Proximity proteomics reveals a co-evolved LRRK2-regulatory network linked to centrosomes

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Abstract

The Leucine-rich repeat kinase 2 (LRRK2) not only plays a vital role in familial forms of Parkinson's

disease (PD) but is also considered as a risk factor for idiopathic PD. Its multi-domain architecture

enables a fine-tuned regulation of its biological function by orchestrating intra- and inter-molecular

interactions. Here, we present BioID proximity proteomes of LRRK2 revealing new interactors, which

we further characterized by a novel evolutionary and structural bioinformatics pipeline. Co-

evolutionary analysis of the protein-protein interaction network identified a structural and functional

module enriched in cytoskeletal components linked to the centrosome and microtubules.

Furthermore, structural modelling of binary interactions via AlphaFold-multimer revealed distinct

groups of interactors engaging with LRRK2 dependent on specific conformations and epitopes.

Furthermore, we identified distinct changes in the LRRK2 proximity proteome induced by the type I

kinase inhibitor MLi-2 or by co-expression of the LRRK2 upstream effector RAB29. Dependent on its

state of activity and conformation, these protein-protein interactions link LRRK2 to defined cellular

sub-compartments, including centriolar satellites as well as vesicular sub-compartments.

Introduction

Parkinson's disease (PD) is the second most common neurodegenerative disease after Alzheimer's, it occurs mostly sporadically and is distributed with high frequency in the older generations. PD is characterized by a loss of dopaminergic neurons in the *substantia nigra* leading to the motor symptoms associated with the disease (Fahn, 2003). Several therapies targeting the symptoms, such as dopamine substitution, are practiced but so far, no causative therapy exists. About 5-10% of PD cases are inheritable (Gasser et al., 2011). *LRRK2*, localized within the PARK8 locus and coding for the Leucine-rich repeat kinase 2, is one of the genes associated with familial forms of the disease but also represents a risk factor in idiopathic PD (Paisan-Ruiz et al., 2004; Simon-Sanchez et al., 2009; Zimprich et al., 2004). The multi-domain protein LRRK2 is an active protein kinase and, at least, mutations with confirmed disease segregation, including the most common G2019S variant within its kinase domain, augment its kinase activity (Kalogeropulou et al., 2022). This makes LRRK2 an important drug target (Kluss et al., 2022b). ATP-competitive LRRK2-specific kinase inhibitors for PD-treatment are under development and in clinical trials (Jennings et al., 2022). Nevertheless, structural changes of the lung seen in non-human primates indicate a considerable small therapeutic window and additionally challenge a long-term treatment with these compounds necessary to allow a life-long reduction of augmented LRRK2 pathway activation (Miller et al., 2023). For this reason, the identification of alternative drug targets within the LRRK2-associated pathways would be a potential solution for a better control of pathophysiological alterations associated with pathogenic forms of this protein, minimizing the risk of drug-induced lung pathologies. In addition, a larger group of patients beyond LRRK2 mutation carriers might benefit from targeting more common PD-associated pathways.

Various functional studies have demonstrated a function of LRRK2 in vesicle trafficking, including presynaptic vesicles as well as autophagy and lysosomal pathways ((Piccoli et al., 2011); reviewed in: (Follett and Farrer, 2021)). In agreement with these cellular functions, LRRK2 phosphorylates a subset of Rab proteins, small G-proteins involved in vesicular transport, such as Rab8a and Rab10, which is elevated by PD-associated LRRK2 variants (Steger et al., 2017).

In common with other Roco proteins, LRRK2 features a remarkable complexity (Wauters et al., 2019). In fact, LRRK2 contains multiple scaffolding domains, not only being involved in keeping inactive LRRK2 in an auto-inhibited state, as recently demonstrated by high-resolution structures of LRRK2 (Deniston et al., 2020; Myasnikov et al., 2021), but are also involved in protein-protein interactions making LRRK2 a signaling hub possibly integrating multiple up- and down-stream signals (reviewed in: (Manzoni et al., 2015)). These studies suggest that the Roco protein LRRK2 is a tightly controlled protein kinase serving as hub and signal integrator and is most likely involved in context-specific, spatially controlled cellular responses that appear evolutionarily conserved. Quite some effort has therefore been spent on the investigation of the protein-protein interaction networks of LRRK2, mainly by using affinity-

based methods. Data of various studies were curated in public databases such as IntAct or STRING and

were also subject to meta-analysis to dissect functional LRRK2 signaling networks (Gloeckner and

Porras, 2020; Manzoni et al., 2015; Zhao et al., 2024; Zhao et al., 2023). However, affinity-based

methods face limits, given their bias for rather stable complexes which miss dynamic/ transient

interactions. We therefore used the BioID proximity labeling method to detect new interaction

partners of LRRK2. By systematically applying the BioID approach with three different BioID tags, we

describe the proximity proteome of LRRK2 under basal conditions. In addition, by a newly designed

bioinformatics workflow that estimates co-evolution of interacting partners based on their co-

presence in sequenced genomes, we could demonstrate that the resulting BioID datasets are

characterized by higher co-evolution values similarly to confirmed direct interactions. Through this

approach, we have identified a cluster of interactors featuring a high degree of co-evolution to LRRK2

which is enriched in cytoskeletal proteins involved in centrosomal and ciliary dynamics. We predicted

the 3D complexes mediated by LRRK2 via AF-multimer, which also revealed distinct conformations and

interfaces associated to specific processes. Finally, to further stratify the interactome of LRRK2, we

looked at experimentally-defined conformational states of LRRK2, e.g. by investigating the context-

specific proximity proteome upon co-expression of the LRRK2 upstream effector RAB29 and upon

stabilization of an active-like kinase conformation by inhibition of LRRK2 with a type I ATP-competitive

inhibitor (Taylor et al., 2020; Wauters et al., 2019). Inhibition of LRRK2 with the type I inhibitor MLi-2

but not with the type II inhibitor GZD-824 lead to an enrichment of a highly specific sub interactome

associated with centriolar satellites.

Methods

Cloning and generation of the stable miniTurbo-LRRK2 cell line

All used proximity tags have been ordered as synthetic genes (Eurofins) and subcloned via *Eco*RI and

*Xho*I into pcDNA3.0 (Invitrogen) allowing the expression of fusion proteins in mammalian cells under

the strong viral CMV promoter. Gateway Destination clones were generated by cloning a Gateway

cassette (Invitrogen) into an *Eco*RV site at the 3' end and in frame of the different BioID-tags [BioID

(Roux et al., 2012) (referred as BioID1), BioID2 (Kim et al., 2016), miniTurbo (Branon et al., 2018)]. For

better and comparable detection of the fusion proteins, all tags contained an N-terminal HA-tag. The

LRRK2 expression clones were subsequently generated by LR cloning (Invitrogen) based on ENTRY

clones described earlier (Gloeckner et al., 2010). The CDS of human of RAB29 has been ordered as

synthetic gene strand (Eurofins) with flanking attB sites to allow its direct cloning into pDONR201 by

BP cloning and subsequently cloned into an in-house (N)FLAG-HA Gateway destination clone by the LR

reaction. The HEK293 cell line (ACC-305) stably expressing miniTurbo-LRRK2 has been generated as

previously described (Gloeckner et al., 2007). Briefly, single colonies have been selected and tested

after transient transfection and selection via G418 (Sigma). The Flag-HA-CYLD vector was kindly

provided by Wade Harper (plasmid #22544, Addgene) (Sowa et al., 2009), the pCMV poly HA-ubiquitin

construct was kindly provided by D. Bohmann (Treier et al., 1994) and the pRK5-HA-Ubiquitin-K63

vector was kindly provided by T. Dawson (plasmid #17606, Addgene) (Lim et al., 2005).

Cell culture and proximity labelling

HEK239T (ACC-635) or HEK293 stably expressing miniTurbo-LRRK2 were grown in 14cm dishes in full

Dulbecco's modified medium (DMEM) supplemented with penicillin streptomycin and if not

mentioned otherwise with 10% FBS at 37°C in 5% CO2. The cells were grown for 8 hrs to 1 day reaching

a density of 50-80% until transfection with a self-made PEI (Polyethylenimine) reagent (Guaitoli et al.,

2016). For each condition, 4-6 plates have been used. If not stated otherwise, for biotinylation, cells

have been incubated with 50μM biotin (Sigma) in serum-free medium for up to 16 hours. In case of

the mini turbo tag, cells have been incubated with 50 μM biotin for 2h after o/n starvation. Starvation

was used to synchronize the cells and, in light of the physiological functions of LRRK2, to enrich for

post-mitotic states in which its centrosomal and ciliary roles are more prominent. For the

determination of the LRRK2-inhibitor dependent proximity proteomes, a HEK293 line stably expressing

mT-LRRK2 was either incubated with 1μM MLi-2, 1μM GZD-824 or DMSO (vehicle) 1h prior and during

2hrs biotin incubation. For the analysis of RAB29-dependent proximity proteomes, mT-LRRK2 was

either co-expressed with HA-RAB29 or an empty pcDNA3.0 vector. See also the point-to-point protocol

deposited on Zenodo (DOI: 10.5281/zenodo.17045153).

Affinity purification of biotinylated proteins

For cell lysis, cell culture dishes were transferred to ice and culture media was removed. The cells were

rinsed two times with 5ml ice-cold PBS. The cells were covered with 1ml lysis buffer [0.5% Nonidet P-

40 Substitute (Merck), protease inhibitor cocktail (Roche), PhosSTOP (Roche) in TBS (30mM Tris-HCl,

pH7.4 and 150mM NaCl)] per 14cm dish and detached with a cell scraper. The cell suspension was

collected and incubated for 20-40min on an end-over-end shaker at 4°C. Lysate were cleared by

centrifugation (15,000xg for 10min at 4°C). The protein concentration of cleared lysates was

determined photometrically using a nanodrop (Thermo-Fisher Scientific) or with the BCA assay

(Thermo-Fisher Scientific) following the manufacturer's instructions. Biotinylated proteins were

purified by affinity enrichment using either Strep-Tactin Superflow resin (IBA) or MagStrep Strep-Tactin

beads (IBA) in case of small-scale enrichments. To this end, the combined lysates for each condition

were incubated with an adjusted resin volume of 25μl packed beads per 14cm cell culture dish. The

resin was equilibrated by washing 3-4x with 500µl lysis buffer. The lysates were incubated with the

resin overnight at 4°C under constant agitation. The resin was sedimented by centrifugation at 5,000xg

for 3min at 4°C, transferred to microSpin columns (Cytiva) and subsequently washed 3x in 500µl lysis

buffer followed by two washes in 500µl 1x TBS. Biotinylated proteins were eluted by incubation with

300-400µl desthiobiotin elution buffer (2.5M, IBA). The desthiobiotin eluates were precipitated by

chloroform/ methanol and directly subjected to tryptic proteolysis following standard protocols

(Gloeckner et al., 2009) and stored at -80°C. The BioID protocol was constantly improved during the

work. See also the point-to-point protocol deposited on Zenodo (DOI: 10.5281/zenodo.17045153).

CYLD co-expression experiment

For the analysis of CYLD co-expression, cells were co-transfected with either Strep/FLAG (SF-TAP)

tagged LRRK2 WT or LRRK2 Y1699C, CYLD, and ubiquitin (either in WT form or K63) plasmids. As

negative control, CYLD plasmid was replaced by pCDNA3 empty vector, to maintain the plasmid

concentration constant among the different conditions. For 6-wells, a total of 500ng of plasmid DNA

was used for each well; for 14cm plates, a total of 8µg of plasmid DNA was used for each plate. The

plasmids ratios were LRRK2:CYLD:Ubiquitin 1:3:1. Transfection and cell lysis were performed as

described above.

siRNA knockdown

24h after splitting, cells were transfected with RNAiMAX (Life Technologies) following manufacturer's

instructions. The following siRNAs were used (Silencer Select®, Life Technologies): PCM1 (#s10128,

CAAAGACUCCACAUACGUUtt), Negative Control No. 2 (#4390846). 360pmol of siRNA were used for

each technical replicate. Cells were kept in culture for another 48h before proceeding with treatment

with 10µM MLi-2 or DMSO. Cells were lysed as described above. Three technical replicates for each

condition (PCM1 or scrambled, MLi-2 or DMSO) were used. The effects of PCM1 knockdown were

evaluated via SDS-PAGE and Western Blot. For each replicate, each individual protein was normalized

against total biotinylation. Statistical analysis was performed using GraphPad Prism 10.2.3. Unpaired

two-tailed t-tests were used to compare group means; each protein of interest was normalized against

total biotinylation.

SDS PAGE and Western Blots

For functional evaluation of the bait proteins and to determine the biotinylation efficiency, samples

were subjected to SDS PAGE and analyzed by Western-blotting on PVDF membranes. To assess blotting

efficiency and equal loading, the membranes were stained with Ponceau S. Prior to blocking of the

membranes in 5% nonfat dry milk powder (BioRad) or 1-5% BSA (Carl Roth) in TBS with 0.1% Tween 20

(TBST) for 1-3h at RT, membranes were de-stained in TBST. Membranes were afterwards incubated

with primary antibody solutions overnight at 4°C under mild agitation. The following primary

antibodies were used: rat α-LRRK2 mAb (24D8 clone, hybridoma supernatant, generated in house

(Carrion et al., 2017)) 1:1000 in 5% milk in TBST), rabbit α-pS935-LRRK2 mAb (Abcam, ab133450;

1:100,000 in 5% BSA in TBST), rabbit α-RAB10 mAb (Abcam, ab181367; 1:5,000 in 5% milk in TBST),

rabbit α-pT73-RAB10 mAb (Abcam, ab230261; 1:1,000 in 5% BSA in TBST), rat α-HA-tag (Roche, 3F10

clone, Rat 1gG1; 1:1,000 in 5% milk in TBST), mouse α-FLAG-tag (Sigma, F1804-200UG, 1:2,000 in 5%

milk in TBST), rabbit α-PCM1 (Proteintech, 19856-1-AP, 1:24,000 in 5% milk in TBST), mouse α-CYLD

(Proteintech, 66858-1-Ig, 1:1,500 in 5% milk in TBST), rabbit α-SSX2IP (Thermo Fisher Scientific, PA5-

31495, 1:1,000 in 5% milk in TBST). Membranes were washed 5x 5min in TBST prior to incubation with

the HRP-conjugated secondary antibodies α-rat (Jackson Immuno-Research, 112-036-062; 1:7,500), α-

rabbit (Jackson Immuno-Research, 111-036-045; 1:7,500), and α-mouse (Jackson Immuno-Research,

115-036-062; 1:7,500), respectively. For the detection of biotinylated proteins, membranes were

incubated with Pierce™ High Sensitivity Streptavidin-HRP (Thermo Fisher Scientific, 1:20,000 in 1% BSA

in TBST) overnight at 4°C and washed the next day 5x 5min in TBST. Blots were developed with Pierce

ECL Plus Western Blotting Substrate following the manufacturer's protocol (Thermo Fisher Scientific)

and exposed to Amersham Hyperfilm™ ECL (Cytiva). For quantification of LRRK2 in presence of CYLD,

the JESS capillary Western system (Bio-Techne) was used. Data analysis was performed with the vendor

software (Compass for SW, v7.0.0) and GraphPad Prism 10.2.3 Unpaired two-tailed t-tests were used

to compare group means; each protein of interest was normalized against beta-actin.

Mass spectrometry

In solution proteolysis was performed as previously described (Gloeckner et al., 2009). Briefly,

chloroform/ methanol precipitates were dissolved in 50mM ammonium bicarbonate pH8.0. Cysteines

were reduced and alkylated by treatment with 100mM DTT and 300mM Iodoacetamide, respectively

following the incubation with trypsin (Sigma-Aldrich) overnight at 37°C. Samples were desalted by

micropipette C-18 solid-phase extraction tips (StageTip, Thermo Fisher Scientific) prior to LC-MSMS

analysis. Peptides were subsequently analyzed by LC-MSMS using a nanoflow HPLC system (Ultimate

3000 RSLC; Thermo-Fisher Scientific) coupled to an Orbitrap Q-Exactive plus (Thermo Fisher Scientific)

tandem mass spectrometer. Peptides were separated by reversed C-18 chromatography and 120-

minute gradients. MS1 spectra were acquired in the Orbitrap at 70K resolution. After selection of the

10 most intense precursor ions from the MS1 scans for HCD fragmentation (Top 10 method), MS2

spectra were acquired at 17.5K. RAW MS files were analyzed with Maxquant (version 2.0.3.1), enabling

label-free-quantification (LFQ with no normalization) and "match between runs" (Cox et al., 2014). For

the embedded Andromeda search engine, trypsin has been selected as enzyme and carbamylation was

set as fixed modification N-terminal acetylation was set as variable modification. The search was

performed against the human subset of the Uniprot/Swissprot database (release 2021_04, 20375

entries). The MaxQuant output was further processed by Perseus and SaintExpress.

Perseus post-processing

For post-analysis, the protein groups were further processed by Perseus (v. 1.6.7). LFQ intensities of

the individual were assigned to the conditions and potential contaminants, reverse hits as well as IDs

only identified by site were filtered out. For further analysis, only IDs with a maximum number of

missing values of n-1 in at least one condition were considered. Remaining missing values were

replaced by group-wise imputation based on a normal distribution (log2 values) around the detection

limit (width=0.3; down-shift=1.8). Intensities of the individual experiments were normalized by their

medians and significantly enriched proteins (condition vs control) were determined by a T-test

combined with an FDR filter (p=0.05, S0=0.1).

SaintExpress post-processing

We followed a previously described protocol (http://proteomics.fi (Liu et al., 2020)) to select the most

confident interactors using the statistical framework of SAINTExpress (Teo et al., 2014), as well as a

reference knowledgebase for contaminants (i.e. CRAPome (Mellacheruvu et al., 2013)), to filter out

most likely false negative interactors. In particular, we generated the "Bait", "Prey" and "Interaction"

files to be subjected to the analysis pipeline from the MaxQuant "*proteinGroups.txt*" file by employing

the following columns: "Bait" file (IP name=EV_* or LRRK2_*, bait name=LRRK2, indicator for test and

negative control= C=EV_* ;T=LRRK2_*); "Interaction" file (IP name = EV_*/LRRK_*, bait name=LRRK2,

prey name=interaction partner gene symbol, spectral count or intensity values=LFQ intensity); "Prey"

file (prey (protein)name= interaction partner Uniprot accession, prey protein length=Sequence length,

prey gene name= gene symbol). The proteins that were not classified through the CRAPome filter,

were retained together with those passing the filter.

We unified the interactor list from the three BioID experiments. Finally, we retained a list of 208

interactors which passed the two filters. Only 45 proteins in the filtered interactor list are already

present in IntAct.

Graphical representation and network annotation

PPI networks were analyzed by Cytoscape version 3.8.2 (Shannon et al., 2003) and were mapped to

available interaction from IntAct using the Cytoscape IntAct app (Ragueneau et al., 2021). Networks

were annotated with STRINGDB terms via the 'Stringify' function of the StringApp (Doncheva et al.,

2019) and functional enrichment retrieved via the same app. We annotated the networks with the

AutoAnnotate plug-in (Kucera et al., 2016), using the GLay clustering method, and generated the

cluster labels with the Adjacent Words option of the WordCloud algorithm, based on significantly
enriched GO Biological Process and Cellular Component terms associated with the genes in each
network. The Venn diagram shown in Figure 1A has been generated with JVenn provided by the
ProteoRE server (Galaxy Version 2021.05.12) (Galaxy, 2024).

Co-evolution based PPI stratification

235 Co-evolution of two proteins A and B was measured as the Jaccard similarity coefficient of the set of
236 genomes containing an orthologous sequence of A and the set of genomes containing an orthologous
237 sequence of B. Orthologous sequences were collected from the Orthologous MAtrix (OMA) database
238 (Altenhoff et al., 2021) using a Python script written in-house using the PyOMADB client (Kaleb et al.,
239 2019) (Kaleb et al., 2019) (December 13, 2023, database version July 2023).

AF-multimer prediction of binary interactions

We used Alphafold-Multimer v2.3.2 (Evans et al., 2021) to generate 3D models of full-length LRRK2
binary interaction complexes for all the interactors shorter than 2000 residues. The databases required
to run AlphaFold-Multimer were downloaded on 12 January 2023 and the predictions were run with
the flag –max template date=01-01-1900, to avoid the usage of any available experimental templates.
Among the 5 models generated for each LRRK2 binary complex, only the best one was considered for
further analysis. The score used to evaluate the models was the default one used by AlphaFold-
Multimer (0.2*pTM + 0.8*ipTM).

Interaction fingerprint

AlphaFold produces probabilities for the C_β-C_β distances between residues i and j to fall into a series of
distance bins. For each pair of residues, we calculated the expected value of such distance distribution,
yielding a matrix of dimension l_{LRRK2} x l_n, corresponding to the sequence lengths of LRRK2 and the nth
interacting partner, respectively. Subsequently, we pooled the distogram on the interaction partner's
side and we assigned to each residue of LRRK2 the minimum expected distance to any residue of the
interactor. We refer to this vector of minimum expected distances as the interaction fingerprint of a
given interacting protein.

RMSD calculation

257 We compared LRRK2 conformations by performing Root Mean Square Deviation (RMSD)-based
258 clustering. To calculate the RMSD we fitted the structures using the C_α atoms of the LRRK2 chains from
259 position 983 to the end of the sequence. The N-terminal portion of the protein was excluded because
260 it is very flexible. Calculations were performed using the Superimposer function of the PDB Biopython
261 module (Cock et al., 2009) (version 1.83) through customized scripts and pipeline (Matic et al., 2023).

Clustering

The clustering was performed using the Ward method with the *clustermap* function from *seaborn* python library (version 0.11.2). Results were displayed through matplotlib (https://matplotlib.org/) and seaborn (https://seaborn.pydata.org/) libraries using customized python scripts. The number of clusters chosen was one that maximized the mean Silhouette Coefficient of all samples (Rousseeuw, 1987) , as computed by the *silhouette_score* function in *scikit-learn* (Pedregosa et al., 2011)*.*

Results

Definition of cellular LRRK2 proximity proteomes with BioID1, BioID2 and miniTurbo.

The elucidation of comprehensive protein-protein networks represents a powerful tool not only to identify signaling pathways associated with a protein of interest but also to connect different pathogenic mutations in a disease network or identify novel disease-associated proteins through proximity to known proteins (Boldt et al., 2016; Kiel et al., 2011). This approach might ultimately lead to the discovery of new drug targets as well as disease biomarkers. For this reason, various studies addressed the interactomic networks of LRRK2 in a different cellular context or model organism (Islam et al., 2016; Meixner et al., 2011; Piccoli et al., 2011; Piccoli et al., 2014). However, most of the used technologies rely on immuno-precipitation or pull-downs, which have a bias for considerable stable PPIs while transient interactions remain underrepresented. In the last years, proximity-based methods, e.g. BioID, have been developed allowing the identification of PPIs under physiological conditions and within cells also covering transient interactions. The BioID approach is based on a variant of the bacterial biotin ligase BirA, which has been converted to a promiscuous labeling tool by a single point mutation within its active center allowing the release of the reactive intermediate biotinyl-AMP. To determine the LRRK2 proximity proteome, we have generated different N-terminal fusion proteins. Besides the original BioID (BioID1) tag (Roux et al., 2012), we used the BioID2 (Kim et al., 2016) as well as miniTurbo, a shorter version of the TurboID tag, lacking the BirA DNA binding domain (Branon et al., 2018). For reasons of comparability, all tags have been cloned in the same way, containing an N-terminal HA tag for detection and a 3x(GGGS) linker between the tag and the protein of interest inserted C-terminally of the tag by Gateway-cloning. All constructs were functionally assessed to demonstrate biotinylation activity and to ensure the tag does not interfere with the activity of LRRK2. Western blot analysis shown in Supplemental Figure 1 demonstrates that all constructs allowed sufficient biotinylation in cells. As expected, BioID2 and miniTurbo showed a higher labeling efficiency compared to BioID1. We also confirmed that the fusion proteins are still able to phosphorylate the physiological LRRK2 substrates by probing for Rab10 phosphorylated on Threonine-73 (see: Supplemental Figure 1).

After initial functional tests, all three fusion constructs were used to determine their interactomes

after transient expression in HEK293T cells. To increase the stringency of the approach, a tag-only

control (corresponding BioID tag) was used as reference. For the relative comparison of the LRRK2 vs.

the tag-only control, the expression levels were adjusted via pre-experiments and always confirmed

by Western blot analysis of the cell lysates. In contrast to comparing plus/minus biotin conditions, the

tag-only control allows to efficiently exclude natively biotinylated proteins as their degree of

biotinylation is also increased by the biotin treatment of the cells already by the experimental design.

The data of all three proximity proteomes were analyzed by SaintExpress and, alternatively, by

labelfree quantification (LFQ) combined with an FDR-controlled T-test as implemented in Perseus

(Tyanova et al., 2016). To further filter and analyze the data set, we implemented a bioinformatics

pipeline based on interacting genes co-evolution and structural modeling by Alphafold-Multimer

(Figure 1A). With BioID1, BioID2 and miniTurbo, 168, 312 and 241 proteins were identified,

respectively, as being significantly enriched in the LRRK2-containing condition over the tag-only control

by the LFQ approach (Figure 1B; Table S1). SaintExpress (Teo et al., 2014) and subsequent filtering

against the Crapome data of commonly found false-positive hits (Mellacheruvu et al., 2013), however,

appeared to be the more stringent analysis compared to MS1-based label-free quantification. Based

on this analysis, we created a non-redundant set of 208 unique interactors, hereinafter referred to as

"BioID set" which was submitted to the bioinformatics post-processing pipeline.

Functional-enrichment analysis of the LRRK2 proximity proteome

We compared the LRRK2 BioID interactome with a reference LRRK2 interactome available from IntAct

(Orchard et al., 2014). We found that only 54 interactors in the filtered BioID set were previously

reported in association with LRRK2, while the vast majority (154) were not yet reported for their

interaction with LRRK2 (Figure 2A,B). By integrating the LRRK2 BioID interactome with the network of

known interactions from Intact it is possible to define several modules of proteins interacting with

each other in addition to LRRK2 (Figure 2A). Among the enriched intracellular and cytosolic

compartment, we found terms related to microtubules (e.g. Microtubule Cytoskeleton, FDR=5.83E-15)

and Centrosome (FDR= 1.43E-13), emphasizing the importance of LRRK2 in microtubule organization

(Figure 2C) and representing the largest module of interacting proteins in the network (Figure 2A). In

addition, we also found processes related to cell cycle and mitosis (e.g. "Regulation of mitotic cell

cycle", FDR=9.71E-08) (Figure 2D).

Multiple protein domains instances are significantly enriched within the list of identified LRRK2

interaction partners (Figure 2E). In particular, we found that the Armadillo-type fold instance is the

most enriched one (FDR= 9.58E-06), which is found in 21 distinct interactors, followed by other

domains with lower, but still significant, degree of enrichment, such as "Forkhead-associated (FHA)

domain" (Almawi et al., 2017), found in six distinct proteins, the "DIX domain" (Ehebauer and Arias, 2009; Kafka et al., 2014) found in four proteins, the "K Homology domain, type 1" (Grishin, 2001) found in six proteins and "Ubiquitin-like domain superfamily" (FDR<0.05), the latter being found in 11 distinct interacting proteins. Interestingly, LRRK2 and its Drosophila orthologue dLRRK have been demonstrated to functionally interact with the Forkhead box transcription factor FoxO1/dFoxO. In particular, LRRK2-mediated phosphorylation of FoxO1 enhances its transcriptional activity leading to neuronal toxicity (Kanao et al., 2010).

Analysis of co-evolution reveals a cluster of proteins involved in centrosome/ microtubular dynamics

We analyzed the extent of co-evolution of LRRK2 with its interactors to prioritize the most conserved relationships associated to LRRK2's functions. We derived a metric for the co-evolution between two interacting proteins based on the degree of co-presence, assessed through a Jaccard score, of interacting gene pairs in sequenced genomes available from a reference orthology database (e.g. OMA, see Methods). We found that such simple co-evolutionary metric is informative about non-enzymatic, direct interactions as well as stable complexes at a human interactome-wide level (Meldal et al., 2022; Miglionico et al., 2024; Orchard et al., 2014) (Table S2A). We found similar trends when focusing on LRRK2 interactions reported in IntAct, with "non-enzymatic direct" PPIs being characterized by a significantly higher Jaccard score compared to those categorized as "association" or "physical interaction" (Figure 3A). Intriguingly, we found that the distribution of co-evolution metrics calculated for BioID interactors is significantly higher than most IntAct categories (i.e. "association", "physical", "enzymatic direct") and it is as high as the non-enzymatic direct interactions (Figure 3A). This result suggests that interactors captured within the BioID dataset are more likely to represent stable, direct interaction partners of LRRK2 with whom they co-evolved. Likewise, simple co-evolution scores are predictive for LRRK2 direct interactions, particularly the non-enzymatic direct ones (Figure 3B; ROC AUC=0.715). In the case of the LRRK2 interactome, we found that this co-evolutionary metric is more predictive of non-enzymatic direct interaction compared to another co-evolution estimate, i.e. Mutual Information (ROC AUC=0.66; Supplemental Figure 2), which is instead more predictive at a human proteome scale (Miglionico et al., 2024).

We used the co-evolution metric to cluster LRRK2 and its interactors, which revealed groups of proteins characterized by reciprocal high co-evolution (Figure 3C). The different groups of interactors are also characterized by specific biological processes, which are overall mutual exclusive to each other, suggesting that clusters of co-evolved proteins underpin specific functional interactions and biological processes (Supplemental Figure 3). Notably, the cluster containing LRRK2 (i.e. Cluster 2) is characterized by on average higher co-evolution and several highly co-evolved proteins (Figure 3C). The top 10 proteins of Cluster 2 showing highest co-evolution are summarized in Table 1. By mapping

members of this cluster onto the STRING functional interaction network (Figure 3D, see Methods), we found functional interactions for most of the members, including LRRK2. Overall, STRING yielded a highly significant PPI enrichment (P-value=1.0E-16), suggesting that these proteins are more likely to interact with themselves than with randomly selected proteins. Moreover, network topology analysis delineated two distinct submodules: the first encompassing LRRK2 together with proteins predominantly linked to Wnt signaling (FDR = 0.0013; Figure 3E), and the second consisting of proteins implicated in cytoskeletal processes ("Protein localization to microtubule cytoskeleton"; FDR=2.18E-05), centrosome dynamics ("Loss of Nlp from mitotic centrosomes"; FDR = 0.00086) and centriole organization ("Centriole"; FDR = 1.11E-05). Consistently, Cluster 2 displayed strong enrichment for cellular component terms, most notably "Microtubule Cytoskeleton" (FDR=2.10E-12) and "Centrosome" (FDR=1.01E-12) (Table S2B). Interestingly, the single most co-evolved protein to LRRK2 is CYLD, a deubiquitinase that acts as a negative regulator of dopamine neuron survival in Parkinson's disease (Pirooznia et al., 2022), but that has not yet been reported in direct interaction with LRRK2. Interestingly, CYLD emerges as a critical bridging element that connects the two principal submodules within Cluster 2's interactome (Figure 3D).

On the other hand, Cluster 1 is characterized by distinct processes, such as FAR/SIN/STRIPAK complex (P-value= 0.0012; Table S2B). Taken together, these results suggest that our co-evolution-based approach was able to identify a group of LRRK2 interacting proteins, mostly associated to microtubule and centrosome functions via the formation of stable, supramolecular complexes.

Likewise, co-evolution-based clustering of LRRK2 interactors from IntAct shows interactors such as RAB29, RAB12, LRRK1 or SNCA among the most co-evolved partners which clustered with LRRK2 (Supplemental Figure 4), confirming the tight evolutionary constraint on LRRK2's functional regulation.

After confirming the proximity of LRRK2 and CYLD (Figure 3F) by Western blot, we investigated the functional relevance of the interaction between LRRK2 and CYLD by examining whether the latter influences the stability of the former. Notably, co-expression of CYLD led to a marked increase in LRRK2 stability. This effect was observed for both wild-type LRRK2 and the pathogenic Y1999C variant, which is characterized by elevated Rab phosphorylation levels but reduced protein stability, as quantified using capillary-based Western blotting (Figure 3G-H). Given that CYLD is a deubiquitinase with specificity for K63-linked ubiquitin chains, we co-expressed HA-tagged ubiquitin constructs—one encoding wild-type ubiquitin and another engineered to exclusively support K63-linked polyubiquitination (with all other lysines mutated to arginine, as described by Kaleb et al. (Kaleb et al., 2019)). Consistent with CYLD's known substrate specificity, the stabilizing effect on LRRK2 was significantly enhanced, while only a trend was observed with wild-type ubiquitin. A second, independently conducted biological replicate confirmed these trends (Supplemental Figure 5).

In summary, our data demonstrate that CYLD co-expression enhances LRRK2 protein stability,
supporting a functional role for CYLD-mediated deubiquitination in regulating LRRK2 turnover.

AF-based modelling reveals a discrete repertoire of conformations and interfaces shaping the LRRK2
interactome

We predicted 3D complexes for binary interactions of LRRK2 and 196 partners in the BioID dataset via
AlphaFold-multimer (AF) and characterized their interfaces and conformations (Figure 4A).
We structurally aligned LRRK2 conformations in the different complexes by superimposing the
monomeric LRRK2 chains and then calculating the pairwise distance of all the possible pairs via Root
Mean Squared Deviation (RMSD) (Figure 4A; Methods). We performed hierarchical clustering of LRRK2
structures based on RMSD, through which we identified two main clusters (Figure 4B). A smaller one
is characterized by conformations with lower deviation from the inactive LRRK2 structure (PDB ID:
7LHW), which we termed "locked" (Figure 4B-D). The main structural hallmarks characterizing the
"locked" conformation and inherited by the experimental inactive LRRK2 structure (PDB:7LHW)
(Myasnikov et al., 2021), are the LRR wrapping around the catalytic domains, particularly the kinase,
and a contact formed by the extended loop within the LRR repeats, which includes the hinge helix, and
the C-terminal WD40 (Myasnikov et al., 2021). A bigger cluster comprised conformations that are
characterized by a larger deviation from the inactive structure, due to a major rearrangement of the
LRR, ROC and COR domains, which we termed "unlocked" (Figure 4B-D). Such conformations are
reminiscent of the ones that we observed in the predicted LRRK2-RAB10 complex, which we validated
as the AF-predicted conformation best accommodating experimental XL-MS data (Guaitoli et al.,
2023). Intriguingly, we found that proteins interacting with LRRK2 in these two alternative
conformations are associated with distinct evolutionary and functional properties. Indeed, LRRK2's
complexes in the "locked" conformations are characterized by a significantly lower RMSD (P-
value=3.81E-20; Figure 4D) and higher co-evolution (P-value=1.33E-3; Figure 4E) than the complexes
in the "unlocked" conformations. The two groups are also characterized by largely exclusively enriched
processes (Figure 4F). In particular, the interactors engaging with LRRK2 in the locked conformation
are significantly associated with processes linked to centriolar satellites (P-value=4.76E-10) and cilium
assembly (P-value= 0.0001) (Figure 4G; Table S2C). Those interacting with the unlocked conformations
are associated with a higher number of enriched processes, including cluster-specific ones such as
cytoskeleton (P-value= 4.53E-09) or the microtubule organizing center MTOC (P-value= 6.40E-07).
Hence, the predicted LRRK2 conformational switch might be relevant in orchestrating the interaction
with partners associated with distinct functions.
We also explored patterns of engagement of interactors with the different regions of the LRRK2 multi-
domain architecture. To this end, we considered the distribution of distance probabilities (i.e.

distograms) generated by AF for each binary interaction and mapped the most likely interacting

residues of the partners with respect to LRRK2 positions. We used these partner-specific interaction

fingerprints to cluster the interactome and identified four major groups of proteins, each named

according to the LRRK2 3D interface involved in the interaction (Figure 5A; Table S2D; see Methods).

The Core cluster comprises partners that interact with the LRR–ROC–COR–KIN domains and was

named accordingly, as these domains represent the most evolutionarily conserved regions of LRRK2

(Figure 5A,B). The WD40 cluster includes complexes involving the homonymous domain (Figure 5A,C),

while the ARM cluster comprises proteins engaging a newly identified site within the Armadillo domain

(Figure 5A,D). Lastly, the "Other" cluster contains proteins with alternative interaction modalities or

no clearly defined binding interface (Figure 5A). Complexes predicted to interact with the Core group

displayed significantly higher AlphaFold confidence scores compared to those interacting with the

WD40 domain (P-value=1.14E-8) or via "Other" interaction modalities (P-value=4.86E-14; Figure 5E).

Complexes of the Core group are also characterized by significantly higher RMSD from interactors of

the WD40 one (P-value=4.05E-3; Figure 5F), which bind LRRK2 in both locked and unlocked

conformations (see below). In the Core cluster we found established LRRK2 interactors, such as DVL1-

2, confirming the known interaction mechanism through the ROC-COR domains (Sancho et al., 2009).

In the same cluster, we also found novel interactors discovered within the BioID set, such as e.g. CYLD

interacting through the Core interface in a similar way. A second cluster displayed higher likelihood of

interaction with the WD40 domain (Figure 5A,C), with predicted complexes characterized by a similar

interface involving the binding of a short, disordered peptide stretch to a conserved crevice formed by

β-propeller loops of the WD40 domain (Figure 5C). WD40 domain binders might bind to LRRK2 in either

locked or unlocked conformation (Figure 5A, F), as also suggested by a bi-modal distribution of the

LRRK2 chain RMSD (Figure 5F). These interactors also displayed significantly higher co-evolution with

respect to "Other" LRRK2 interactors (P-value=5.57E-4; Figure 5G). The ARM cluster involves proteins

showing binding preferences for a new docking site on the Armadillo domain, centered around

residues 500-600 and mediating the interaction with proteins such as AXIN1 (Figure 5A,D). The larger

cluster (Other) contains proteins interacting with weaker likelihood and variable docking sites (Figure

5A). Among these, we also found proteins engaging with two more ARM docking sites, including one

(around aa 240) which has been recently shown to interact with RAB12 (Dhekne et al., 2023; Li et al.,

2024) and mediate a RAB29-independent LRRK2 Kinase activity activation (Purlyte et al., 2018; Zhu et

al., 2023), which is predicted to interact with proteins such as KIF16B. A third ARM binding region,

encompassing aa 354-444, corresponds to the RAB29 interaction interface found in the recently

determined RAB29-bound, tetrameric LRRK2 structure (Zhu et al., 2023), and is the predicted interface

for proteins such as NBR1 and MON2.

Overall, these results further support the hypothesis that interactors *de-novo* identified by proximity

labelling may physically interact with LRRK2 domains through specific conformations and contact

interfaces. Interactors lacking a strong interaction probability in the distogram might represent

proteins interacting with LRRK2 either transiently or indirectly.

Type I but not type II inhibitors lead to conformational changes directing LRRK2 to centriolar satellites

As mentioned above, LRRK2 adopts multiple conformations through which it engages with its

interactors. Effector-mediated oligomerization to an asymmetric tetramer has been demonstrated to

switch LRRK2 to an active conformation (Zhu et al., 2023). In addition, ATP-competitive kinase

inhibitors can stabilize different conformations. While type I inhibitors stabilize the kinase domain in

an active-like conformation, type II inhibitors stabilize the inactive conformation of the kinase domain

(Deniston et al., 2020; Raig et al., 2024; Zhu et al., 2024). Interestingly, type I but not type II inhibitors

lead to a rapid dephosphorylation of LRRK2 phospho-sites in an interdomain space downstream of the

hinge-helix which locks LRRK2 in a compact inactive fold by binding the C-terminal WD40 domain

(Kalogeropulou et al., 2022; Myasnikov et al., 2021). We hypothesized that specific inhibitors could

distinguish the proximity proteome of LRRK2 in its active versus inactive conformation. To test this, we

compared the LRRK2 proximity proteome in HEK293 cells stably expressing miniTurbo-tagged LRRK2

following treatment with the type I kinase inhibitor MLi-2, the type II inhibitor GZD-824 (previously

shown to inhibit LRRK2 in cells (Kalogeropulou et al., 2022; Tasegian et al., 2021)), or a DMSO control.

Western blot analysis of LRRK2 pS935 and pT73-Rab10 confirmed effective inhibition under these

conditions (Supplemental Figure 6A). Quantitative label-free analysis combined with an FDR-controlled

t-test revealed that only MLi-2 treatment led to the enrichment of a distinct set of proteins associated

with cilia assembly and centriolar satellites in the LRRK2 proximity proteome (Figure 6A–E,

Supplemental Figure 6B), a pattern not observed with GZD-824. Intriguingly, interactors determined in

presence of the MLi-2 inhibitor are characterized by a significantly higher co-evolution (Figure 6F)

compared to other BioID interactors. The complete list of significantly enriched proteins (MLi-2 vs.

DMSO) is shown in Table 2. Additional information is provided in Table S1.

To functionally validate the interaction between LRRK2 and centriolar satellites, we first confirmed

that LRRK2 associates with key components of these membrane-less organelles, including the central

scaffold protein PCM1 and the centrosome maturation factor SSX2IP (Barenz et al., 2013).

Both proteins were enriched in LRRK2 proximity proteomes following biotin treatment, with their

interaction further enhanced by the LRRK2 inhibitor MLi-2. To further substantiate LRRK2's association

with centriolar satellites, we employed RNA interference to deplete PCM1, a strategy known to disrupt

satellite integrity (Dammermann and Merdes, 2002). As expected, PCM1 depletion led to its absence

from the LRRK2 proximity proteome and also abolished the association between LRRK2 and SSX2IP.

Remarkably, SSX2IP was entirely absent from LRRK2 proteomes regardless of MLi-2 treatment,

indicating that LRRK2's interaction with SSX2IP is dependent on intact centriolar satellites. Finally, we

asked whether the MLi-2–induced dephosphorylation of serine 935—located near the LRRK2 hinge

helix—is affected by satellite disruption. However, PCM1 knock-down did not influence serine 935

dephosphorylation, suggesting that this post-translational modification occurs independently of

centriolar satellite integrity (Figure 6D, Supplemental Figure 6C).

RAB29 co-expression directs LRRK2 to well established lysosomal protein networks

It has been previously shown that LRRK2 interacts with the RAB29/RAB32/RAB38 family of small G-

proteins and subsequently locates to post-Golgi endomembranes (Beilina et al., 2014; McGrath et al.,

2021; Vides et al., 2022; Waschbusch et al., 2014). Furthermore, RAB29 not only recruits LRRK2 to

lysosomal membranes but also induces its tetramerization (Zhu et al., 2023). For this reason, we

performed the BioID experiment in dependence of RAB29 co-expression to determine PPI networks

associated with the activated oligomeric state of LRRK2. In addition to mT-LRRK2, either RAB29 or an

empty vector were co-expressed. The resulting differential proximity proteome showed differentially

enriched proteins only for the condition where RAB29 was co-expressed, demonstrating that a small

portion of LRRK2 was recruited by RAB29 with lesser impact on the total proximity proteome (Figure

6G). While we did not observe any significant difference in terms of co-evolution of the RAB29

differential proteome with respect to the BioID one (Figure 6F), we found significant enrichment in

proteins involved in vesicular transport, including the post-Golgi/ lysosomal compartment as revealed

by an LFQ analysis combined with an FDR-controlled T-test (Figure 6H,I). In particular, known LRRK2

effectors, namely RAB8A (Steger et al., 2017) and SPAG9 (Boecker and Holzbaur, 2021; Bonet-Ponce

et al., 2020; Kluss et al., 2022a) were among the enriched proteins (Figure 6G,H). Furthermore, VAMP7,

which is part of a SNARE complex mediating autophagosome-lysosome fusion (Jian et al., 2024), was

among the significantly enriched proteins (Figure 6G,H). VAMP7-LRRK2 physical interaction has

recently been described (Filippini et al., 2023). The complete list of significantly enriched proteins is

shown in Table 3. Additional information is provided in Table S1.

MLi-2 and RAB29 differently modulate the shape the LRRK2 interactome

We constructed 3D models of binary complexes between LRRK2 and its partners from the MLi-2– and

RAB29-dependent interactomes, and examined both the complex interfaces and the conformational

states of LRRK2 (Vides et al., 2022).

Interface-based clustering suggests that MLi-2–dependent interactors engage with LRRK2 through

specific domain regions, particularly the ones involving terminal domains (Supplemental Figure 7A),

while they are predicted to engage little with the catalytic domains, particularly the kinase, which is

consistent with the MLi-2 inhibitor function, while in the RAB29-dependent LRRK2 interactome we see
a more variable pattern of engaged interfaces. In particular, predictions for this interactome revealed
a cluster of proteins interacting with LRRK2 through the ROC-COR-KIN domains, including the RAB8A
substrate (Supplemental Figure 7B), suggesting that in the context of RAB29 overexpression several
partners might bind in a substrate-like manner. Notably, this interaction modality is absent in the
predicted binary complexes of the MLi-2–bound interactome, confirming that the inhibitor prevents
LRRK2 from binding substrate-like partners. Moreover, we further validated the capability of the
distogram-based interface-fingerprint to discriminate functionally relevant LRRK2 interactors. We
predicted the 3D complexes of LRRK2 with known interacting Rabs, and we found out that Rab
substrates clustered separately from interacting RABs with distinct functions (e.g. RAB29 and RAB32;
Supplemental Figure 7B).

From a conformational perspective, we pooled the RAB29- and MLi-2–dependent interactomes and
clustered LRRK2 structures based on RMSD, which likewise revealed distinct locked and unlocked
clusters. Interactors from the RAB29-dependent interactome are predicted to bind LRRK2 in both
conformations, with a particular enrichment in the unlocked state. (Figure 7A). Conversely, MLi-2–
dependent interactors were significantly enriched in the locked LRRK2 conformation (Fisher's exact
test, P = 0.023), further suggesting that the inhibitor-bound state favors engagement with partners
that, unlike substrates, do not require interaction with the kinase domain (Figure 7A). Also for these
interactors, the locked conformation has a significantly lower RMSD (P-value=3.85E-6; Figure 7D) and
a tendency towards higher co-evolution (Figure 7E). Notably, enrichment analysis of interaction
partners grouped by conformational cluster revealed even stronger cluster-specific associations, with
only a single process shared between partners in the locked and unlocked conformations (Figure 7F,
G; Table S2E).

Discussion

The systematic analysis of the interactomes of disease-associated proteins is a powerful tool to dissect
molecular networks and to identify pathways relevant for the molecular pathophysiology, for example
demonstrated by the comprehensive interactomic network underlying the assembly and homeostasis
of primary cilia (Boldt et al., 2016). Given its complex domain structure and tight regulation, LRRK2 is
most likely representing a scaffold protein orchestrating distinct signaling complexes, which makes the
analysis of its PPI networks an attractive approach to dissect context-specific signaling pathways
(Gloeckner and Porras, 2020). While a wealth of data is available for affinity-based identification of
protein-protein interactions, which have a bias for rather stable complexes, there is an unmet need
for the dissection of less stable and transient PPIs. The rapid development of biotin-based proximity

labelling technologies has made it possible to explore previously inaccessible regions of the
interactome (Qin et al., 2021). Consequently, APEX and BioID have emerged as widely used techniques
for dissecting protein–protein interaction (PPI) networks (Li et al., 2017). For LRRK2, the first APEX2
dataset only recently became available recently (Bonet-Ponce et al., 2020). In the present study, we
systematically investigated the proximity interactome of LRRK2 using three different BioID tags—
BioID1, BioID2 and miniTurbo. The resulting data showed the expected overlap with previously
reported interactors, while also uncovering numerous novel candidates. However, several anticipated
interactors, including Rab proteins, were not detected in the initial BioID experiments. Nevertheless,
Rab8a was identified together with Rab29 upon co-expression of Rab29, which may reflect
conformational constraints that reduce accessibility for BioID-mediated biotinylation. Consistent with
these observations, a recent comparison of BioID and co-immunoprecipitation in the context of the
nuclear pore complex demonstrated that the two approaches provide functionally relevant yet
complementary interaction profiles (Moreira et al., 2023).

To identify direct LRRK2 interactors with high confidence from the proximity proteomes, we applied a
newly developed bioinformatic pipeline relying on co-evolution, co-expression and protein domain
architecture to predict interaction type (Miglionico et al., 2024). This pipeline allowed the identification
of a group of proteins within the proximity proteome showing a high degree of co-evolution with LRRK2
which is highly enriched in proteins associated with the centrosome. This is particularly interesting as
pathogenic LRRK2 has previously been shown to lead to a centrosome cohesion defect as well as to a
shortening of primary cilia (Dhekne et al., 2018; Khan et al., 2024; Lara Ordonez et al., 2022).
Furthermore, defects in primary cilia emerge as a new pathophysiological scheme connecting different
genes involved in mendelian forms of Parkinson's disease (Mohd Rafiq et al., 2024; Schmidt et al.,
2022). Strikingly, in the cluster of proteins with higher co-evolution to LRRK2 (Cluster 2), the single
protein showing the highest co-evolutionary association was the deubiquitinase CYLD. The K-63
selective deubiquitinase CYLD is a key effector in the biology of primary cilia. Furthermore, preliminary
functional data indicate that CYLD, when co-expressed, is able to increase expression levels of the
pathogenic variant LRRK2 Y1699C which shows a decreased protein stability. In fact, this DUB has been
shown to control apical docking of basal bodies in ciliated epithelial cells (Eguether et al., 2014). It also
regulates centrosomal satellites proteostasis by counteracting the E3 ubiquitin ligase MIB1 (Douanne
et al., 2019). Furthermore, CYLD has been shown to deubiquitinate Cep70 and to inactivate HDAC6
thereby mediating ciliogenesis (Yang et al., 2014). Interestingly, a missense mutation has recently been
shown to be a causative gene for FTD-ALS (Dobson-Stone et al., 2020). Most importantly, as the
deubiquitinase CYLD is part of the PINK1/Parkin pathway to regulate cellular levels of PARIS (ZNF746),
it is a critical regulator of mitochondrial biogenesis thereby acting as a negative regulator of dopamine
neuron survival in Parkinson's disease. In fact, CYLD inhibition has been shown to ameliorate

mitochondrial pathologies in cellular PD models which makes CYLD a promising target for therapeutic intervention for PD and other neurodegenerative diseases (Pirooznia et al., 2022). The E3 ligase Parkin preserves mitochondrial homeostasis via the PARIS/PGC-1a axis (Stevens et al., 2015; Zheng et al., 2017). Noteworthy, the PD risk variant LRRK2 G2385R has recently been shown to alter mitochondrial biogenesis via the PGC-1a pathway, leading to reduced PGC-1a levels similarly to a Parkin knock-out in DA neurons (Kumar et al., 2020; Xue et al., 2023). Parkin-deficient DA neurons show increased PARIS and decreased PGC-1a levels (Kumar et al., 2020). In conclusion, this indicates that LRRK2/CYLD signaling, at one hand, and Parkin signaling at the other, might converge on PARIS.

The second most co-evolved interactor found in Cluster 2 was the cilia-associated protein CEP85, which plays a role in centrosome disjunction (Chen et al., 2019). This is of particular interest as pathogenic LRRK2 variants lead to a centrosomal cohesion phenotype in patient-derived cells (Lara Ordonez et al., 2019). Notably, the cohesion phenotype is also seen in EBV-transformed lymphoblastoid cell lines from iPD patients which can be reverted by LRRK2 kinase inhibition (Naaldijk et al., 2024). Importantly, Cluster 2 also contains the autophagy cargo protein p62/sequestosome-1. This protein has previously been identified as LRRK2 interactor and substrate (Kalogeropulou et al., 2018; Park et al., 2016). SQSTM1 is also linked to PD pathophysiology via the Pink1/Parkin pathway (Shin et al., 2020). In addition, Cluster 2 points towards other pathways which have been previously linked to PD pathology. In particular, SNAP47 is part of the STX17-SNAP47-VAMP7/8 pathway, which has been shown to be involved in selective autophagy – a process not only essential for the homeostasis of primary cilia but also for mitophagy (Jian et al., 2024). Phospho-ubiquitin chains generated by PINK1/Parkin activity can recruit autophagy adaptors including the ubiquitin adaptor SQSTM1/p62 linking the PINK1/Parkin-dependent pathway to mitophagy (Geisler et al., 2010). All three proteins participate in the recruitment of the autophagy machinery to aged or dysfunctional mitochondria, which are eventually enclosed inside an autophagosome and subjected to the lysosome for degradation (Wilhelm et al., 2022).

Interestingly, also the kinesin motor protein and PX-domain containing protein KIF16B was among the top hits which is involved in vesicle trafficking and receptor recycling. It binds to PdIns(3)P by its unique PX domain and promotes the circulating cargo movement from early endosomes to the plasma membrane (Li et al., 2020). Interestingly, it has been found to be associated with lipid droplets (Bersuker et al., 2018). In turn, increased RAB8a phosphorylation by pathogenic LRRK2 variants, has been shown to promote lipid storage (Yu et al., 2018), indicating that the functional interaction of LRRK2 and KIF16B might contribute to alterations in the neuronal lipid homeostasis associated with PD pathology (Hallett et al., 2019).

Furthermore, conformational changes induced by MLi-2 binding—leading to rapid and nearly complete

dephosphorylation of LRRK2 in a disordered domain downstream of the hinge helix—positioned LRRK2

in proximity to a molecular network around PCM1, OFD1, and CEP131, strongly supporting a link

between LRRK2 and centriolar satellites. Strikingly, the proteins interacting with LRRK2 in presence of

MLi-2 displayed the highest degree of co-evolution, suggesting that the connection between LRRK2

and centriolar satellites is highly conserved. The conformational changes induced by binding of type I

inhibitors might result in a phase transition of LRRK2 into centriolar satellites preventing 14-3-3 binding

and thereby shifting the equilibrium towards the dephosphorylated protein. A similar mechanism has

been shown for CEP131, which is extracted from centriolar satellites by phosphorylation and

subsequent binding of 14-3-3 (Tollenaere et al., 2015). Centriolar satellites not only play an essential

role in the biogenesis of primary cilia (Hall et al., 2023) but also play a role in the degradation of

proteins by driving the assembly of aggresomes (Prosser et al., 2022). In fact, also LRRK2 has been

shown to interfere with aggresome formation preceding autophagic clearance (Bang et al., 2016).

Nevertheless, one caveat might be seen in the fact that LRRK2 inhibition with type I inhibitors induces

its destabilization and degradation (Lobbestael et al., 2016).

Structural prediction of LRRK2 binary complexes with AF revealed distinct interfaces within each

determined interactome. While several interactors preferentially engage with the catalytic core of the

protein, including the ROC and KIN domains, others are more likely to interact on terminal domains

(e.g. ARM or WD40). We predicted the interaction on the kinase domain for several interactors of the

BioID interactome as well as the one obtained in the presence of RAB29 over-expression, including the

known substrate RAB8A. The engagement in a substrate-like fashion with the catalytic core entails the

opening of the LRRK2 structure, with the detachment of the LRR and rearrangement of the entire N-

term domains. We previously showed that this conformation, that we similarly predicted for the

LRRK2:RAB10 complex, is compatible with distance restraints obtained by mass spectrometric mapping

of chemical crosslinking sites via XL-MS (Guaitoli et al., 2023). Moreover, if superimposed on the

active-like conformation of the most recent LRRK2 tetrameric structure, such predicted complexes fit

very well on the rest of the oligomer, with no clashes between the modelled LRRK2 chain and the rest

of the tetramer, further supporting the plausibility of such complexes (Guaitoli et al., 2023). We

speculate that partners engaging in this manner may represent putative substrates or function as

regulators of kinase domain activity. Strikingly, we rarely observed such interaction interface for

interactors enriched in presence of the MLi-2 inhibitor, which bind more frequently to the locked

conformation of LRRK2.

In conclusion, combining proximity proteomics with co-evolutionary and structural information helped

identifying distinct molecular networks linking LRRK2 to centrosomal biology and primary cilia. The

data may foster future functional analysis to further dissect these pathways to get a detailed picture

of the molecular pathophysiology underlying the cilia phenotypes caused by genetic variants in

different genes underlying familial forms of Parkinson's disease.

Limitations of the study

While the HEK293T system enables large-scale experimental throughput, it may not fully reflect the

neuronal context. Nonetheless, many neuronal proteins are known to be expressed in this cell line,

providing partial relevance. Another limitation is the low endogenous expression of LRRK2 in HEK293T

cells. As a result, the study relies on the overexpression of LRRK2 BioID fusion proteins, which may

introduce artifacts not present under physiological conditions.

Data availability

The mass spectrometry proteomics data have been deposited to the ProteomeXchange Consortium

via the PRIDE (Perez-Riverol et al., 2022) partner repository with the dataset identifier PXD048832 and

https://doi.org/10.6019/PXD048832 (BioID1, BioID2, miniTurbo), PXD048806 and

https://doi.org/10.6019/PXD048806 (LRRK2 proximity-proteomes MLi-2 vs. GZD-524) as well as

PXD048808 and https://doi.org/10.6019/PXD048808 (RAB29 co-expression). The data will be released

with the final publication of the peer-reviewed manuscript. The BioID interaction network and

associated co-evolutionary and structural analysis can accessed at the following URL:

(https://dip.bioinfolab.sns.it/explore/). The code used for the analysis is available at the following

links: https://github.com/raimondilab/distogram_analysis ,

https://github.com/raimondilab/Stable_complex_predictor. All data presented in the preprint version

of the manuscript, including MS RAW files, can be obtained upon request. A point-to-point protocol

for the BioID approach was deposited on Zenodo (DOI: 10.5281/zenodo.17045153).

Author contributions

M.E. and F.I. performed BioID experiments. M.E., F.I., P.M., F.R. and C.J.G. analyzed the primary BioID data. P.M. and F.R. performed the post-analysis of primary BioID data. P.M., N.D.R. and F.R. wrote the code for the bioinformatic workflow. C.J.G., F.R. and M.U. designed the study. M.E., F.I., P.M., F.R. and C.J.G. wrote the manuscript.

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1116

Tables:

Accession	Gene name	Protein name	Jaccard score
Q5S007	LRRK2	Leucine-rich repeat serine/threonine-protein kinase 2 (bait)	1.000000
Q9NQC7	**CYLD**	**Ubiquitin carboxyl-terminal hydrolase CYLD**	**0.848485**
Q6P2H3	CEP85	Centrosomal protein of 85 kDa	0.846626
Q96JH7	VCPIP1	Deubiquitinating protein VCPIP1	0.834286
Q96L93	KIF16B	Kinesin-like protein KIF16B	0.832335
P78312	FAM193A	Protein FAM193A	0.817647
Q8NEY1	NAV1	Neuron navigator 1	0.817073
Q7Z2T5	TRMT1L	TRMT1-like protein	0.815951
Q5SZL2	CEP85L	Centrosomal protein of 85 kDa-like	0.810127
O95684	CEP43	Centrosomal protein 43	0.809816
Q96RS0	TGS1	Trimethylguanosine synthase	0.798742

Table 1: 10 IDs with the highest Jaccard score underlying identified "Cluster 2". The DUB CYLD (bold) shows highest co-evolution to LRRK2.

1121

Accession	Protein name	Gene name	p-value	Log2 T-test difference
P81605	Dermcidin	DCD	2.91844	3.3
Q9UPN4	Centrosomal protein of 131 kDa	CEP131	4.15494	2.2
Q49A88	Coiled-coil domain-containing protein 14	CCDC14	5.39373	2.1
Q9Y2D8	Afadin- and alpha-actinin-binding protein	SSX2IP	2.98866	2.1
O75665	Oral-facial-digital syndrome 1 protein	OFD1	7.92350	2.0
Q15154	Pericentriolar material 1 protein	PCM1	6.62987	1.8
Q9UHB6	LIM domain and actin-binding protein 1	LIMA1	2.96451	1.7
Q8WXW3	Progesterone-induced-blocking factor 1	PIBF1	4.62801	1.5
Q6WCQ1	Myosin phosphatase Rho-interacting protein	MPRIP	7.40967	1.5
Q14596	Next to BRCA1 gene 1 protein	NBR1	3.34347	1.3
Q5TB80	Centrosomal protein of 162 kDa	CEP162	3.46643	1.3
Q8WWI1	LIM domain only protein 7	LMO7	5.84664	1.3

1122

1123 **Table 2:** Enriched proteins in the LRRK2 proximity proteome upon MLi-2 treatment (FDR-controlled T-

1124 test).

Accession	Protein name	Gene name	p-value (-log10)	Log2 T-test difference
O14966	Ras-related protein Rab-7L1	RAB29	9,72545	10,5
Q8IZA0	Dyslexia-associated protein KIAA0319-like protein	KIAA0319L	5,00209	4,9
Q6WKZ4	Rab11 family-interacting protein 1	RAB11FIP1	4,56815	4,6
O15126	Secretory carrier-associated membrane protein 1	SCAMP1	4,09385	3,5
Q9UIQ6	Leucyl-cystinyl aminopeptidase;Leucyl-cystinyl aminopeptidase, pregnancy serum form	LNPEP	4,33552	3,4
Q9H2J7	Sodium-dependent neutral amino acid transporter B(0)AT2	SLC6A15	5,06972	2,9
Q96QD8	Sodium-coupled neutral amino acid transporter 2	SLC38A2	2,89271	2,9
Q13586	Stromal interaction molecule 1	STIM1	3,12115	2,8
P98194	Calcium-transporting ATPase type 2C member 1	ATP2C1	5,00347	2,6
Q9H2H9	Sodium-coupled neutral amino acid transporter 1	SLC38A1	2,37723	2,6
O75976	Carboxypeptidase D	CPD	3,20952	2,6
Q14126	Desmoglein-2	DSG2	4,31421	2,5
P51809	Vesicle-associated membrane protein 7	VAMP7	4,02209	2,5
Q8NDI1	EH domain-binding protein 1	EHBP1	2,37346	2,5
Q8TBA6	Golgin subfamily A member 5	GOLGA5	3,0259	2,4
Q9BXF6	Rab11 family-interacting protein 5	RAB11FIP5	2,9101	2,4
Q7L804	Rab11 family-interacting protein 2	RAB11FIP2	3,33154	2,2
Q6IAA8	Ragulator complex protein LAMTOR1	LAMTOR1	3,23604	2,1
P61006	Ras-related protein Rab-8A	RAB8A	3,56486	2,1
P11717	Cation-independent mannose-6-phosphate receptor	IGF2R	5,71543	2,0
Q13190	Syntaxin-5	STX5	3,88884	2,0
Q15836;P63027	Vesicle-associated membrane protein 3;Vesicle-associated membrane protein 2	VAMP3;VAMP2	3,67019	2,0
Q92692	Nectin-2	PVRL2	3,06334	2,0
Q6NXT4	Zinc transporter 6	SLC30A6	3,47208	1,8
Q96A65	Exocyst complex component 4	EXOC4	2,51947	1,7
Q14141	Septin-6	Sept6	3,87782	1,7
Q04656	Copper-transporting ATPase 1	ATP7A	2,77415	1,7
Q7Z2W4	Zinc finger CCCH-type antiviral protein 1	ZC3HAV1	2,62567	1,6
P78310	Coxsackievirus and adenovirus receptor	CXADR	2,87956	1,4
P08195	4F2 cell-surface antigen heavy chain	SLC3A2	2,43497	1,4
P02786	Transferrin receptor protein 1;Transferrin receptor protein 1, serum form	TFRC	3,21618	1,3
Q07960	Rho GTPase-activating protein 1	ARHGAP1	2,65929	1,3
O60271	C-Jun-amino-terminal kinase-interacting protein 4	SPAG9	3,80826	1,3
P42167	Lamina-associated polypeptide 2, isoforms beta/gamma;Thymopoietin;Thymopentin	TMPO	3,54449	1,2
Q9BSJ8	Extended synaptotagmin-1	ESYT1	3,59064	1,0
Q14739	Lamin-B receptor	LBR	3,12029	0,8

1125

1126 **Table 3:** Enriched proteins in the RAB29-depended LRRK2 proximity proteome (FDR-controlled T-test).

1127

  Figures

Figure 1: (A) Workflow of the BioID analysis and post processing of the data. (B) Overlap between the different BioID approaches. Enriched proteins were determined by relative label-free quantification (combined with an FDR-controlled T-test, p=0.05) of BioID LRRK2 vs the corresponding tag-only control: BioID1-LRRK/BioID1-tag: N=8, BioID2-LRRK2/BioID2: N=9, miniTurbo-LRRK2/miniTurbo tag: N=8.

Figure 2: LRRK2 interactome functional enrichment. A) the LRRK2 BioID interaction network was mapped on known PPIs from IntAct via Cytoscape. The edges' color and thickness for mapped interactions are proportional to IntAct MI score, else they are colored in gray with constant thickness. Modules within the network have been identified through the Glay method of the Autoannotate app and nodes colored accordingly. Labels were calculated using the 'Adjacent Words' setting of the WordCloud algorithm, based on significantly enriched (FDR < 0.01) Gene Ontology (GO) Biological Processes and Cellular Component terms associated with each gene in the network ; B) Venn diagram showing the overlap between the BioID interaction sets and known LRRK2 interactions in IntAct;

1144   barplots showing the top 15 most significant terms obtained through String's functional enrichment of

1145   C) Compartments, D) GO Biological Process and D) Interpro domains.

Figure 3: Co-evolution analysis of LRRK2 interactome. (A) Distribution of co-evolution metric (i.e. Jaccard score) among the different LRRK2 interaction sets, including IntAct interaction types and BIOID. P-values have been computed with the Wilcoxon rank-sum, two-sided test with Bonferroni correction (* $P < 0.05$, ** $P < 0.001$, *** $P < 0.001$, **** $P < 0.0001$). Violin plots show the median as the white central dot and first and third quartiles as bounds of the thicker black line (box); the whiskers extend to the last data point within 1.5 times the interquartile range (IQR)

from the box's boundaries. (B) predictiveness of LRRK2 interaction type via co-evolution (C)

Heatmap showing the hierarchical clustering based on the degree of co-evolution (Jaccard score) of

the LRRK2 proximity proteome. (D) String network of the proteins most co-evolved with LRRK2

participating to cluster 2. Nodes are colored in gray or light-gray according to modules membership

determined with the MCL approach available from the Cytoscape clusterMaker app. Node rims are

colored according to enriched processes. (E) Dot plot showing top 15 significantly enriched (FDR < 0.01)

categories from the string network of cluster 2. Values on the x axis display the ratio of genes of the

term participating to the network over the total number of genes in that category. The ratio of the dot

is proportional to the number of genes of that category in the network, while the node color is the

darker the more significant is the category enrichment. Only categories with at least 5 genes have been

displayed. Representative, enriched categories have been colored with the same color scheme as in

panel (D). (F) Validation of CYLD as part of the LRRK2 proximity proteome. Due to the lack of suitable

antibodies for detecting endogenous CYLD, a FLAG-HA–tagged CYLD construct was recombinantly

expressed for this experiment, in contrast to the mass spectrometry–based approach, which relied on

endogenous protein levels. (G, H) CYLD co-expression enhances LRRK2 protein stability in HEK293T

cells. HEK293T cells were transfected with either LRRK2 wild-type (WT) or the Y1699C variant, along

with HA-tagged ubiquitin constructs: wild-type ubiquitin (WT-Ub) or a K63-only ubiquitin mutant (Ub-

K63), which permits conjugation exclusively at lysine 63. CYLD or an empty vector control (pcDNA3.0)

was co-expressed in each condition. Protein levels were assessed by capillary-based Western blotting

with fluorescence detection. (G) Reconstructed Western blot signals. (H) Quantification of LRRK2

protein levels from three technical replicates (same biological samples analyzed in three independent

capillary Western runs). Statistical significance was assessed using unpaired two-tailed t-tests.

Figure 4: 3D complexes prediction and analysis of LRRK2 interactome (A) workflow of the structural

bioinformatics analysis of LRRK2 interactome; (B) on the left the heatmap showing the hierarchical

clustering based on pairwise RMSD calculation of LRRK2's chain in the different complexes. On the

right two representative conformations of LRRK2 in the "locked" and "unlocked" conformation to mark

the two main clusters; (C) cartoon representation of "locked" (tomato) and "unlocked" (light green)

superimposed LRRK2 conformations; (D) violin plot of the distribution of RMSD calculated with respect

to the reference LRRK2 experimental structure (PDB 7LHW) for the locked and unlocked clusters. Color

scheme as in (C); (E) distribution of the Jaccard score for the locked and unlocked clusters. Color

scheme as in (C). P-values for violin plots in D and E have been computed with the Wilcoxon

rank-sum, two-sided test (* $P < 0.05$, ** $P < 0.001$, *** $P < 0.001$, **** $P < 0.0001$). Violin plots

show the median as the white central dot and first and third quartiles as bounds of the thicker

black line (box); the whiskers extend to the last data point within 1.5 times the interquartile

range (IQR) from the box's boundaries; (F) Venn diagram of the enriched terms in the locked and

unlocked clusters. Color scheme as in (C); (E) Top 10 exclusively enriched (FDR<0.01) terms in the

locked and unlocked clusters.

Figure 5: Interface analysis of LRRK2's 3D interactome (A) interface fingerprint of the LRRK2 complexes

via distogram calculation. Cells contain the minimum expected distance obtained from the distograms

of any residue of each interactor (row) to every residue in LRRK2 (column). Rows are color annotated

based on confidence score, Jaccard and RMSD from reference LRRK2 structure (PDB: 7LHW); (B)

representative structure of the Core cluster (i.e. LRRK2-ANAPC5 complex); (C) representative

structures of the WD40 cluster; (D) representative structures of the ARM binding mode cluster (i.e.

LRRK2- DCUN1D4 complex); distribution among different interface cluster of the (E) confidence score,

1199   (F) RMSD and (G) Jaccard score descriptors. P-values have been computed with the Wilcoxon

1200   rank-sum, two-sided test with Bonferroni correction (* $P < 0.05$, ** $P < 0.001$, *** $P < 0.001$,

1201   **** $P < 0.0001$). Violin plots show the median as the white central dot and first and third

1202   quartiles as bounds of the thicker black line (box); the whiskers extend to the last data point

1203   within 1.5 times the interquartile range (IQR) from the box's boundaries.

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Figure 6: Different effects of type I and type II inhibitors and RAB29 on the LRRK2 proximity proteome.
(A) MLi-2 vs DMSO. (B) Stringified network of the MLi-2 interactome from (A) mapped to IntAct known
interactions. Edges' color and thickness for mapped interactions are proportional to IntAct MI score,
else they are colored in gray with constant thickness. Node rims are colored according to enriched
processes; (C) Dot plot showing top 15 significantly enriched (FDR < 0.01) categories from the LRRK2-
MLi-2 interactome. Values on the x axis display the ratio of genes of the term participating to the
network over the total number of genes in that category. The ratio of the dot is proportional to the
number of genes of that category in the network, while the node color is the darker the more
significant is the category enrichment. Only categories with at least 5 genes have been displayed.
Representative, enriched categories have been colored with the same color scheme as in panel C. (D)
Scheme highlighting the function of centriolar satellites and the core protein PCM1 in ciliation ((Hori
and Toda, 2017), modified) (E) BioID analysis following PCM1 knockdown and MLi-2 treatment. Shown
are eluates obtained after streptavidin-based enrichment of biotinylated proteins. (F) Violin plot
reporting the Jaccard Score distribution in different data sets. P-values have been computed with
the Wilcoxon rank-sum, two-sided test with Bonferroni correction (* $P < 0.05$). Violin plots

show the median as the white central dot and first and third quartiles as bounds of the thicker

black line (box); the whiskers extend to the last data point within 1.5 times the interquartile

range (IQR) from the box's boundaries; (G) mt-LRRK2 proximity proteome in dependence of RAB29

co-expression: RAB29 redirects LRRK2 to lysosomal PPI networks; (H) Stringified network of the RAB29

over-expression interactome from (G) mapped to IntAct known interactions. Edges' color and thickness

for mapped interactions are proportional to IntAct MI score, else they are colored in gray with constant

thickness. Node rims are colored according to enriched processes; (I) Dot plot showing top 15

significantly enriched (FDR < 0.01) categories from the LRRK2-RAB29 over-expression interactome.

Values on the x axis display the ratio of genes of the term participating to the network over the total

number of genes in that category. The ratio of the dot is proportional to the number of genes of that

category in the network, while the node color is the darker the more significant is the category

enrichment. Only categories with at least 5 genes have been displayed. Representative, enriched

categories have been colored with the same color scheme as in panel H.

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Figure 7: AF-structural modeling of MLi-2 and Rab29 interactomes. (A) Hierarchical clustering of pairwise RMSD of LRRK2 chain in the different complexes; B) LRRK2-RAB8A predicted complex; (C) LRRK2-RAB29 predicted complex; (D) violin plot of the distribution of RMSD calculated with respect to the reference LRRK2 experimental structure (PDB 7LHW) for the locked and unlocked clusters. Color scheme as in 4C; (E) distribution of the Jaccard score for the locked and unlocked clusters. Color scheme as in 4C. P-values in violin plots in (D) and (E) have been computed with the Wilcoxon rank-sum, two-sided test (* $P < 0.05$, ** $P < 0.001$, *** $P < 0.001$, **** $P < 0.0001$). Violin plots show the median as the white central dot and first and third quartiles as bounds of the thicker black line (box); the whiskers extend to the last data point within 1.5 times the interquartile range (IQR) from the box's boundaries; (F) Venn diagram of the enriched terms in the locked and unlocked clusters. Color scheme as in 4C; (G) Top exclusively enriched (FDR<0.01) terms in the locked and unlocked clusters. Color scheme as in 4C.

Supplemental Information

Supplemental Tables

Table S1: LRRK2 proximity proteomes obtained with BioID1, BioID2 and miniTurbo as well as the

Rab29-dependent LRRK2-mT and the inhibitor-dependent LRRK2-mT proximity proteomes.

Table S2: (A) Co-evolutionary and structural features of BioID interactors. (B) Over-representation

analysis of protein clusters defined via co-evolution. (C) Over-representation analysis of protein

clusters defined via RMSD clustering of AF-multimer predicted complexes. (D) Over-representation

analysis of protein clusters defined via Interaction fingerprint clustering of AF-multimer predicted

complexes. (E) Over-representation analysis of protein clusters defined via RMSD clustering of AF-

multimer predicted complexes of the RAB29-overexpression and MLi-2 interactomes.

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Supplemental Figure 1: Western blot, pRab10, pS935 and biotinylation – comparison BioID1, BioID2,

miniTurbo. To ensure functional BioID-fusion proteins, LRRK2 phosphorylation at S935 and Rab10

phosphorylation was assessed. (A) The different BioID LRRK2 constructs show sufficient biotinylation

of proteins as well as MLi-2-responsive Rab10 phosphorylation. (B) All constructs are phosphorylated

at S935 (top panel). Replication of LRRK2-mediated Rab10 phosphorylation (last panel) demonstrating

functional BioID fusion proteins.

Mutual Information ROC curve

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1264 **Supplemental Figure 2:** ROC Curve of prediction with Mutual Information (MI) of LRRK2's IntAct
1265 interaction types.

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Supplemental Figure 3: heatmap with cluster specific, significantly enriched processes (FDR < 0.01; StringDB) obtained from co-evolution (Jaccard)-based clustering of the LRRK2 BioID interactome.

Interaction type

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1271   **Supplemental Figure 4:** heatmap displaying the coevolution (Jaccard)-based hierarchical clustering of

1272   the LRRK2's IntAct interactome. Colored bars provide IntAct's interaction type annotation.

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Supplemental Figure 5: Effect of CYLD co-expression on the stability of LRRK2 WT and Y1699C variants. (A) Western blot analysis (ECL detection) of two independent biological replicates. HEK293T cells were transfected with either LRRK2 wild-type (WT) or the Y1699C mutant, co-expressed with CYLD and either a 6xUbiquitin (6xUb) construct or a K63-only ubiquitin mutant (Ub-K63), which restricts polyubiquitin chain formation to K63 linkages. (B, C) Quantification of LRRK2 protein levels from three technical replicates of the second biological replicate (same biological samples analyzed in three independent capillary Western runs), confirming the trend observed in the first replicate (shown in Figure 3). Statistical significance was assessed using unpaired two-tailed t-tests.

A

B

BioID: MLi-2 vs GZD824

C

PCM1 knock-down/ Lysate controls

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Supplemental Figure 6: (A) Western blot analysis demonstrating effective LRRK2 inhibition by MLi-2

(type I inhibitor) and GZD-824 (type II inhibitor) in HEK293 cells stably expressing miniTurbo-tagged

LRRK2. Both compounds at 1 µM significantly reduce Rab10 phosphorylation. As expected, MLi-2

induces a marked loss of LRRK2 pS935 phosphorylation, a hallmark of type I kinase inhibition. In

contrast, GZD-824 does not affect pS935 phosphorylation, consistent with its classification as a type II

inhibitor(Tasegian et al., 2021). (B) BioID-based proximity labeling comparing the interactome changes

upon treatment with MLi-2 versus GZD-824. Volcano plot illustrates differential protein interactions,

highlighting the distinct molecular effects of type I and type II LRRK2 inhibition. (C) Validation of PCM1

knock-down by western blot. Efficient depletion of PCM1 is confirmed, with no observed changes in

total protein levels of LRRK2, Rab10, or SSX2IP. LRRK2 inhibition by MLi-2 remains effective in the

context of PCM1 knock-down.

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Supplemental Figure 7: (A) interaction fingerprint clustering of LRRK2's modulators (MLi-2 and RAB29) interactomes. Cells contain the minimum expected distance obtained from the distograms of any residue of each interactor (row) to every residue in LRRK2 (column). Rows are color annotated based on confidence score, Jaccard, RMSD from reference LRRK2 structure (PDB: 7LHW) and experiment type (either MLi-2 or RAB29 over-expression); (B) interaction fingerprint of LRRK2 and interacting RABs (from IntAct).