

Mapping the theory space of the flavor puzzle: the Froggatt-Nielsen case

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Based on: 2306.08026 with D. Curtin, E. Neil, J. Thompson 2501.00629, with D. Curtin, G. Krnjaic, M. Mellors

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Flavor puzzle = a series of puzzling observations:

3 copies of each species, identical from the point of view of gauge interactions, yet:

- 12 orders of magnitude from neutrinos to the top mass
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Common approach: describe in terms of a symmetry and its breaking

- "Deeper" origin can vary:
 - flavor-dependent gauge interactions
 - geometry, e.g. localisation of fermions in extra dimension
- Many examples: discrete, U(2)ⁿ from flavor deconstruction, Froggatt-Nielsen

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- When it works "locally": $\exists \underline{a} \text{ set of } c_{ij} \sim \mathcal{O}(1)$ that fits data
- When it works "globally": generic $c_{ij} \sim \mathcal{O}(1)$ reproduces data



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We'll use the global definition to explore which FN ansätze are good solutions to the quark & lepton flavor puzzle(s)

A quick review of the Froggatt-Nielsen mechanism

(in its simplest version)

• The SM is extended by a U(1)_x symmetry, spontaneously broken by a scalar ϕ , the *flavon*, with $\epsilon = \frac{\langle \phi \rangle}{\Lambda_F} \ll 1$

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- Yukawas arise from higher-dim. operators suppressed by powers of ϵ

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• The same selection rules hold for other higher-dimensional operators

$$\frac{c_{ijkl}}{\Lambda^2} \left(\bar{\psi}_i \psi_j \right) \left(\bar{\psi}_k \psi_l \right) \epsilon^{n_{ijkl}}$$
$$n_{ijkl} \equiv |X_{\psi_i} - X_{\psi_j}| + X_{\psi_k} - X_{\psi_l}$$

 \Rightarrow predictions for flavor-violating processes bear the fingerprint of textures!

What we want to do

• Define a notion of <u>global</u> goodness for a given texture

[we know one can "always" fit almost any FN model to the SM, but that's not the point;The model should 'want' to look like the SM with 'O(1)' parameters]

- Use it to rank textures for quarks & leptons [separately]
- Identify predictions across viable textures
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Previous works have tackled FN textures from various angles, e.g.

- Bayesian analysis of specific textures: Altarelli, Feruglio, Masina [0210342, 1207.0587]
- Finding "locally" good textures with small charges: Fedele, Mastrodii, Valli [2009.05587]
- Leptonic FN + CPV in MSSM Aloni et al. [2104.02679]
- coincidentally with our lepton paper, Ibe, Shirai, Watanabe [2412.19484] performed a Bayesian scan very similar in spirit to our work

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$$C_{ij}^{u} \left(\frac{\langle \phi \rangle}{\Lambda_F}\right)^{|X_{Q_i} - X_{u_j}|} \bar{Q}_i \tilde{H} u_j$$

For each choice of coefficients:

- compute masses and mixings 0
- compute masses and mixings quantify how well they reproduce data via $\delta_{\max} \equiv \max_{\mathcal{O}} \left| \frac{\mathcal{O}_{FN}}{\mathcal{O}_{exp}}, \frac{\mathcal{O}_{exp}}{\mathcal{O}_{FN}} \right|$ 0

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• Define: $F_a \equiv$ fraction of models $|\delta_{\max} < a$

 F_a estimates "how much" of the parameter space is data-like for a given texture $F_2 = 0.5 \rightarrow$ half of models are within a factor 2 from data

Rank textures by F_2 [or smaller deviations, if statistics permits]

15%93% $X_Q = \{3, 2, 0\}$ $X_u = \{-4, -2, 0\}$ $X_Q = \{1, 1, 0\}$ 67% 0.09% 15% 2.7%0% $X_u = \{-2, 0, 0\}$ 10% $X_d = \{-2, -2, -2\}$ $X_d = \{-3, -3, -3\}$ 10%5%5% δ_{\max}^{5} 23 5 $\overline{7}$ 10 20 30 50100 20 10 30 1 23 71 $\delta_{
m max}$

Example: distribution of $\delta_{\rm max}$ for a "good" and a "bad" texture

[Obviously $F_a \to 0$ as $a \to 1$, but top charge assignments tend to give $F_5 \sim 50 \%$, $F_2 \sim \text{few \%}$]

F_a as a global "goodness" criterium

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- ✓ A "high" F_5 (50%) implies "high" F_2 (few %)
 - useful in practice: trying ~ 10 models often gives a rough idea of quality
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What is a "natural fit"?

1. It should involve only O(1) coefficients -i.e. reproduce masses and mixings due to FN charges & ϵ , without relying on accidental hierarchies in the c_i

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2. It should be collectively untuned, i.e. stable under collective deformations of the c_k

A standard approach is to use the Barbieri-Giudice tuning measure:

$$\Delta_{BG}^{K} = \max_{k} \left| \frac{\partial \log \mathcal{O}_{K}}{\partial \log c_{k}} \right| \Rightarrow \Delta_{BG} = \sum_{K} \Delta_{BG}^{K}$$

...but not suitable for models that derive from a UV competition where jiggling one UV parameter will jiggle all the IR coefficients.

Instead:

$$\Delta_{\text{tot}}^{K} = \sqrt{\sum_{s} (\lambda_{s}^{K})^{2}} \quad , \ \lambda_{s}^{K} \in \text{Eig}\left(\frac{\partial^{2} \log \mathcal{O}_{K}}{\partial \log c_{k} \partial \log c_{l}}\right) \implies \Delta_{\text{tot}} = \sum_{K} \Delta_{\text{tot}}^{K}$$

We now have, for a given type of FN model, a way of ranking charge assignments by how data-like they want to be.

Next: interrogate the "top" textures for their predictions (e.g. for flavor-violating processes)

- $^\circ\,$ find an ensemble of "natural" fits starting from the coefficient choices with small $\delta_{
 m max}$
- for each fit, generate random O(1) coefficients for SMEFT operators
- get distributions of predictions

Setup
$$L_Y \supset -c_{ij}^u \left(\frac{\langle \phi \rangle}{\Lambda_F}\right)^{|X_{Q_i} - X_{u_j}|} \bar{Q}_i \tilde{H} u_j - c_{ij}^d \left(\frac{\langle \phi \rangle}{\Lambda_F}\right)^{|X_{Q_i} - X_{d_j}|} \bar{Q}_i H d_j$$

- ° assume $X_H = 0$ [the $X_H \neq 0$ case can always be mapped to the $X_H = 0$ one]
- ° scan all integer $\{X_{Q_i}, X_{u_i}, X_{d_i}\}_{i=1,2,3}$ textures with $|X|_{max} = 4$
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Phenomenologically viable textures for $|X|_{max} = 4$

Num.	X_{Q_1}	X_{Q_2}	X_{u_1}	X_{u_2}	X_{d_1}	X_{d_2}	X_{d_3}	$\left \mathcal{F}_{2}\left(\% ight) ight $	$\left \mathcal{F}_{5} \left(\% ight) ight $	ϵ
1	3	2	-4	-2	-3	-3	-3	2.7	67	0.17
2	3	2	-4	-2	-4	-3	-3	2.5	66	0.18
3	3	2	-3	-1	-3	-2	-2	1.9	56	0.12
4	3	2	-4	-1	-3	-3	-3	1.5	65	0.16
5	4	3	-4	-2	-4	-3	-3	1.2	52	0.23
6	3	2	-4	-1	-3	-3	-2	1.1	63	0.15
7	4	2	-4	-2	-4	-3	-3	1.1	47	0.21
8	3	2	-3	-1	-2	-2	-2	0.9	41	0.11
9	3	2	-3	-1	-3	-3	-2	0.9	55	0.14
10	3	2	-4	-2	-3	-3	-2	0.9	59	0.16
11	2	1	-3	-1	-2	-2	-2	0.8	52	0.06
12	4	3	-4	-1	-4	-3	-3	0.8	52	0.22
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 [Leurer, Nir, Seiberg]
 - textures with minimal charges (0,1) are at most locally good, never globally

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 good textures have near degenerate charges in the down sector

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large right-handed rotations are needed to diagonalize down-type Yukawas

 \rightarrow large FCNCs (of BSM origin)

 \rightarrow K mixing bounds push $\Lambda_F \gtrsim 10^5 \,\text{TeV}$, close to the flavor anarchic case!



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Example: predictions for D and Bd mixing for two different textures, fixing $\Lambda_F = \Lambda_K$



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Caveat: this does <u>not</u> mean that a "good" low-scale explanation of quark flavor masses and mixings is not possible within FN.

Need to go beyond this simple setup, e.g.

- ° going much higher up in charges ($X_{\rm max} \sim 10$)
- ° add additional discrete symmetries, e.g. some \mathbb{Z}_N that makes Yukawa matrices upper triangular to begin with [see e.g. Greljio, Smolkovič, Valenti,2407.02998]
- have multiple flavons

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...but also **opportunities**:

- ° cLFV especially in the μ − e sector can be tested with extreme precision \Rightarrow potential access even to high-scale flavor models
- Cosmology measurements and $0\nu\beta\beta$ searches are set for big advancements

Setup

For charged leptons, implement FN analogously to quarks.

For neutrinos, two options:

smallness of m_{ν} comes entirely from FN

 $...\Lambda_W$ helps!

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For charged leptons, implement FN analogously to quarks.

For neutrinos, two options:

smallness of m_{ν} comes entirely from FN

 $....\Lambda_W$ helps!

Scan

Scan all integer lepton flavour charges up to $|X_{\ell'}|_{max} = 7(9)$ for Dirac (Majorana) Compare to $\mathcal{O} = \left\{ m_{\ell} , \Delta m_{ij}^2 , |V_{ij}| , \sum m_{\nu} \right\}$ Note: cosmological or lab input has no effect on results

Best textures for Dirac and Majorana neutrinos

		Top Dirac textures									Top Majorana textures										
L_1	L_2	L_3	$ar{e}_1$	\bar{e}_2	\bar{e}_3	N_1	N_2	N_3	ϵ	NO	L_1	L_2	$_2 L$	$_3 \bar{e}$	1	\bar{e}_2	$ar{e}_3$	ϵ	$\log \Lambda_{\!_{V}}$,NO	
6	5	5	-3	-2	0	9	8	8	0.10	96	2	0	-]	. 7	7	6	4	0.24	15	91	
3	3	3	2	-1	-6	9	9	8	0.07	99	5	5	-2	2 7	7.	-2 ·	-3	0.08	12	3	
3	3	3	2	-5	-6	9	9	8	0.07	99	4	4	3	L U	5	2	0	0.23	11	96	
7	7	6	-4	-2	0	9	9	9	0.14	99	7	6	5	7	7	3	0	0.39	11	97	
7	7	6	-4	-3	-1	9	7	7	0.11	99	6	6	5		5	1 .	-1	0.30	10	96	
3	3	3	2	0	-5	9	9	8	0.07	99	7	7	6	2	2	-1 ·	-3	0.23	7.6	96	
3	3	3	2	0	-1	9	9	8	0.07	99	5	5	4	6	3	2	0	0.30	11	96	
6	5	5	-3	-2	0	9	7	7	0.08	97	7	7	6	4	1	0 .	-2	0.30	9	96	
7	3	3	2	0	-5	9	9	9	0.08	93	5	5	-2	2 7	7.	-2 ·	-7	0.08	12	3	% of "models" predicting
6	6	6	-4	-3	-1	9	6	5	0.07	99	1	1	-1	'	7	-5 -	-4	0.18	15	2	normal ordering

Textures reproducing masses and mixings within a factor $\delta_{max} < 5, 2, 1.35$ for $\sim 50\%$, 2-5% and 0.03% of O(1) coefficient choices.

- Dirac requires large FN charges for RH neutrinos.
- FN favours normal ordering. Majorana neutrinos can also have inverted ordering.

0

CLFV predictions from Dirac FN models



Average predicted cLFV decay rates (relative to current constraint) for the 100 top Dirac FN textures

CLFV predictions from Dirac FN models



Average predicted cLFV decay rates (relative to current constraint) for the 100 top Dirac FN textures

CLFV predictions from Dirac FN models



Average predicted cLFV decay rates (relative to current constraint) for the 100 top Dirac FN textures

For this best-case scenario, Dirac FN textures yield **measurable** $\mu \rightarrow 3e$ and μ -e conversion in nuclei, the latter with significant spread between different textures

CLFV predictions from Majorana FN models

Average predicted cLFV decay rates (relative to current constraint) for the 100 top Majorana FN textures



CLFV predictions from Majorana FN models

Average predicted cLFV decay rates (relative to current constraint) for the 100 top Majorana FN textures



 $\mu \to e\gamma, \mu \to 3e$ and $\mu - e$ conversion are the most promising observables with $\Lambda = \Lambda_W$, only some textures give signals measurable in the near future

Discriminating FN textures via cLFV correlation

Correlations between $\mu \rightarrow 3e$ and μ –e conversion for two good Dirac textures



Different textures predict not just different signal strengths, but also different **correlations** between observables.

 \Rightarrow Detecting **multiple** cLFV signals could help **favour or exclude** certain textures.

(for the 100 best Majorana FN textures)



(for the 100 best Majorana FN textures)



(for the 100 best Majorana FN textures)



(for the 100 best Majorana FN textures)



All these textures are either detectable in next-gen $0\nu\beta\beta$ or within a 10x improvement \Rightarrow opens the door to a **complete "near"-future test** of the FN Majorana scenario

Conclusions

- We aimed at giving a bird's eye view of FN models in the quark & lepton sectors, ranking textures by how "generically" they reproduce data with O(1) coefficients and studying their phenomenology.
- In the quark sector:
 - Top textures involve quasi-degenerate RH down-quark charges \Rightarrow large RH rotations, hence large FCNCS, hence $\Lambda_F \gtrsim 10^5 \,\text{TeV}$
 - Direct collider phenomenology unlikely in the minimal setup without additional hypothesis (higher charges, discrete symmetries...)
- In the lepton sector:
 - FN favors normal ordering and predicts interesting cLFV patterns
 - Most top Majorana FN textures predict measurable rates for 0vββ
- These ideas can be extended to other symmetry-based solutions to the flavor puzzle