_\pi = m_\pi^{\text{phys}} \sim 140$ MeV and $\bar{m}_b(m_b) \approx 4.2$ GeV:

$$L \gtrsim 5.6 \text{ fm} \quad a^{-1} \gg 4.2 \text{ GeV} \approx (0.05 \text{ fm})^{-1}$$

Requires $N \equiv L/a \gg 120 \Rightarrow N^3 \times (2N) \gg 4 \times 10^8$ lattice sites.

very expensive to satisfy both constraints simultaneously. . .

. . . needs to be repeated for different values of a .

Currently computationally impossible at physical quark masses!

How to simulate the b -quark?

Effective action for b

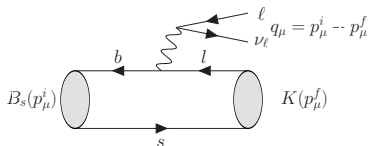
- Can tune to $m_b \sim m_b^{\text{phys}}$
- comes with **systematic errors** which are hard to estimate/reduce

Relativistic action for b

- Theoretically cleaner and systematically improvable
- $m_b < m_b^{\text{phys}}$: **control extrapolation to m_b^{phys}**

- relativistic will win in the long term
- for now, settle on a compromise
- very different systematics \Rightarrow complementary results!

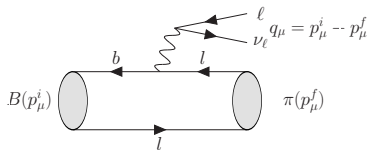
Complication: semileptonic form factors depend on momentum transfer q^2 :



$$q^2 = (E_{B_s} - E_K)^2 - (\vec{p}_i - \vec{p}_f)^2$$

Interplay between varying m_b in q^2 , M_{B_s} and $(am_b)^n$ highly non-trivial!

Challenges in computing $f_X(q^2)$: example $B \rightarrow \pi \ell \nu$



- $q^\mu = p_B^\mu - p_\pi^\mu$
- $M_B \approx 5.28 \text{ GeV}$, $M_\pi \approx 0.14 \text{ GeV}$
- Semileptonic region $q^2 \in [0, q_{\text{max}}^2]$
- $q_{\text{max}}^2 \equiv (M_B - M_\pi)^2 \sim 26.4 \text{ GeV}^2$

- physical kinematics in the B rest-frame: $q^2 = 0 \Leftrightarrow |p_\pi|^2 = 6.96 \text{ GeV}^2$
- Assuming $M_\pi L = 4$ and physical pion masses implies:
 \Rightarrow final state momentum of $\vec{p}_\pi \approx \frac{2\pi}{L}(7, 7, 7)$ to reach $q^2 \sim 0$.
- typical simulations cannot achieve (i.e. control) this
 \Rightarrow compromise in at least one of the following:
 - $M_\pi > M_\pi^{\text{phys}}$ (\Rightarrow need chiral extrapolation)
 - $m_b < m_b^{\text{phys}}$ (\Rightarrow need heavy quark mass extrapolation)
 - $q_{\text{min}}^2 \gg 0$ (\Rightarrow need kinematic extrapolation)

From correlators to the physical world

Extrapolations are based on theoretical foundations...

- Extraction of ground state parameters see also → Oliver Bär's talk
- M_π^{phys} (chiral) extrapolation guided by heavy meson chiral perturbation theory (HM χ PT)
- m_b^{phys} (heavy quark) extrapolation guided by HQET
- $a \rightarrow 0$ (continuum limit) extrapolation guided by Symanzik E.T.
- $q^2 = 0$ (kinematic) extrapolation guided by model independent z-expansion (BGL, BCL)
 - Physical q^2 dependence can be mapped to interval $z(q^2) \in [-z_{\text{max}}, z_{\text{max}}]$ with $0 < z_{\text{max}} \ll 1$
 - BGL expansion: $f_X(z) = \frac{1}{B_X \phi_X} \sum_i a_i z^i$, unitarity bounds $\sum_i |a_i|^2 < 1$.

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 - BGL expansion: $f_X(z) = \frac{1}{B_X \phi_X} \sum_i a_i z^i$, unitarity bounds $\sum_i |a_i|^2 < 1$.
... but they are intertwined and difficult

and all of them come with systematic uncertainties - are they controlled?

An example: $B_s \rightarrow K \ell \nu$ [JTT, RBC/UKQCD PRD 107 (2023) 114512]

Set up of our calculation and required extrapolations

Simulations

- 3 $a \in [0.07, 0.11]$ fm.
- finite volume $L \Rightarrow$ discrete momenta $\mathbf{p} = \frac{2\pi}{L} \mathbf{n}$.
- heavier-than-physical $M_\pi^{\text{sim}} \in [270, 430]$ MeV.

Physics

- Need to recover $a \rightarrow 0$ limit.
- Need continuous description of $q^2 = M_{B_s}^2 + M_K^2 - 2M_{B_s} E_K$.
- Recover physical light quark masses.

Use $\text{HM}\chi\text{PT}$ [NP B812 64,NP 840 54,PRD 67 054010] for extrapolation:

$$f_X^{B_s \rightarrow K}(M_\pi, E_K, a^2) = \frac{\Lambda}{E_K + \Delta_X} \left[c_{X,0} \left(1 + \frac{\delta f(M_\pi^s) - \delta f(M_\pi^p)}{(4\pi f_\pi)^2} \right) + c_{X,1} \frac{\Delta M_\pi^2}{\Lambda^2} + c_{X,2} \frac{E_K}{\Lambda} + c_{X,3} \frac{E_K^2}{\Lambda^2} + c_{X,4} (a\Lambda)^2 \right]$$

(+vary fit ansatz + estimate missing/H.O. terms)

Choice of ff basis: (f_0, f_+) vs $(f_{\parallel}, f_{\perp})$

- Interested in matrix elements of the vector current $\langle B_s | \mathcal{V}^\mu | K \rangle$
- Can be decomposed as

$$\langle K | \mathcal{V}^\mu | B_s \rangle = \sqrt{2M_{B_s}} [v^\mu f_{\parallel}(E_K) + p_{\perp}^\mu f_{\perp}(E_K)]$$

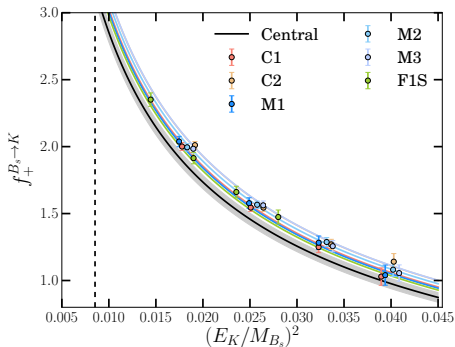
- trivially related to f_+, f_0 via

$$f_0(q^2) = \frac{\sqrt{2M_{B_s}}}{M_{B_s}^2 - M_K^2} [(M_{B_s} - E_K)f_{\parallel}(E_K) + (E_K^2 - M_K^2)f_{\perp}(E_K)]$$
$$f_+(q^2) = \frac{1}{\sqrt{2M_{B_s}}} [f_{\parallel}(E_K) + (M_{B_s} - E_K)f_{\perp}(E_K)]$$

⇒ Convenient: \mathcal{V}_0 vs \mathcal{V}_i to isolate f_{\parallel} and f_{\perp} from correlators

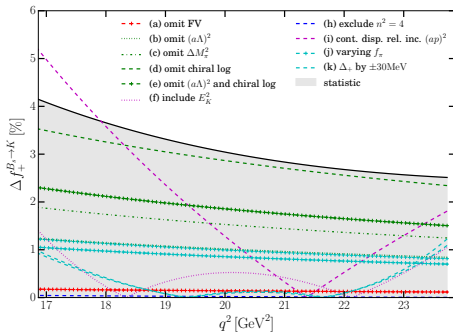
RBC/UKQCD 23 Fit results f_+

HM χ PT fit to lattice data

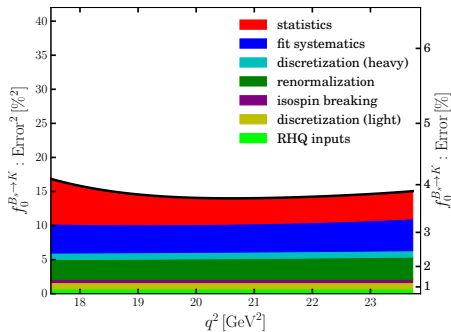
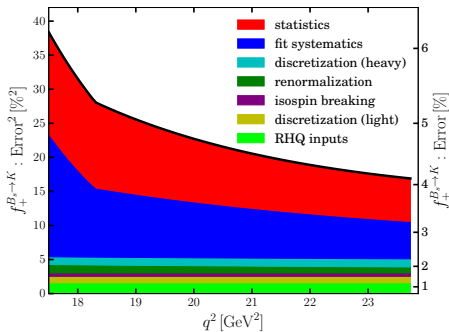


\Rightarrow take maximal deviation between the chosen fit and fit variation as fit-systematic value.

Fit systematics



Assembling the error budget



- Dominated by statistical and fit systematic uncertainties \Rightarrow both improvable!
- Most precise near q_{\max}^2
- Data covers range $q^2 > 17$ GeV²

Some insights from $B_s \rightarrow K$ (1)

$$f_X^{B_s \rightarrow K} = \frac{\Lambda}{E_K + \Delta_X} \times [\chi(M_\pi^2) + k(E_K) + d((a\Lambda)^2)]$$

where Δ_X is the “relevant pole mass”

$$\Delta_+ = M_{B^*(1-)} - M_{B_s}, \quad M_{B^*(1-)} = 5.32471 \text{ GeV} \quad (\text{exp.})$$

$$\Delta_0 = M_{B^*(0+)} - M_{B_s}, \quad M_{B^*(0+)} = 5.63 \text{ GeV} \quad (\text{the.})$$

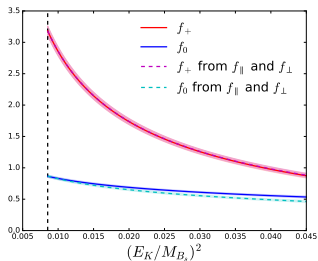
RBC/UKQCD'15 and FNAL/MILC'19 strategy:

1. **Assume** f_{\parallel} dominated by f_0 and f_{\perp} dominated by f_+ .
2. HM χ PT fit to f_{\parallel} , f_{\perp} using $\Delta_{\parallel} \sim \Delta_0$, $\Delta_{\perp} \sim \Delta_+$
3. converting to f_+ , f_0 in the continuum

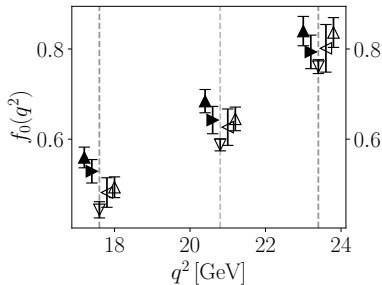
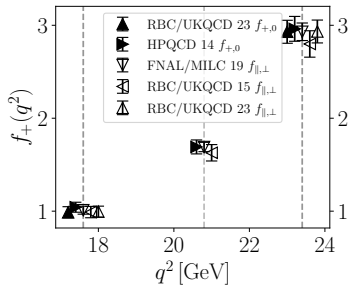
Is this justified?

And how to deal with poles when $m_h \neq m_b$?

Some insights from $B_s \rightarrow K$ (2)



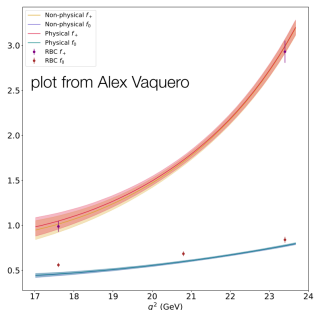
- ← All fine for f_+ (red vs magenta) ✓
- ← Several (stat) sigmas difference for f_0 ✗!!
- ← Discrepancy gets worse with increasing energy \Rightarrow easy to miss!
- ↓ picture persists with full error budget



\Rightarrow Not unique to $B_s \rightarrow K$, same strategy was used for $B \rightarrow \pi$

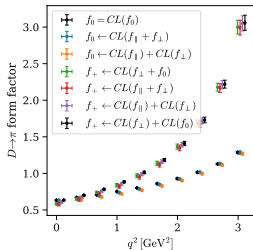
It remains puzzling

[from A. Kronfeld's talk last week]

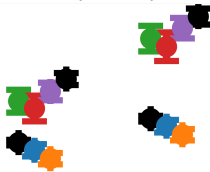


- (↑) Re-analysis of FNAL/MILC'19 doesn't reproduce what we see.
- But we see an effect in our data (prev. slide).

[FNAL/MILC, Phys.Rev.D 107 (2023) 9, 094516]



↓ (zoom)



There is an effect in $D_{(s)} \rightarrow \pi/K$

z-expansions and Bayesian inference

Theory inputs

- Lattice results *after* chiral-continuum limit typically provide
 - ffs in restricted kinematical range $q^2 \in [q_{\min, \text{sim}}^2, q_{\max}^2]$.
 - continuous description of ffs in terms of small number of basis functions (e.g. from $\text{HM}\chi\text{PT}$)
 - $q_{\max}^2 \equiv (M_i - M_f)^2$, so particularly relevant for $B \rightarrow \pi$ and $B_s \rightarrow K$.
 - e.g. for $B_s \rightarrow K$ $17 \text{ GeV}^2 \lesssim q_{\min, \text{sim}}^2$
- Sum rule results typically individual point(s) near $q^2 \approx 0$.
- Ff's satisfy kinematic constraint $f_+(0) = f_0(0)$.

GOAL: Want form factors over full range $[0, q_{\max}^2]$.

Extrapolating over the full kinematic range: z -expansion

- Map $q^2 \in [0, q_{\max}^2]$ to $z \in [z_{\min}, z_{\max}]$ with $|z| < 1$ and branch cut t_* .

$$z(q^2; t_0) = \frac{\sqrt{t_* - q^2} - \sqrt{t_* - t_0}}{\sqrt{t_* - q^2} + \sqrt{t_* - t_0}}$$

- Form factor is a polynomial in z after poles have been removed
e.g. BGL: Boyd, Grinstein, Lebed [PRL 74 4603]:

$$f_X(q^2) = \left(\prod_{\text{poles}} \frac{1}{B_X(q^2)} \right) \frac{1}{\phi_X(q^2)} \sum_{n \geq 0} a_{X,n} z^n$$

Goal: Determine some “un-truncated” number of coefficients $a_{X,n}$ to obtain model independent parameterisation.

Unitarity constraint

Substituting BGL into the unitarity constraint

$$\frac{1}{2\pi i} \oint_C \frac{dz}{z} |B_X(q^2)\phi_X(q^2, t_0)f_X(q^2)|^2 \leq 1,$$

gives

$$|\mathbf{a}_X|^2 \leq 1$$

(see [Gubernari, Reboud, van Dyk, Virto, JHEP 09 (2022) 133] and [Flynn, Jüttner, JTT, JHEP 12 (2023) 175] for modified versions of this constraint)

- **MUST** be satisfied for any believable fit!
- Provides strong constraint on allowed size of coefficients $a_{X,i}$.
- In practice, any z -expansion fit must necessarily be truncated:

$$B_X\phi_X f_X = \sum_{n=0}^{K_X-1} a_{X,n} z^n$$

- **How can we best make use of this?**

Frequentist z -expansion fit: pro's and con's

Typically only small number of coefficients in $\text{HM}_{\chi^2\text{PT}}$:

$$f_X^{B_s \rightarrow K} = \frac{\Lambda}{E_K + \Delta_X} \left(c_{X,0} + c_{X,1} \frac{E_K}{\Lambda} + c_{X,2} \frac{E_K^2}{\Lambda^2} \right)$$

- ⇒ limited number of independent data points N_X - typically $\approx 2-4$ per ff.
- ⇒ Maximum meaningful truncation K_X of the z -expansion requires

$$\underbrace{N_+ + N_0}_{\text{input data}} - \underbrace{(K_+ + K_0)}_{\text{fit parameters}} + \underbrace{1}_{\text{kinematic constraint}} = N_{\text{dof}} \geq 1$$

Goal: Determine \mathbf{a} , given data \mathbf{f} , covariance C_f and Z (encoding ϕ_X, B_X , kinematic constraint)

$$\chi^2(\mathbf{a}, \mathbf{f}) = [\mathbf{f} - Z\mathbf{a}]^T C_f^{-1} [\mathbf{f} - Z\mathbf{a}]^T .$$

- pro** Clear measure to assess quality of fit (p -value).
- con** No satisfactory way to assess systematic effect of truncation.
- con** Can only check unitarity constraint *a posteriori*.

Bayesian ansatz: Our *prior knowledge*

Our prior knowledge about probability distribution for

1. data points $\pi_{\mathbf{f}}(\mathbf{f}|\mathbf{f}_p, C_{\mathbf{f}_p}) \propto \exp\left(-\frac{1}{2}(\mathbf{f} - \mathbf{f}_p)^T C_{\mathbf{f}_p}^{-1}(\mathbf{f} - \mathbf{f}_p)\right)$
2. BGL ansatz: $\Theta(\mathbf{f}, \mathbf{a}|Z) \propto \delta(|\mathbf{f} - Z\mathbf{a}|)$
3. unitarity: $\pi_{\mathbf{a}}(\mathbf{a}|\text{unitarity}) \propto \theta(1 - |\mathbf{a}_+|^2) \theta(1 - |\mathbf{a}_0|^2)$

Marginalising over \mathbf{f} (using 1. and 2.) gives

$$\pi_{\mathbf{a}}(\mathbf{a}|\mathbf{f}_p, C_{\mathbf{f}_p}) \propto \exp\left(-\frac{1}{2}\chi^2(\mathbf{a}, \mathbf{f}_p)\right)$$

Now combining these

$$\begin{aligned}\pi_{\mathbf{a}}(\mathbf{a}|\mathbf{f}_p, C_{\mathbf{f}_p})\pi_{\mathbf{a}}(\mathbf{a}|\text{unitarity}) &\propto \theta(\mathbf{a}) \exp\left(-\frac{1}{2}\chi^2(\mathbf{a}, \mathbf{f}_p)\right) \\ &= \theta(\mathbf{a}) \exp\left(-\frac{1}{2}(\mathbf{a} - \tilde{\mathbf{a}})^T C_{\tilde{\mathbf{a}}}^{-1}(\mathbf{a} - \tilde{\mathbf{a}})\right)\end{aligned}$$

with $\tilde{\mathbf{a}} = \tilde{C}_{\mathbf{f}_p}(Z^T C_{\mathbf{f}_p}^{-1} \mathbf{f}_p)$ and $C_{\tilde{\mathbf{a}}}^{-1} = Z^T C_{\mathbf{f}_p}^{-1} Z$

Bayesian Inference regulated by unitarity and analyticity

Compute the expectation value $g(\mathbf{a})$ in the presence of “*prior knowledge B*” via Bayes’ theorem as (with normalisation $\mathcal{Z} = \int d\mathbf{a} \pi(\mathbf{a}|B)$)

$$\begin{aligned}\langle g(\mathbf{a}) \rangle &= \frac{1}{\mathcal{Z}} \int d\mathbf{a} g(\mathbf{a}) \pi(\mathbf{a}|B), \\ &= \frac{1}{\mathcal{Z}} \int d\mathbf{a} g(\mathbf{a}) \theta(1 - |\mathbf{a}_+|^2) \theta(1 - |\mathbf{a}_0|^2) \exp\left(-\frac{1}{2}(\mathbf{a} - \tilde{\mathbf{a}})^T C_{\tilde{\mathbf{a}}}^{-1}(\mathbf{a} - \tilde{\mathbf{a}})\right)\end{aligned}$$

⇒ Computable via Monte-Carlo integration by drawing from multivariate normal distribution $\mathcal{N}(\tilde{\mathbf{a}}, C_{\tilde{\mathbf{a}}})^1$.

con No clear measure to assess quality of fit (p -value).

pro Unitarity constraint automatically satisfied

pro Remove truncation error by computing as many terms as relevant

¹See Flynn, Jüttner, JTT, JHEP 12 (2023) 175 for details of efficient algorithm

Application I: RBC/UKQCD'23 [JTT et al., PRD 107, 114512]

$(N_+, N_0) = (2, 3)$ independent datapoints as input for z-exp.

Frequentist: limited to $K_+ + K_0 \leq 5$ truncation uncertainty unquantified!

Bayesian: unlimited order \Rightarrow Can investigate effect of truncation! \checkmark

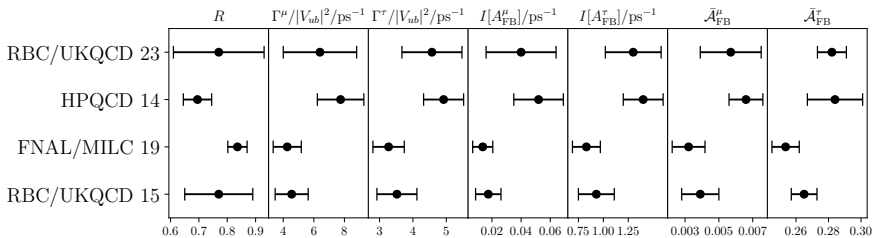
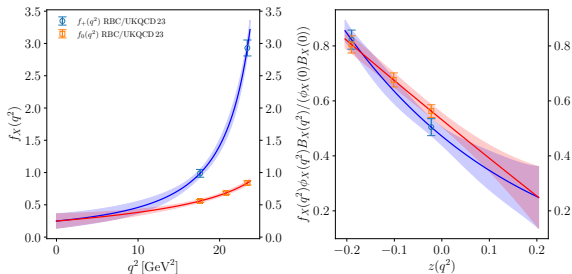
K_+	K_0	$a_{0,0}$	$a_{0,1}$	$a_{0,2}$	$a_{0,3}$	$a_{0,4}$	$a_{0,5}$	$a_{0,6}$	$a_{0,7}$
2	2	0.0981(36)	-0.286(14)						
2	3	0.0917(39)	-0.331(19)	-0.211(53)					
3	2	0.0950(37)	-0.263(15)						
3	3	0.0953(43)	-0.254(41)	0.02(13)					
4	4	0.0953(42)	-0.254(42)	0.02(21)	-0.02(60)				
5	5	0.0954(44)	-0.254(41)	0.02(21)	-0.01(55)	-0.00(62)			
6	6	0.0957(42)	-0.251(41)	0.04(21)	-0.01(52)	-0.06(65)	0.07(65)		
7	7	0.0955(44)	-0.250(40)	0.06(20)	0.05(50)	-0.13(72)	0.17(79)	-0.12(69)	
8	8	0.0954(43)	-0.250(41)	0.06(22)	0.06(50)	-0.18(84)	0.2(1.0)	-0.21(99)	0.10(74)
K_+	K_0	$a_{+,0}$	$a_{+,1}$	$a_{+,2}$	$a_{+,3}$	$a_{+,4}$	$a_{+,5}$	$a_{+,6}$	$a_{+,7}$
2	2	0.0293(11)	-0.0871(46)						
2	3	0.0249(16)	-0.0999(57)						
3	2	0.0245(16)	-0.0799(50)	0.093(21)					
3	3	0.0245(15)	-0.078(12)	0.101(49)					
4	4	0.0246(17)	-0.077(32)	0.100(68)	-0.03(70)				
5	5	0.0246(17)	-0.074(31)	0.099(70)	-0.08(67)	0.05(70)			
6	6	0.0247(16)	-0.073(32)	0.101(69)	-0.10(69)	0.09(74)	-0.05(71)		
7	7	0.0247(17)	-0.071(33)	0.107(70)	-0.11(72)	0.08(89)	-0.04(89)	0.03(73)	
8	8	0.0248(17)	-0.068(35)	0.102(74)	-0.18(77)	0.2(1.1)	-0.2(1.3)	0.1(1.2)	-0.06(82)

\Rightarrow Stable results for $(K_+, K_0) \sim (5, 5)$

Results I: RBC/UKQCD'23 [JTT et al., PRD 107, 114512]

- B.I. converges from $(K_+, K_0) \sim (5, 5)$
- $10^3 \times |V_{ub}| = 3.78(61)$
- $R_{B_s \rightarrow K} = 0.77(16)$
- $R_{B_s \rightarrow K}^{\text{impr}} = 1.72(11)$

⇒ more pheno in paper



Status of the Literature for $B_s \rightarrow K\ell\nu$

Joint frequentist fit to RBC/UKQCD23, HPQCD14 and FNAL/MILC19:

K_+	K_0	$a_{0,0}$	$a_{0,1}$	$a_{0,2}$	$a_{0,3}$	$a_{0,4}$	p	χ^2/N_{dof}	N_{dof}
2	2	0.0854(17)	-0.2565(75)	-	-	-	0.00	5.15	14
3	3	0.0864(18)	-0.2379(95)	0.061(28)	-	-	0.00	3.89	12
4	4	0.0887(27)	-0.08(17)	2.2(2.4)	7.0(7.9)	-	0.00	4.53	10
5	5	0.0887(28)	0.07(20)	6.1(3.3)	41.5(19.0)	93.3(44.0)	0.00	5.04	8

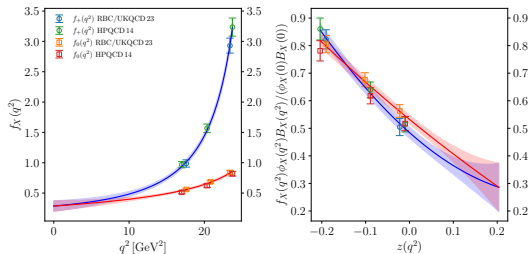
K_+	K_0	$a_{+,0}$	$a_{+,1}$	$a_{+,2}$	$a_{+,3}$	$a_{+,4}$	p	χ^2/N_{dof}	N_{dof}
2	2	0.02641(58)	-0.0824(26)	-	-	-	0.00	5.15	14
3	3	0.02534(73)	-0.0792(31)	0.062(12)	-	-	0.00	3.89	12
4	4	0.02592(97)	-0.033(50)	0.69(69)	2.1(2.3)	-	0.00	4.53	10
5	5	0.0266(10)	0.052(65)	2.21(97)	11.1(5.6)	17.2(15.1)	0.00	5.04	8

This data is incompatible \Rightarrow should not be fitted simultaneously
Frequentist fit ideally suited to detect this!

Exploit Complementarity of Frequentist and Bayesian!

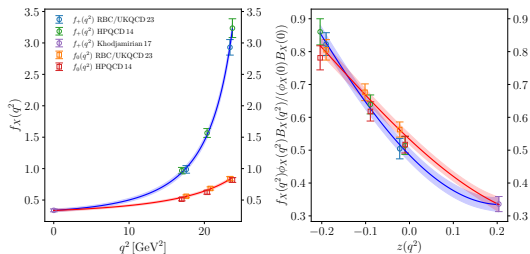
1. Assess compatibility of data first (Frequentist)
2. Perform truncation independent z-expansion to compatible datasets

Results II: Joint analysis to multiple datasets



Joint fit to

- RBC/UKQCD23
- HPQCD14



Joint fit to

- RBC/UKQCD23
- HPQCD14
- Khodjamirian17 (sum rules)

Easy to jointly fit multiple datasets!

⇒ more pheno in [Flynn, Jüttner, JTT, JHEP 12 (2023) 175]

Code: BFF

Flynn, A.J., Tsang, arXiv:2303.11285

Python3 available via [github](https://github.com/andreasjuettner/BFF)/[Zenodo](https://zenodo.org/record/7799543#.ZEezTy8Ro80)

<https://github.com/andreasjuettner/BFF>

<https://zenodo.org/record/7799543#.ZEezTy8Ro80>

```
#####  
# specify input for BGL fit  
#####  
input_dict = {  
    'decay':      'Btopi',      # initial-state mass  
    'Ml':         pc.mBphys,    # final-state mass  
    'Ma':         pc.mBphys,    # final-state mass  
    'sigma':      5,           # sigma for prior in algorithm  
    'Kp':         4,           # target Kp (BGL truncation) - can be changed later  
    'K0':         4,           # target K0 (BGL truncation) - can be changed later  
    'tstar':      '29.349570696829012', # value of t*  
    't0':         'self.tstar - np.sqrt(self.tstar*(self.tstar-self.tm))', # definition of t0  
    'chip':       pc.chip_Btopi, # susceptibility fp  
    'chi0':       pc.chi0_Btopi, # susceptibility f0  
    'mpolep':     [pc.mBstar],   # fplus pole  
    'mpole0':     [],           # fzero pole (no pole for Bstak)  
    'N':          N,           # number of desired samples  
    'outer_p':    [3./2, '48*np.pi', 3, 2], # specs for outer function fp  
    'outer_0':    [3./2, '16*np.pi/(self.tp*self.tm)', 1, 1], # specs for outer function f0  
    'seed':       123,        # RNG seed  
}
```

```
input_data = {  
    'RBCUKQCD 23 lat':  
    {  
        'data type': 'ff',  
        'label'      : 'RBC/UKQCD 23',  
        'Np':        2,  
        'N0':        3,  
        'qsqp':      np.array([17.60, 23.40]),  
        'qsq0':      np.array([17.60, 20.80, 23.40]),  
        'fp':        fpararray,  
        'f0':        f0array,  
        'Cff':       cov_array  
    }  
}
```

[taken from A.Jüttner's talks at LHCb Implications 2023]

Improvements and Summary

Literature overview $B \rightarrow \pi, B_s \rightarrow K$

	heavy	light	N_{ens}	N_a	$M_{\pi}^{\text{min}}/\text{MeV}$
$B \rightarrow \pi$					
HPQCD'06	NRQCD	asqtad	6	2	400
RBC/UKQCD'15	RHQ	DWF	5	2	300
FNAL/MILC'15	Fermilab	asqtad	12	4	255
JLQCD'22	DWF²	DWF	11	3	230
$B_s \rightarrow K$					
HPQCD'14	NRQCD	asqtad	5	2	260
FNAL/MILC'19	Fermilab	asqtad	6	3	255
RBC/UKQCD'23	RHQ	DWF	6	3	270

- work in progress by several collaborations
- typically improving over **continuum limit, i.e.** $N_a, M_{\pi}^{\text{min}}$
- expect more fully-relativistic results to come

$$^2 m_h^{\text{max}} \approx 2.44 m_c$$

Example: Taming discretisation effects via massive NPR?

Strategy:

1. renormalise $\mathcal{O}_{\text{bare}}$ in *regularisation independent* scheme
2. take the continuum limit
3. perturbatively match from this scheme to $\overline{\text{MS}}$

Caveat: Typical schemes (RI/(S)MOM) defined in massless limit of QCD

- mass independent renormalisation constants
- introduces discretisation effects scaling with $(am_q)^n$.
- on typical lattices $am_c \sim 0.2$ $am_b \lesssim 1$. Large cut-off effects!

for HQ: $a \rightarrow 0$ often limit hardest to control:

Way out: RI/mSMOM (defined at finite $\overline{m} \geq 0$)? [Boyle et al., 2016]

ADVANTAGE: Different masses at which the scheme is defined.

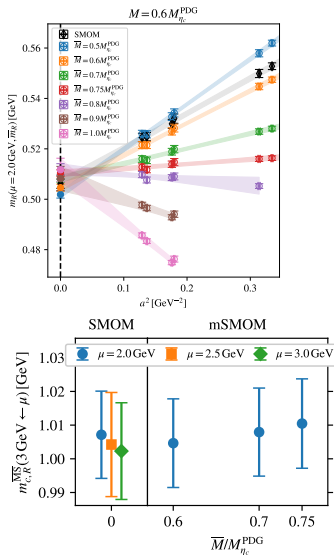
Tunable! - different approaches to $a \rightarrow 0$?

Possible to choose this to reduce cut-off effects?

JTT et al.'24: First numerical implementation of mSMOM.

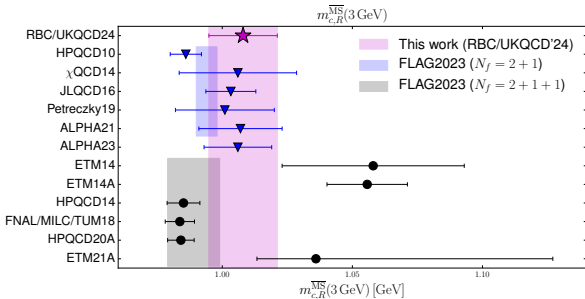
modified approach to the continuum

[JTT et al., PRD 110, (2024) 5, 054512]



← continuum limit results still in different schemes! Values cannot be directly compared from plot.

- Very different CL approaches
- Free choice of tunable parameter



⇒ Many possible applications in flavour and beyond!

Summary

The status quo is not great:

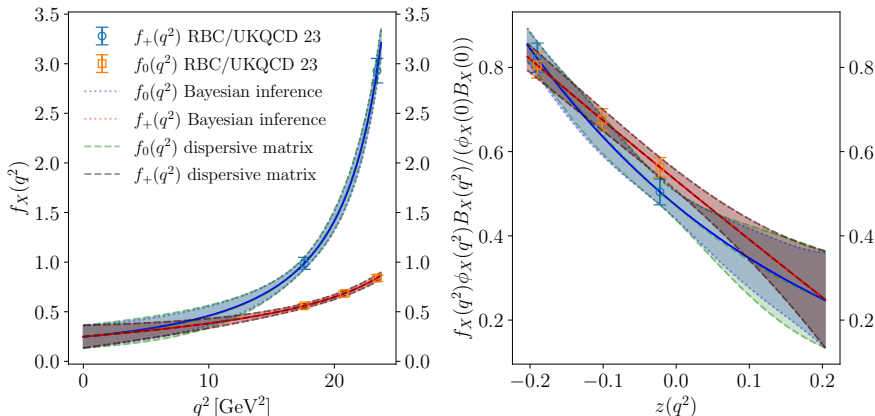
Theory calculations of the same quantities should not disagree!

BUT:

- These are hard and involved calculations.
- Existing literature is **highly complementary**.
- **Many works in progress** \Rightarrow might resolve tensions/provide understanding.
- Suggestion of **benchmark quantities** to scrutinise less complicated intermediate results [JTT, Della Morte, EPJ-ST 233 (2024) 2, 253-27].
- Expect **improvement and scrutiny** of all aspects of these calculations (HQET-inspired vs fully relativistic; excited states; chiral-continuum extrapolation; z -expansions,...).

ADDITIONAL SLIDES

Results III: Bayesian Inference vs Dispersive Matrix Method



- BI and DM produce equivalent results
- BI simpler to implement + easily to combine with multiple data sets
- BI provides numerical value of the coefficients