Feature Review
Comparative Connectomics

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We introduce comparative connectomics, the quantitative study of cross-species commonalities and variations in brain network topology that aims to discover general principles of network architecture of nervous systems and the identification of species-specific features of brain connectivity. By comparing connectomes derived from simple to more advanced species, we identify two conserved themes of wiring: the tendency to organize network topology into communities that serve specialized functionality and the general drive to enable high topological integration by means of investment of neural resources in short communication paths, hubs, and rich clubs. Within the space of wiring possibilities that conform to these common principles, we argue that differences in connectome organization between closely related species support adaptations in cognition and behavior.

Comparing Brains

Comparative biology and comparative neuroscience generally aim to discover common plans of organization while also accounting for diversity among species. A key objective of comparative studies of brain architecture is to achieve an understanding of the neurobiological basis for the emergence of complex brain structure and function. For example, several classic studies on the cellular composition of the primate cortex have addressed cross-species homologies [1–3] and contemporary comparative analyses have highlighted common cortical phenotypes and important roles of genetic and epigenetic interactions in development for creating cross-species diversity [4,5]. Together, these and many other comparative studies have laid the foundations for our understanding of mammalian brain anatomy and function.

One major focus has been on the growing size of brains from smaller to larger animals [6–9] and, in particular, the significant increase in volume required by the expansion of anatomical connections [10–12]. A seminal observation is that the proportion of brain mass spent on cortical white matter follows an **allometric scaling** (see Glossary) relation between body and brain size across the entire spectrum of simpler to higher-order mammalian species [13,14]. In small mammals, such as the mouse, only approximately 11% of total cortical volume comprises white matter, in contrast to 27% in the macaque monkey, 40% in chimpanzees, and 41% in humans (data from [13]). However, despite a larger volume of white matter, maintaining constant connection density among an increasing number of neurons and regions in larger brains will quickly outstrip the volume that can be allocated to long-distance neural wiring [15–17]. Thus, the scaling between brain size and white matter volume implies a lower proportion of directly connected neural elements in larger-sized brains [15,17,18], making it increasingly difficult for neural elements to communicate via direct connections. Maintaining fast and efficient neural communication brings significant benefits to brain function, arguably leading to a fundamental tension or tradeoff [19] between the conservation of neural resources that can be spent on long-distance connectivity and the promotion of efficient communication to support complex neural processing. These apparent opposing or competitive pressures highlight the importance of the **topological** organization of nervous systems that must provide an arrangement of neural

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elements and connections to balance the amount of neural resources used for connectivity while simultaneously enabling effective information transfer in the service of brain function.

Comparative Connectomics

The examination of brain network topology is a core element of the field of **connectomics** [20], the emerging science of structural and functional brain networks [21–24]. The increasing availability of connectomes of multiple animal species (Figure 1, Key Figure) provides a new opportunity for the comparative analysis of network architecture across species. In this review, we introduce '**comparative connectomics**', defined as the comparison of the topological layout of nervous systems across species, with the aim of identifying common principles and variations in network features. Comparative connectomics can provide insight into general principles of neural wiring that apply across species and can examine to what extent variations in connectivity between species may form the basis for differences in brain function. As we discuss, connectomes of different species reconstructed by a broad range of methodologies (Box 1) can be compared by applying a consistent set of network analysis measures and graph analytical

Key Figure

Connectomes across Species

Connectomes

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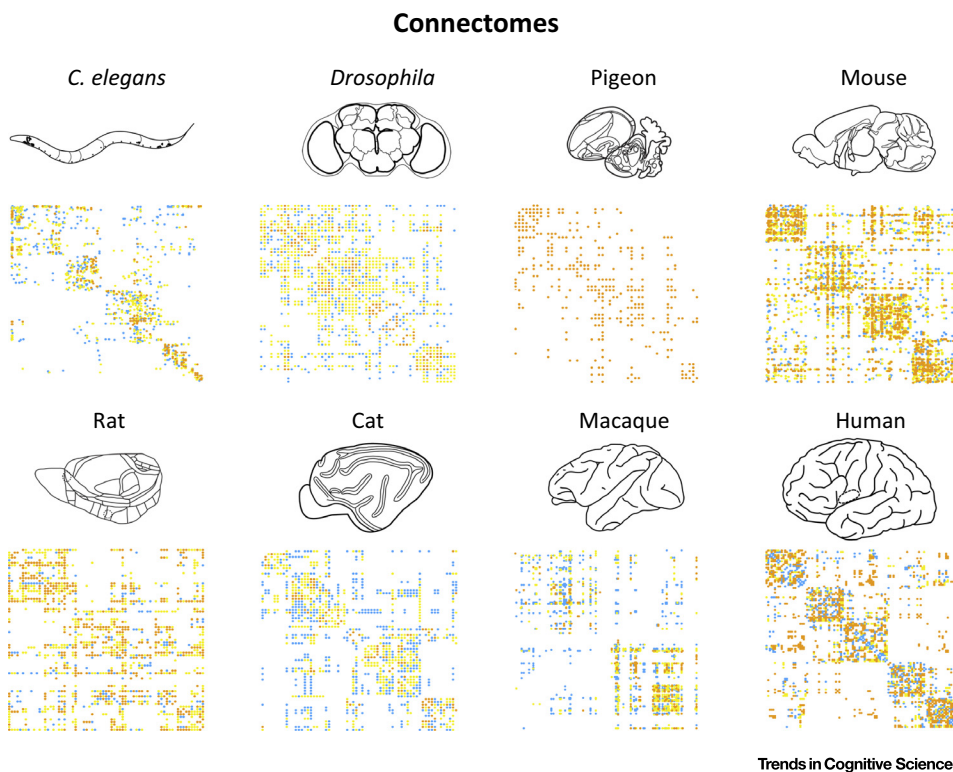
Figure 1. The Figure displays reconstructed connectomes of eight different species: *Caenorhabditis elegans* (roundworm) [26], *Drosophila* (fruit fly) [40], pigeon [44], mouse [45], rat [49], cat [51], rhesus monkey (macaque, FE91 atlas) [53,160], and human (Human Connectome Project data, [128]). Connectomes are represented as connectivity matrices with rows and columns depicting source and target regions (grouping regions participating in the same community together) and with the elements of the connectivity matrices showing the reconstructed projections. Pathways are grouped accordingly to whether they are weak (blue), medium (yellow), or strong (orange).

Box 1. Connectome Reconstruction Methods

Different methodologies can be applied for the reconstruction of connectome maps. Electron microscopy techniques have enabled reconstruction of the *Caenorhabditis elegans* nervous system at the synaptic level [25]. Tracer injection techniques combined with high-throughput serial tomography optical imaging have enabled the high-resolution reconstruction of the mouse connectome [45] and systematic collations of data across high numbers of tracing experiments have resulted in grouped consensus matrices of the cat, macaque, and rat brains. Advances in *in vivo* diffusion MRI techniques have made it increasingly feasible to reconstruct macroscale connectomes of individual brains of great apes [144] and humans. The development of techniques such as CLARITY [161] and 3D Polarized Light Imaging (3D-PLI) may bring unprecedented high-resolution reconstructions of animal and human connectome wiring in the near future.

Today, the field of comparative connectomics is limited by the fact that connectome maps of different species are being acquired with different types of methodology, restricting direct comparison of connectomes across a range of species. For example, the *C. elegans* connectome describes neuron-to-neuron interactions at the microscale, while the macaque and cat connectomes describe region-to-region connections at the macroscale level of nervous system organization. In addition, while electron microscopy, tract-tracing and MRI-derived connectome reconstructions can provide information about the strength of connections (often referred to in graph theoretical analysis as 'weighted networks'), connectome maps derived from cumulative collation of data across published literature are often limited in only providing information about the presence or absence of connections (referred to in the connectome literature as 'binary networks') or coarse categorization of the magnitude of pathways.

Thus, we argue that there are two aspects that are critical for enabling comparative studies in the future. First, it is crucial to establish ways to interpolate across different methodologies, ideally by directly comparing brain connectivity patterns derived by multiple techniques in the same organism, for example tract-tracing and noninvasive neuroimaging [76,162,163]. Second, the application of the same technique to multiple species (e.g., [28,144]) facilitates cross-species comparisons.

Network: in nervous systems, a network describes the set of neural elements (e.g., neurons or brain regions) and their relations (e.g., synaptic connections, macroscale pathways, or functional interactions).

Rich club organization: the property of a network to display an increasing level of connectivity between subsets of highly connected nodes, a level of connectivity higher than would be expected on the basis of the individual degree of the subset of nodes alone.

Scale-free organization: a class of networks with a degree distribution that follows a power law, or, more generally, a type of broad-scale or fat-tailed degree distribution.

Small-world organization: a class of networks that shows both a high level of clustering (such as a regular lattice) and short characteristic paths (such as a random graph).

Spectral graph theory: a branch of graph theory that studies networks by examining the spectrum of eigenvalues and eigenvectors of the adjacency matrix of the network.

Topology: the topological structure of a network describes the arrangement of connections in a network and is invariant to any continuous spatial deformation of the system.

tools (Box 2) to enable quantitative comparison of the topological architecture of the nervous systems across and between species. We begin with an overview of currently available systems-level connectome maps, ranging from invertebrates, such as *Caenorhabditis elegans* (a nematode or roundworm) and *Drosophila melanogaster* (fruit fly), to the macroscale connectome maps of mammalian species, such as rodents, Old World monkeys, great apes, and humans (Figure 1). Using **graph theory** as a general framework to quantify topological features of **network organization**, we first identify common aspects of brain network topology. We then turn to differences in neural wiring and brain network organization among closely related species, suggesting that variations in connectome topology subserve species-specific behavioral and cognitive adaptations.

Connectomes from Simpler to More Advanced Animals

Nearly 30 years ago, serial electron microscope reconstructions of the nervous system of the hermaphrodite nematode *C. elegans* [25] served as the foundation for the assembly of a near-complete neuronal connectivity network of 890 gap junctions, 6393 chemical synapses, and 1410 neuromuscular junctions among 279 neurons of the somatic nervous system of the worm [26]. This seminal work inspired later efforts to map the posterior nervous system of the *C. elegans* adult male [27], as well as the reconstruction of (partial) neuronal connectomes of other worm species, including that of the roundworm *Pristionchus pacificus* [28] and the marine worm *Platynereis dumerilii* [29].

Early graph theoretical analyses showed the *C. elegans* nervous system to exhibit a **small-world** network organization [30] with high **clustering** (Box 2) and short paths, indicating preferred connectivity into locally clustered circuits, combined with the presence of relatively few long-distance connections that mediate topological short-cuts and facilitate efficient global communication [31]. Further analyses disclosed preferential formation of specific local motifs [26] and a **community structure** of densely intraconnected and sparsely interconnected subnetworks or modules [32], an organization that roughly aligns with a classification of neurons into sensory, motor, and interneuronal categories [26] (Figure 2).

Box 2. Graph Theory

The connectomes of different species can be compared by applying a consistent set of network analysis methods. These methods are principally drawn from the mathematical field of graph theory, enabling quantitative comparison between species using the same mathematical language. Within this approach, a nervous system is described as a graph, comprising a collection of nodes (e.g., neurons or brain areas) and a collection of edges describing the pairwise relations between nodes (e.g., synaptic connections or macroscopic axonal projections) (Figure I). Once such a mathematical description of a network is made, graph theoretical metrics can be used to describe and, across species, compare topological properties of the network. The metric of degree represents the number of connections attached to a node of the network. Clustering describes the tendency of nodes to form closed triangles, such that the nearest neighbors of a node are also directly connected to each other. The metric of clustering reflects the tendency of a network to form topologically local circuits, and is often interpreted as a metric of information segregation in networks. A path describes a route of information transfer between two nodes in a network, with the metric of path length describing the number of steps (comprising unique edges) crossed when traveling from one node to another node in a network. The shortest path length expresses the minimal number of steps needed to travel between nodes, and is often interpreted as a metric of the efficiency of information transfer between nodes of a network. Hubs are nodes with a high degree and a topologically central position in the overall network, with the core of a network describing a set of highly connected nodes that are mutually densely connected. The related concept of rich club organization describes the propensity of highly connected nodes to be more densely connected to each other than expected based on the individual degree of the nodes. The community structure of a network refers to the tendency of a network to form densely connected subgraphs or modules, corresponding to a subset of nodes that are densely connected to each other and sparsely connected to nodes in other modules. In nervous systems, anatomical modules often overlap with known functional systems.
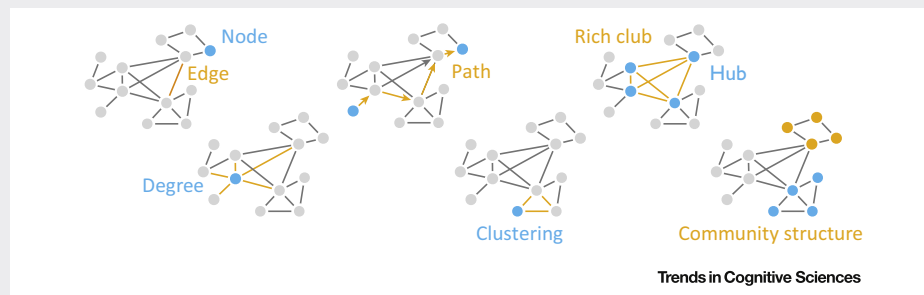
Figure I. Graph Metrics. Figure illustrates (from left to right) the mathematical description of a network as a graph comprising nodes and edges, the metric of degree, a communication path, the metric of clustering, the formation of a densely connected rich club or core, and the community structure of a network, reflecting the formation of densely connected subgraphs or modules within the network.

Network analyses of the *C. elegans* connectome have had a prominent role in discussions of wiring minimization models [33,34], which advocate that the conservation of wiring governs the spatial placement of neuronal elements and accounts for the formation of spatially localized circuits. However, these analyses have also revealed that some neurons and their connections deviate from optimal spatial placement, as predicted by a strict minimization of wiring cost [33,35,36], which argues for other, nongeometric biological factors to have a role in the topology of the *C. elegans* nervous system. Indeed, one of the most prominent features of the connectome of the adult worm is the non-uniform distribution of synaptic connectivity (i.e., **node degree**) across neurons [26,37], with most synaptic connections maintained by a small set of highly connected neurons. These **hub** neurons maintain dense interconnectivity with each other despite being spatially distributed in both the anterior and posterior extremities of the body of the worm. These findings are indicative of a biologically expensive **rich club** organization of the adult *C. elegans* connectome in which high-degree elements form a densely centralized core [32]. Interestingly, hub neurons constituting the rich club of the adult *C. elegans* had previously been categorized as command interneurons with known functional importance for information integration [26], coordinated movement, and adaptive behaviors of the animal [32]. Furthermore, in *C. elegans*, these hub neurons have been noted to appear as one of the first elements of the nervous system [38], suggesting a central role of rich club architecture in nervous system development [32,39].

Community structure

(A) *C. elegans*

(B) *Drosophila*

(C) Macaque

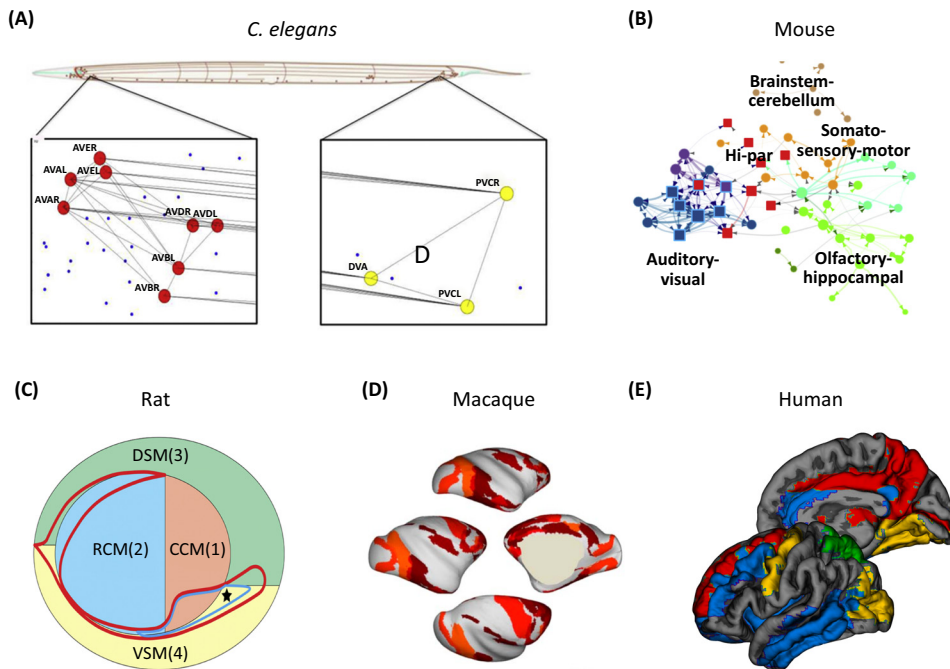**(D)** Human
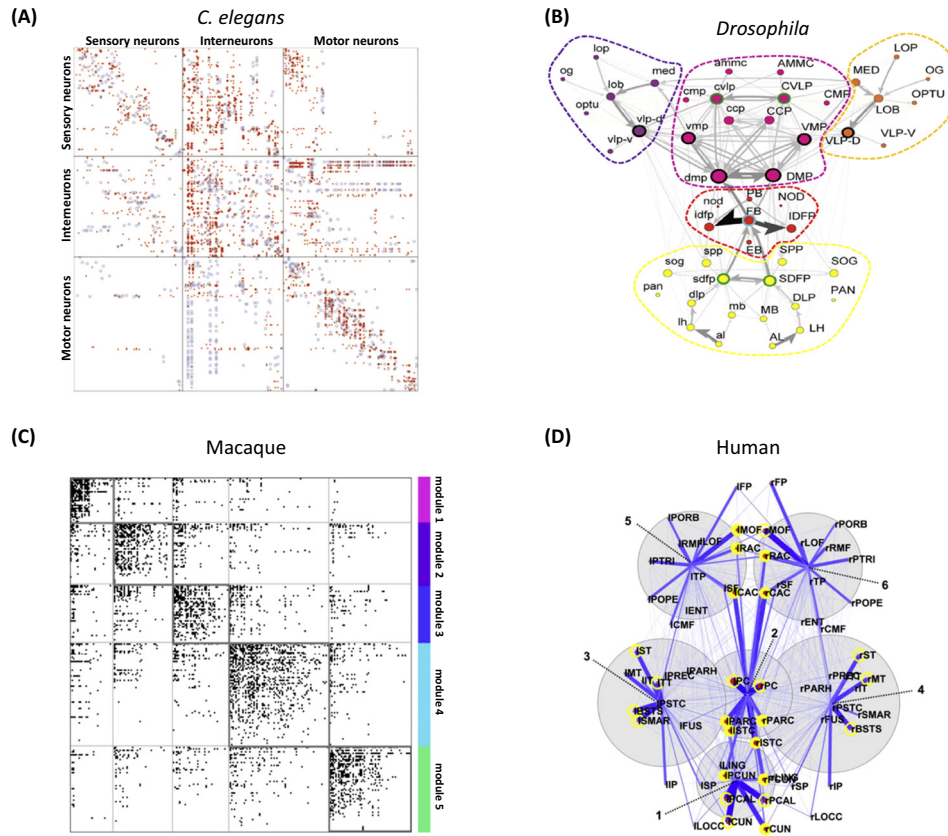
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Figure 2. Community Structure. Studies have shown consistent community organization of nervous systems across species, including (A) the *Caenorhabditis elegans* neuronal network, and (B) *Drosophila*, (C) macaque, and (D) human connectomes. Anatomical communities obtained by graph theoretical analysis often with known functional domains, as for example the olfactory (yellow), visual (purple and orange, left and right), auditory/mechanosensory (magenta), and premotor (red) functional systems of the fly brain (B) [41]. Modular decomposition of the human connectome as derived from diffusion-weighted imaging revealed the formation of at least six anatomical communities overlapping spatial domains of the human brain. Reprinted under the creative Commons Attribution License from [26] (A), [62] (B), and [68] (C); adapted and reproduced with permission from [41] (B).

Other projects aiming to establish neuron-scale connectivity maps have focused on the reconstruction of the connectome of *Drosophila*, for example by imaging a sample of approximately 12 000 of the >100 000 neurons that comprise the adult fly brain [40]. Grouping of individually imaged neurons into over 40 morphologically distinguishable brain areas resulted in a detailed interregional connectome map [40]. Subsequent refinement of this map and detailed graph theoretical analysis showed several highly connected hub regions as well as five network communities (**modules**) of densely interconnected regions [41]. These five communities largely correspond to known olfactory, visual, mechanosensory/auditory, and premotor systems of the fly brain, suggesting that their topological structure forms the anatomical substrate for specialized functional processing (Figure 2). Wiring economy has been suggested to have a major role in local neuron and connectivity placement in the *Drosophila* brain [42]. However, as in *C. elegans*, a modular structure minimizing the expense of wiring is supplemented by the presence of a highly connected rich club [41]. Rich club regions are found distributed across all modules of the *Drosophila* nervous system and their remote geometric placement implies a high cost of wiring of their interconnecting pathways [41] (Figure 3). Parallel mapping efforts using

Hubs and rich club organization

(A) *C. elegans*

(B) Mouse

Brainstem-cerebellum
Somato-sensory-motor
Hi-par
Auditory-visual
Olfactory-hippocampal

(C) Rat
DSM(3)
RCM(2) CCM(1)
VSM(4)

(D) Macaque

(E) Human

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Figure 3. Hubs and Rich Club Organization across Species. Connectome studies have shown hub and rich club organization for (A) the microscale *Caenorhabditis elegans* nervous system, as well as for the macroscale brain networks of (B) mouse, (C) rat, (D) macaque, and (E) human. Analysis of the *C. elegans* connectome revealed a small set of highly connected rich club hub neurons distributed across anterior (red nodes) and posterior (yellow nodes) parts of the animal. (B) Network analysis of the mouse connectome showed the existence of highly connected and highly central connector hubs (red nodes), which mediated most of the intermodular connections between functionally specialized modules of the community structure (nodes are color coded by modular affiliation). (C) Network analysis of the rat connectome has shown the rich club (red line) to participate across multiple functional domains (depicted as colored blocks). (D) Connectome analysis of the macaque brain has revealed rich club members (red nodes) to be spatially distributed across the cortex. (E) Network studies of the human connectome have similarly shown the existence of highly connected hubs in the human brain (depicted as red, yellow, blue, and green regions). These hub areas are distributed across multiple functional domains and show strong rich club organization. Adapted and reproduced from [32] (A) and [72] (E) under the Creative Commons Attribution-Noncommercial-Share License; adapted and reproduced, with permission, from [47] (B) and [49] (C); reprinted under the Creative Commons License from [66] (D).

electron microscopy serial sectioning and reconstruction of neuronal circuits in the *Drosophila* larva have provided detailed wiring diagrams of multisensory circuits that are similar to those of the adult fly brain in demonstrating complex patterns of convergence to enable multimodal information integration [43]. These and other studies have established important links between the topology of neuronal circuits and the sensory/behavioral functions of the brain.

Reconstructions of whole-brain connectome maps of more complex species predominantly involve the tracking of macroscale axonal projections between large-scale brain areas. Such reconstructions include mappings of the avian pigeon brain, revealing a modular network architecture and densely connected hub areas [44], as well as detailed mappings of the macroscale systems of several rodent and primate species. Recent comprehensive mapping efforts utilized a large number of injections of anterograde tracers combined with high-through-put serial tomography optical imaging for a detailed reconstruction of the mouse connectome [45]. Although such automated reconstruction procedures will inevitably be prone to some level

of measurement error and statistical noise [45], standardized experimental conditions allowed tracking of an unprecedented number (>15 000) of directed projections among more than 200 areas of the mouse brain. A parallel effort involved the detailed mapping of over 600 macroscale corticocortical pathways combining both anterograde and retrograde tracer experiments [46]. Network analysis of these mouse connectome maps revealed dense local clustering of wiring and modular organization, combined with the existence of topologically short pathways [45,46]. Furthermore, the strength of anatomical connectivity was noted to approximate a log-normal distribution ranging over five orders of magnitude [45]. Projection strengths of anatomical pathways generally were strongest between spatially neighboring regions and decayed monotonically as a function of increasing connection distance. This argues in favor of an important role of geometric factors in shaping the topology of mouse wiring [45]. However, similar to the invertebrate species considered earlier, the wiring cost of the mouse connectome is not strictly minimized and its modular organization is complemented by the existence of high-degree rich club hubs mediating biologically expensive (long-distance) connections between modules [47,48] (Figure 3). Generative modeling approaches proved successful in reproducing the modular aspects of connectome topology of the mouse brain simply by minimizing wiring cost, but these models failed to account for the emergence of long-distance connections until the penalty imposed on spatial distance was adaptively relaxed for higher degree hubs [47].

A different approach aiming to build a rat cortical connectome involved collating data across >16 000 literature reports of tract-tracing experiments. Collectively, these data captured over 1900 cortical association macroconnections spanning >70 distinct gray-matter areas of the rat cerebral cortex, together with an ordinal assessment of connection strength [49]. Graph theoretical examination revealed short communication paths, and high clustering, connected communities largely coinciding with functionally specialized systems [49], and a central rich club [49,50] comprising a set of highly connected, spatially distributed, and multimodal cortical areas [49] (Figure 3).

Consensus matrices obtained by collating data across large numbers of tracing experiments have also been built for the cat [51], ferret [52], and macaque brains [53]. One of the earliest mappings of a mammalian cerebral system involved the compilation of the macroscale cat connectome, describing 65 cortical areas and >1000 corticocortical pathways [51]. Network analysis of the cat connectome showed short-length paths and strong clustering of anatomical wiring within functionally specialized motor, visual, auditory, and frontolimbic subsystems [54]. Follow-up analysis identified different types of hub area in the cat cortex [55], and defined a central, reciprocally interconnected core module [56,57] corresponding to a rich club of cortical hubs [58].

The systematic analysis of anatomical connectivity patterns in the macaque brain was propelled forward by the early connectomics pioneer Rolf Kötter (1961–2010) who founded the open-access Collation of Connectivity Data for the Macaque (CoCoMac) database, which aggregated data from hundreds of macaque tract-tracing experiments [53,59]. Graph theoretical analyses of macaque connectome maps revealed high clustering, a hierarchical ordering of wiring from primary to multimodal areas [60], short path lengths, pronounced modular organization [61–63], and the presence of spatially distributed but centrally connected cortical hub areas [55,62] (Figure 3). As already noted by early computational analyses of macaque interareal connectivity [64], inverse correlations between projection density and geometric length suggest an important role of spatial or geometric factors in shaping the topology of macaque interareal projections [65]. However, modeling analyses of macaque connectivity have also pointed out an important role for topological factors in shaping macaque connectivity, such as the drive to enable efficient communication pathways by means of the formation of high-cost long-distance projections [35] and the high investment of connectivity around hubs and a rich club core [55,66,67]. These network attributes are difficult to account for by purely geometric factors.

The development of diffusion-weighted imaging combined with tractography has enabled the assembly of the macroscale human connectome [68–72]. Despite the caveats and limitations regarding the interpretation of the diffusion MRI signal [73–75] and its use in connectome reconstruction (discussed in [68,76–78]), diffusion-weighted imaging is currently one of the most widely used methods for the assessment of anatomical connectivity in the human brain. Several large-scale efforts, including the Human Connectome Project [79], are delivering unprecedented amounts of high-quality human connectome data, and advances in *ex vivo* imaging techniques might provide even more-detailed maps of postmortem animal and human wiring in the future (Box 1). Paralleling key features of connectome topology already reviewed for other animals, network analysis of the human connectome has shown an organization of densely connected communities that form the anatomical wiring skeleton of known functional domains [80–85] (Figure 2). This modular structure is complemented by anatomically long-distance projections that support topologically short-distance global communication paths [68,69], combined with densely connected and topologically central communication hubs [68,71,86–88] that form a central rich club in the human brain [39,89–92] (Figure 3).

Having surveyed our current knowledge of connectome topology across a range of species, we are now in a position to examine the potential principles that may drive common themes in wiring patterns, as well as the important roles of cross-species variations in differentiating behavioral and cognitive adaptations.

Common Principles of Connectome Wiring

We first discuss which general principles may underlie common themes of connectome organization observed across species. Across the range of species studied so far, we observe a strong tendency at both the micro- and macroscale of network organization for neurons (micro) or brain regions (macro) to connect to their spatial and topological neighbors, favoring the formation of spatially colocalized, topologically clustered cliques or communities [80] (Figure 2). This community architecture largely coincides with the formation of physiologically specialized functional domains in nervous systems, consistent with the idea that anatomical wiring has an important role in the functional differentiation of cortical areas [93–95]. Cross-species comparison shows a high level of consistency of functional domains across species, with human, macaque, and rodents showing several homologous primary and higher-order associative limbic and cognitive networks [96–99]. The conservation of wiring has long been proposed as a fundamental rule governing the local layout of circuits [34,47,64,65,100–102]. Geometric constraints promote the emergence of functionally specialized network communities in nervous systems. It has been hypothesized that the growth of brain volume across species favors the existence of local modules [103,104] while penalizing the formation of long-range connectivity because such connections become increasingly expensive in terms of neural resources. Support for such notions comes from studies examining the volume of white matter connectivity tracts across a range of primate species, revealing long-distance (and, thus, costly) connectivity in the corpus callosum to be lower in larger primate brains, while intrahemispheric white matter volume is enlarged [105]. A more local organization of cortical wiring in larger brains may promote the emergence of specialized brain functionality in larger animals [103,104,106–108]. For example, a stronger modularity structure of connectome wiring may have had a role in the formation of more spatially localized cortical fields for visual, sensory, and motor information processing in larger mammalian brains [5], as well as increased brain lateralization, preferred hand use, and the development of specialized traits, such as language processing, in humans [109–111].

A second general theme of connectome topology includes the drive of nervous systems to invest resources in network attributes that result in topological integration (Figure 3). Across the range of examined species, all nervous systems show topologically short and efficient paths, regardless of increasing brain network size. Although modeling studies have generally succeeded in

Box 3. Nerve Nets

The nematode, insect, avian, and mammalian species discussed in this review are all species that display a central brain system. Others, such as jellyfish, hydra, and starfish, do not have a centralized brain. Instead, their nervous system comprises a 'nerve net' (Figure IA), with neurons distributed across most parts of the animal, a type of system organization reflecting their distinct body plan [164,165]. The nervous systems of jellyfish and hydra are often described as a simple network in which neurons are only connected to their spatial neighbors (in network terms, this would make a simple 'regular graph' or spatial 'grid' with no modules, short paths, or hubs).

However, recent studies have suggested that the nervous systems of adult jellyfish and hydra may exhibit a more diverse architecture, including potential concentration of neural elements into complex structures facilitating information integrative processes [165–167]. The nervous systems of adult jellyfish comprise multiple components (see [164] for an introduction to jellyfish nervous systems), including sensory structures (rhopalia, Figure IB) that are connected via an interconnecting nerve ring(s) system running around the bell (Figure IB). This nerve ring is involved in integrating the swimming, visual, and tentacle system [165] and is argued to represent a rudimentary central nervous system [168,169].

Thus, although the basic plan of the nerve nets of species such as jellyfish and hydra may be laid out differently from the nervous systems of the species discussed in this review, we argue that there may be common topological attributes, for example a centralized system of interconnected neuronal structures. This suggestion calls for further expansion of the currently available set of connectomes beyond those discussed here, and emphasizes the need for comparative connectomics to examine commonalities and differences in nervous system architecture across a (much) wider range of species.
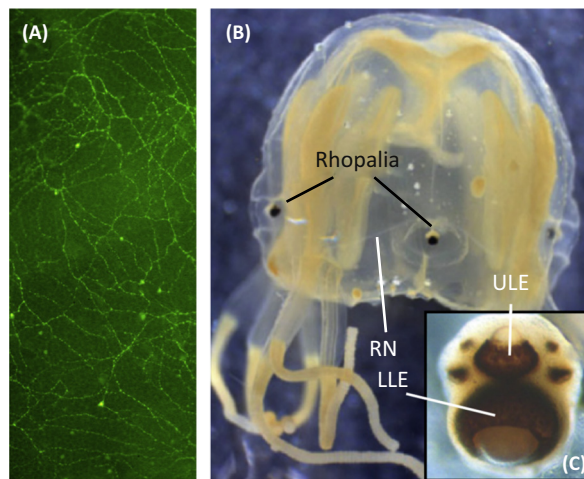
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Figure I. Jellyfish Nerve Net. (A) The nervous system of a jellyfish involves a nerve net. (B) Sensory structures (rhopalia, with the insert showing a magnification with details of lower [LLE] and upper [ULE] lens eyes [164]) in several jellyfish species include concentrated neural structures that are interconnected by an organized nerve ring [RN in (B) and also shown in (C)], potentially supporting multiple communication pathways. Adapted and reproduced with permission from [165] (A) and [164] (B). Abbreviations: LLE, Lower Lens Eye; ULE Upper Lens Eye.

accounting for parsimonious local wiring and cost-controlled community structure, they have difficulty explaining the formation of long-distance pathways (essential for efficient communication paths) unless constraints on wiring cost are relaxed [47,101,102,112]. Across species, the consistent presence of long-distance connections that violate strict minimization of wiring suggests a general drive of neural systems to invest neural resources in network attributes that maintain short communication relays.

The strong drive to invest costly resources in network attributes that result in topological integration may become even more apparent from the consistent allocation of neural assets to the formation of a densely connected core or rich club [32,87,113]. This general tendency to centralize connectivity appears to be ubiquitous: it is observed across spatial scales (e.g., from patterns of information flow in microcircuits [114–116] to whole-brain systems), as well as across small and large nervous systems of vastly different species (Box 3).

Communication in nervous systems goes beyond the simple relaying of messages along shortest paths and involves continual processing of information at each neural step. Thus, the centralization of neural connectivity into a connective core may have important benefits with regard to information integration [32,87,117,118], potentially facilitating higher-order brain functionality that thrives on integrated information, such as coordinated movement in the worm [32] and/or executive functioning in humans [119]. Due to their central embedding in network topology, hubs and their connections can attract and disseminate a large proportion of all neural communication [89,118], forming an anatomical substrate for the exchange of information between otherwise segregated domains. This has led to the theory that hubs and rich clubs provide an anatomical infrastructure [87] for the formation of a 'global workspace' [120–122] or 'connective core' [123], a functional construct in which information from different parts of the system is integrated. These ideas are supported by observations demonstrating that hub regions in mammalian species tend to overlap with multimodal association cortex [124,125], display an intricate neuronal architecture with heavily branched pyramidal neurons [126–130], pose high energy demands [19,90,131], and form convergence zones between different functional networks [132–134].

The common theme across nervous systems to invest neural resources in network attributes that bring topological integration argues for a more diverse set of principles than strict conservation of neural resources to shape the connectivity layout of nervous systems [135,136]. Rather, the drive to invest neural resources in dense connectivity around an integrative core appears to favor the emergence of connectome attributes that are advantageous for enabling short pathways and transmodal communication. Offsetting their greater biological cost in terms of neural and metabolic resources, these topological features may bring strong potential benefits for integrative neural processing and cognitive brain function and behavior.

Variations Support Behavioral and Cognitive Adaptations

Although we argue for common cross-species themes in connectome organization, a proposed trade-off between pressures to conserve wiring cost and pressures to maximize topological integration [19,137] cannot, by itself, account for the abundant diversity in brain connectivity and nervous system organization across species. Multiple factors, including variations in environmental factors, genes, and genetic regulatory networks operating during development, combine to generate a diverse set of brain networks even across closely related species. A natural next question, then, is how differences in connectome wiring between species relate to species-specific differences in behavior and information processing. This question is most productively explored by comparing connectomes of relatively closely related species that share a basic anatomical plan and an overlapping behavioral repertoire.

One example is offered by comparative studies of the nervous systems of roundworms. Comparison between reconstructions of microscale neural networks of the pharyngeal systems of *C. elegans* and *P. pacificus* showed significant differences in neural connectivity that could be linked to their distinct feeding behaviors [28]. In contrast to the bacterial feeder *C. elegans*, *P. pacificus* is a predator with a distinct anatomy of its mouth areas. Side-by-side comparison of the layout of synaptic connectivity of the pharyngeal systems of the two species revealed significant differences [28] (Figure 4), with *P. pacificus* showing increased connectivity around neural elements controlling tooth-like denticles. Comparative network centrality analysis further showed that presumptive communication paths and information flow in *P. pacificus* are more tuned towards the control of pharyngeal motor neurons. Thus, differences in connectome topology appear to constitute at least part of the biological basis for the substantial divergence in feeding behavior between the two worm species [28]. In addition, a second comparative example in the roundworm (not across species but gender) is provided by dimorphic differences in neural wiring of the reproductive system of *C. elegans*. Male worms have 383 neurons

(A)

(B)

Human Chimpanzee Macaque

(C)

Node degree
group difference
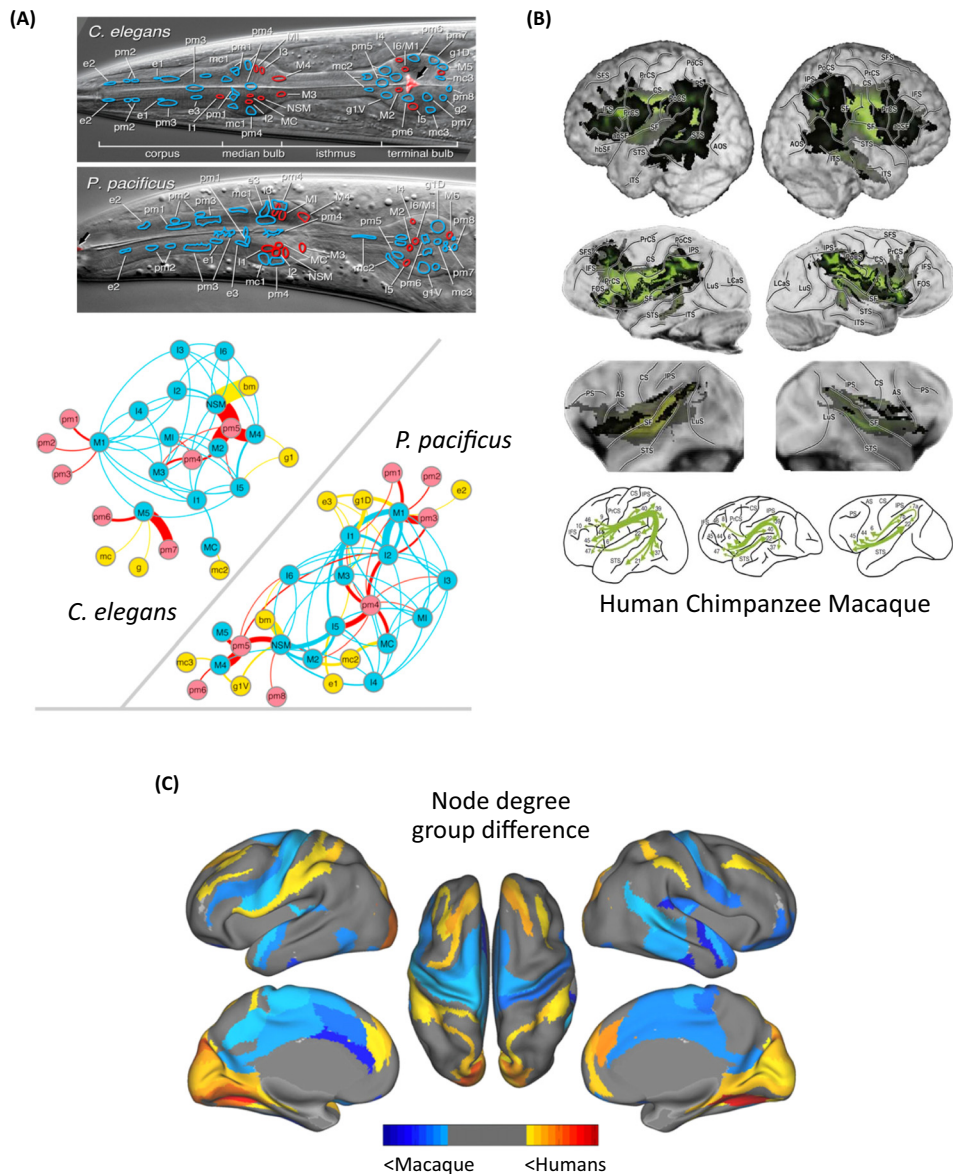
<Macaque   <Humans

Figure 4. Connectivity Comparison across Species. (A) A side-by-side comparison of the neuronal organization (upper panel) and wiring (graphs in the lower panel) of the pharyngeal system across two worm species, *Pristionchus pacificus* (right) and *Caenorhabditis elegans* (left). A comparison between the two systems (comprising the same number and same types of neuron) reveals wiring differences: the predator *P. pacificus* shows higher complexity of connectivity around neural elements controlling the tooth-like denticles compared with *C. elegans* [28]. (B) Comparison of macroscale connectivity between macaque (top), chimpanzees (middle), and humans (bottom) revealed more elaborate connectivity of the arcuate fasciculus in humans, a tract important for complex language processing [109]. (C) Comparison of whole-brain functional connectivity patterns between macaques and humans revealed areas of both weaker (blue areas) and stronger functional connectivity across the cortex (red areas) in humans compared with macaques, as well as an overall more spatially diffuse hub structure in humans [145]. Adapted and reproduced with permission from [28] (A) and [109] (B); adapted and reproduced under the Creative Commons Attribution-Noncommercial-Share License from [145] (C).

(in comparison to 302 in hermaphrodites) with dimorphic changes in neurons and connectivity mostly focused around the formation of circuits related to mating and reproduction [25,27]. Male worms, but not hermaphrodites, are capable of sexual adaptive learning that facilitates effective mate finding and, interestingly, a recent study reported a key role of specific neurons and their wiring in the emergence of this behavior [138]. Detailed reconstruction of synaptic connectivity revealed two newly discovered neurons to be incorporated in existing circuits during late development, with the newly added connectivity and accompanying change in network structure allowing the male worm to add sex-specific learning plasticity to its functional repertoire [138].

Comparative connectivity analyses between primate species have similarly argued for a relation between subtle differences in connectome wiring and divergent behavior [139]. For example, bonobos and chimpanzees share a recent common ancestor, but they show distinct social interactive behavior, with bonobos showing less aggression and higher social tolerance compared with chimpanzees (discussed in [139,140]). Comparative analysis of MRI-derived reconstructions of brain connectivity between the two species showed anatomical connectivity between the amygdala and anterior cingulate regions (a network of subcortical and cortical areas involved in the modulation of social and emotional behavior [141,142]) to be stronger in bonobos than in chimpanzees [139]. Furthermore, comparative MRI studies among macaques, chimpanzees, and humans [109] have linked elaboration of anatomical connectivity of the arcuate fasciculus connecting temporal and frontal cortical areas to the development of language and speech processing in humans (Figure 4) [109,143].

In addition to evidence for behaviorally related connectivity differences in specific circuits, comparative connectome examinations between primates and humans have further offered support for a role of variations in system-level connectivity patterns in the evolutionary emergence of specific advanced brain functions. Although studies have reported a general homology of connectome organization across macaque, chimpanzee, and human, there is evidence for significant species-specific differences in the wiring of parietal and medial prefrontal regions [67,144]. Comparative connectome analyses suggest that areas of the frontal cortex in humans exhibit a less central position compared with homologous areas in macaques and chimpanzees [67,104]. In addition, functional connectivity hubs in the human brain have been argued to exhibit a more spatially distributed organization compared with nonhuman primates [145], differences that may contribute to the increasing functional specialization of frontal cortical areas [104]. Potentially more, including more centrally connected, hubs in other parts of the cortex may confer higher robustness and bring greater support for the functional involvement of other areas in global processing [145]. These hypotheses are supported by recent comparisons between chimpanzees and humans, which revealed strong expansion of the precuneus in humans [146]. Indeed, the precuneus is one of the most central and most connected hub areas of the human brain [68,113]. These ideas parallel observations that variations in connectome organization and hub wiring relate to individual variation in cognitive processing in humans [119,147–151].

Concluding Remarks and Future Directions
The central idea of this review is that nervous systems of different species exhibit both common themes and important variations in connectome organization. We argue that connectome architecture may follow general principles of wiring, shaped by forces that minimize the expenditure of biological resources and forces that favor functionally important topological attributes that benefit efficient communication and global integration. Competition and trade-offs between these opposing forces can account for the ubiquity of connectomes that are generally parsimoniously wired, while allowing for features that can promote topological integration. We further argue that, within the envelope of wiring solutions that conform to these general principles, subtle variations in connectome organization support species-specific adaptations in behavior and cognitive functioning.

The ideas put forth in this review, in particular our hypotheses on adaptive changes in network attributes to form biological underpinnings of variations in brain function, of course need thorough empirical testing. Comparative studies that use uniform acquisition methodology to map (Box 1) and study (Box 2) connectomes across species are needed to rigorously map commonalities and differences of connectivity patterns. Furthermore, investments in studies that explicitly address the diversity of brain architectures are required to draw more secure inferences about the putatively universal principles shaping connectome evolution. This would necessitate examinations across a wider range of species. For example, besides the work of reconstructing connectomes of nematode, insect, avian, and mammalian species, as summarized in this review, comparative connectomics could be extended further to encompass the nervous systems of animals with a completely different body plan but that may nevertheless share universal attributes of connectome organization (Box 3). In addition to the collection of more comprehensive and consistent empirical data, this burgeoning field would benefit from computational modeling studies that simulate generative mechanisms behind cross-species network evolution and test the roles of network attributes with respect to functional diversity [117], dynamics [83,152], resilience, and integration [153,154].

By focusing on patterns of brain connectivity, comparative connectomics is subject to several important limitations. First, comparative connectomics is currently constrained by the divergent methodologies used to assess connectivity at different scales and in different species (e.g., electron microscopy, tract-tracing, and MRI; Box 1). This limitation underscores the importance of finding ways to reconstruct connectomes more uniformly across different nervous systems. Second, the architecture of a nervous system must be considered in the context of the overall structural anatomy, physiology, and ecological embedding of the organism [155]. Among many other factors, the body plan of an organism, the arrangement and physiology of its sensory apparatus, and the geometry of its musculoskeletal system, are important factors that shape, constrain, and enable behavioral (and cognitive) function, as well as the layout of the nervous system. The common themes of connectome organization highlighted in this review cut across vast differences in body plan and 'evolutionary complexity'. This suggests that the observed

Box 4. Comparative Analysis of Different Classes of Network

Going beyond the examination of shared topological features across nervous systems, the generalized mathematical language of graph theory also offers tools for the comparison of the organization of brain networks to other classes of network studied by different scientific disciplines. Many real-world systems operate as some sort of interaction or communication network, including, for example, social networks, gene regulatory networks, computer networks, and transportation networks. Similar to brain networks, many of these real-world networks display an efficient small-world organization, a pronounced community structure with densely connected modules, as well as the formation of hubs and rich clubs [30,37,170]. Going beyond the comparison of networks within the class of nervous systems, the field of 'comparative network analysis' examines commonalities and differences across a range of network classes.

A core concept in this context is that of network morphospace [156], which examines underlying morphological characteristics of networks by describing common and differentiating aspects across networks compared with the total space of possible networks. Applications of morphospace analysis include examinations of the capacity of brain networks to diffuse and route signals in comparison to social, gene regulatory, and email networks [171]. A complementary approach is based on applications of spectral graph theory [172], with the spectrum describing the multiset of eigenvalues of the **adjacency matrix** of a network. Networks showing overlapping spectra display common organizational and functional features. For example, spectral examinations have shown similar features between nervous systems and the organization of ecological networks, an observation that may provide clues to selection pressures with a role in the evolution of both systems [173].

Comparative network analysis allows for the examination of commonalities and differences between classes of network. In particular, it provides a powerful approach to answer the question of which topological network attributes are specific to nervous systems and which represent more universal properties of network organization [174]. As such, it provides a unique tool to examine which network properties are unique to brain networks, and which properties are shared with other types of communication and interaction network.

principles of wiring reflect universal features of efficient and effective network organization that cause common connectome patterns to be found across a range of species. Viewed from another perspective, they may define hard limitations to connectome formation that cannot easily be contravened by natural selection. The competitive pressures of parsimonious wiring and topological integration impose general trajectories and boundaries that define the space of network topologies that are geometrically and functionally possible. This space is embedded within a 'theoretical **morphospace**' of biological forms [156] and defines the realm within which connectome topologies can vary, at least in principle, and contribute to adaptations that support different behavioral and cognitive specialization. The size and shape of this common morphospace, how it constrains the possible behavioral and cognitive repertoire [157–159], and how it compares to the total morphospace of natural and human-engineered networks (Box 4) all remain important open questions (see Outstanding Questions).

We hope that, with growing access to connectome data from a range of species, comparative connectomics may become a useful addition to the spectrum of approaches aiming to account for cross-species commonalities and differences in brain structure and function.

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