

# Testing an ansatz for the quark mass matrices

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## Abstract

The quark mass matrices are defined by two matrices whose eigenvalues are the quark masses (one matrix for each of the up-like and down-like quarks):

$$M'_i = \begin{bmatrix} \frac{(r'_i)^2}{s'_i} & p'_i & 0 \\ p_i & r'_i & q'_i \\ 0 & q_i & s_i \end{bmatrix} = U_i^T D'_i U_i, \quad (1)$$

where

$$D'_i = \begin{bmatrix} \lambda'_1 & 0 & 0 \\ 0 & \lambda'_2 & 0 \\ 0 & 0 & \lambda'_3 \end{bmatrix}_i = \begin{bmatrix} m_{u,d} & 0 & 0 \\ 0 & m_{c,s} & 0 \\ 0 & 0 & m_{t,b} \end{bmatrix}, \quad (2)$$

and  $U_i$  is the unitary matrix that diagonalises  $M'_i$  to  $D'_i$  whose rows are the eigenvectors of  $M'_i$ . The eigenvectors make up the columns of  $U^\dagger$  and therefore make up the rows of  $U^*=U$  in the real case.

We investigate the normalised mass matrices,  $M_i = \frac{M'_i}{s'_i}$  (with eigenvalues approximately equal to the ratio of quark masses):

$$M_i = \begin{bmatrix} r_i^2 & p_i & 0 \\ p_i & r_i & q_i \\ 0 & q_i & 1 \end{bmatrix} = U_i^T D_i U_i, \quad (3)$$

Since  $M_i$  is only changed by a scale factor from  $M'_i$  it is diagonalised by the same  $U_i$ .

Hence, setting  $i=u$ :

$$\begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix}_u = D_u = U_u M_u U_u^T = U_u \frac{M'_u}{s'_u} U_u^T = D'_u / s'_u = \begin{bmatrix} m_u/s'_u & 0 & 0 \\ 0 & m_c/s'_u & 0 \\ 0 & 0 & m_t/s'_u \end{bmatrix}, \quad (4)$$

Thus:

$$\frac{1}{\lambda_3^u} D_u = \begin{bmatrix} \frac{\lambda_1}{\lambda_3} & 0 & 0 \\ 0 & \frac{\lambda_2}{\lambda_3} & 0 \\ 0 & 0 & 1 \end{bmatrix}_u = \begin{bmatrix} m_u/m_t & 0 & 0 \\ 0 & m_c/m_t & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad (5)$$

and since  $\lambda_3 \approx 1$ :

$$D_u \approx \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & 1 \end{bmatrix}_u \approx \begin{bmatrix} m_u/m_t & 0 & 0 \\ 0 & m_c/m_t & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad (6)$$

Hence,

$$D_u = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix}_u \approx \begin{bmatrix} m_u/m_t & 0 & 0 \\ 0 & m_c/m_t & 0 \\ 0 & 0 & 1 \end{bmatrix}. \quad (7)$$

Similarly for  $M_d$

The purpose of this project is to start with some "known" values of the eigenvalues of the  $M_i$  (i.e. the approximations  $\lambda_1 = m_u/m_t$  etc.) and to construct reasonable estimates for the  $M_i$  ( $i=u,d$ ) by comparing values of the constructed CKM matrix  $V = U_u U_d^T$  to previously measured values. And then to do the same allowing the eigenvalues to vary, and for the case that the  $p_i$  and  $q_i$  are complex.

## Fixed eigenvalues

### Set up

We first want to find equations for the  $p_i$  and  $q_i$  as functions of the  $\lambda_i$  and  $r_i$ . Start by noting that, by the unitarity of  $U_i$  (dropping the  $i$  for simplicity):

$$\text{Det}(M) = \text{Det}(D) \Rightarrow r^3 - q^2 r^2 - p^2 = \prod_{n=1}^3 \lambda_n \approx \lambda_1 \lambda_2, \quad (8)$$

$$\text{Tr}(M) = \text{Tr}(D) \Rightarrow r^2 + r + 1 = \sum_{n=1}^3 \lambda_n \approx \lambda_1 + \lambda_2 + 1, \quad (9)$$

$$\sum_{ij=11,22,33} [M]_{ij} = \sum_{ij=11,22,33} [D]_{ij} \Rightarrow r^3 + r^2 + r - q^2 - p^2 = \lambda_1 \lambda_2 + \lambda_2 \lambda_3 + \lambda_1 \lambda_3 \approx \lambda_1 \lambda_2 + \lambda_2 + \lambda_1. \quad (10)$$

Hence,

$$\sum_{ij=11,22,33} [M]_{ij} - \text{Det}(M) = q^2(r^2-1) + r^2 + r \approx \lambda_1 + \lambda_2, \quad (11)$$

$$q^2 \approx \frac{\lambda_1 + \lambda_2 - r^2 - r}{r^2 - 1}. \quad (12)$$

And substituting  $q^2$  into  $\text{Det}(M)$ :

$$r^3 - \left( \frac{\lambda_1 + \lambda_2 - r^2 - r}{r^2 - 1} \right) r^2 - p^2 \approx \lambda_1 \lambda_2, \quad (13)$$

$$p^2 \approx \frac{\lambda_1 \lambda_2 - (\lambda_1 + \lambda_2 + \lambda_1 \lambda_2) r^2 + r^4 + r^5}{r^2 - 1}. \quad (14)$$

Next we want to find the  $U_i$ , we start by finding the ratios of the elements of the eigenvectors. Again dropping the  $i$  suffix and using  $(M - \lambda_k I) \vec{v}_k = \vec{0}$  ( $k=1,2,3$ ):

$$\begin{bmatrix} r^2 - \lambda_k & p & 0 \\ p & r - \lambda_k & q \\ 0 & q & 1 - \lambda_k \end{bmatrix} \begin{bmatrix} x_k \\ y_k \\ z_k \end{bmatrix} = \vec{0}. \quad (15)$$

Thus,

$$(r^2 - \lambda_k)x_k + py_k = 0, \quad (16)$$

$$\Rightarrow \frac{x_k}{y_k} = \frac{p}{\lambda_k - r^2}, \quad (17)$$

$$px_k + (r - \lambda_k)y_k + qz_k = 0, \quad (18)$$

$$qy_k + (1 - \lambda_k)z_k = 0, \quad (19)$$

$$\Rightarrow \frac{y_k}{z_k} = \frac{\lambda_k - 1}{q}. \quad (20)$$

And multiplying (11) and (14) gives:

$$\frac{x_k}{z_k} = \frac{p(\lambda_k - 1)}{q(\lambda_k - r^2)}. \quad (21)$$

Recall that the  $\vec{v}_k$  make up the rows of  $U$  so:

$$U = \begin{bmatrix} x_1 & y_1 & z_1 \\ x_2 & y_2 & z_2 \\ x_3 & y_3 & z_3 \end{bmatrix}. \quad (22)$$

And as  $U$  is a unitary matrix we can write (since we have assumed real parameters i.e. the phase  $\delta=0$  and therefore  $e^{i\delta}=1$ ):

$$U = \begin{bmatrix} c_{12}c_{13} & s_{12}c_{13} & s_{13} \\ -s_{12}c_{23} - c_{12}s_{23}s_{13} & c_{12}c_{23} - s_{12}s_{23}s_{13} & s_{23}c_{13} \\ s_{12}s_{23} - c_{12}c_{23}s_{13} & -c_{12}s_{23} - s_{12}c_{23}s_{13} & c_{23}c_{13} \end{bmatrix}, \quad (23)$$

where  $s_{ij} = \sin\theta_{ij}$ ,  $c_{ij} = \cos\theta_{ij}$  and the angles  $\theta_{ij}$  can lie in any of the four quadrants, so  $-1 \leq s_{ij} \leq 1$ ,  $-1 \leq c_{ij} \leq 1$ .

Now we have:

From (11):

$$\frac{\lambda_1 - r^2}{p} = \frac{y_1}{x_1} = \frac{s_{12}c_{13}}{c_{12}c_{13}} = \tan\theta_{12}, \quad (24)$$

$$\Rightarrow \theta_{12} = \arctan\left(\frac{\lambda_1 - r^2}{p}\right), \quad (25)$$

From (15):

$$\frac{q(\lambda_1 - r^2)}{p(\lambda_1 - 1)} = \frac{z_1}{x_1} = \frac{s_{13}}{c_{12}c_{13}}, \quad (26)$$

$$\Rightarrow \theta_{13} = \arctan\left(\frac{c_{12}q(\lambda_1 - r^2)}{p(\lambda_1 - 1)}\right), \quad (27)$$

From (14):

$$\frac{q}{\lambda_2 - 1} = \frac{z_2}{y_2} = \frac{s_{23}c_{13}}{c_{12}c_{23} - s_{12}s_{23}s_{13}} \approx \frac{s_{23}c_{13}}{c_{12}c_{23}} = \frac{c_{13}\tan\theta_{23}}{c_{12}}, \quad (28)$$

$$\Rightarrow \theta_{23} = \arctan\left(\frac{c_{12}q}{c_{13}(\lambda_2 - 1)}\right). \quad (29)$$

So now we can substitute the  $\theta_{ij}$  into the  $U_i$  to get the  $U_i$  as functions of  $\lambda_1^i$ ,  $\lambda_2^i$ , and  $r_i$  and we can easily construct the CKM matrix:

$$V(r_u, r_d, \lambda_1^u, \lambda_2^u, \lambda_1^d, \lambda_2^d) = U_u(r_u, \lambda_1^u, \lambda_2^u) U_d(r_d, \lambda_1^d, \lambda_2^d)^T. \quad (30)$$

## Tests

To test the accuracy of our constructed CKM matrix,  $V$ , we evaluate the  $\chi^2$  of the elements of the matrix against their values given by the PDG [1]. Due to the orthogonality of  $V$  we only test  $V_{12}$ ,  $V_{13}$ , and  $V_{23}$ . We take the observed values of the CKM matrix to be:

$$|V_{ij}^{measured}| = \begin{bmatrix} 0.97427 & 0.22536 & 0.00355 \\ 0.22522 & 0.97343 & 0.0414 \\ 0.00886 & 0.0405 & 0.99914 \end{bmatrix}. \quad (31)$$

We also define:

$$\sigma_{12}=0.0025, \sigma_{13}=0.0006, \text{ and } \sigma_{23}=0.004. \quad (32)$$

And now

$$\chi^2 = \sum_{ij} \frac{(|V_{ij}| - |V_{ij}^{measured}|)^2}{\sigma_{ij}^2} \quad (ij = 12, 13, 23). \quad (33)$$

We take the quark masses as known values from the PDG [1]:

$$m_u=0.0023, m_c=1.275, m_t=173.21, m_d=0.0048, m_s=0.095, m_b=4.66. \quad (34)$$

Now complications arise as we do not know the signs of the eigenvalues or the  $p_i$  and  $q_i$ . So there are 256 possible combinations of signs. But due to the invariance of the  $D_i$  and  $V$  under unitary transformations on the  $M_i$  of the form  $M'_i = U^\dagger M_i U$  (see Appendix: A), we have that  $V$  is unchanged by two transformations: i) flipping the signs of both  $p_u$  and  $p_d$ , and ii) flipping the signs of both  $q_u$  and  $q_d$ . Therefore every solution that we obtain will be in a set of four equivalent solutions (the three extra solutions gained by the transformations i), ii), i) and ii) ).

Since we have assumed  $p_i, q_i \in \mathbb{R}$  we have  $p_i^2 > 0, q_i^2 > 0$ . We can use this to obtain allowed ranges for the  $r_i$  (we also assume  $|r_i| \leq 0.5$ ). We use Wolfram Mathematica to gain solutions of  $Min(\chi^2) < 20$  within these bounds, keeping  $p_u$  and  $q_u$  positive and varying the signs of  $\lambda_1^u, \lambda_2^u, \lambda_1^d, \lambda_2^d, p_d, q_d$ , so that each returned solution represents a set of 4 solutions gained by the above described transformations.

Recall, we have the assumed values:  $|\lambda_1^u| = \frac{m_u}{m_t} = 0.0000132787, |\lambda_2^u| = \frac{m_c}{m_t} = 0.00736101, |\lambda_1^d| = \frac{m_d}{m_b} = 0.00103004, |\lambda_2^d| = \frac{m_s}{m_b} = 0.0203863$ .

The headings (e.g. (+, -, +, -, +, +)) refer to the signs of ( $\lambda_1^u$  -ve,  $\lambda_2^u$  +ve,  $\lambda_1^d$  +ve,  $\lambda_2^d$  +ve,  $p_d$  -ve,  $q_d$  +ve) respectively.

Table 1: Results

	(-, +, +, -, +, +)	(+, -, -, +, -, +)	(+, -, +, +, +, +)	(+, +, +, +, -, +)
$p_u$	$1.23 \times 10^{-3}$	$2.67 \times 10^{-4}$	$2.51 \times 10^{-4}$	$1.16 \times 10^{-3}$
$q_u$	0.0832	0.0731	0.0976	0.0835
$p_d$	$-1.07 \times 10^{-3}$	$-5.33 \times 10^{-3}$	$3.57 \times 10^{-3}$	$-1.26 \times 10^{-3}$
$q_d$	0.113	0.0289	0.148	0.114
$Min(\chi^2)$	17.9	10.3	12.4	15.3
Min point ( $r_u, r_d$ )	(0.0141, 0.0330)	( $-2.01 \times 10^{-3}$ , 0.0198)	( $2.20 \times 10^{-3}$ , 0.0414)	(0.0141, 0.0334)
rubounds	$7.30 \times 10^{-3} \leq r_u \leq 0.082$	$-3.64 \times 10^{-3} \leq r_u \leq 3.64 \times 10^{-3}$	$-3.64 \times 10^{-3} \leq r_u \leq 3.64 \times 10^{-3}$	$7.32 \times 10^{-3} \leq r_u \leq 0.0825$
rdbounds	$0.0321 \leq r_d \leq 0.134$	$0.0190 \leq r_d \leq 0.134$	$0.0321 \leq r_d \leq 0.134$	$0.0321 \leq r_d \leq 0.134$
$V_{12}$	0.225	0.223	-0.224	0.225
$V_{13}$	$-5.50 \times 10^{-3}$	$-1.70 \times 10^{-3}$	$2.30 \times 10^{-3}$	$-5.40 \times 10^{-3}$
$V_{23}$	-0.0304	0.0429	-0.0530	-0.0320
$V_{31}$	$-1.50 \times 10^{-3}$	0.0112	$9.60 \times 10^{-3}$	$-1.90 \times 10^{-3}$

## Treating the eigenvalues as variables

### Set up

Now, instead of holding the eigenvalues as fixed values, we wish to find ansatza for

$$M_i = \begin{bmatrix} r_i^2 & p_i & 0 \\ p_i & r_i & q_i \\ 0 & q_i & 1 \end{bmatrix}, \quad i=u,d \quad (35)$$

by allowing the eigenvalues to vary. The approximation  $\lambda_3^i = 1$  is automatically no longer used.

We take the  $\lambda_k$  as functions of  $p$ ,  $q$ , and  $r$ ,  $\lambda_1$  being the smallest eigenvalue and  $\lambda_3$  being the biggest. Now we want to find the  $U_i$ . Recall

$$U = \begin{bmatrix} x_1 & y_1 & z_1 \\ x_2 & y_2 & z_2 \\ x_3 & y_3 & z_3 \end{bmatrix} = \begin{bmatrix} c_{12}c_{13} & s_{12}c_{13} & s_{13} \\ -s_{12}c_{23} - c_{12}s_{23}s_{13} & c_{12}c_{23} - s_{12}s_{23}s_{13} & s_{23}c_{13} \\ s_{12}s_{23} - c_{12}c_{23}s_{13} & -c_{12}s_{23} - s_{12}c_{23}s_{13} & c_{23}c_{13} \end{bmatrix}, \quad (36)$$

where  $s_{ij} = \sin\theta_{ij}$ ,  $c_{ij} = \cos\theta_{ij}$  and the angles  $\theta_{ij}$  can lie in any of the four quadrants, so  $-1 \leq s_{ij} \leq 1$ ,  $-1 \leq c_{ij} \leq 1$ .

Again we have:

$$\theta_{12} = \arctan\left(\frac{\lambda_1 - r^2}{p}\right), \quad (37)$$

$$\theta_{13} = \arctan\left(\frac{c_{12}q(\lambda_1 - r^2)}{p(\lambda_1 - 1)}\right), \quad (38)$$

$$\theta_{23} = \arctan\left(\frac{c_{12}q}{c_{13}(\lambda_2 - 1)}\right). \quad (39)$$

So now, remembering that the  $\lambda_k^i$  are function of  $p_i$ ,  $q_i$ , and  $r_i$  we can substitute the  $\theta_{ij}$  into the  $U_i$  to get the  $U_i$  as functions of  $p_i$ ,  $q_i$ , and  $r_i$  and we can easily construct the CKM matrix:

$$V(r_u, p_u, q_u, r_d, p_d, q_d) = U_u(r_u, p_u, q_u)U_d(r_d, p_d, q_d)^T. \quad (40)$$

### Tests

To test the accuracy of our constructed CKM matrix,  $V$ , we again evaluate the  $\chi^2$  of the elements of the matrix against their values given by the PDG. Due to the orthogonality of  $V$  we only test  $V_{12}$ ,  $V_{13}$ , and  $V_{23}$ . We take the observed values of the CKM matrix to be:

$$|V_{ij}^{measured}| = \begin{bmatrix} 0.97427 & 0.22536 & 0.00355 \\ 0.22522 & 0.97343 & 0.0414 \\ 0.00886 & 0.0405 & 0.99914 \end{bmatrix}. \quad (41)$$

We also define:

$$\sigma_{12} = 0.0025, \sigma_{13} = 0.0006, \text{ and } \sigma_{23} = 0.004. \quad (42)$$

And now

$$\chi_v^2 = \sum_{ij} \frac{(|V_{ij}| - |V_{ij}^{measured}|)^2}{\sigma_{ij}^2} \quad (ij = 12, 13, 23). \quad (43)$$

We also test the accuracy of our values for the eigenvalues:

$$\chi_{\lambda_k}^2 = \sum_k \frac{(|\lambda_k| - |\lambda_k^{measured}|)^2}{\sigma_{\lambda_k}^2} \quad (k=1, 2, 3, 4), \quad (44)$$

where  $\lambda_1 = \lambda_1^u$ ,  $\lambda_2 = \lambda_2^u$ ,  $\lambda_3 = \lambda_1^d$ ,  $\lambda_4 = \lambda_2^d$ , and:

$$\sigma_{\lambda_k} = (0.000002, 0.0003, 0.00005, 0.0008).$$

We now use two methods to minimise  $\chi^2 = \chi_v^2 + \chi_{\lambda_k}^2$  with Wolfram Mathematica.

We can recover the solutions from the previous case:

Table 2: Recovered solutions

	1	2	3	4
$\lambda_1^u$	$-1.31 \times 10^{-5}$	$1.39 \times 10^{-5}$	$1.39 \times 10^{-5}$	$1.35 \times 10^{-5}$
$\lambda_2^u$	$7.36 \times 10^{-3}$	$-7.36 \times 10^{-3}$	$-7.45 \times 10^{-3}$	$7.37 \times 10^{-3}$
$\lambda_3^u$	1.01	1.01	1.01	1.01
$r_u$	0.0136	$-1.91 \times 10^{-3}$	$2.19 \times 10^{-3}$	0.0141
$p_u$	$1.19 \times 10^{-3}$	$2.75 \times 10^{-4}$	$2.60 \times 10^{-4}$	$1.16 \times 10^{-3}$
$q_u$	0.0798	0.0740	0.0985	0.0827
$\lambda_1^d$	$1.01 \times 10^{-3}$	$-1.06 \times 10^{-3}$	$9.98 \times 10^{-4}$	$1.01 \times 10^{-3}$
$\lambda_2^d$	0.0207	0.0201	0.0205	0.0204
$\lambda_3^d$	1.01	1.00	1.02	1.01
$r_d$	0.0329	0.0195	0.0413	0.0332
$p_d$	$-1.24 \times 10^{-3}$	$-5.33 \times 10^{-3}$	$3.67 \times 10^{-3}$	$-1.32 \times 10^{-3}$
$q_d$	0.110	0.0290	0.145	0.113
$Min(\chi^2)$	16.4	9.36	9.87	14.7
$V_{12}$	0.225	0.225	-0.225	0.225
$V_{13}$	$-5.33 \times 10^{-3}$	$-1.80 \times 10^{-3}$	$2.24 \times 10^{-3}$	$-5.19 \times 10^{-3}$
$V_{23}$	-0.0307	0.0438	-0.0497	-0.0308
$V_{31}$	$-1.72 \times 10^{-3}$	0.0116	$8.95 \times 10^{-3}$	$-1.87 \times 10^{-3}$

And obtain new solutions:

Table 3: New solutions

	5	6	7	8
$\lambda_1^u$	$1.40 \times 10^{-5}$	$1.44 \times 10^{-5}$	$1.40 \times 10^{-5}$	$1.41 \times 10^{-5}$
$\lambda_2^u$	$7.20 \times 10^{-3}$	$-7.32 \times 10^{-3}$	$7.22 \times 10^{-3}$	$-7.36 \times 10^{-3}$
$\lambda_3^u$	1.00	1.01	1.00	1.01
$r_u$	$8.31 \times 10^{-3}$	$3.56 \times 10^{-4}$	$7.21 \times 10^{-3}$	$1.35 \times 10^{-3}$
$p_u$	$-6.26 \times 10^{-4}$	$3.24 \times 10^{-4}$	$5.22 \times 10^{-4}$	$-3.00 \times 10^{-4}$
$q_u$	0.0340	0.0878	$5.09 \times 10^{-3}$	-0.0936
$\lambda_1^d$	$9.26 \times 10^{-4}$	$1.03 \times 10^{-3}$	$9.42 \times 10^{-4}$	$-1.04 \times 10^{-3}$
$\lambda_2^d$	-0.0221	-0.0204	-0.0218	0.0203
$\lambda_3^d$	1.00	1.00	1.00	1.00
$r_d$	-0.0216	-0.0178	-0.0198	0.0212
$p_d$	$-3.22 \times 10^{-3}$	$-3.84 \times 10^{-3}$	$3.50 \times 10^{-3}$	$5.45 \times 10^{-3}$
$q_d$	$-4.85 \times 10^{-3}$	0.0437	-0.0390	-0.0487
$Min(\chi^2)$	10.4	6.32	7.96	6.64
$V_{12}$	-0.227	-0.225	0.227	-0.225
$V_{13}$	$-3.37 \times 10^{-3}$	$-2.13 \times 10^{-3}$	$2.99 \times 10^{-3}$	$-2.06 \times 10^{-3}$
$V_{23}$	0.0388	0.0440	0.0430	-0.0430
$V_{31}$	$-5.52 \times 10^{-3}$	$-7.84 \times 10^{-3}$	$6.83 \times 10^{-3}$	0.0120



## Complex case

Now we test

$$M_i = \begin{bmatrix} r_i^2 & p_i & 0 \\ p_i^* & r_i & q_i \\ 0 & q_i^* & 1 \end{bmatrix} = U_i^\dagger D_i U_i, \quad (45)$$

with

$$D_i = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix}_i, \quad (46)$$

where the  $r_i, \lambda_k^i$  are real and the  $p_i, q_i$  are complex.

We have  $D_i = U_i M_i U_i^\dagger$ . Let

$$\Phi_i = \begin{bmatrix} e^{i\phi_1} & 0 & 0 \\ 0 & e^{i\phi_2} & 0 \\ 0 & 0 & e^{i\phi_3} \end{bmatrix}_i, \quad (47)$$

then, since the  $\Phi_i$  are diagonal

$$D_i = \Phi_i D_i \Phi_i^\dagger = \Phi_i U_i M_i U_i^\dagger \Phi_i^\dagger = U_i' M_i (U_i')^\dagger. \quad (48)$$

I.E. the  $D_i$  are invariant under transformations of the form  $U_i \rightarrow \Phi_i U_i$ . So we can choose the  $\Phi_i$  such that the diagonal elements of the  $U_i$  are real.

Note that  $V = U_u U_d^\dagger$  is not invariant under such transformations, instead we have:  $V' = \Phi_u U_u U_d^\dagger \Phi_d^\dagger$ . But the physics are unchanged (observable quantities such as  $|V|$ )...

We parameterise the complex (transformed)  $U_i$  with the general parameterisation (using the orthogonality of  $U_i$ )

$$U_i \approx \begin{bmatrix} 1 & U_{12} & U_{13} \\ -U_{12}^* & 1 & U_{23} \\ U_{31} & -U_{23}^* & 1 \end{bmatrix}_i. \quad (49)$$

## Set up

We want to find the  $U_i$ , we start by finding the ratios of the elements of the eigenvectors. Dropping the  $i$  suffix and using  $(M - \lambda_k I) \vec{v}_k = \vec{0}$  ( $k=1,2,3$ ):

$$\begin{bmatrix} r^2 - \lambda_k & p & 0 \\ p^* & r - \lambda_k & q \\ 0 & q^* & 1 - \lambda_k \end{bmatrix} \begin{bmatrix} x_k \\ y_k \\ z_k \end{bmatrix} = \vec{0}. \quad (50)$$

Thus,

$$\frac{x_k}{y_k} = \frac{p}{\lambda_k - r^2}, \quad (51)$$

$$\frac{y_k}{z_k} = \frac{\lambda_k - 1}{q^*}, \quad (52)$$

$$\frac{x_k}{z_k} = \frac{p(\lambda_k - 1)}{q^*(\lambda_k - r^2)}. \quad (53)$$

The eigenvectors make up the columns of  $U^\dagger$  and therefore make up the rows of  $U^*$ :

$$U^* = \begin{bmatrix} x_1 & y_1 & z_1 \\ x_2 & y_2 & z_2 \\ x_3 & y_3 & z_3 \end{bmatrix} \Rightarrow U = \begin{bmatrix} x_1 & y_1 & z_1 \\ x_2 & y_2 & z_2 \\ x_3 & y_3 & z_3 \end{bmatrix}^* \quad (54)$$

Hence,

$$U \approx \begin{bmatrix} 1 & U_{12} & U_{13} \\ -U_{12}^* & 1 & U_{23} \\ U_{31} & -U_{23}^* & 1 \end{bmatrix} \approx \begin{bmatrix} x_1 & y_1 & z_1 \\ x_2 & y_2 & z_2 \\ x_3 & y_3 & z_3 \end{bmatrix}^*, \quad (55)$$

where  $U_{12} = |U_{12}|e^{i\theta_{12}}$  etc.

So, From (51):

$$-U_{12}^* = \left(\frac{x_2}{y_2}\right)^* = \frac{p^*}{\lambda_2 - r^2} = \frac{|p|e^{-i\theta_p}}{\lambda_2 - r^2}, \quad (56)$$

$$\Rightarrow |U_{12}|e^{-i\theta_{12}} = \frac{|p|e^{i(\pi - \theta_p)}}{\lambda_2 - r^2}, \quad (57)$$

and since  $\lambda_2 \gg r^2$ :

$$|U_{12}| = \frac{|p|}{\lambda_2 - r^2}, \text{ and } \theta_{12} = \pi + \theta_p. \quad (58)$$

From (52):

$$U_{23} = \left(\frac{z_2}{y_2}\right)^* = \frac{q}{\lambda_2 - 1} = \frac{|q|e^{i\theta_q}}{\lambda_2 - 1} = \frac{|q|e^{i(\pi + \theta_q)}}{1 - \lambda_2}, \quad (59)$$

$$\Rightarrow |U_{23}|e^{i\theta_{23}} = \frac{|q|e^{i(\pi + \theta_q)}}{1 - \lambda_2}, \quad (60)$$

and since  $1 \gg \lambda_2$ :

$$|U_{23}| = \frac{|q|}{1 - \lambda_2}, \text{ and } \theta_{23} = \pi + \theta_q. \quad (61)$$

Now:

$$\frac{U_{13}}{U_{12}} = \left(\frac{z_1}{y_1}\right)^* = \frac{q}{\lambda_1 - 1} = \frac{|q|e^{i(\pi + \theta_q)}}{1 - \lambda_1}, \quad (62)$$

hence:

$$|U_{13}|e^{i\theta_{13}} = \frac{|q|e^{i(\pi + \theta_q)}}{1 - \lambda_1} \frac{|p|}{\lambda_2 - r^2} e^{i(\pi + \theta_p)}, \quad (63)$$

and since  $1 \gg \lambda_1$  and  $\lambda_2 \gg r^2$ :

$$|U_{13}| = \frac{|q||p|}{(1 - \lambda_1)(\lambda_2 - r^2)}, \text{ and } \theta_{13} = \theta_q + \theta_p. \quad (64)$$

Furthermore:

$$\frac{U_{31}}{-U_{23}^*} = \left(\frac{x_3}{y_3}\right)^* = \frac{p^*}{\lambda_3 - r^2} = \frac{|p|e^{-i\theta_p}}{\lambda_3 - r^2}, \quad (65)$$

hence:

$$|U_{31}|e^{i\theta_{31}} = \frac{|p|e^{-i\theta_p}}{\lambda_3 - r^2} \left( -\frac{|q|e^{-i(\pi+\theta_q)}}{1-\lambda_2} \right) = \frac{|p||q|e^{-i(\theta_q+\theta_p)}}{(\lambda_3 - r^2)(1-\lambda_2)}, \quad (66)$$

and since  $1 \gg \lambda_2$  and  $\lambda_3 \gg r^2$ :

$$|U_{31}| = \frac{|p||q|}{(\lambda_3 - r^2)(1-\lambda_2)}, \text{ and } \theta_{31} = -(\theta_q + \theta_p). \quad (67)$$

Finally:

$$U \approx \begin{bmatrix} 1 & -\frac{|p|}{\lambda_2 - r^2}e^{\theta_p} & \frac{|q||p|}{(1-\lambda_1)(\lambda_2 - r^2)}e^{i(\theta_q+\theta_p)} \\ -U_{12}^* & 1 & -\frac{|q|}{1-\lambda_2}e^{i\theta_q} \\ \frac{|p||q|}{(\lambda_3 - r^2)(1-\lambda_2)}e^{-i(\theta_q+\theta_p)} & -U_{23}^* & 1 \end{bmatrix} = \begin{bmatrix} 1 & -\frac{p}{\lambda_2 - r^2} & \frac{qp}{(1-\lambda_1)(\lambda_2 - r^2)} \\ -U_{12}^* & 1 & -\frac{q}{1-\lambda_2} \\ \frac{p^*q^*}{(\lambda_3 - r^2)(1-\lambda_2)} & -U_{23}^* & 1 \end{bmatrix}, \quad (68)$$

thus, as  $\lambda_3 \approx 1$ ,  $\lambda_2 \gg r^2$ ,  $\lambda_1 \ll 1$ :

$$U \approx \begin{bmatrix} 1 & -\frac{p}{\lambda_2} & \frac{qp}{\lambda_2} \\ \frac{p^*}{\lambda_2} & 1 & -q \\ p^*q^* & q^* & 1 \end{bmatrix}. \quad (69)$$

We can now construct the CKM matrix,  $V = U_u U_d^\dagger$ :

$$V \approx \begin{bmatrix} 1 & -\frac{p}{\lambda_2} & \frac{qp}{\lambda_2} \\ \frac{p^*}{\lambda_2} & 1 & -q \\ p^*q^* & q^* & 1 \end{bmatrix}_u \begin{bmatrix} 1 & \frac{p}{\lambda_2} & qp \\ -\frac{p^*}{\lambda_2} & 1 & q \\ \frac{p^*q^*}{\lambda_2} & -q^* & 1 \end{bmatrix}_d, \quad (70)$$

$$\approx \begin{bmatrix} 1 & \frac{p_d}{\lambda_2^d} - \frac{p_u}{\lambda_2^u} & \frac{p_u(q_u - q_d)}{\lambda_2^u} \\ \frac{p_u^*}{\lambda_2^u} - \frac{p_d^*}{\lambda_2^d} & 1 & q_d - q_u \\ \frac{p_d^*(q_d - q_u)}{\lambda_2^d} & q_u^* - q_d^* & 1 \end{bmatrix}, \quad (71)$$

hence,

$$V \approx \begin{bmatrix} 1 & U_{12}^u - U_{12}^d & U_{12}^u(V_{23}) \\ -V_{12}^* & 1 & U_{23}^u - U_{23}^d \\ (-U_{12}^d(V_{23}))^* & -V_{23}^* & 1 \end{bmatrix}. \quad (72)$$

## Tests

To test the accuracy of our constructed CKM matrix,  $V$ , we again evaluate the  $\chi^2$  of the elements of the matrix against their values given by the PDG. Due to the orthogonality of  $V$  we only test  $V_{12}$ ,  $V_{13}$ ,  $V_{31}$ , and  $V_{23}$ . We take the observed values of the CKM matrix to be:

$$|V_{ij}^{measured}| = \begin{bmatrix} 0.97427 & 0.22536 & 0.00355 \\ 0.22522 & 0.97343 & 0.0414 \\ 0.00886 & 0.0405 & 0.99914 \end{bmatrix}. \quad (73)$$

We calculate numerical exact values for our constructed  $V$  using mathematica.

We also define:

$$\sigma_{12} = 0.0025, \sigma_{13} = 0.0006, \sigma_{31} = 0.001 \text{ and } \sigma_{23} = 0.004. \quad (74)$$

And now

$$\chi_v^2 = \sum_{ij} \frac{(|V_{ij}| - |V_{ij}^{measured}|)^2}{\sigma_{ij}^2} \quad (ij = 12, 13, 23, 31). \quad (75)$$

We also test the accuracy of our values for the eigenvalues:

$$\chi_{\lambda_k}^2 = \sum_k \frac{(|\lambda_k| - |\lambda_k^{measured}|)^2}{\sigma_{\lambda_k}^2} \quad (k=1, 2, 3, 4), \quad (76)$$

where  $\lambda_1 = \lambda_1^u$ ,  $\lambda_2 = \lambda_2^u$ ,  $\lambda_3 = \lambda_1^d$ ,  $\lambda_4 = \lambda_2^d$ , and:

$$\sigma_{\lambda_k} = (0.000002, 0.0003, 0.00005, 0.0008).$$

We now minimise  $\chi^2 = \chi_v^2 + \chi_{\lambda_k}^2$ . Note that, since we have two unphysical phase degrees of freedom, we can fix a phase convention such that  $Arg(p_u) = -Arg(p_d)$  and  $Arg(q_u) = -Arg(q_d)$  (see Appendix: A).

And obtain new solutions:

Solution 9 -  $\chi^2 = 5.68 \times 10^{-1}$

$$|M_u| = \begin{bmatrix} 7.8 \times 10^{-5} & 6.9 \times 10^{-4} & 0 \\ 6.9 \times 10^{-4} & 8.8 \times 10^{-3} & 3.8 \times 10^{-2} \\ 0 & 3.8 \times 10^{-2} & 1 \end{bmatrix}, \quad |M_d| = \begin{bmatrix} 1.7 \times 10^{-4} & 4.3 \times 10^{-3} & 0 \\ 4.3 \times 10^{-3} & -1.3 \times 10^{-2} & 8.1 \times 10^{-2} \\ 0 & 8.1 \times 10^{-2} & 1 \end{bmatrix} \quad (77)$$

Solution 10 -  $\chi^2 = 1.56$

$$|M_u| = \begin{bmatrix} 5.5 \times 10^{-5} & 7.1 \times 10^{-4} & 0 \\ 7.1 \times 10^{-4} & 7.4 \times 10^{-3} & 1.1 \times 10^{-2} \\ 0 & 1.1 \times 10^{-2} & 1 \end{bmatrix}, \quad |M_d| = \begin{bmatrix} 4.6 \times 10^{-4} & 5.4 \times 10^{-3} & 0 \\ 5.4 \times 10^{-3} & 2.1 \times 10^{-2} & 4.8 \times 10^{-2} \\ 0 & 4.8 \times 10^{-2} & 1 \end{bmatrix} \quad (78)$$

Solution 11 -  $\chi^2 = 2.97$

$$|M_u| = \begin{bmatrix} 7.7 \times 10^{-5} & 6.8 \times 10^{-4} & 0 \\ 6.8 \times 10^{-4} & 8.8 \times 10^{-3} & 4.0 \times 10^{-2} \\ 0 & 4.0 \times 10^{-2} & 1 \end{bmatrix}, \quad |M_d| = \begin{bmatrix} 3.9 \times 10^{-4} & 3.6 \times 10^{-3} & 0 \\ 3.6 \times 10^{-3} & -2.0 \times 10^{-2} & 4.1 \times 10^{-3} \\ 0 & 4.1 \times 10^{-3} & 1 \end{bmatrix} \quad (79)$$

Solution 12 -  $\chi^2 = 1.43$

$$|M_u| = \begin{bmatrix} 7.6 \times 10^{-5} & 6.8 \times 10^{-4} & 0 \\ 6.8 \times 10^{-4} & 8.7 \times 10^{-3} & 3.7 \times 10^{-2} \\ 0 & 3.7 \times 10^{-2} & 1 \end{bmatrix}, \quad |M_d| = \begin{bmatrix} 3.6 \times 10^{-4} & 5.3 \times 10^{-3} & 0 \\ 5.3 \times 10^{-3} & 1.9 \times 10^{-2} & 7.1 \times 10^{-4} \\ 0 & 7.1 \times 10^{-4} & 1 \end{bmatrix} \quad (80)$$

Solution 13 -  $\chi^2 = 1.41$

$$|M_u| = \begin{bmatrix} 5.3 \times 10^{-5} & 7.0 \times 10^{-4} & 0 \\ 7.0 \times 10^{-4} & 7.3 \times 10^{-3} & 3.9 \times 10^{-3} \\ 0 & 3.9 \times 10^{-3} & 1 \end{bmatrix}, \quad |M_d| = \begin{bmatrix} 4.1 \times 10^{-4} & 5.4 \times 10^{-3} & 0 \\ 5.4 \times 10^{-3} & 2.0 \times 10^{-2} & 3.4 \times 10^{-2} \\ 0 & 3.4 \times 10^{-2} & 1 \end{bmatrix} \quad (81)$$

Table 4: New solutions

	9	10	11	12	13
$\lambda_1^u$	$1.31 \times 10^{-5}$	$-1.33 \times 10^{-5}$	$1.37 \times 10^{-5}$	$1.33 \times 10^{-5}$	$-1.33 \times 10^{-5}$
$\lambda_2^u$	$7.42 \times 10^{-3}$	$7.35 \times 10^{-3}$	$7.26 \times 10^{-3}$	$7.37 \times 10^{-3}$	$7.36 \times 10^{-3}$
$r_u$	$8.82 \times 10^{-3}$	$7.41 \times 10^{-3}$	$8.77 \times 10^{-3}$	$8.71 \times 10^{-3}$	$7.3 \times 10^{-3}$
$r_u^2$	$7.78 \times 10^{-5}$	$5.49 \times 10^{-5}$	$7.69 \times 10^{-5}$	$7.59 \times 10^{-5}$	$5.34 \times 10^{-5}$
$Re(p_u)$	$-4.96 \times 10^{-4}$	$-6.18 \times 10^{-4}$	$-5.68 \times 10^{-4}$	$-5.8 \times 10^{-4}$	$6.05 \times 10^{-4}$
$Im(p_u)$	$4.79 \times 10^{-4}$	$-3.39 \times 10^{-4}$	$-3.64 \times 10^{-4}$	$3.48 \times 10^{-4}$	$3.48 \times 10^{-4}$
$ p_u $	$6.9 \times 10^{-4}$	$7.05 \times 10^{-4}$	$6.75 \times 10^{-4}$	$6.76 \times 10^{-4}$	$6.98 \times 10^{-4}$
$Arg(p_u)$	2.37	3.64	3.71	2.6	$5.22 \times 10^{-1}$
$Re(q_u)$	$3.8 \times 10^{-2}$	$-1.09 \times 10^{-2}$	$-1.61 \times 10^{-2}$	$8.98 \times 10^{-3}$	$4.63 \times 10^{-4}$
$Im(q_u)$	$-3.08 \times 10^{-3}$	$2.55 \times 10^{-4}$	$-3.61 \times 10^{-2}$	$3.63 \times 10^{-2}$	$3.88 \times 10^{-3}$
$ q_u $	$3.82 \times 10^{-2}$	$1.09 \times 10^{-2}$	$3.95 \times 10^{-2}$	$3.74 \times 10^{-2}$	$3.91 \times 10^{-3}$
$Arg(q_u)$	$-8.08 \times 10^{-2}$	3.12	4.29	1.33	1.45
$\lambda_1^d$	$1.05 \times 10^{-3}$	$-1.02 \times 10^{-3}$	$1.02 \times 10^{-3}$	$-1.02 \times 10^{-3}$	$-1.02 \times 10^{-3}$
$\lambda_2^d$	$-2.02 \times 10^{-2}$	$2.04 \times 10^{-2}$	$-2.04 \times 10^{-2}$	$2.04 \times 10^{-2}$	$2.04 \times 10^{-2}$
$r_d$	$-1.29 \times 10^{-2}$	$2.13 \times 10^{-2}$	$-1.98 \times 10^{-2}$	$1.91 \times 10^{-2}$	$2.02 \times 10^{-2}$
$r_d^2$	$1.66 \times 10^{-4}$	$4.55 \times 10^{-4}$	$3.9 \times 10^{-4}$	$3.63 \times 10^{-4}$	$4.07 \times 10^{-4}$
$Re(p_d)$	$-3.06 \times 10^{-3}$	$-4.77 \times 10^{-3}$	$-3.05 \times 10^{-3}$	$-4.52 \times 10^{-3}$	$4.64 \times 10^{-3}$
$Im(p_d)$	$-2.95 \times 10^{-3}$	$2.62 \times 10^{-3}$	$1.96 \times 10^{-3}$	$-2.72 \times 10^{-3}$	$-2.67 \times 10^{-3}$
$ p_d $	$4.25 \times 10^{-3}$	$5.44 \times 10^{-3}$	$3.63 \times 10^{-3}$	$5.28 \times 10^{-3}$	$5.36 \times 10^{-3}$
$Re(q_d)$	$8.09 \times 10^{-2}$	$-4.81 \times 10^{-2}$	$-1.68 \times 10^{-3}$	$1.7 \times 10^{-4}$	$3.98 \times 10^{-3}$
$Im(q_d)$	$6.55 \times 10^{-3}$	$-1.12 \times 10^{-3}$	$3.78 \times 10^{-3}$	$-6.87 \times 10^{-4}$	$-3.34 \times 10^{-2}$
$ q_d $	$8.11 \times 10^{-2}$	$4.81 \times 10^{-2}$	$4.14 \times 10^{-3}$	$7.07 \times 10^{-4}$	$3.37 \times 10^{-2}$
$Min$	$5.68 \times 10^{-1}$	1.56	2.97	1.43	1.41
$ V_{12} $	$2.25 \times 10^{-1}$	$2.25 \times 10^{-1}$	$2.25 \times 10^{-1}$	$2.25 \times 10^{-1}$	$2.25 \times 10^{-1}$
$ V_{13} $	$3.9 \times 10^{-3}$	$3.53 \times 10^{-3}$	$3.94 \times 10^{-3}$	$3.51 \times 10^{-3}$	$3.53 \times 10^{-3}$
$ V_{23} $	$4.18 \times 10^{-2}$	$3.78 \times 10^{-2}$	$4.24 \times 10^{-2}$	$3.81 \times 10^{-2}$	$3.8 \times 10^{-2}$
$ V_{31} $	$8.87 \times 10^{-3}$	$9.72 \times 10^{-3}$	$7.34 \times 10^{-3}$	$9.71 \times 10^{-3}$	$9.67 \times 10^{-3}$
$x =  V_{13}/V_{23} $	$9.34 \times 10^{-2}$	$9.33 \times 10^{-2}$	$9.29 \times 10^{-2}$	$9.2 \times 10^{-2}$	$9.3 \times 10^{-2}$
$y =  V_{31}/V_{23} $	$2.13 \times 10^{-1}$	$2.57 \times 10^{-1}$	$1.73 \times 10^{-1}$	$2.55 \times 10^{-1}$	$2.55 \times 10^{-1}$
$ p_u / q_u $	$1.81 \times 10^{-2}$	$6.47 \times 10^{-2}$	$1.71 \times 10^{-2}$	$1.81 \times 10^{-2}$	$1.78 \times 10^{-1}$
$ p_d / q_d $	$5.24 \times 10^{-2}$	$1.13 \times 10^{-1}$	$8.77 \times 10^{-1}$	7.46	$1.59 \times 10^{-1}$
$ p_u / p_d $	$1.62 \times 10^{-1}$	$1.3 \times 10^{-1}$	$1.86 \times 10^{-1}$	$1.28 \times 10^{-1}$	$1.3 \times 10^{-1}$
$ q_u / q_d $	$4.7 \times 10^{-1}$	$2.27 \times 10^{-1}$	9.56	$5.29 \times 10^1$	$1.16 \times 10^{-1}$
$\sqrt{ p_u  q_u }$	$5.13 \times 10^{-3}$	$2.77 \times 10^{-3}$	$5.16 \times 10^{-3}$	$5.03 \times 10^{-3}$	$1.65 \times 10^{-3}$
$\sqrt{ p_d  q_d }$	$1.86 \times 10^{-2}$	$1.62 \times 10^{-2}$	$3.87 \times 10^{-3}$	$1.93 \times 10^{-3}$	$1.34 \times 10^{-2}$

### Reducing parameters

We now aim to reduce the number of parameters by setting our ansatz to be:

$$M_u = \begin{bmatrix} (z\bar{\lambda}^3)^2 & z\bar{\lambda}^5 & 0 \\ z\bar{\lambda}^5 & z\bar{\lambda}^3 & \bar{\lambda}^2 \\ 0 & \bar{\lambda}^2 & 1 \end{bmatrix}, \quad M_d = \begin{bmatrix} (-\bar{\lambda}^3)^2 & -i\bar{\lambda}^4 c/z & 0 \\ i\bar{\lambda}^4 c/z & -\bar{\lambda}^3 & \bar{\lambda}^2/z \\ 0 & \bar{\lambda}^2/z & 1 \end{bmatrix} \quad (82)$$

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Or equivalently:

$$M_u = \begin{bmatrix} (z\bar{\lambda}^3)^2 & 0 & 0 \\ 0 & z\bar{\lambda}^3 & 0 \\ 0 & 0 & 1 \end{bmatrix} + \bar{\lambda}^2 \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} + z\bar{\lambda}^5 \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \quad (83)$$

$$M_d = \begin{bmatrix} (-\bar{\lambda}^3)^2 & 0 & 0 \\ 0 & -\bar{\lambda}^3 & 0 \\ 0 & 0 & 1 \end{bmatrix} + \frac{\bar{\lambda}^2}{z} \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} + \frac{\bar{\lambda}^4 c}{z} \begin{bmatrix} 0 & -i & 0 \\ i & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}. \quad (84)$$

Essentially we have set:

$$r_u = z\bar{\lambda}^3, \quad r_d = -\bar{\lambda}^3, \quad (85)$$

$$|p_u| = z\bar{\lambda}^5, \quad |p_d| = \frac{\bar{\lambda}^4 c}{z}, \quad (86)$$

$$|q_u| = \bar{\lambda}^2, \quad |q_d| = \frac{\bar{\lambda}^2}{z}, \quad (87)$$

$$Arg(p_u) = \frac{\pi}{2}, \quad Arg(p_d) = 0, \quad (88)$$

$$\Rightarrow \Delta\theta = Arg(p_u) - Arg(p_d) = \frac{\pi}{2}, \quad (89)$$

$$Arg(q_u) = 0, \quad Arg(q_d) = 0, \quad (90)$$

$$\Rightarrow \Delta\phi = Arg(q_u) - Arg(q_d) = 0, \quad (91)$$

and reduced the number of parameters from eight to three.

### Tests

Now we test this ansatz using the  $\chi^2$  as before. But now we test the quantities:

$$\lambda_1^u, \quad \lambda_2^u, \quad \lambda_1^d, \quad \lambda_2^d, \quad \lambda, \quad A, \quad \sin(2\beta), \quad \alpha, \quad (92)$$

where

$$\lambda = |V_{us}|, \quad (93)$$

$$A = \frac{|V_{cb}|}{\lambda^2}, \quad (94)$$

$$\beta = Arg\left(-\frac{V_{cd}V_{cb}^*}{V_{td}V_{tb}^*}\right), \quad (95)$$

$$\alpha = Arg\left(-\frac{V_{td}V_{tb}^*}{V_{ud}V_{ub}^*}\right). \quad (96)$$

We take the centered values and errors as (from the PDG):

$$\lambda_1^u = (6.3 \pm 0.5) \times 10^{-6}, \quad (97)$$

$$\lambda_2^u = (3.2 \pm 0.1) \times 10^{-3}, \quad (98)$$

$$\lambda_1^d = (1.08 \pm 0.05) \times 10^{-3}, \quad (99)$$

$$\lambda_2^d = (2.2 \pm 0.06) \times 10^{-2}, \quad (100)$$

$$\lambda = 0.2251 \pm 0.0005, \quad (101)$$

$$A = 0.95 \pm 0.05, \quad (102)$$

$$\sin(2\beta) = 0.691 \pm 0.017, \quad (103)$$

$$\alpha = 87.6^\circ \pm 3.5^\circ. \quad (104)$$

So we minimise the  $\chi^2$ :

$$\begin{aligned} \chi^2 = & \frac{(|\lambda_1^u| - 6.3 \times 10^{-6})^2}{(0.5 \times 10^{-6})^2} + \frac{(|\lambda_2^u| - 3.2 \times 10^{-3})^2}{(0.1 \times 10^{-3})^2} + \frac{(|\lambda_1^d| - 1.08 \times 10^{-3})^2}{(0.05 \times 10^{-3})^2} + \frac{(|\lambda_2^d| - 2.2 \times 10^{-2})^2}{(0.06 \times 10^{-2})^2} \\ & + \frac{(\lambda - 0.2251)^2}{(0.0005)^2} + \frac{(A - 0.95)^2}{(0.05)^2} + \frac{(|\sin(2\beta)| - 0.691)^2}{(0.017)^2} + \frac{(\alpha - \frac{73\pi}{150})^2}{(\frac{7\pi}{360})^2}, \end{aligned} \quad (105)$$

We minimise the  $\chi^2$  with the parameters as functions of  $z$ ,  $\bar{\lambda}$ , and  $c$  to find:

$$\chi^2 = 5.68 \quad \text{at} \quad (z, \bar{\lambda}, c) = (0.504, 0.223, 0.949). \quad (106)$$

For this solution the mass matrices become:

$$M_u = \begin{bmatrix} 3.11 \times 10^{-5} & 2.77 \times 10^{-4} & 0 \\ 2.77 \times 10^{-4} & 5.58 \times 10^{-3} & 4.97 \times 10^{-2} \\ 0 & 4.97 \times 10^{-2} & 1 \end{bmatrix}, \quad M_d = \begin{bmatrix} 1.22 \times 10^{-4} & 4.64 \times 10^{-3}i & 0 \\ -4.64 \times 10^{-3}i & -1.11 \times 10^{-2} & 9.85 \times 10^{-2} \\ 0 & 9.85 \times 10^{-2} & 1 \end{bmatrix}. \quad (107)$$

And the CKM matrix is:

$$|V| = \begin{bmatrix} 0.974 & 0.225 & 4.17 \times 10^{-3} \\ 0.225 & 0.973 & 4.62 \times 10^{-2} \\ 1.01 \times 10^{-2} & 4.53 \times 10^{-2} & 0.999 \end{bmatrix}, \quad (108)$$

and the eigenvalues become:

$$\lambda_1^u = 6.42 \times 10^{-6}, \quad (109)$$

$$\lambda_2^u = 3.13 \times 10^{-3}, \quad (110)$$

$$\lambda_1^d = 1.11 \times 10^{-3}, \quad (111)$$

$$\lambda_2^d = -2.16 \times 10^{-2}. \quad (112)$$

## Examining Parameter Coefficients

Now we test the mass matrices:

$$M_u = \begin{bmatrix} (a\bar{\lambda}^3)^2 & b\bar{\lambda}^5 & 0 \\ b\bar{\lambda}^5 & a\bar{\lambda}^3 & c\bar{\lambda}^2 \\ 0 & c\bar{\lambda}^2 & 1 \end{bmatrix}, \quad M_d = \begin{bmatrix} (-d\bar{\lambda}^3)^2 & -ei\bar{\lambda}^4 & 0 \\ ei\bar{\lambda}^4 & -d\bar{\lambda}^3 & f\bar{\lambda}^2 \\ 0 & f\bar{\lambda}^2 & 1 \end{bmatrix}, \quad (113)$$

with the extra quantity  $\frac{m_b}{m_t} = g\bar{\lambda}^3$ .

And we minimise the  $\chi^2$  with respect to the quantities:

$$\lambda_1^u, \quad \lambda_2^u, \quad \lambda_1^d, \quad \lambda_2^d, \quad \lambda, \quad A, \quad \sin(2\beta), \quad \alpha, \quad \frac{m_b}{m_t}. \quad (114)$$

We take the centered values and errors as (from the PDG):

$$\lambda_1^u = (7.15 \pm 0.56) \times 10^{-6}, \quad (115)$$

$$\lambda_2^u = (3.49 \pm 0.091) \times 10^{-3}, \quad (116)$$

$$\lambda_1^d = (9.55 \pm 0.42) \times 10^{-4}, \quad (117)$$

$$\lambda_2^d = (1.93 \pm 0.15) \times 10^{-2}, \quad (118)$$

$$\frac{m_b}{m_t} = (1.59 \pm 0.016) \times 10^{-2}, \quad (119)$$

$$\lambda = 0.2251 \pm 0.0005, \quad (120)$$

$$A = 0.823 \pm 0.013, \quad (121)$$

$$\sin(2\beta) = 0.691 \pm 0.017, \quad (122)$$

$$\alpha = 87.6^\circ \pm 3.5^\circ. \quad (123)$$

We now fix  $\bar{\lambda} = 0.2251$  and minimise the  $\chi^2$  with respect to the coefficients  $(a, b, c, d, e, f, g)$  to find:

$$\chi^2 = 1.76 \quad \text{at} \quad (a, b, c, d, e, f, g) = (0.5, 0.51, 0.94, 0.85, 1.58, 1.8, 1.39). \quad (124)$$

Now we note that:

$$a \approx b \approx \sin\left(\frac{\pi}{6}\right) = \frac{1}{2}, \quad (125)$$

$$c \approx d \approx \cos\left(\frac{\pi}{6}\right) = \frac{\sqrt{3}}{2} \approx 0.866, \quad (126)$$

$$e \approx \frac{\cos^2\left(\frac{\pi}{6}\right)}{\tan\left(\frac{\pi}{6}\right)} = \frac{3}{4}\sqrt{3} \approx 1.3, \quad (127)$$

$$f \approx \frac{\cos\left(\frac{\pi}{6}\right)}{\tan\left(\frac{\pi}{6}\right)} = \frac{3}{2}, \quad (128)$$



$$g \approx \frac{1}{\cos^2(\frac{\pi}{6})} = \frac{2}{\sqrt{3}} \approx 1.16, \quad (129)$$

$\chi$  **ansatz**

From these coefficient values we now set the ansatz:

$$M_u = \begin{bmatrix} (\sin(\chi)\bar{\lambda}^3)^2 & \sin(\chi)\bar{\lambda}^5 & 0 \\ \sin(\chi)\bar{\lambda}^5 & \sin(\chi)\bar{\lambda}^3 & \cos(\chi)\bar{\lambda}^2 \\ 0 & \cos(\chi)\bar{\lambda}^2 & 1 \end{bmatrix}, \quad M_d = \begin{bmatrix} (-\cos(\chi)\bar{\lambda}^3)^2 & -\frac{\cos^2(\chi)}{\tan(\chi)}i\bar{\lambda}^4 & 0 \\ \frac{\cos^2(\chi)}{\tan(\chi)}i\bar{\lambda}^4 & -\cos(\chi)\bar{\lambda}^3 & \frac{\cos(\chi)}{\tan(\chi)}\bar{\lambda}^2 \\ 0 & \frac{\cos(\chi)}{\tan(\chi)}\bar{\lambda}^2 & 1 \end{bmatrix}, \quad (130)$$

and minimise the  $\chi^2$  with respect to the observables from (114) and measured values and errors as in (115) to (123), allowing  $\chi$  and  $\bar{\lambda}$  to vary, to find:

$$\chi^2 = 12.46 \quad \text{at} \quad (\chi, \bar{\lambda}) = (0.483, 0.231). \quad (131)$$

The mass matrices then become:

$$M_u = \begin{bmatrix} 3.3 \times 10^{-5} & 3.07 \times 10^{-4} & 0 \\ 3.07 \times 10^{-4} & 5.74 \times 10^{-3} & 4.73 \times 10^{-2} \\ 0 & 4.73 \times 10^{-2} & 1 \end{bmatrix}, \quad M_d = \begin{bmatrix} 1.2 \times 10^{-4} & 4.28 \times 10^{-3}i & 0 \\ 4.28 \times 10^{-3}i & -1.09 \times 10^{-2} & 9.03 \times 10^{-2} \\ 0 & 9.03 \times 10^{-2} & 1 \end{bmatrix}, \quad (132)$$

with the CKM matrix:

$$|V| = \begin{bmatrix} 0.974 & 0.225 & 3.62 \times 10^{-3} \\ 0.225 & 0.974 & 4.07 \times 10^{-2} \\ 8.89 \times 10^{-3} & 3.99 \times 10^{-2} & 0.999 \end{bmatrix}, \quad (133)$$

and we have the observable quantities:

$$\lambda_1^u = 5.99 \times 10^{-6}, \quad (134)$$

$$\lambda_2^u = 3.52 \times 10^{-3}, \quad (135)$$

$$\lambda_1^d = 1.03 \times 10^{-3}, \quad (136)$$

$$\lambda_2^d = -1.99 \times 10^{-2}, \quad (137)$$

$$\frac{m_b}{m_t} = 1.58 \times 10^{-2}, \quad (138)$$

$$A = 0.807, \quad (139)$$

$$\sin(2\beta) = 0.707, \quad (140)$$

$$\alpha = 82.92, \quad (141)$$

# Appendices

## Proof of invariance of D and V under unitary transformations on M

We have:

$$D_i = U_i M_i U_i^\dagger \Rightarrow M_i = U_i^\dagger D_i U_i, \quad (142)$$

and:

$$V = U_u U_d^\dagger. \quad (143)$$

Suppose we choose some other matrix  $M'_i = U^\dagger M_i U$ , where  $U$  is some arbitrary unitary matrix.

Then

$$(U'_i)^\dagger D'_i U'_i = M'_i = U^\dagger M_i U = U^\dagger U_i^\dagger D_i U_i U = (U'_i)^\dagger D_i U'_i. \quad (144)$$

Hence  $D'_i = D_i$  i.e. the matrix that is obtained by diagonalising  $M'_i$  is the same as the matrix obtained by diagonalising  $M_i$ . That is,  $D_i$  is invariant under the transformation  $M_i \rightarrow M'_i$ .

Now,

$$V = U_u U_d^\dagger. \quad (145)$$

Suppose we have  $V' = U'_u (U'_d)^\dagger$ .

Then

$$V' = U'_u (U'_d)^\dagger = U_u U U^\dagger U_d^\dagger = U_u I U_d^\dagger = U_u U_d^\dagger = V. \quad (146)$$

Thus  $V$  is also invariant under this change of mass matrices.

Recall

$$M_i = \begin{bmatrix} r_i^2 & p_i & 0 \\ p_i & r_i & q_i \\ 0 & q_i & 1 \end{bmatrix}, \quad (147)$$

and take

$$U = \begin{bmatrix} -1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad (148)$$

then

$$M'_i = U^\dagger M_i U = \begin{bmatrix} -1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_i^2 & p_i & 0 \\ p_i & r_i & q_i \\ 0 & q_i & 1 \end{bmatrix} \begin{bmatrix} -1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} r_i^2 & -p_i & 0 \\ -p_i & r_i & q_i \\ 0 & q_i & 1 \end{bmatrix}, \quad (149)$$

so  $V$  is invariant under the transformation  $p_i \rightarrow -p_i$  i.e. changing the signs of both  $p_u$  and  $p_d$ . Similarly for

$$U = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{bmatrix} \equiv q_i \rightarrow -q_i, \quad (150)$$

and

$$U = \begin{bmatrix} 1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \equiv p_i \rightarrow -p_i \text{ and } q_i \rightarrow -q_i. \quad (151)$$

Similarly  $D$  and  $V$  are invariant under transformations of the form  $M'_i = \Phi^\dagger M_i \Phi$  and if we set:

$$\Phi = \begin{bmatrix} e^{i\phi_1} & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & e^{i\phi_2} \end{bmatrix}, \quad (152)$$

then, with  $p_i$  and  $q_i$  complex:

$$M'_i = \begin{bmatrix} r_i^2 & p_i e^{i\phi_1} & 0 \\ p_i^* e^{-i\phi_1} & r_i & q_i e^{-i\phi_2} \\ 0 & q_i^* e^{i\phi_2} & 1 \end{bmatrix} = \Phi^\dagger M_i \Phi = \Phi^\dagger U_i^\dagger D_i U_i \Phi = (U'_i)^\dagger D_i U'_i. \quad (153)$$

Hence,

$$V' = U'_u (U'_d)^\dagger = U_u \Phi \Phi^\dagger U_d^\dagger = V. \quad (154)$$

So we have some choice of phase convention, for example, setting  $\phi_1 = -\frac{(Arg(p_u) + Arg(p_d))}{2} \Rightarrow Arg(p_u) = -Arg(p_d)$ . Similarly we can set  $Arg(q_u) = -Arg(q_d)$ .

### Approximate observable quantities

In the complex case:

$$Det(M) = Det(D) \Rightarrow r^3 - |q|^2 r^2 - |p|^2 = \prod_{n=1}^3 \lambda_n \approx \lambda_1 \lambda_2, \quad (155)$$

$$\sum_{ij=11,22,33} [M]_{ij} = \sum_{ij=11,22,33} [D]_{ij} \Rightarrow r^3 + r^2 + r - |q|^2 - |p|^2 = \lambda_1 \lambda_2 + \lambda_2 \lambda_3 + \lambda_1 \lambda_3 \approx \lambda_1 \lambda_2 + \lambda_2 + \lambda_1. \quad (156)$$

Hence,

$$\sum_{ij=11,22,33} [M]_{ij} - Det(M) = |q|^2 (r^2 - 1) + r^2 + r \approx \lambda_1 + \lambda_2, \quad (157)$$

and since  $\lambda_1 \ll \lambda_2$  and  $r^2 \ll r$  and  $1 - r^2 \approx 1$ :

$$\lambda_2 \approx r - |q|^2, \quad (158)$$

and substituting  $\lambda_2$  into (90):

$$\lambda_1 \approx r^2 - \frac{|p|^2}{r - |q|^2}. \quad (159)$$

And since:

$$V \approx \begin{bmatrix} 1 & \frac{p_d}{\lambda_2^d} - \frac{p_u}{\lambda_2^u} & \frac{p_u(q_u - q_d)}{\lambda_2^u} \\ \frac{p_u^*}{\lambda_2^u} - \frac{p_d^*}{\lambda_2^d} & 1 & q_d - q_u \\ \frac{p_d^*(q_d^* - q_u^*)}{\lambda_2^d} & q_u^* - q_d^* & 1 \end{bmatrix}, \quad (160)$$

we have:

$$V \approx \begin{bmatrix} 1 & \frac{p_d}{r_d - |q_d|^2} - \frac{p_u}{r_u - |q_u|^2} & \frac{p_u(q_u - q_d)}{r_u - |q_u|^2} \\ \frac{p_u^*}{r_u - |q_u|^2} - \frac{p_d^*}{r_d - |q_d|^2} & 1 & q_d - q_u \\ \frac{p_d^*(q_d^* - q_u^*)}{r_d - |q_d|^2} & q_u^* - q_d^* & 1 \end{bmatrix}. \quad (161)$$

Hence:

$$|V_{12}| \approx \left| \frac{p_d}{r_d - |q_d|^2} - \frac{p_u}{r_u - |q_u|^2} \right|, \quad (162)$$

$$|V_{13}| \approx \left| \frac{p_u(q_u - q_d)}{r_u - |q_u|^2} \right|, \quad (163)$$

$$|V_{23}| \approx |q_d - q_u|, \quad (164)$$

$$|V_{31}| \approx \left| \frac{p_d^*(q_d^* - q_u^*)}{r_d - |q_d|^2} \right|. \quad (165)$$

### Proof of Commutator/J relation

First let the commutator,  $C$ , be defined:

$$C = [M_u, M_d] = M_u M_d - M_d M_u, \quad (166)$$

for any two hermitian matrices,  $M_u$  and  $M_d$ . Then  $C$  is anti-hermitian, i.e.  $C = -C^\dagger$ .

Proof:

$$-C^\dagger = -(M_u M_d - M_d M_u)^\dagger = -M_d^\dagger M_u^\dagger + M_u^\dagger M_d^\dagger = M_u M_d - M_d M_u = C \quad \square \quad (167)$$

Now define the Jarlskog invariant,  $J$ , as

$$J = \pm \text{Im}(V_{i\alpha} V_{i\beta}^* V_{j\beta} V_{j\alpha}^*). \quad (168)$$

Then,

$$\text{Det}(C) = 2i \Delta_u \Delta_d J, \quad (169)$$

where  $\Delta_i = (\lambda_1^i - \lambda_2^i)(\lambda_2^i - \lambda_3^i)(\lambda_3^i - \lambda_1^i)$ .

### Proof

First we perform a unitary transformation on  $M_u$  so that:

$$M_u \rightarrow M'_u = U M_u U^\dagger = D_u, \quad (170)$$

where  $D_u$  is diagonal. To keep  $\text{Det}(C)$  invariant we perform the same transformation on  $M_d$ :

$$M_d \rightarrow M'_d = U M_d U^\dagger = U U_d^\dagger D_d U_d U^\dagger = V D_d V^\dagger, \quad (171)$$

as  $U = U_u$  is the unitary matrix which diagonalises  $M_u$  and  $V = U_u U_d^\dagger$ .

So  $C$  becomes:

$$C = D_u V D_d V^\dagger - V D_d V^\dagger D_u \quad (172)$$

Recall:  $D_i = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix}_i$ .

Hence,  $C_{ij} = (\lambda_i^u - \lambda_j^u) \sum_{\alpha=1}^3 V_{i\alpha} V_{j\alpha}^* \lambda_\alpha^d$ .

Now, by the orthogonality of  $V$  we have:

$$\sum_{\alpha=1}^3 V_{i\alpha} V_{j\alpha}^* = 0 \Rightarrow \sum_{\alpha=1}^3 V_{i\alpha} V_{j\alpha}^* \lambda_\alpha^d = 0 \quad (173)$$

So

$$C_{ij} = (\lambda_i^u - \lambda_j^u) \left( \sum_{\alpha=1}^3 V_{i\alpha} V_{j\alpha}^* \lambda_\alpha^d - \sum_{\alpha=1}^3 V_{i\alpha} V_{j\alpha}^* \lambda_\beta^d \right) = (\lambda_i^u - \lambda_j^u) \left( \sum_{\alpha=1}^3 (\lambda_\alpha^d - \lambda_\beta^d) V_{i\alpha} V_{j\alpha}^* \right) \quad (174)$$

, for any  $\beta$  so we choose  $\beta \in 1, 2, 3$  such that  $\beta \neq i, \beta \neq j$ . Clearly if  $i = j$  or  $\alpha = \beta$  then  $C_{ij} = 0$ .

Hence,

$$\text{Det}(C) = C_{12}C_{23}C_{31} + C_{21}C_{32}C_{13} = C_{12}C_{23}C_{31} - (C_{12}C_{23}C_{31})^* \quad (175)$$

as  $C = -C^\dagger$ . So,

$$\text{Det}(C) = 2i \text{Im}(C_{12}C_{23}C_{31}) \quad (176)$$

From (109):

$$C_{12} = (\lambda_1^u - \lambda_2^u) [(\lambda_1^d - \lambda_3^d) V_{11} V_{21}^* + (\lambda_2^d - \lambda_3^d) V_{12} V_{22}^*], \quad (177)$$

$$C_{23} = (\lambda_2^u - \lambda_3^u) [(\lambda_2^d - \lambda_1^d) V_{22} V_{32}^* + (\lambda_3^d - \lambda_1^d) V_{23} V_{33}^*], \quad (178)$$

$$C_{31} = (\lambda_3^u - \lambda_1^u) [(\lambda_1^d - \lambda_2^d) V_{31} V_{11}^* + (\lambda_3^d - \lambda_2^d) V_{33} V_{13}^*], \quad (179)$$

Hence,

$$\begin{aligned} \text{Det}[C] &= 2i \text{Im}(C_{12}C_{23}C_{31}) = 2i \text{Im}((\lambda_1^u - \lambda_2^u)(\lambda_2^u - \lambda_3^u)(\lambda_3^u - \lambda_1^u) [(\lambda_1^d - \lambda_3^d) V_{11} V_{21}^* \\ &\quad + (\lambda_2^d - \lambda_3^d) V_{12} V_{22}^*] [(\lambda_2^d - \lambda_1^d) V_{22} V_{32}^* + (\lambda_3^d - \lambda_1^d) V_{23} V_{33}^*] [(\lambda_1^d - \lambda_2^d) V_{31} V_{11}^* + (\lambda_3^d - \lambda_2^d) V_{33} V_{13}^*]) \\ &= 2i \Delta_u \text{Im}([( \lambda_3^d - \lambda_1^d)(\lambda_1^d - \lambda_2^d)^2 |V_{11}|^2 V_{21}^* V_{22} V_{32}^* V_{31} - (\lambda_1^d - \lambda_2^d)(\lambda_2^d - \lambda_3^d)(\lambda_3^d - \lambda_1^d) V_{11} V_{21}^* V_{22} V_{32}^* V_{33} V_{13}^* \\ &\quad - (\lambda_1^d - \lambda_2^d)(\lambda_3^d - \lambda_1^d)^2 |V_{11}|^2 V_{21}^* V_{23} V_{33}^* V_{31} \\ &\quad + (\lambda_2^d - \lambda_3^d)(\lambda_3^d - \lambda_1^d)^2 |V_{33}|^2 V_{11} V_{21}^* V_{23} V_{13}^* - (\lambda_2^d - \lambda_3^d)(\lambda_1^d - \lambda_2^d)^2 |V_{22}|^2 V_{12} V_{32}^* V_{31} V_{11}^* \\ &\quad + (\lambda_1^d - \lambda_2^d)(\lambda_2^d - \lambda_3^d)^2 |V_{22}|^2 V_{12} V_{32}^* V_{33} V_{13}^* + (\lambda_1^d - \lambda_2^d)(\lambda_2^d - \lambda_3^d)(\lambda_3^d - \lambda_1^d) V_{12} V_{22}^* V_{23} V_{33}^* V_{31} V_{11}^* \\ &\quad - (\lambda_3^d - \lambda_1^d)(\lambda_2^d - \lambda_3^d)^2 |V_{33}|^2 V_{12} V_{22}^* V_{23} V_{13}^*]), \\ &= 2i \Delta_u \text{Im}((\lambda_1^d - \lambda_2^d)(\lambda_2^d - \lambda_3^d)(\lambda_3^d - \lambda_1^d) [V_{12} V_{22}^* V_{23} V_{33}^* V_{31} V_{11}^* - V_{11} V_{21}^* V_{22} V_{32}^* V_{33} V_{13}^*] \\ &\quad + |V_{11}|^2 (\lambda_1^d - \lambda_2^d)(\lambda_3^d - \lambda_1^d) [(\lambda_1^d - \lambda_2^d) V_{22} V_{21}^* V_{31} V_{32}^* - (\lambda_3^d - \lambda_1^d) V_{23} V_{21}^* V_{31} V_{33}^*] \\ &\quad + |V_{22}|^2 (\lambda_1^d - \lambda_2^d)(\lambda_2^d - \lambda_3^d) [(\lambda_2^d - \lambda_3^d) V_{12} V_{32}^* V_{33} V_{13}^* - (\lambda_1^d - \lambda_2^d) V_{12} V_{32}^* V_{31} V_{11}^*] \\ &\quad + |V_{33}|^2 (\lambda_2^d - \lambda_3^d)(\lambda_3^d - \lambda_1^d) [(\lambda_3^d - \lambda_1^d) V_{11} V_{21}^* V_{23} V_{13}^* - (\lambda_2^d - \lambda_3^d) V_{12} V_{22}^* V_{23} V_{13}^*]). \quad (180) \end{aligned}$$

Ignore the  $2i \Delta_u$  for now and split this into four terms:

$$t1 = \text{Im}(\lambda_1^d - \lambda_2^d)(\lambda_2^d - \lambda_3^d)(\lambda_3^d - \lambda_1^d) [V_{12} V_{22}^* V_{23} V_{33}^* V_{31} V_{11}^* - V_{11} V_{21}^* V_{22} V_{32}^* V_{33} V_{13}^*], \quad (181)$$

$$t2 = |V_{11}|^2(\lambda_1^d - \lambda_2^d)(\lambda_3^d - \lambda_1^d)[(\lambda_1^d - \lambda_2^d)V_{22}V_{21}^*V_{31}V_{32}^* - (\lambda_3^d - \lambda_1^d)V_{23}V_{21}^*V_{31}V_{33}^*], \quad (182)$$

$$t3 = |V_{22}|^2(\lambda_1^d - \lambda_2^d)(\lambda_2^d - \lambda_3^d)[(\lambda_2^d - \lambda_3^d)V_{12}V_{32}^*V_{33}V_{13}^* - (\lambda_1^d - \lambda_2^d)V_{12}V_{32}^*V_{31}V_{11}^*], \quad (183)$$

$$t4 = |V_{33}|^2(\lambda_2^d - \lambda_3^d)(\lambda_3^d - \lambda_1^d)[(\lambda_3^d - \lambda_1^d)V_{11}V_{21}^*V_{23}V_{13}^* - (\lambda_2^d - \lambda_3^d)V_{12}V_{22}^*V_{23}V_{13}^*]. \quad (184)$$

Now recall  $\sum_{\alpha=1}^3 V_{i\alpha}V_{j\alpha}^* = 0 \Rightarrow V_{31}V_{33}^* = -V_{21}V_{23}^* - V_{11}V_{13}^*$  and  $-V_{11}V_{13}^* = V_{21}V_{23}^* + V_{31}V_{33}^*$ . Hence,  $t1$  becomes:

$$\begin{aligned} t1 &= \Delta_d \text{Im}[-V_{12}V_{22}^*V_{23}V_{11}^*(V_{21}V_{23}^* - V_{11}V_{13}^*) + V_{21}^*V_{22}V_{32}^*V_{33}(V_{21}V_{23}^* + V_{31}V_{33}^*)] \\ &= \Delta_d[-|V_{23}|^2 \text{Im}(V_{12}V_{22}^*V_{21}V_{11}^*) - |V_{11}|^2 \text{Im}(V_{12}V_{22}^*V_{23}V_{13}^*) + |V_{21}|^2 \text{Im}(V_{22}V_{23}^*V_{33}V_{32}^*) + |V_{33}|^2 \text{Im}(V_{22}V_{21}^*V_{31}V_{32}^*)] \\ &= \Delta_d[-|V_{23}|^2(-J) - |V_{11}|^2 J + |V_{21}|^2 J + |V_{33}|^2(-J)] \\ &= \Delta_d J[|V_{23}|^2 - |V_{11}|^2 + |V_{21}|^2 - |V_{33}|^2]. \end{aligned} \quad (185)$$

Now since  $J = \pm \text{Im}(V_{i\alpha}V_{i\beta}^*V_{j\beta}V_{j\alpha}^*)$ ,

$$\begin{aligned} t2 &= |V_{11}|^2(\lambda_1^d - \lambda_2^d)(\lambda_3^d - \lambda_1^d)[(\lambda_1^d - \lambda_2^d)V_{22}V_{21}^*V_{31}V_{32}^* - (\lambda_3^d - \lambda_1^d)V_{23}V_{21}^*V_{31}V_{33}^*] \\ &= |V_{11}|^2(\lambda_1^d - \lambda_2^d)(\lambda_3^d - \lambda_1^d)[(\lambda_1^d - \lambda_2^d)(-J) - (\lambda_3^d - \lambda_1^d)J] = |V_{11}|^2 \Delta_d J. \end{aligned} \quad (186)$$

Similarly:

$$t3 = |V_{22}|^2 \Delta_d J, \quad (187)$$

and

$$t4 = |V_{33}|^2 \Delta_d J, \quad (188)$$

Hence,

$$\text{Det}[C] = 2i\Delta_u \Delta_d J(|V_{23}|^2 - |V_{11}|^2 + |V_{21}|^2 - |V_{33}|^2 + |V_{11}|^2 + |V_{22}|^2 + |V_{33}|^2) = 2\Delta_u \Delta_d J(|V_{21}|^2 |V_{22}|^2 + |V_{23}|^2), \quad (189)$$

and since  $\sum_{k=1}^3 |V_{i\alpha}|^2 = 1$

$$\text{Det}[C] = 2i\Delta_u \Delta_d J \quad \square \quad (190)$$

## Approximation for J

Recall from (70) we have:

$$V \approx \begin{bmatrix} 1 & -\frac{p}{\lambda_2} & \frac{qp}{\lambda_2} \\ \frac{p^*}{\lambda_2} & 1 & -q \\ p^*q^* & q^* & 1 \end{bmatrix}_u \begin{bmatrix} 1 & \frac{p}{\lambda_2} & qp \\ -\frac{p^*}{\lambda_2} & 1 & q \\ \frac{p^*q^*}{\lambda_2} & -q^* & 1 \end{bmatrix}_d, \quad (191)$$

$$\approx \begin{bmatrix} 1 + \frac{p_u p_d^* + q_u p_u q_d^* p_d}{\lambda_2^u \lambda_2^u} & \frac{p_d}{\lambda_2^d} - \frac{p_u}{\lambda_2^u} - \frac{q_u p_u q_d^*}{\lambda_2^u} & \frac{p_u(q_u - q_d)}{\lambda_2^u} + p_d q_d \\ \frac{p_u^*}{\lambda_2^u} - \frac{p_d^*}{\lambda_2^d} - \frac{q_u p_d^* q_d^*}{\lambda_2^d} & 1 + \frac{p_u^* p_d}{\lambda_2^u \lambda_2^d} + q_u q_d^* & q_d - q_u + \frac{p_u^* p_d q_d}{\lambda_2^u} \\ \frac{p_d^*(q_d^* - q_u^*)}{\lambda_2^d} + p_u^* q_u^* & q_u^* - q_d^* + \frac{p_u^* q_u^* p_d}{\lambda_2^d} & 1 + q_u^* q_d(p_u^* p_d + 1) \end{bmatrix}. \quad (192)$$

Hence we can construct J, from (108):

$$J = \text{Im}(V_{12}V_{13}^*V_{23}V_{22}^*), \quad (193)$$

$$J \approx \tilde{J} = \text{Im}\left[\left(\frac{p_d}{\lambda_2^d} - \frac{p_u}{\lambda_2^u} - \frac{q_u p_u q_d^*}{\lambda_2^u}\right)\left(\frac{p_u(q_u - q_d)}{\lambda_2^u} + p_d q_d\right)^*(q_d - q_u + \frac{p_u^* p_d q_d}{\lambda_2^u})\left(1 + \frac{p_u^* p_d}{\lambda_2^u \lambda_2^d} + q_u q_d^*\right)^*\right], \quad (194)$$

$$\tilde{J} = Im[\frac{(p_d \lambda_2^u - p_u \lambda_2^d - q_u p_u q_d^* \lambda_2^d)(p_u(q_u - q_d) + \lambda_2^u p_d q_d)^*((q_d - q_u)\lambda_2^u + p_u^* p_d q_d)(p_u^* p_d + \lambda_2^u \lambda_2^d(1 + q_u q_d^*))^*}{(\lambda_2^u)^4 (\lambda_2^d)^2}], \quad (195)$$

Multiplying the leading order terms we get:

$$\tilde{J} = Im[\frac{(p_d \lambda_2^u - p_u \lambda_2^d) p_u^* (q_u - q_d)^* (q_d - q_u)}{(\lambda_2^u)^2 \lambda_2^d}], \quad (196)$$

$$\tilde{J} = Im[\frac{|q_u - q_d|^2 (|p_u|^2 \lambda_2^d - p_u^* p_d \lambda_2^u)}{(\lambda_2^u)^2 \lambda_2^d}], \quad (197)$$

$$\tilde{J} = Im[\frac{-|q_u - q_d|^2 (|p_u| |p_d| e^{-i\Delta\theta})}{\lambda_2^u \lambda_2^d}], \quad (198)$$

$$\tilde{J} = \frac{|q_u - q_d|^2 |p_u| |p_d| \sin \Delta\theta}{\lambda_2^u \lambda_2^d}, \quad (199)$$

And hence,

$$\tilde{J} = \frac{|V_{cb}|^2 |p_u| |p_d| \sin \Delta\theta}{\lambda_2^u \lambda_2^d}, \quad (200)$$

$$\tilde{J} = |V_{ub}| |V_{td}| \sin \Delta\theta, \quad (201)$$

Now, the next highest order terms are:

$$a = Im[\frac{p_u^* p_d q_d (\lambda_2^u p_d - \lambda_2^d p_u) \lambda_2^u \lambda_2^d (p_u^* (q_u^* - q_d^*))}{\lambda_2^u \lambda_2^d}], \quad (202)$$

$$b = -Im[\frac{\lambda_2^d p_u q_u q_d^* \lambda_2^u (q_d - q_u) \lambda_2^u \lambda_2^d p_u^* (q_u^* - q_d^*)}{\lambda_2^u \lambda_2^d}], \quad (203)$$

$$c = Im[\frac{(p_u p_d^* + \lambda_2^u \lambda_2^d q_u^* q_d) (\lambda_2^u (q_d - q_u)) (\lambda_2^u p_d - \lambda_2^d p_u) (p_u^* (q_u^* - q_d^*))}{\lambda_2^u \lambda_2^d}], \quad (204)$$

$$d = Im[\frac{\lambda_2^u p_d^* q_d^* \lambda_2^u (q_d - q_u) (\lambda_2^u p_d - \lambda_2^d p_u) \lambda_2^u \lambda_2^d}{\lambda_2^u \lambda_2^d}]. \quad (205)$$

Which become:

$$a = \frac{|p_u|^2 |p_d| |q_d| (\lambda_2^u |p_d| |q_d| \sin(2\Delta\theta) - \lambda_2^d |p_u| |q_d| \sin(\Delta\theta) + \lambda_2^d |p_u| |q_d| \sin(\Delta\theta + \Delta\phi) - \lambda_2^u |p_d| |q_u| \sin(2\Delta\theta + \Delta\phi))}{(\lambda_2^u)^3 \lambda_2^d}, \quad (206)$$

$$b = -\frac{|p_u|^2 |q_u| |q_d| (|q_u|^2 + |q_d|^2 - 2|q_u| |q_d| \cos(\Delta\phi)) \sin(\Delta\theta)}{(\lambda_2^u)^2}, \quad (207)$$

$$c = \frac{|p_u| (|q_u|^2 + |q_d|^2 - 2|q_u| |q_d| \cos(\Delta\phi)) (|p_u|^2 |p_d| \sin(\Delta\theta) + \lambda_2^u |q_u| |q_d| (\lambda_2^u |p_d| \sin(\Delta\theta + \Delta\phi) - \lambda_2^d |p_u| \sin(\Delta\phi)))}{(\lambda_2^u)^3 \lambda_2^d}, \quad (208)$$

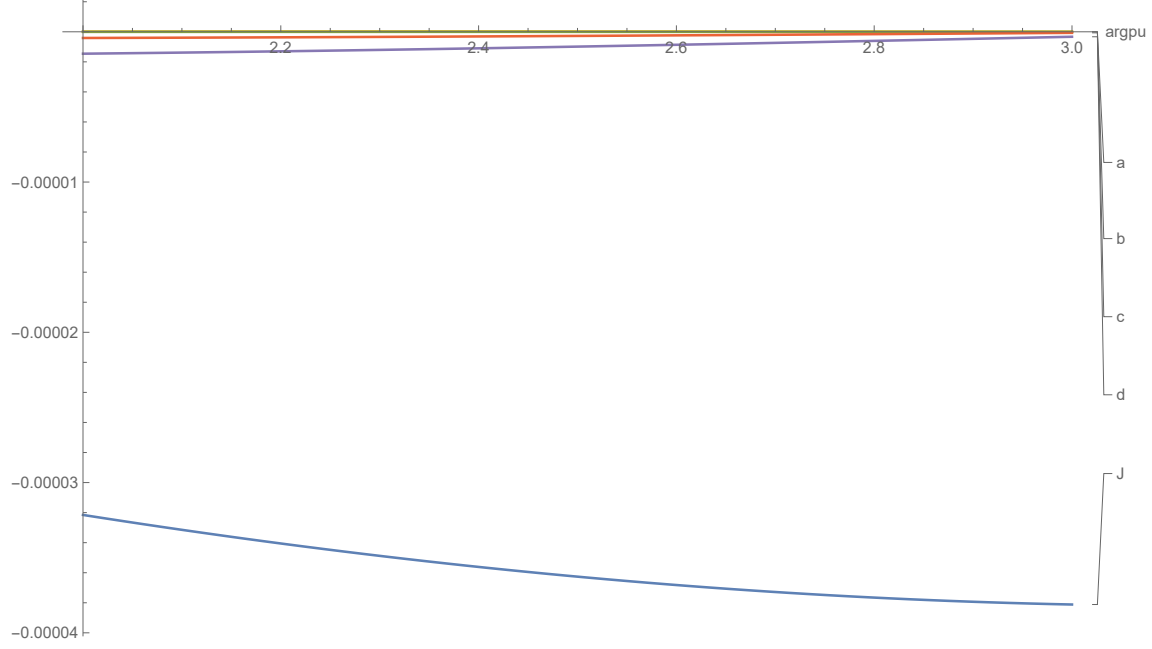
$$d = -\frac{|p_d| |q_d| (\lambda_2^u |p_d| |q_u| \sin(\Delta\phi) + \lambda_2^d |p_u| |q_d| \sin(\Delta\theta) - \lambda_2^d |p_u| |q_u| \sin(\Delta\theta + \Delta\phi))}{\lambda_2^u \lambda_2^d}. \quad (209)$$

Numerically, for our complex solutions, we can show that these higher order terms are negligible.

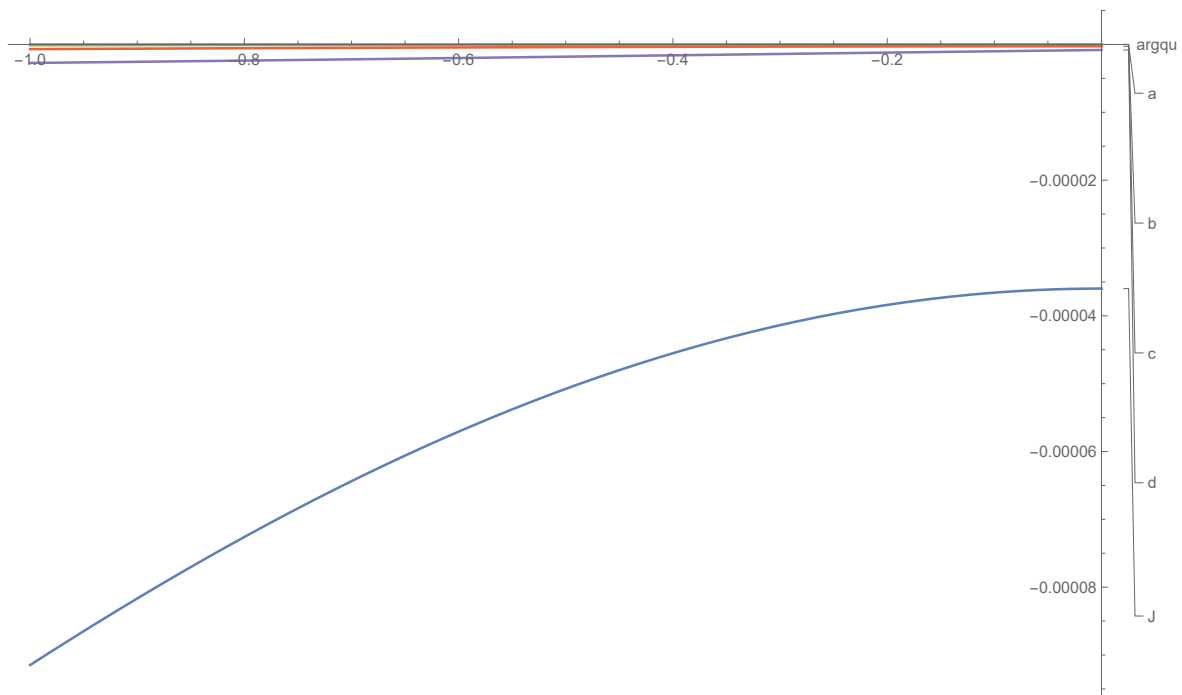
We can also see that the higher order terms in the approximation of J are negligible in the space around our solutions. For example:

Table 5: New solutions

	9	10	11	12	13
$J$	$3.36 \times 10^{-5}$	$2.89 \times 10^{-5}$	$-2.57 \times 10^{-5}$	$-2.9 \times 10^{-5}$	$2.91 \times 10^{-5}$
$\tilde{J}$	$3.94 \times 10^{-5}$	$3.23 \times 10^{-5}$	$-2.8 \times 10^{-5}$	$-3.27 \times 10^{-5}$	$3.27 \times 10^{-5}$
$\tilde{J}+a+b+c+d$	$4.08 \times 10^{-5}$	$3.19 \times 10^{-5}$	$-2.82 \times 10^{-5}$	$-3.29 \times 10^{-5}$	$3.24 \times 10^{-5}$

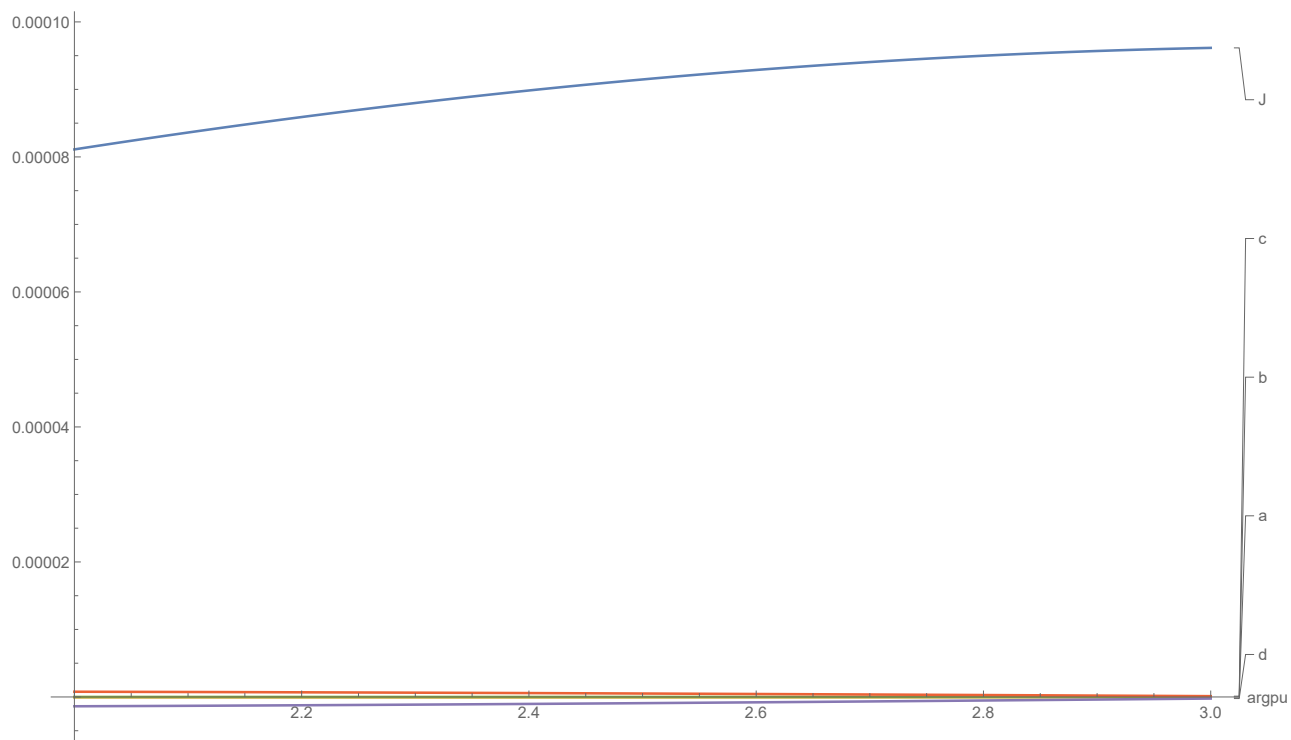
Solution 9J and higher order terms varying  $Arg(p_u)$ :Varying  $Arg(q_u)$ :





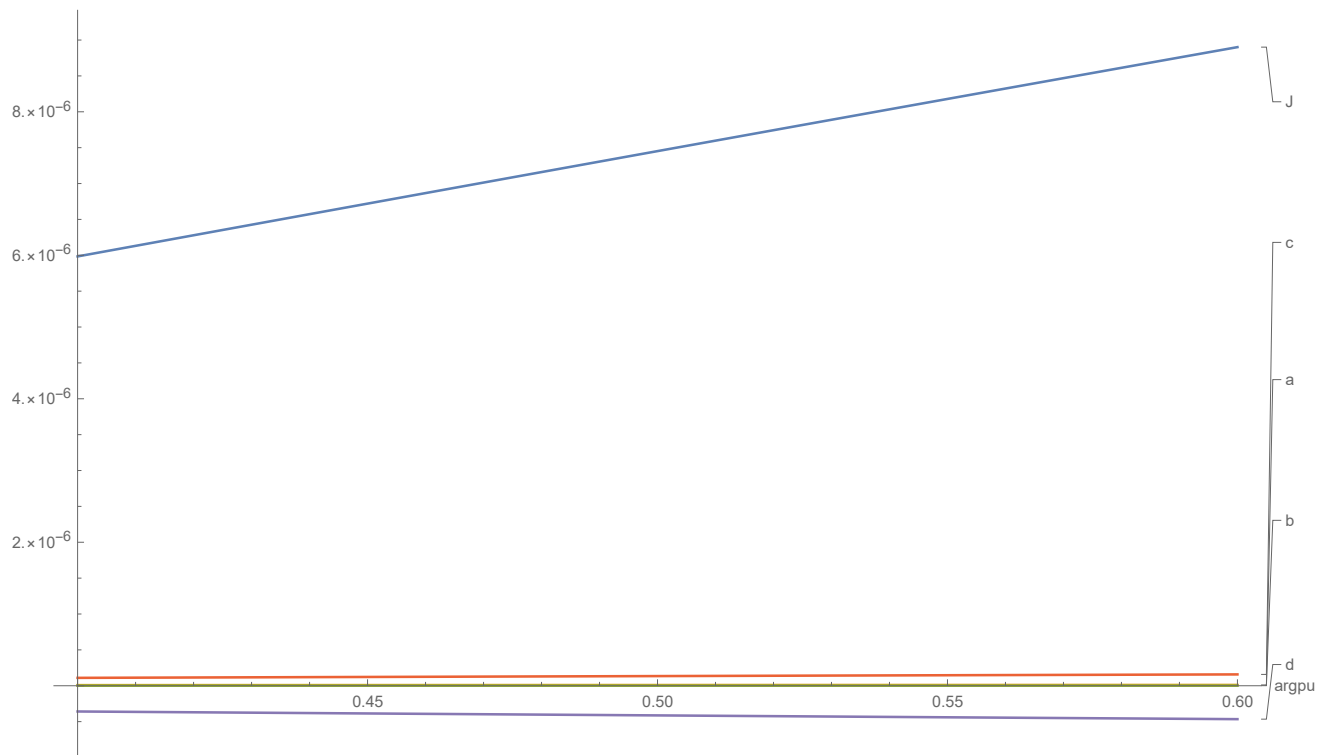
### Solution 10

Varying  $\text{Arg}(p_u)$ :

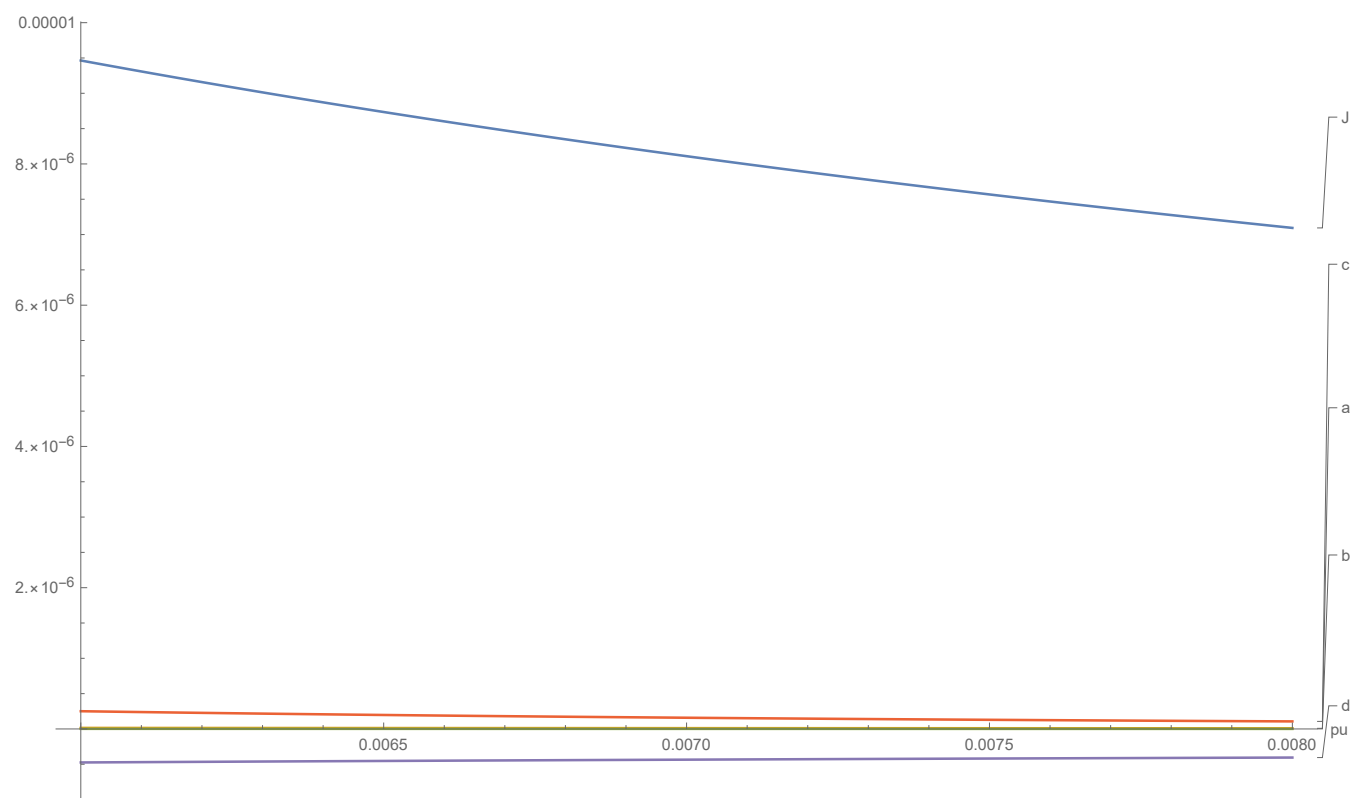


### Solution 13

Varying  $\text{Arg}(p_u)$ :



Varying  $|p_u|$ :



## Pseudocode

For the complex case we use Wolfram Mathematica to obtain numerically exact  $V$  matrices from the inputted parameters  $|p_u|$ ,  $|p_d|$ ,  $Arg(pu)$ ,  $|q_u|$ ,  $|q_d|$ ,  $Arg(qu)$ ,  $r_u$ , and  $r_d$  (recall that we set the phase convention  $Arg(p_u) = -Arg(p_d)$ ) as follows:

- Define

$$M(r, |p|, argp, |q|, argq) = \begin{bmatrix} r^2 & |p|e^{iargp} & 0 \\ |p|e^{-iargp} & r & |q|e^{iargq} \\ 0 & |q|e^{-iargq} & 0 \end{bmatrix}; \quad (210)$$

- Create a list of the eigenvectors of  $M$ , ordering the list by the  $x$  elements of the eigenvectors as  $x_1 \gg x_2 \gg x_3$ :

$$Egv(r, |p|, argp, |q|, argq) = [v_1, v_2, v_3]; \quad (211)$$

- Now create the matrix  $Ut$  whose rows are the conjugates of the eigenvectors:

$$Ut(r, |p|, argp, |q|, argq) = \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix}^* = \begin{bmatrix} x_1^* & y_1^* & z_1^* \\ x_2^* & y_2^* & z_2^* \\ x_3^* & y_3^* & z_3^* \end{bmatrix}; \quad (212)$$

- Create the transformation matrix  $\Phi$ :

$$\Phi(Ut) = \begin{bmatrix} e^{-I*Arg(Ut[1,1])} & 0 & 0 \\ 0 & e^{-I*Arg(Ut[2,2])} & 0 \\ 0 & 0 & e^{-I*Arg(Ut[3,3])} \end{bmatrix}, \quad (213)$$

and the Matrix  $U$  with real diagonal elements:

$$U(r, |p|, argp, |q|, argq) = \Phi(Ut(r, |p|, argp, |q|, argq)).Ut(r, |p|, argp, |q|, argq) \quad (214)$$

Recall that the  $D_i$  and observable quantities are invariant under such a transformation.

- Now construct  $V$ :

$$V(r_u, |p_u|, argp, |q_u|, argq, r_d, |p_d|, |q_d|) = U(r_u, |p_u|, argpu, |q_u|, argqu).U(r_d, |p_d|, -argpu, |q_d|, -argqu)^\dagger. \quad (215)$$

We now wish to minimise  $\chi^2 = \chi_v^2 + \chi_{\lambda_k}^2$  from equations (75) and (76) where the  $\lambda_k$  are the numerically exact eigenvalues of  $M$  corresponding to the eigenvectors  $v_k$ . To minimise:

- Create a random population of 10 sets of arguments within reasonable bounds for the arguments.
- For each set in the start population create a new population of, for example, 99 sets of arguments taken from the normal distribution with mean centered at the argument start value and variance 5% of the start value. So now we have a list of 100 sets of arguments close to the starting arguments (including the start set).
- Compare the  $\chi^2$  value for each set in the new population and replace the starting set in the original population with the set with the smallest  $\chi^2$ .
- Iterate this process over the population to produce 10 minimised  $\chi^2$  values and the minimum points.

## References

- [1] K. A. Olive *et al.* [Particle Data Group Collaboration], Chin. Phys. C **38** (2014) 090001. doi:10.1088/1674-1137/38/9/090001