

Supplementary Tables and Figures for Manuscript

A cellular assay to determine the fusion capacity of MFN2 variants linked to Charcot-Marie-Tooth disease of type 2A

Chloe Barsa¹, Julian Perrin¹, Claudine David¹, Arnaud Mourier^{1*} and Manuel Rojo^{1*#}

¹Institut de Biochimie et Génétique Cellulaires (IBGC), Université de Bordeaux, CNRS, IBGC, UMR 5095, F-33000 Bordeaux, France

* co-last authors

Correspondence to: manuel.rojo@ibgc.cnrs.fr

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Supplementary References

Supplementary Tables

Reference	year	Variants				
Züchner et al. ¹	2004	p.V69F: 5-15 (f)	p.L76P: 7-44 (f)	p.R94Q: 6-17 (f) 3-15 (f)		
		p.R94W: <10 (f)	p.T105M: 3-15 (f)	p.W740S: 5-52 (f), <10 (f), 7-47 (f)		
Zhu et al. ² (Vucic et al. ³ 2003)	2005	p.H165D: 4, 8, 14, 10 (3), 20 (3)				
Chung et al. ⁴	2006	p.R94W: <10 (2)	p.T105M: >10	p.H165R: >10 (2)	p.R364W: <10 (4)	p.M376T: >10
		p.L76P: unknown	p.R94W: 2, 3	p.R94Q: 2, 3 (2), 5, 7 (2)		p.H165R: 6
Verhoeven et al. ⁵	2006	p.H165Y: 12	p.R250W: 4	p.R250Q: 21		p.H361Y: 2
		p.M376I: 35	p.W740S: unknown			
Engelfried et al. ⁶	2006	p.M376I: 22	p.R468H: 26			
Calvo et al. ⁷	2009	p.R94W: <10 (2)	p.R364P: <5 (2)	p.R364Q: >10	p.W740C: <10 (2)	
Casasnovas et al. ⁸	2010	p.R94W: <10	p.R94Q: <10	p.R364Q: >20	p.M376V: >20	
		p.R468H: >20 (2) >40 (4)				
Feely et al. ⁹	2011	p.R94G: 1 (3)	p.R94Q: 4	p.R94W: 1 (2), 4	p.T105M: 1	p.L248V: 1 (2)
		p.H361Y: 1	p.364P: 2	p.R364W: 1 (3) 2 (2)	p.W740S: 5, 15, 16, 33	
Bombelli et al. ¹⁰	2014	p.R94Q: 2, 6	p.R94W: 2, 3, 4, 5 (2)		p.T105M: 1	p.R364P: 3 (2), 4
		p.R364W: <10	p.M376R: 3			
Stuppia et al. ¹¹	2015	p.V69F: 5-15	p.L76P: 7-44	p.R94G: 1	p.R94Q: 2, 4 (2), 3-15, <10	
		p.R94W: 1 (2), 4 (2), 5, 3-8, 8 (3)		p.T105M: 1	p.H165L: 14	p.H165R: 6, 7, 16
		p.H165Y: 12	p.L248V: 1 (2)	p.R250Q: 21, 12	p.H361Y: 1, 1	
		p.364P: 2, <5 (2), 22		p.364Q: >10, >20		
		p.R364W: 1 (4), 2 (6), 3 (2), 4, 5, 6, 8 (2)			p.M376I: 22, 35	
		p.M376L: 26	p.M376V: >20	p.R468H: 2, <10, 26, >20 (2), >50 (3)		
		p.W740C: <10 (2)	p.W740S: 5, 15, 16, 5-52, 33			
Lv et al. ¹²	2015	p.R94W: 2, 3, 4 (3), 6, 7		p.R94Q: inf, 1, 18	p.T105M: inf	p.R364W: inf, 6
		p.W740R: 5, 7				
Choi et al. ¹³	2015	p.R94W: 1, 4, 5, 8 (4), 9		p.T105M: 11, 25	p.H165R: 5, 10, 14, 50	
		p.R364W: 1 (4), 2, 3, 4 (2), 5, 8 (2)		p.M376T: 39		
Xie et al. ¹⁴	2016	p.R94W: 3, 5, 6 (2)	p.R94Q: 2	p.T105M: 4	p.M376V: 3, 9	
Pipis et al. ¹⁵	2020	p.L76--: 35.5 (2)	p.R94--: 4.7 (31)	p.T105--: 3.5 (2)	p.H165--: 19.7 (3)	
		p.248--: 2.0 (6)	p.R250--: 3.0 (2)	p.H361--: 1.8 (2)	p.R364--: 3.6 (16)	
		p.M376--: 11.6 (9)		p.W740--: 9.4 (19)		

Supplementary table 1: Age at onset reported for patients carrying *MFN2* variants characterized in this study. The average age at onset in Table 2 was calculated with the ages summarized in this table. The age at onset is labeled in red (early onset, <10 years) or green (late onset, ≥ 10 years). The number of patients (#) or families (# f) is indicated between parenthesis; (f) denotes a single family. For CMT2A onset in infants we used an age of 1. For patients/families with upper and lower age limits (e.g. <10, 5-15), we used the median age (i.e. 5, 10). For patients/families with lower age limits (e.g. >20) we used the lower age limit + 5 (i.e. 25). Amino acid positions labeled with "--" include all reported amino acid substitutions.

frequency ranking	allele category	protein consequence	allele frequency	allele count	allele number	number of homozygotes	clinical significance
1	rare	p.Val705Ile	5.08e-3	8193	1614164	81	B/LB
2	rare	p.Gly298Arg	3.86e-3	6225	1614002	18	B/LB
3	rare	p.Arg468His	2.76e-3	4448	1613718	9	CIP
4	rare	p.Leu392Val	1.50e-3	228	152362	0	
5	very rare	p.Arg707Trp	4.53e-4	731	1614174	1	P/LP
6	very rare	p.Arg250Gln	3.54e-4	571	1614068	1	CIP
7	very rare	p.Met393Ile	2.95e-4	476	1614260	1	B/LB
8	very rare	p.Ala716Thr	2.63e-4	425	1614182	1	CIP
9	very rare	p.Cys281Ser	2.22e-4	358	1613878	2	CIP
10	very rare	p.Arg510Gln	1.93e-4	311	1614156	7	CIP
11	very rare	p.Arg663Cys	1.38e-4	223	1614192	0	CIP
12	very rare	p.Asn525Ser	1.13e-4	182	1614128	0	CIP
13	very rare	p.His20Tyr	1.09e-4	176	1614124	9	CIP
14	very rare	p.Gly548Arg	1.07e-4	173	1614016	0	VUS
15	very rare	p.Pro737Ser	1.05e-4	16		0	
16	very rare	p.Ala54Thr	8.30e-5	134	1613812	0	CIP
17	very rare	p.Thr484Met	6.07e-5	98	1613444	0	CIP
18	very rare	p.Thr60Met	6.01e-5	97	1614150	0	CIP
19	very rare	p.Gln367His	5.76e-5	93	1614150	1	CIP
20	very rare	p.Pro587Ser	5.58e-5	90	1614130	0	VUS
≥513	very rare	p.Val69Phe	--	0	--	--	P
≥363	very rare	p.Leu76Pro	1.24e-6	2	1614100	0	P
≥513	very rare	p.Arg94Gln	--	0	--	--	P
≥513	very rare	p.Thr105Met	--	0	--	--	P/LP
≥513	very rare	p.His165Asp	--	0	--	--	P
≥513	very rare	p.Leu248Val	--	0	--	--	LP
≥513	very rare	p.His361Tyr	--	0	--	--	P
≥513	very rare	p.Arg364Trp	--	0	--	--	P
≥513	very rare	p.Met376Val	--	0	--	--	P/LP
≥513	very rare	p.Trp740Ser	--	0	--	--	P/LP

Supplementary Table 2. Allele frequency and classification of missense variants of *MFN2* in the gnomAD database. Shown are the 20 most frequent missense SNVs described in the gnomAD database and, labeled in bold, the SNVs subjected to functional analysis in this work. Frequent SNVs are ranked according to their allele frequency in gnomAD. The variants p.Arg468His and p.Arg250Gln represent the 3rd and 6th most frequent missense SNVs and have been identified in homozygous state in 9 or 1 cases, respectively. The p.Leu76Pro variant has been identified in 2 individuals in heterozygous state. The other SNVs characterized in this study were absent from the gnomAD database. Clinical significance: B: Benign, LB: Likely Benign, P: Pathogenic, LP: Likely Pathogenic, VUS: Variant of unknown significance, CIP: Conflicting interpretations of pathogenicity. The gnomAD v4.1.0 release (730,947 exomes and 76,215 genomes /GRCh38) depicts 512 missense variants within the coding sequence of *MFN2*. Variants with allele frequencies $\geq 0,1\%$, $\geq 0,01\%$ or $\geq 0,05\%$ are underlined in green, light green or light blue, respectively.

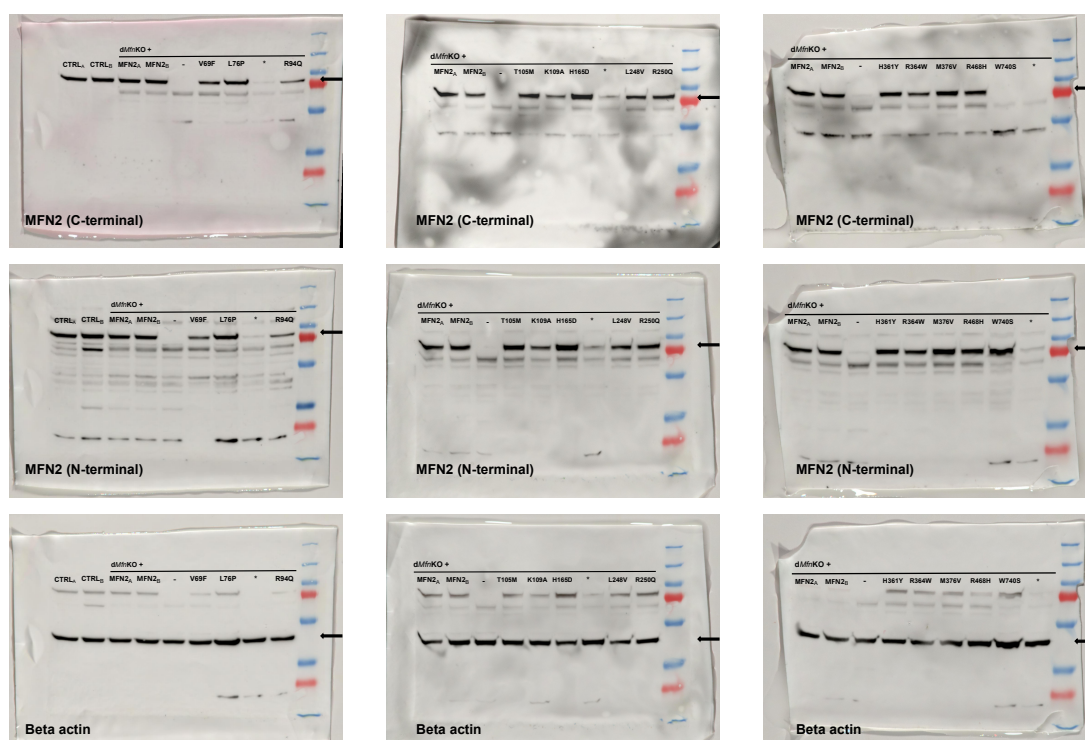
database or tool	data	Website accessed
ClinVar ¹⁶	variants	https://www.ncbi.nlm.nih.gov/clinvar/
Human Gene Mutation Database/HGMD ¹⁷	variants	https://www.hgmd.cf.ac.uk0
Inherited Neuropathy Variant Browser ¹⁸	variants	https://neuropathybrowser.zuchnerlab.net/#/
Leiden Open Variation Database ¹⁹	Variants	https://databases.lovd.nl
gnomAD ²⁰	variants	https://gnomad.broadinstitute.org/
Ensembl ²¹	VEP tools	http://www.ensembl.org/index.html
dbNSFP v4 ^{22,23}	VEP tools	http://database.liulab.science/dbNSFP
Polyphen2_HVAR_score ²⁴	protein	http://www.ensembl.org/index.html
EVE_score ²⁵	protein	http://database.liulab.science/dbNSFP
alpha-missense ²⁶	protein	https://github.com/google-deepmind/alphamissense
ESM1b ²⁷	protein	https://huggingface.co/spaces/ntranoslab/esm_variants
SIFT ²⁸	protein	http://database.liulab.science/dbNSFP
SIFT4G_score ²⁹	protein	http://database.liulab.science/dbNSFP
ENTPRISE ³⁰	protein	https://sites.gatech.edu/cssb/entprise/
PROVEAN ³¹	protein	http://provean.jcvi.org/seq_submit.php
phyloP100way vertebrate ³²	nucleotide	http://database.liulab.science/dbNSFP
phyloP470way mammalian ³²	nucleotide	http://database.liulab.science/dbNSFP
SiPhy_29way_logOdds_rankscore ³³	nucleotide	http://database.liulab.science/dbNSFP
GERP_RS_rankscore ³⁴	nucleotide	http://database.liulab.science/dbNSFP
CADD v1.7 ³⁵	integration	https://cadd.bihealth.org/snv
REVEL_score ³⁶	integration	http://database.liulab.science/dbNSFP
MetaLR ³⁷	integration	http://database.liulab.science/dbNSFP
FATHMM_XF ³⁸	integration	http://database.liulab.science/dbNSFP
BayesDel_addAF/BayesDel_noAF ³⁹	integration	http://database.liulab.science/dbNSFP
UMD-Predictor ⁴⁰	integration	https://umd-predictor.genomnis.com/

Supplementary table 3: Databases and websites hosting clinical, genetic and genomic data (ClinVar, HGMD, gnomAD) or Variant Effect Predictor (VEP) tools (Ensembl, dbNSFP v4). The data column indicates the data analysed by VEP tools to predict conservation, dysfunction and/or pathogenicity. Integration indicates that predictions are based on the integration of clinical findings and/or the predictions of several VEP tools. Unless otherwise indicated, predictions rely on the default threshold values proposed by dbNSFP, Ensembl or the dedicated web pages. For ESM1B, we applied a threshold (-7.5) yielding a true-positive rate of 81% and a true-negative rate of 82% in both datasets²⁷. We applied the threshold values proposed by Dong et al.³⁷ for GERP++RS (>4.4), Siphy (>12.17) and phyloP470way mammalian (>1.6) and a threshold value equivalent to purifying selection by ‘loss of function’ (> 7.5) for phyloP100way vertebrate⁴¹. For CADD, we either applied a common threshold value (≥ 15) with a sensitivity/specificity of 93,6%/57,1 or a higher threshold (≥ 25) shown to lower sensitivity to 71,5% and increase specificity to 85,3%^{37,42}.

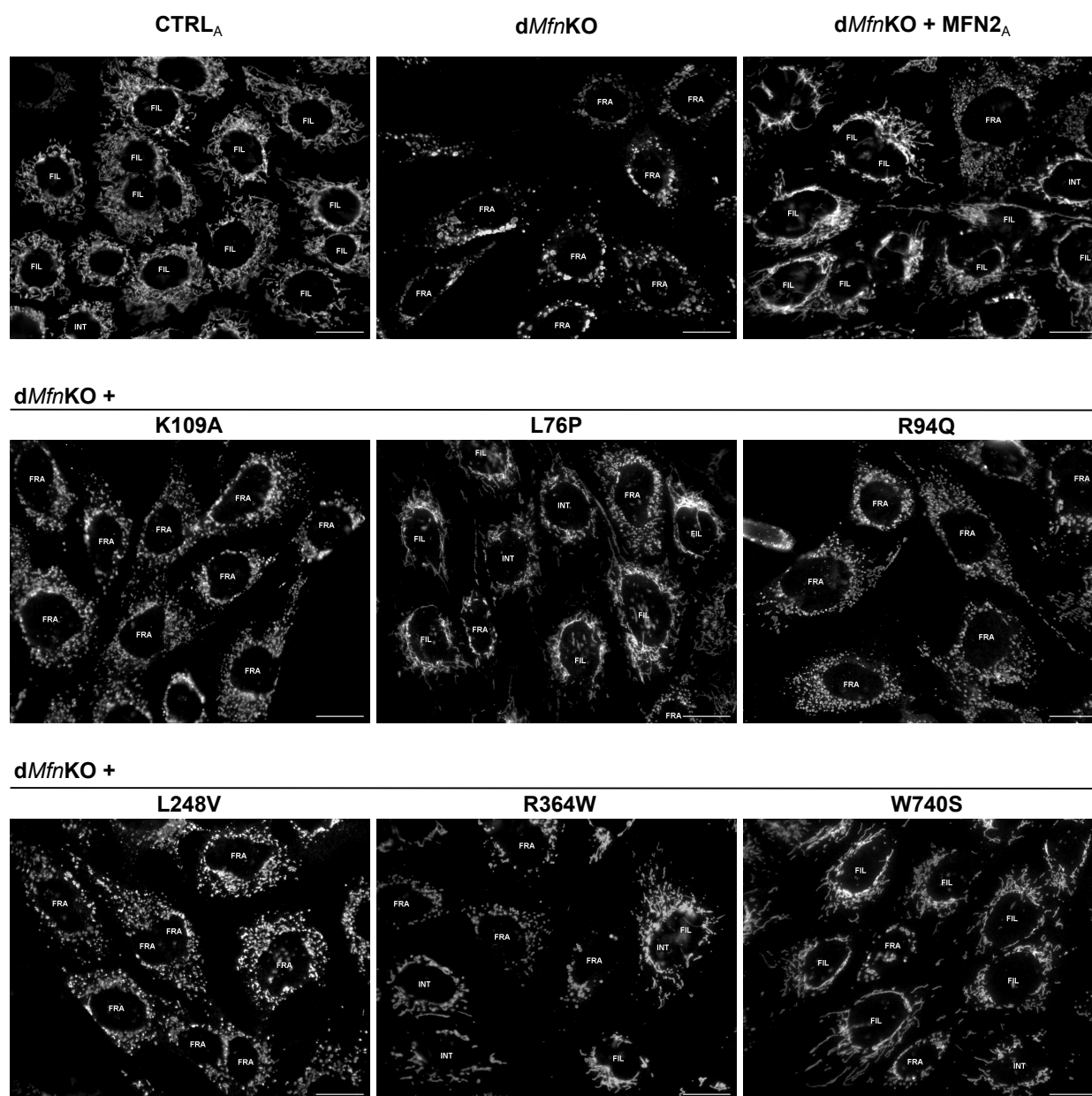
Couple	Forward primer (5' – 3')	Reverse primer (5' – 3')
1	ACGCCATCCACGCTGTTTTGACCT	CGTCTGCATCAGGGTGGACTCTGAG
2	CATCCAGGAGAGCGCCACCTTCCTTG	CGTCTGCATCAGGGTGGACTCTGAG
3	GTCTGGATGCTGATGTGTTTGTGC	GTAACCATGGAAACCATGAACTCCTC
4	CAGCATGCCCCCACTGCCACAGGGC	CTGCAGGTACTGGTGTGTGAACATG

Supplementary table 4. Oligonucleotides for amplification and sequencing of MFN2 cDNA.

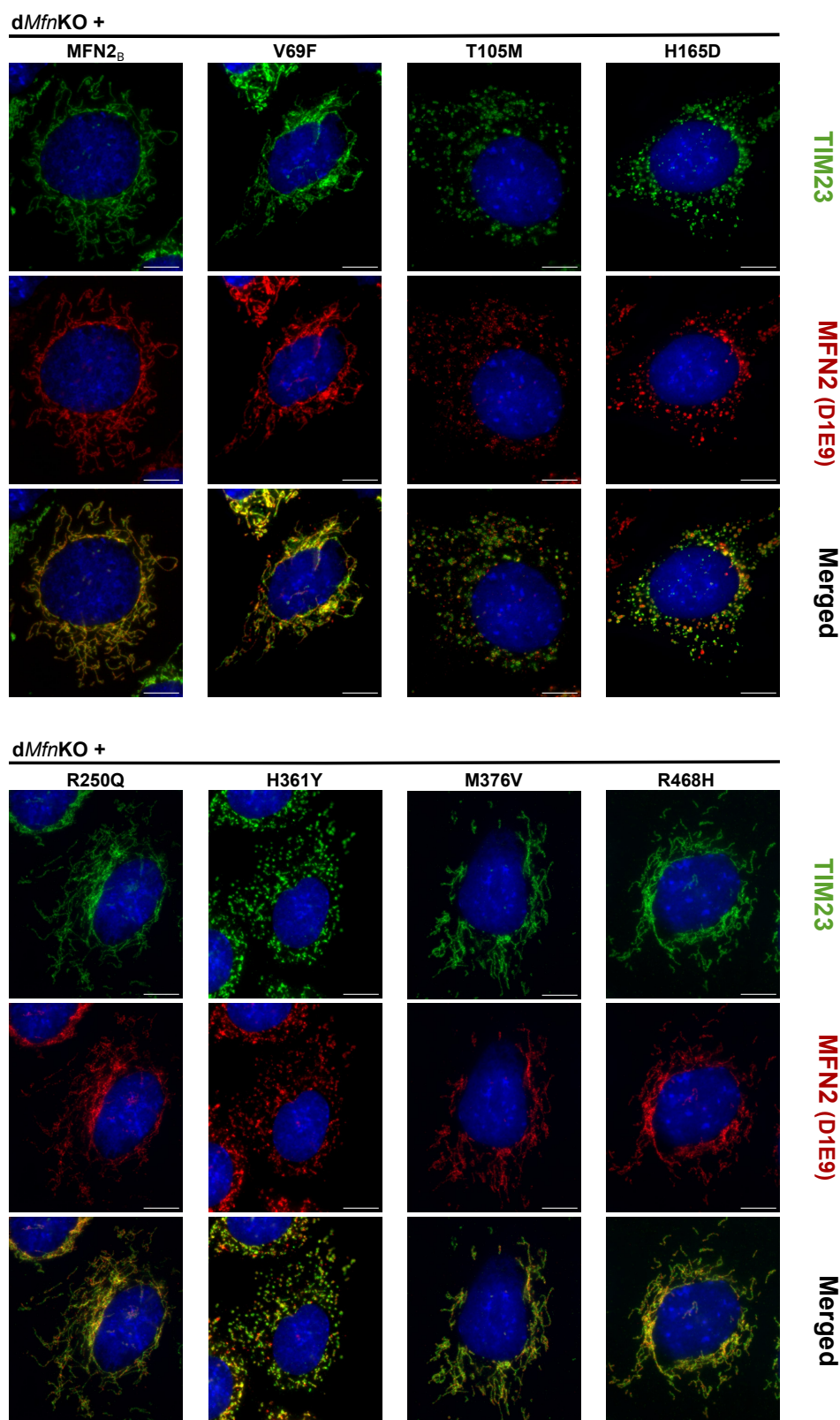
Supplementary Figures



Supplementary Figure 1. Uncropped images of the Western-blots shown in Figure 1B. Representative Western blots of MFN2 (mouse and rabbit antibodies targeting the C-terminal and N-terminal domain, respectively) and of a loading control (Beta actin) in the indicated MEF lines. Arrows point to the position of MFN2 or beta actin. The sizes of proteins in the prestained Protein Ladder are: ~250, ~130, ~100, ~70 (red), ~55, ~35, ~25 (red) and ~15 kDa. CTRL_A, CTRL_B: two different control MEF lines. MFN2_A, MFN2_B: two independently transduced *dMfn*KO MEF lines expressing wild-type MFN2.



Supplementary Figure 2. Visualization and assessment of mitochondrial morphology by immunofluorescence microscopy. Representative immunofluorescence images of wild-type MEFs (CTRL_A), untransduced dMfnKO MEFs and dMfnKO MEFs transduced with wild-type *MFN2* (MFN2_A) or the indicated variants. The overall mitochondrial morphology of cells stained with antibodies against the mitochondrial marker VDAC was classified as fragmented (FRA), filamentous (FIL) or intermediate (INT). Bar: 20 μ m.



Supplementary Figure 3. Mitochondrial localization of human MFN2 variants expressed in dMfnKO MEFs. Representative immunofluorescence images of cells stained with antibodies against the mitochondrial marker TIM23 (green) and MFN2 (D1E9, red) and with the nuclear stain DAPI (blue). MFN2, undetectable in untransduced dMfnKO MEFs, localizes to TIM23-positive mitochondria in transduced dMfnKO MEFs, independent of the fusion-capacity of the expressed MFN2 variant. Bar: 10 μ m.

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