# Panorama of ancient metazoan macromolecular complexes 

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#### Abstract

Macromolecular complexes are essential to conserved biological processes, but their prevalence across animals is unclear. By combining extensive biochemical fractionation with quantitative mass spectrometry, here we directly examined the composition of soluble multiprotein complexes among diverse metazoan models. Using an integrative approach, we generated a draft conservation map consisting of more than one million putative high-confidence co-complex interactions for species with fully sequenced genomes that encompasses functional modules present broadly across all extant animals. Clustering reveals a spectrum of conservation, ranging from ancient eukaryotic assemblies that have probably served cellular housekeeping roles for at least one billion years, ancestral complexes that have accrued contemporary components, and rarer metazoan innovations linked to multicellularity. We validated these projections by independent co-fractionation experiments in evolutionarily distant species, affinity purification and functional analyses. The comprehensiveness, centrality and modularity of these reconstructed interactomes reflect their fundamental mechanistic importance and adaptive value to animal cell systems.


## Introduction

Elucidating the components, conservation and functions of multiprotein complexes is essential to understand cellular processes ${ }^{1,2}$, but mapping physical association networks on a proteome-wide scale is challenging. The development of high-throughput methods for systematically determining protein-protein interactions (PPIs) has led to global molecular interaction maps for model organisms including E. coli, yeast, worm, fly and human ${ }^{3-10}$. In turn, comparative analyses have shown that PPI networks tend to be conserved ${ }^{11,12}$, evolve more slowly than regulatory networks ${ }^{13}$, and closely mirror function retention across orthologous groups ${ }^{111,14,15}$. Yet fundamental questions arise ${ }^{16,17}$. Here we define: (i) the extent to which physical interactions are preserved between phyla; (ii) the identity of protein complexes that are evolutionarily stable across animals; and (iii) the unique attributes of macromolecule composition, phylogenetic distribution and phenotypic significance.

## Generating a high-quality conserved interaction dataset

As previous cross-species interactome comparisons, based on experimental data from different sources and methods, show limited overlap ${ }^{12,18}$, we sought to produce a more comprehensive and accurate map of protein complexes common to metazoa by applying a standardized approach to multiple species. We employed biochemical fractionation of native macromolecular assemblies followed by tandem mass spectrometry to elucidate protein complex membership (Fig. 1; see Supplementary Methods). Previous application of this co-fractionation strategy to human cell lines preferentially identified vertebrate-specific protein complexes ${ }^{6}$, so we selected eight additional species for study on the basis of their relevance as model
organisms, spanning roughly a billion years of evolutionary divergence (Fig. 1a). The resulting co-fractionation data (Fig. 1b) acquired for Caenorhabditis elegans (worm), Drosophila melanogaster (fly), Mus musculus (mouse), Strongylocentrotus purpuratus (sea urchin), and human were used to discover conserved interactions (Fig. 1c), while the data obtained for Xenopus laevis (frog), Nematostella vectensis (sea anemone), Dictyostelium discoideum (amoeba) and Saccharomyces cerevisiae (yeast) were used for independent validation. Details on the cell types, developmental stages and fractionation procedures used are provided in Supplementary Table 1.

We identified and quantified (see Supplementary Methods) 13,386 protein orthologues across 6,387 fractions obtained from 69 different experiments (Fig. 2a), an order of magnitude expansion in data coverage relative to our original (H. sapiens only) study ${ }^{6}$. Individual pair-wise protein associations were scored based on the fractionation profile similarity measured in each species. Next, we used an integrative computational scoring procedure (Fig. 1c; see Supplementary Methods) to derive conserved interactions for human proteins and their orthologues in worm, fly, mouse and sea urchin, defined as high pair-wise protein co-fractionation in at least two of the five input species. The support vector machine learning classifier used was trained (using fivefold cross-validation) on correlation scores obtained for conserved reference annotated protein complexes (see Supplementary Methods), and combined all of the input species co-fractionation data together with previously published human ${ }^{6,19}$ and fly interactions ${ }^{5}$ and additional supporting functional association evidence ${ }^{20}$ (HumanNet). Measurements of overall performance showed high precision with reasonable recall by the co-fractionation data alone (Fig. 2b), with external data sets serving only to increase

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Figure 1 Workflow. a, Phylogenetic relationships of organisms analysed in this study. We fractionated soluble protein complexes from worm (C. elegans) larvae, fly (D. melanogaster) S2 cells, mouse (M. musculus) embryonic stem cells, sea urchin (S. purpuratus) eggs and human (HEK293/HeLa) cell lines. Holdout species ('T', for test) likewise analysed were frog (X. laevis), an amphibian; sea anemone ( $N$. vectensis), a cnidarian with primitive eumetazoan tissue organization; slime mould (D. discoideum), an amoeba; and yeast (S. cerevisiae), a unicellular eukaryote. b, Protein fractions were digested and
analysed by high-performance liquid chromatography tandem mass spectrometry (LC-MS/MS), measuring peptide spectral counts and precursor ion intensities. c, Integrative computational analysis. After orthologue mapping to human, correlation scores of co-eluting protein pairs detected in each 'input' species were subjected to machine learning together with additional external association evidence, using the CORUM complex database as a reference standard for training. High-confidence interactions were clustered to define co-complex membership.
precision and recall as we required all derived interactions to have extensive biochemical support (see Supplementary Methods). Co-fractionation data of each input species affected overall performance, in each case increasing precision and recall (Extended Data Fig. 1a). The final filtered interaction network consists of 16,655 high-confidence co-complex interactions in human (Supplementary Table 2). All of the interactions were supported by direct biochemical evidence in at least two input species, with half $(8,121)$ detected in three or more (Extended Data Fig. 1b), enabling cross-species modelling and functional inference.

## Benchmarking protein complexes

Multiple lines of evidence support the quality of the network: reference complexes withheld during training were reconstructed with higher precision and recall (Fig. 2b; see Extended Data Fig. 1c) relative to our human-only map ${ }^{6}$. The interacting proteins were also sixfold enriched (hypergeometric $P<1 \times 10^{-24}$ ) for shared subcellular localization annotations in the Human Protein Atlas Database ${ }^{21}$, 21-fold enriched $\left(P<1 \times 10^{-56}\right)$ for shared disease associations in $\mathrm{OMIM}^{22}$, and showed highly correlated human tissue proteome abundance profiles ${ }^{23}$ (Extended Data Fig. 2a).
To independently verify the reliability of these projections, we examined the co-fractionation profiles of putatively interacting orthologues (interologues) in the four holdout species, as obtained by protein quantification across 1,127 biochemical fractions (see Supplementary Methods). Whereas sequence divergence changed absolute chromatographic retention times (Extended Data Fig. 2b), most of the predicted interactors showed highly correlated co-fractionation profiles among the holdout test species to a degree comparable to those of the input species used for learning (Fig. 2c). The biochemical data obtained for frog and sea anemone showed slightly better agreement than that for Dictyostelium and yeast that was proportional to evolutionary distance ${ }^{24}$.
Besides indicating stably associated proteins, our multispecies biochemical profiles faithfully recapitulated the architecture of
multiprotein complexes of known three-dimensional structure, with a general trend for most correlated protein pairs to be spatially closer (Extended Data Fig. 2c). For example, hierarchical clustering of 30 S proteasome subunits according to chromatographic elution profiles of all five input species correctly separated the 20 S and 19S particles and the regulatory lid from the base sub-complex (Fig. 2d), reflecting known hierarchies of complex formation and disassembly.

## Landscape of interaction conservation across species

Because most of the interacting components were phylogenetically conserved across vast evolutionary timescales, we were able to predict over one million high-confidence co-complex interactions among orthologous protein pairs for 122 extant eukaryotes with sequenced genomes (Supplementary Table 3). The number of interactions ranged from $\sim 8,000$ to $\sim 15,000$ per species depending on phyla (Fig. 2e), with more projected among Deuterostomes, Protostomes and Cnidaria, which show high component retention, and fewer in Fungi, Plants and, especially, Protists, where the relative paucity of co-complex conservation probably reflects inherent clade diversity, especially in parasite genomes (for example, gene loss among Apicomplexa). While largely congruent with previous smaller-scale studies of PPI conservation ${ }^{25}$, the majority of conserved co-complex interactions are novel (less than one-third curated in CORUM, STRING and GeneMANIA databases; Fig. 2e). This markedly increases the number of metazoan protein interactions reported to date (Supplementary Table 3), covering roughly $10 \%-25 \%$ of the estimated conserved animal cell interactome ${ }^{26,27}$, opening up many new avenues of inquiry.

To systematically define evolutionarily conserved functional modules, we partitioned the interaction network using a two-stage clustering procedure (Fig. 1c; see Supplementary Methods) that allowed proteins to participate in multiple complexes (that is, moonlighting) as merited (Extended Data Fig. 3a). The 981 putative multiprotein groupings (Fig. 3a; see Supplementary Table 4) include both


Figure $2 \mid$ Derivation and projection of protein co-complex associations across taxa. a, Expanded coverage via experimental scale-up relative to our previous human study ${ }^{6}$. Chart shows number of proteins detected, most ( $63 \%$ ) in two or more species. b, Performance benchmarks, measuring precision and recall of our method and data in identifying known co-complex interactions (annotated human complexes from CORUM ${ }^{39}$ ). Complexes were split into training and withheld test sets; fivefold cross-validation against 4,528 interactions derived from the withheld test set shows strong performance gains, beyond baselines achieved using only co-fractionation or external evidence alone. TP, true positive; FP, false positive; FN, false negative. c, Plots showing high enrichment (probability ratio of interacting) of predicted interacting orthologous protein pairs (relative to non-interacting pairs) among highly
correlated fractionation profiles, in both the holdout validation (test, T) and input species (colours reflect clade memberships). d, Left, representative co-fractionation data (normalized spectral counts shown for portions of 3 of 42 experimental profiles) from human, fly and sea urchin showing characteristic profiles of proteasome core, base and lid sub-complexes. Hierarchical clustering (right) of pan-species pairwise Pearson correlation scores (centre) is consistent with accepted structural models (Protein Data Bank ID: 4CR2; core, red; base, blue; lid, green; out-clusters, white). e, Projection of conserved co-complex interactions across 122 eukaryotic species, indicating overlap with leading public PPI reference databases ${ }^{39-41}$. STRING bars indicate excess over CORUM; GeneMANIA bars indicate excess over both; component and interaction occurrences across clades indicated at bottom.
many well-known and novel complexes linked to diverse biological processes (Extended Data Fig. 3b). The complexes have estimated component ages spanning from $\sim 500$ million (metazoan-specific, or 'new') to over one billion years (ancient, or 'old') of evolutionary divergence. Details of species, orthologues, taxonomic groups, protein ages and evolutionary distances are provided in Supplementary Tables 3 and 5 and Supplementary Methods.

Although proteins arising in metazoa (by gene duplication or other means) account for about three quarters of all human gene products,
they form only about a third (39\%; 147) of the clusters (Fig. 3a). These 'new' complexes tend to be smaller ( $\leq 3$ components; Fig. 3b) and specific (components not present in 'mixed' complexes). This indicates that although protein number and diversity greatly increased with the rise of animals ${ }^{25}$, most stable protein complexes were inherited from the unicellular ancestor and subsequently modified slightly over time (Fig. 3c and Supplementary Table 5). Indeed, the dominant phylogenetic profile of complexes across Eukarya (Fig. 3d) is composed either entirely (344 old complexes) or predominantly (490


Figure $3 \mid$ Prevalence of conservation of protein complexes across Metazoa and beyond. a, Conserved multiprotein complexes, identified by clustering, arranged according to average estimated component age (see Supplementary Methods and ref. 25). Proteins (nodes) classified as metazoan (green) or ancient (orange); assemblies showing divergent phylogenetic trajectories termed 'mixed'. b, Example complexes with different proportions of old and new subunits. c, Presumed origins of metazoan (new), mixed and old complexes; '?' indicates variable origins of new genes. d, Heat map showing prevalence of selected complexes across phyla. Colour reflects fraction of components with detectable orthologues (absence, dark blue). Sea anemone ( $N$. vectensis) is the most distant metazoan (cnidarian) analysed biochemically.
mixed complexes) of ancient subunits ubiquitous among eukaryotes (Extended Data Fig. 4a; see Supplementary Table 5 for details), the latter presumably reflecting preferential accretion of additional components to pre-existing macromolecules (Fig. 3c) ${ }^{28}$.

These primordial complexes are present throughout the Opisthokonta supergroup (animals and fungi), estimated to be more than one billion years old ${ }^{29}$, and plants (and presumably lost/significantly diverged among parasitic protists). Reflecting this central importance, these complexes tend strongly to be ubiquitously expressed throughout all cell types and tissues (Extended Data Fig. 5a), are abundant (Extended Data Fig. 5b), and are enriched for associations to human disease and perturbation phenotypes in $C$. elegans (Supplementary Table 6). In comparison with other proteins in the 16,655 interactions, the older, conserved proteins present in these stable complexes have lower average domain complexity
( $P<0.02$; see Supplementary Methods), suggesting multi-domain architectures underlie more transient or tissue-specific interactions. Whereas mixed and old complexes are enriched for functional associations with core cellular processes, such as metabolism (Extended Data Fig. 4c), the strictly metazoan complexes were far more likely to be linked to cell adhesion, organization and differentiation, consistent with roles in multicellularity. Reflecting these different evolutionary trajectories, new clusters are substantially more enriched for cancer-related proteins ( $42 \% ; 62 / 147$; hypergeometric $P \leq 1 \times 10^{-5}$ ) compared to strictly old ( $15 \% ; 53 / 344 ; P \leq 1 \times 10^{-3}$ ) clusters ( $Z$-test $<0.0001$ ) (Supplementary Table 7), have generally lower annotation rates (Extended Data Fig. 4b), and show different preponderances of protein domains (Extended Data Fig. 4c and Supplementary Table 6).

## Independent biological assessment

We used multiple approaches to assess the accuracy (Fig. 4) and functional significance (Fig. 5) of the predicted complexes. First, we performed affinity purification mass spectrometry (AP/MS) experiments on select novel complexes from the new, old and mixed age clusters, validating most associations in both worm and human (Fig. 4a and Extended Data Fig. 6a). We next performed a global validation by comparing our derived complexes to a newly reported large-scale AP/MS study of 23,756 putative human protein interactions detected in cell culture (E. L. Huttlin et al., BioGRID preprint 166968), and observed a partial, but highly statistically significant, overlap to a degree comparable to literature-derived complexes (Fig. 4b, Extended Data Fig. 6b).
We also observed broad agreement between the derived complexes' inferred molecular weights (assuming 1:1 stoichiometries) and migration by size-exclusion chromatography (Fig. 4c and Extended Data Fig. 7a) and density gradient centrifugation (Extended Data Fig. 7b). A prime example is the coherent profiles of a large $(\sim 500 \mathrm{kDa})$ mixed complex with several un-annotated components (Fig. 4d and Extended Data Fig. 8), dubbed 'Commander', because most subunits share COMM (copper metabolism MURR1) domains ${ }^{30}$ implicated in copper toxicosis ${ }^{31}$, among other roles ${ }^{30,32}$. Commander contains coiled-coil domain proteins CCDC22 and CCDC93 (Figs 4a, d) in addition to ten COMM domain proteins, broadly supported by co-fractionation in human, fly and sea urchin (Extended Data Fig. 9a-c and supporting website, http://metazoa.med.utoronto.ca/ php/view_elution_image.php?id=71\&cond $=\mathrm{ms} 2$ ).

We found an unexpected role in embryonic development for Commander, whose subunits are strongly co-expressed in developing frog (Extended Data Fig. 9d, e). COMMD2/3-knockdown (morpholino) tadpoles showed impaired head and eye development (Fig. 5a and Extended Data Fig. 9f, h), and defective neural patterning and expression changes in brain markers PAX6, EN2 and KROX20/EGR1 (Fig. 5b and Extended Data Fig. 9g, h). Given the recently discovered link ${ }^{33,34}$ between CCDC22 and human syndromes of intellectual disability, malformed cerebellum and craniofacial abnormalities, the deep conservation of the Commander complex suggests COMMD2/3 as strong candidates in the aetiology of these heterogeneous disorders.

Among metazoan-specific protein complexes, we confirmed physical and functional associations of spindle checkpoint protein BUB3 with ZNF207, a zinc-finger protein conspicuously lacking orthologues in cnidarians and fungi. ZNF207 binds Bub3 via a Gle2-binding-sequence (GLEBS) motif ${ }^{35}$ restricted to deuterostomes and protostomes (Extended Data Fig. 10a). As in human, knockdown of the ZNF207 orthologue in C. elegans (B0035.1) enhanced lethality owing to impaired Bub3-mediated checkpoint arrest (Fig. 5c).
Among mixed complexes, we confirmed metazoan-specific coiled-coil domain protein CCDC97 as a sub-stoichiometric component of human and worm SF3B spliceosomal complex involved in branch-site recognition (Fig. 4a). Consistent with a possible role in


Figure $4 \mid$ Physical validation of complexes. a, Verification of complexes from tagged human cell lines and transgenic worms (see Supplementary Methods; complexes drawn as in Fig. 3). Inset reports spectral counts obtained in replicate AP/MS analyses of indicated bait protein (header). MIB2-VPS4 complex confirmed by co-immunoprecipitation (co-IP; Extended Data Fig. 6a). b, Conserved complexes significantly overlap large-scale AP/MS data reported for human cell lines (E. L. Huttlin et al., BioGRID preprint 166968) to a
pre-mRNA splicing, CRISPR-based CCDC97-knockout human cells were slower growing than were control lines (Extended Data Fig. 10b, c) and hypersensitive to pladienolide B (Fig. 5d), a macrolide inhibitor of SF3b ${ }^{36}$.

comparable extent as literature reference sets ${ }^{39,42}$, using three measures of complex-level agreement (see Supplementary Methods, Extended Data Fig. 6 b); ${ }^{* * *} P<0.001$, determined by shuffling (grey distributions). c, Agreement of inferred molecular weights (MW) of human protein complexes with sizeexclusion chromatography profiles (data in $\mathbf{c}$, d, from ref. 43). d, Co-elution of human Commander complex subunits by size-exclusion chromatography consistent with an approximately $500-\mathrm{kDa}$ particle.

## Network perspective into conserved biological systems

 Knowledge of conserved macromolecular associations provides a road map for additional functional inferences. For instance, fractionation profiles can be compared for any pair of proteins in our data set to search for evidence of interactions. We found significant enrichment for interactions among pairs of human proteins acting sequentially in annotated pathways ${ }^{37}$ (Fig. 5e), especially G-protein and MAP-kinase cascades (Supplementary Table 8). Enzymes acting consecutively in core metabolic reactions (Fig. 5f) also showed a higher tendency to interact (Supplementary Table 8), the significance of which decayed with more intervening steps (Fig. 5e). For example, strong consecutiveFigure 5 | Functional validation of complexes. a, Morpholino (MO(ATG), targeting start codon to block translation) knockdown of COMMD2 ( $n=55$ animals, 2 clutches, 1 eye each) or COMMD3 ( $n=64$ ) in X. laevis embryos causes defective head and eye development ( control $n=57$; Extended Data Fig. 9f, h). ${ }^{* * * P<0.0001, ~ 2-s i d e d ~ M a n n-W h i t n e y ~ t e s t . ~ b, ~ C O M M D 2 / 3 ~}$ knockdown animals (five embryos per treatment examined) show altered neural patterning, including posterior shift or loss of expression of mid-brain marker EN2 and KROX20 (EGR1), the latter in rhombomeres R3/R5 (compare to Extended Data Fig. 9g, h). c, Enhanced embryonic lethality (epistasis) following RNAi knockdown in C. elegans of B0035.1 (ZNF207) and bub-3 together (eggs laid: HT115, 1,308; B0035.1, 1,096; bub-3, 445; bub$3+$ B0035.1, 341). d, Enhanced sensitivity (mean $\pm$ s.d. across four cell culture experiments) of two independent CCDC97-knockout lines to the SF3b inhibitor pladienolide B (PB) relative to control HEK293 cells. e, Enrichment (permutation test $P$ value) for interactions among sequential pathway components and metabolic enzymes relative to shuffled controls ( $n$ refers to enzyme index, where $n, n+1$ denotes sequential enzymes, $n, n+2$ sequential-but-one, and so on, as described in Supplementary Information. f, Metabolic channelling as opposed to traditional (typical) two-step cascade model. g, Conserved interactions among consecutively acting enzymes involved in purine biosynthesis (two representative co-fractionation profiles of the 69 total generated are shown).
interactions were apparent within the widely conserved purine biosynthetic pathway, with enzymes (for example, PAICS, GART) eluting in two peaks (Fig. 5g), one coincident with the prior enzyme and the second with the downstream enzyme, suggestive of substrate channelling ${ }^{38}$.
Despite the diversity of multicellular organisms, our study reveals fundamental attributes of the macromolecular machinery of animal cells with near universal pertinence to metazoan biology, development and evolution. Our extremely large set of supporting biochemical fractionation data (via ProteomeXchange with identifiers PXD002319-PXD002328), PPIs (via BioGRID; http://thebiogrid. org/185267/publication/) and interaction network projections are fully accessible (http://metazoa.med.utoronto.ca) to facilitate indepth exploration. Although we focused on global conservation properties, these data can be analysed at the individual animal species or complex levels to assess the variety and functional adaptations of particular protein assemblies across phyla.

Online Content Methods, along with any additional Extended Data display items and Source Data, are available in the online version of the paper; references unique to these sections appear only in the online paper.

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## Supplementary Information is available in the online version of the paper.

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Author Contributions A.E. and E.M.M. designed and co-supervised the project. C.W. performed proteomic experiments, aided by P.C.H. B.B. coordinated data analysis, aided by S.Ph., K.D. and S.S., and guided by E.M.M. E.R.T., G.CI., A.B., J.P., X.X., K.C., G.Cr., C.W. and S.Ph. analysed network and conservation data. C.W., F.T., O.K., J.K., S.Pa., O.P., Z.N., D.R.B., X.G., R.H.M, M.S., J.G., M.B., W.B.D. and J.B.W. contributed validation experiments. S.Ph. designed the web portal. C.W., B.B., E.M.M. and A.E. drafted the manuscript. All authors discussed results and contributed edits.

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Extended Data Figure $1 \mid$ Performance measures. a, Performance benchmarks, measuring the precision and recall of our method and data in identifying known co-complex interactions from a withheld reference set of annotated human complexes (from CORUM ${ }^{39}$; as in Fig. 2b). Fivefold crossvalidation against this withheld set shows strong performance gains, beyond a baseline achieved using only human and mouse co-fractionation data along with additional evidence from independent protein interaction screens ${ }^{5,19}$ and a functional gene network ${ }^{20}$ (far-left curve), made by integrating co-fractionation data from the additional non-human animal species (as indicated). 'All data' and 'Fractionation data only' curves include biochemical fractionation data from all five input species: human, mouse, urchin, fly and worm; the latter curve omits all external data. In all cases, at least two species were required to show supporting biochemical evidence. Recall refers to the fraction of 4,528 total positive interactions derived from the withheld human CORUM complexes. b, All 16,655 interactions were identified at least in two species, half $(49 \%, 8,121)$ found in three or more species. c, Among these high-confidence co-complex interactions, 8,981 (54\%) were not reported in iRefWeb ${ }^{44}$ (v13.0), BioGRID ${ }^{45}$ (v3.2.119) or CORUM reference
44. Turner, B. et al. iRefWeb: interactive analysis of consolidated protein interaction data and their supporting evidence. Database 2010, baq023 (2010).
(Supplementary Table 2) for any of the five input species or in yeast; half ( $46 \%$, 4,128 ) of these novel co-complex interactions display evidence of cofractionation in three or more species. d, Final precision/recall performance on withheld interaction test set. A support vector machine classifier was trained using interactions derived from our training set of CORUM complexes, then $\sim 1$ million protein pairs found to co-elute in at least two of the five input species were scored by the classifier. Black curve shows precision and recall for ranked list of co-eluting pairs, with recall representing the fraction recovered of 4,528 total positive interactions derived from the withheld set of merged human CORUM complexes, and precision measured using co-eluting pairs where both members of the pair are contained in the set of proteins represented in the CORUM withheld set. The top 16,655 pairs, giving a cumulative precision of $67.5 \%$ and recall of $23.0 \%$ on this withheld test set, form the highconfidence set of co-complex protein-protein interactions (blue circle). The highest-scoring interactions were clustered using the two-stage approach described in the Supplementary Methods, yielding a final set of 7,669 interactions, which form the 981 identified complexes (red circle; precision $=90.0 \%$, recall $=20.8 \%$ ).
45. Stark, C. et al. BioGRID: a general repository for interaction datasets. Nucleic Acids Res. 34, D535-D539 (2006).



## Extended Data Figure $2 \mid$ Properties of protein elution profiles.

a, Distribution of global protein tissue expression pattern similarity, measured as the Pearson correlation coefficient of protein abundance across 30 human tissues ${ }^{23}$, showing markedly higher correlations for 16,468 protein-protein pairs of putative co-complex interaction partners compared to the same number of randomized pairs of proteins in the network which were not predicted to interact. $\mathbf{b}$, Heat map illustrating the low to moderate cross-species Spearman's rank correlation coefficients in the elution profiles observed between orthologous proteins during mixed-bed ion exchange
chromatography under standardized conditions, highlighting the shift in absolute chromatographic retention times in different species. This variation indicates that the conservation of co-fractionation by putatively interacting proteins is not merely a trivial result stemming from fixed column-retention times. c, The degree of co-fractionation is measured as the correlation coefficient between elution profiles. Spatial proximity is calculated from the mean of residue pair distances between components of multisubunit complexes with known three-dimensional structures (see Supplementary Methods).

b

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Extended Data Figure $3 \mid$ Derivation of complexes．a，The 2，153 proteins present in the 981 derived metazoan complexes participate in multiple assemblies（＇moonlighting＇）to an extent comparable to the sharing of subunits reported for literature－derived complexes（CORUM）．For comparison，we examined the 1,550 unique proteins from the full CORUM set of 1,216 human complexes passing our selection criteria for supporting evidence（＇Unmerged＇） and the 1,461 unique proteins from the non－redundant set of 501 merged complexes used as the reference for splitting our training and testing sets，with some of the largest complexes removed to avoid bias in training（＇Merged＇；
see＇Optimizing the two－stage clustering＇in Supplementary Methods for details）．b，Schematic of 981 identified complexes containing 2,153 unique proteins．In this graphical representation，7，669 co－complex interactions are shown as lines，and proteins as nodes．Red and green interactions were previously annotated in CORUM．Red interactions were used in training the classifier and／or clustering procedure，while green interactions were held out for validation purposes．Grey interactions were not previously annotated in CORUM．


C


Extended Data Figure $4 \mid$ Properties of new and old proteins and complexes. a, The 2,153 protein components in the conserved animal complexes tend to be more ancient than the 2,301 proteins reported in the CORUM reference complexes or in two recent large-scale protein interaction assays, based on either the 7,062 proteins found by affinity purification/mass spectrometry (AP/MS; E. L. Huttlin et al., BioGRID preprint 166968, http:// thebiogrid.org/166968/publication/) or the 3,667 proteins analysed by yeast two-hybrid assays $(\mathrm{Y} 2 \mathrm{H})^{10}$. Ages are derived from OMA (Orthologous Matrix
database) as in ref. 25.b, Annotation rates (mean count of annotation terms per protein) of old and new proteins in the derived complexes and pairwise PPIs, compared with proteins in the CORUM reference complex set. Old proteins (defined by OMA) from the complexes generally exhibited higher annotation rates than new proteins. c, Differential enrichment of old, mixed and metazoan-specific protein complexes for functional annotations (select GOslim biological process terms shown, top) and protein domains (Pfam, bottom).

## RESEARCH ARTICLE



Extended Data Figure $5 \mid$ Abundance and expression trends for proteins in complexes. Proteins within the identified complexes tend to be ubiquitously expressed across human tissues. $\mathbf{a}, \mathbf{b}$, Pie charts show the proportions of proteins with varying tissue expression patterns, from a recently published human tissue proteome map ${ }^{46}$, comparing the full set of 20,258 human proteins (a) with the 2,131 proteins within the identified complexes (b). Consistent with these observations, $91 \%$ of the protein components in the complexes were expressed in $>15$ tissues in data from a reference human
proteome ${ }^{23}$, compared to less than half ( $46 \%$ ) of the 17,294 proteins in the overall reference set ( $Z$-test $P<0.001$ ). c, $\mathbf{d}$, The distributions of average mRNA (c, data from EBI accession E-MTAB-1733) and protein (d, data from PaxDb integrated data set, 9606-H.sapiens_whole_organism-integrated_data set) abundances for all proteins identified and those within complexes. Evolutionarily old proteins (defined by OMA as described in ref. 25 and mentioned earlier) tend towards higher abundances, even for proteins in reference complexes.
46. Uhlen, M. et al. Tissue-based map of the human proteome. Science 347, 6220 (2015).


Extended Data Figure $6 \mid$ Additional validation data. a, Confirmation of MIB2 interactions by co-immunoprecipitation. Extract ( $\sim 10 \mathrm{mg}$ protein) from cultured human HCT116 cells expressing Flag-tagged MIB2 or control (WT) cells was incubated with $100 \mu$ anti-Flag M2 resin for 4 h while gently rotating at $4{ }^{\circ} \mathrm{C}$. After extensive washing with RIPA buffer, co-purifying proteins bound to the beads were eluted by the addition of $25 \mu \mathrm{l}$ Laemmli loading buffer at $95^{\circ} \mathrm{C}$. Polypeptides were separated by SDS-PAGE and immunoblotted using Flag, VPS4A, VPS4B or IST1 antibodies as indicated
(expanded gel images provided in Supplementary Information). b, Protein co-complex interactions reported in the CYC2008 yeast protein complex database ${ }^{42}$ are reconstructed accurately from the co-fractionation data, regardless of whether the full set of co-fractionation plus external data are used to derive protein interactions ('All data', see also Fig. 4b) or if the external yeast data was specifically excluded from the analyses ('All data, excluding yeast').

b


Extended Data Figure $7 \mid$ Agreement of derived complexes' molecular weights with measurement by HPLC and density centrifugation.
a, CORUM reference complexes' inferred molecular weights (MW) are consistent with their components' average cumulative size-exclusion chromatograms. The molecular weight of each complex was calculated as the sum of putative component molecular weights, assuming $1: 1$ stoichiometry. Data from ref. 43 were analysed as in Fig. 4c and show a similar trend as
for the derived complexes. $\mathbf{b}$, Derived complexes' inferred molecular weights are broadly consistent with their components' average cumulative ultracentrifugation profiles on a sucrose density gradient. Average profiles are plotted for $X$. laevis orthologues, based on a preparation of haemoglobindepleted heart and liver proteins separated on a $7-47 \%$ sucrose density gradient, as described in the Supplementary Methods.

Extended Data Figure $8 \mid$ Distribution of uncharacterized proteins and novel interactions across the 981 derived complexes. Complexes were sorted by median age (defined by OMA). Among 2,153 unique proteins, 293 (red)
lack Gene Ontology (GO) functional annotations, while 1,756 of 7,665 co-complex interactions are novel (light green) (not listed in iRefWeb curation database).

e

| CCDC22 | Commd2 | Commd10 |
| :---: | :---: | :---: |
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g

f




Native transcript PCR 461 bp


Morphant transcript PCR 378 bp

Extended Data Figure $9 \mid$ Properties of the Commander complex. The automatically derived 8 subunit Commander complex (Fig. 3b) was subsequently extended to 13 subunits (COMMD1 to 10, CCDC22, CCDC93, and SH3GLB1) based on combined analysis of AP/MS (Fig. 4a), size-exclusion chromatograms ${ }^{43}$ (Fig. 4d), published pairwise interactions ${ }^{30,47,48}$, and analysis of elution profiles of the remaining COMM-domain-containing proteins, as shown here. Example protein elution profiles are plotted for Commander complex subunits observed from: HEK293 cell nuclear extract (a); sea urchin embryonic (5 days post-fertilization) extract (b); and fly SL2 cell nuclear extract (c); each fractionated by heparin affinity chromatography. d, Co-expression of Commander complex subunits during embryonic development of $X$. tropicalis (plotting mean $\pm$ s.d. of three clutches; data from ref. 49). e, Messenger RNA expression patterns of Commander complex subunits in stage 15 X. laevis embryos. Images show coordinated spatial expression in early vertebrate embryogenesis, as measured by in situ hybridization (three embryos examined). f, Knockdown of Commd2 induced marked head and eye defects in developing X. laevis. Top, Commd2 antisense knockdown significantly decreased eye size, shown for stage 38 tadpoles
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48. Phillips-Krawczak, C. A. et al. COMMD1 is linked to the WASH complex and regulates endosomal trafficking of the copper transporter ATP7A. Mol. Biol. Cell 26, 91-103 (2015).
(from three clutches; control $n=47$ animals, one eye each; $* * * P<0.0001$, two-sided Mann-Whitney test); phenotypes were consistent between translation blocking (MOatg; $n=60$ ) morpholino reagents, splice site blocking (MOsp; $n=50$ ) morpholinos, and knockdowns of interaction partner Commd3 (see Fig. 5a). Bottom, Commd2-knockdown induced altered Pax6 patterning in the embryonic eye (control $n=8$ animals, two eyes each; MO $n=11$ ). $\mathbf{g}$, Commd $2 / 3$-knockdown animals show altered neural patterning. Changes in stage 15 X . laevis embryos, measured by in situ hybridization (assayed in duplicates; five embryos per treatment), seen upon knockdown but not on controls: the forebrain marker PAX6 was expanded, while the mid-brain marker EN2 was strongly reduced. Notably, while expression of KROX20/ EGR1 in rhombomere R3 was shifted posteriorly, expression in R5 was strongly reduced or entirely absent. Panels in Fig. 5b are reproduced from this figure and are directly comparable. $\mathbf{h}$, Confirmation of splice-blocking Commd 2 morpholino activity. Images and schematic show the basis and results of RTPCR and agarose gel electrophoresis obtained with the corresponding X. laevis knockdown tadpoles.
49. Yanai, I., Peshkin, L., Jorgensen, P. \& Kirschner, M. W. Mapping gene expression in two Xenopus species: evolutionary constraints and developmental flexibility. Dev. Cell 20, 483-496 (2011).

b


Extended Data Figure $10 \mid$ Supporting data for BUB3 and CCDC97 experiments. a, Sequence alignment showing conservation of ZNF207 GLEBS domain. b, Targeted CRISPR/Cas9-induced knockout of CCDC97 in two independent lines of human HEK293 cells, as verified by western blotting
(expanded gel images provided in Supplementary Information). c, Loss of CCDC97 impairs cell growth. Lines show growth curves of control versus knockout cell lines in two biological replicate assays.


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