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### A prototype symbolic model of canonical functional neuroanatomy of the motor system

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#### 9 Abstract

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Recent advances in bioinformatics have opened entire new avenues for organizing, integrating and retrieving neuroscientific data, in a digital, machine-processable format, which can be at the same time understood by humans, using ontological, symbolic data representations. Declarative information stored in ontological format can be perused and maintained by domain experts, interpreted by machines, and serve as basis for a multitude of decision support, computerized simulation, data mining, and teaching applications.

We have developed a prototype symbolic model of canonical neuroanatomy of the motor system. Our symbolic model is intended to support symbolic lookup, logical inference and mathematical modeling by integrating descriptive, qualitative and quantitative functional neuroanatomical knowledge. Furthermore, we show how our approach can be extended to modeling impaired brain connectivity in disease states, such as common movement disorders.

In developing our ontology, we adopted a disciplined modeling approach, relying on a set of declared principles, a high-level schema, Aristotelian definitions, and a frame-based authoring system. These features, along with the use of the Unified Medical Language System (UMLS) vocabulary, enable the alignment of our functional ontology with an existing comprehensive ontology of human anatomy, and thus allow for combining the structural and functional views of neuroanatomy for clinical decision support and neuroanatomy teaching applications.

Although the scope of our current prototype ontology is limited to a particular functional system in the brain, it may be possible to adapt this approach for modeling other brain functional systems as well.

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*Keywords:* Functional neuroanatomy; Ontology; Neural network; Motor system; Basal ganglia; Disease model; Parkinson's disease; Chorea; Hemiballism; Neural circuit

#### 29 1. Introduction

The brain is arguably the most complex organ of the human body, and our understanding of its structure and function is fragmentary. The past few decades have seen an enormous accumulation of neuroscientific data, making it impossible for any one individual to comprehend and assimlate more than a fraction of the available data. This fact becomes evident while performing a simple search of the literature databases, such as Medline. For instance, a recent search in the Medline database performed by the authors, entering the term "brain", yielded 1,003,745 entries (http:// www.ncbi.nlm.nih.gov/entrez, accessed on 01/25/2007).

Recent advances in bioinformatics have opened entire new avenues for organizing, integrating and retrieving neuroscientific data, in a digital format, which can be at the same time understood by both humans and machines, using ontological, symbolic data representations [1]. In addition to providing an understanding of the physical organization of the nervous system, neuroanatomy may also serve as a common frame of reference for organizing all types of neuroscience data [2].

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In a general sense, there are two complementary views 50 51 of neuroanatomy: (a) a structural view, concerned with shape, dimensions, spatial location and relationships, and 52 embryologic origin of neural structures, and (b) a func-53 tional view, dealing with functional (physiologic) relation-54 ships between entities assembled into neural functional 55 systems (connections between these entities via neural 56 57 pathways, physiologic actions-e.g. excitation or inhibition-of one entity on another exerted via neural path-58 ways); these entities often do not share a common 59 embryologic origin and may be spatially remote. The two 60 perspectives are by no means mutually exclusive. On the 61 contrary, they must be viewed as complementary, as both 62 are essential for problem solving, in the basic as well as 63 clinical neuroscience domain. 64

The Foundational Model of Anatomy (FMA), devel-65 oped by Rosse and colleagues, is an excellent example of 66 modeling the structural view of a biomedical domain [3]. 67 The FMA is a comprehensive domain ontology describing 68 the concepts and relationships that pertain to the structural 69 organization of the human body, including the nervous sys-70 tem, from the molecular to the macroscopic level. The 71 72 FMA has been successfully used for developing knowledge-based applications that rely on inference to support 73 clinical decision-making, such as reasoning about conse-74 quences of penetrating chest injuries [4-8]. 75

Structural information contained in the FMA enables 76 queries such as: "which structures are adjacent to or contin-77 uous with other structures?", or "what are the parts of a par-78 79 ticular anatomic structure?". However, structural information is insufficient to support logical inferences on 80 the functional consequences of anatomic lesions. 81

The entities in the FMA are grouped according to 82 83 embryologic origin and spatial adjacency criteria. Functional systems in the brain, on the other hand, include mul-84 tiple, often spatially remote, and embryologically unrelated 85 structures. For instance, from a structural point of view, 86 87 the striatum and globus pallidus are part of the telenceph-88 alon, the ventral anterior nucleus of thalamus and the subthalamic nucleus are part of the diencephalon, while 89 substantia nigra is part of the midbrain. From a functional 90 perspective, all these structures are part of a single subcor-91 tical neural network, controlling the initiation of voluntary 92 93 movement (Fig. 1).

The purpose of the present study was to develop a pro-94 totype symbolic model of canonical functional neuroanat-95 omy of the motor system in an ontology, representing 96 normal and disease states, based on a set of declared prin-97 ciples, Aristotelian definitions, and a high-level schema. We 98 99 chose the motor system because it displays little anatomic variability and its function is better characterized than that 100 of other, more complex functional systems in the brain. 101 Furthermore, the motor system is involved in a host of 102 pathologic processes with high impact on public health, 103 104 such as movement disorders.

Our current prototype ontology, based on the principles 105 of functional connectionism, is limited to a single func-106

tional system. However, since functional connectionism 107 applies to the entire brain, it may be possible to extend 108 the modeling approach we describe in this paper to other 109 brain functional systems as well. As we will show in the fol-110 lowing sections, the ontological modeling approach we are 111 proposing can be employed to describe normal, as well as 112 pathologic states. 113

Symbolic models of functional neuroanatomy, alone or 114 in combination with MRI-based digital brain atlases, could 115 open the way for developing knowledge-based applications 116 for clinical decision support (e.g. surgical planning applica-117 tions capable to identify potential targets for functional ste-118 reotactic surgery), and computer applications for 119 neuroanatomy teaching. 120

#### 2. Materials and methods

First, we extracted the relevant functional neuroanatomical information from authoritative neuroscience textbooks [9-11]. In addition, we collected information on movement disorders that result from impaired connectivity between key anatomical components of the motor 126 system.

Next, we created an ontology of functional anatomy of the motor system, based on the anatomic knowledge we 129 extracted in the previous step. The ontology was created 130 using a disciplined modeling approach, inspired by that 131 adopted by the developers of the Foundational Model of 132 Anatomy [3]. Our ontology was implemented using the 133 Protégé suite of tools (http://protege.stanford.edu), a frame-based ontology editor.

In this section, we provide a brief overview of the functional anatomy of the motor system. We then describe the theoretical framework and the principles we employed for developing our prototype functional ontology. 139

#### 2.1. Functional anatomy of the motor system

Older classifications used to make a distinction between 141 the "pyramidal" and "extrapyramidal" motor system. 142 Modern neuroscience has outgrown this rather simplistic 143 view, and it is currently well established that the concerted 144 action of both these strongly interconnected groups of neu-145 ral circuits is required for normal motor function.

#### 2.1.1. The "Pyramidal" system

This functional system is responsible for voluntary movement. From a functional perspective, it is composed of two major types of neurons: upper motor neurons and lower motor neurons.

The upper motor neurons (UMN) are represented by the giant pyramidal neurons (Betz's cells), located in the internal pyramidal layer (layer V) of the primary motor cortex (precentral gyrus, motor strip).

The lower motor neurons (LMN) are located in the brainstem (motor nuclei of cranial nerves), and in the anterior horn of the spinal cord gray matter.

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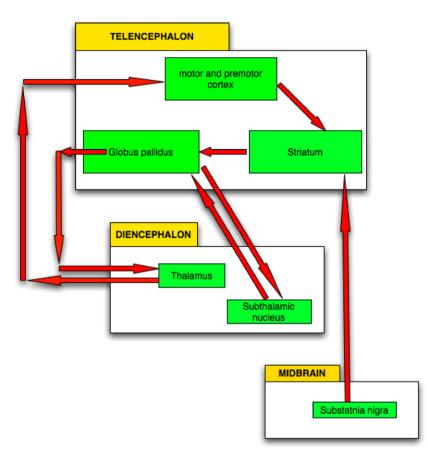


Fig. 1.

The precentral gyrus is located in the posterior frontal 159 lobe. The central sulcus separates it from the postcentral 160 gyrus (site of the primary somatosensory cortex). Medially, 161 it is contiguous with the paracentral lobule, while inferiorly 162 it is separated from the superior temporal gyrus by the lat-163 eral fissure (Sylvius). In the primary motor cortex, the dif-164 165 ferent regions of the body are represented in a somatotopical fashion, with the foot and leg area located 166 on the medial aspect of the cerebral hemisphere, followed 167 by the trunk, upper limb and face areas, from medial to lat-168 eral ("motor homunculus"). The surface of the cortical rep-169 resentation is not proportional with the size of the 170 171 respective body region, but rather with the complexity of movements performed by a particular part. 172

The axons of the upper motor neurons form the cor-173 ticospinal (pyramidal) tract, which descends through the 174 corona radiata and converge in the posterior limb of 175 176 the internal capsule. After passing the internal capsule, these axons continue their descent through the ventral 177 brainstem. At the brainstem level, some of these axons 178 cross-over to the contralateral side and synapse with 179 lower motor neurons located in the motor nuclei of the 180 181 cranial nerves. From a clinical perspective, it is impor-182 tant to point out that the dorsal part of the facial nerve nucleus receives both contralateral and ipsilateral cortical 183 input. 184

In the ventral medulla oblongata, the corticospinal fibers 185 converge into two compact tracts, prominently visible at the 186 surface (pyramids). At the junction between medulla and 187 spinal cord, most (80-90%) of these axons cross-over to the 188 contralateral side (pyramidal decussation), and continue 189 their descent in the lateral spinal cord (lateral corticospinal 190 tract). The uncrossed axons continue their descent in the 191 anterior spinal cord, as the anterior corticospinal cord. This smaller contingent of fibers cross-over at the spinal cord level, shortly before reaching their target lower motor neurons. Most corticospinal axons synapse with their target lower motor neurons via spinal interneurons, while a smaller fiber contingent synapses directly on lower motor neurons.

#### 2.1.2. The "Extrapyramidal" system (motor initiation system)

The motor initiation system consists of a family of parallel circuits linking subcortical structures with the motor cortex. Its principal components are the basal ganglia (striatum, globus pallidus), the subthalamic nucleus, the ventral anterior nucleus of thalamus, the substantia nigra and the motor cortex.

Input from the motor cortex (glutaminergic) reaches the basal ganglia via the striatum (caudate nucleus and putamen). The basal ganglia process this information and project back to the motor cortex, via the internal pallidal segment

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and ventral anterior thalamic nucleus. The only output from
the basal ganglia is inhibitory, and it originates in the
GABA-ergic medium-spiny neurons of the internal pallidal
segment.

The state of activity in the basal ganglia is regulated via 214 two dopaminergic projection systems, originating in the 215 216 substantia nigra pars compacta: the direct and the indirect projection systems (a.k.a. direct and indirect pathways). 217 The facilitating effects on movement of the direct projec-218 tion system are mediated by D1-dopamine receptors. The 219 indirect projection system has an inhibitory effect on move-220 ment, mediated by D2-dopamine receptors (Fig. 2). 221

## 222 2.2. Diseases of the "Extrapyramidal" system (motor223 initiation system)

Lesions of the basal ganglia frequently result in movement disorders:

*Parkinson's disease* is a classical example of *hypokinetic*movement disorder. Clinically, this condition is characterized by impaired initiation of movement (*akinesia*), reduced
velocity and amplitude of movement (*bradykinesia*), and
resting tremor and increased muscle tone (*rigidity*).

The prevalence of idiopathic Parkinson's disease is estimated at 128–168 cases per 100,000 [12]. There is a dramatic increase in Parkinson's cases with increasing age. About 1% of the population aged 50–64 years suffers from Parkinson's disease. This rate increases to 14.9% in the 65–74 years age 235 group and 52.4% of individuals over 85 years old [13]. 236

Pathophysiologically, degeneration of dopamine producing cells in the substantia nigra pars compacta leads to a decrease in the activity of the direct basal ganglia pathway relative to the indirect pathway activity. This, in turn, results in an increased inhibitory output from the internal pallidal segment (globus pallidus pars interna, GP<sub>i</sub>). 242

The therapeutic approaches to Parkinson's disease 243 include pharmacologic agents (dopamine agonists), and 244 stereotactic functional surgery, such as ablation of the subthalamic nucleus or of the internal pallidal segment (GP<sub>i</sub>). 246

Hyperkinetic movement disorders are characterized by 247 excessive, uncontrollable motor activity resulting in abnor-248 mal, involuntary movements of the extremities, head and 249 trunk, and decreased muscle tone (hypotonia). Depending 250 on the anatomic structure affected, the involuntary move-251 ments can take the form of writhing movements of the 252 arms and hands (athetosis), brief, non-rhythmic move-253 ments spreading from one muscle group to the next (cho-254 *rea*), violent, large amplitude movements of the proximal 255 limbs (ballism). 256

*Huntington's disease* is a typical example of hyperkinetic movement disorder. It is an inherited, autosomal-dominant disorder with complete penetrance. Its prevalence in the United States is estimated at about five cases per 100,000. Huntington's disease is a slowly progressive condition, 261

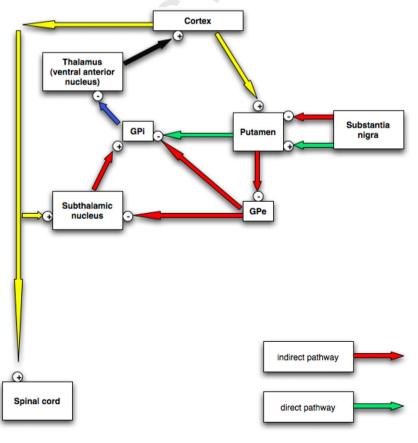


Fig. 2.

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leading to death of the affected individual within 15–20years from symptom onset [14].

The responsible gene is located on chromosome 4. This gene encodes a large protein, *huntingtin*. The function of huntingtin has yet to be characterized. It is theorized, however that huntingtin plays a role in triggering apoptosis (programmed cell death) in certain neuron populations in the central nervous system, including the basal ganglia.

As opposed to Parkinson's disease, the inhibitory internal pallidal output is decreased in hyperkinetic disorders, such as chorea and hemiballism.

Disease-induced alterations of the basal ganglia neural circuits in Parkinson's disease, chorea and hemiballism are presented in Fig. 3.

#### 276 2.3. Ontological modeling

In developing our ontology, we adopted a disciplined
modeling approach, as described by Rosse and colleagues
in their seminal work on the Foundational Model of Anatomy (FMA) [3]. This development was accomplished in a

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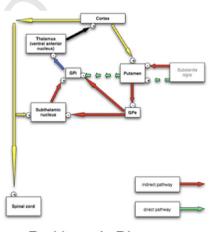


four step process: (1) establishing the appropriate theoretical 281 framework and identification of the biological concepts, 282 attributes and relationships that will form the building 283 blocks of the symbolic representation; (2) defining a rational 284 modeling approach, the elements of the symbolic model 285 structure (high-level schema), and the set of properties and 286 modeling rules to be employed; (3) identification of an 287 appropriate software authoring system, that will not only 288 allow, but also enforce the modeling rules; (4) evaluation 289 of the symbolic model. Our ontology was implemented using 290 the Protégé suite of tools (http://protege.stanford. edu). 291

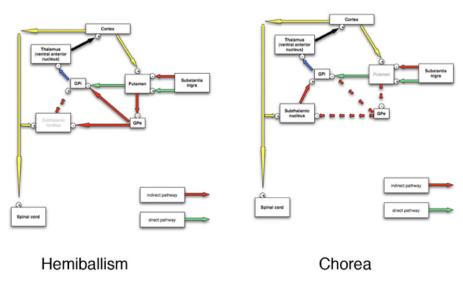
#### 2.3.1. Theoretical framework

The functional organization principles of "cellular connectionism" (Wernicke, Sherrington, Cajal) provide the biological foundation for our symbolic model:

 Neuron doctrine: the elementary signaling unit in the nervous system is the neuron. Each neuron is a distinctive cell with distinctive processes (multiple dendrites, one axon).



Parkinson's Disease



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- 300 (2) Dynamic polarization principle: the signal (action potential) flow in the neuron is unidirectional, from 301 the dendrites to the cell body to the axon. 302
- (3) Connectional specificity principle: each neuron is 303 304 connected with certain other neurons (target cells). but not with others. The connections between neu-305 306 rons provide the physical basis for signal processing, i.e. for brain function. Neurons are arranged in spe-307 cific functional groups or neural networks (e.g. pri-308 mary motor and somatosensory cortex, subcortical 309 nuclei). Each neural network is concerned with spe-310 cific elementary signal processing operations, as part 311 of a specific neurologic function. These groups of 312 neurons are linked via serial and parallel connec-313 tions (neural pathways). Specific brain functions 314 are divisible into elementary signal processing oper-315 ations, performed by specialized neural networks. 316 Consequently, a brain functional system (e.g. motor 317 system) can be viewed as a collection of specialized 318 neural networks. 319
- (4) The signals are transmitted from one neuron to the 320 321 other via synapses. This process is mediated by chemical *neurotransmitters*. Neurotransmitter release may 322 lead to depolarization (excitation) or hyperpolarization 323 (inhibition) of the postsynaptic membrane. For 324 instance, glutamate is the most common excitatory 325 neurotransmitter in the nervous system, while gaba-326 amino-butyric acid (GABA) is the most common 327 inhibitory neurotransmitter. 328
- (5) The action potential is the universal mechanism for 329 signal transmission in the nervous system. The speci-330 ficity of a given signal (e.g. motor, sensory) is deter-331 mined by the neural network(s) it travels through. 332
  - (6) Somatotopic representation: individual parts of the body are represented at specific sites in the motor and somatosensory systems.

337 In summary, elementary signal processing operations are the building blocks of brain function, and neural networks 338 provide the anatomical basis for these operations, i.e. they 339 represent the anatomo-functional units of the nervous sys-340 tem. The defining feature of a network is the link (connec-341 tion) between its nodes. As detailed in the following 342 343 sections, the *neural network* concept plays a central role in our symbolic representation. 344

#### 2.3.2. Disciplined modeling 345

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One important requirement for ontological models is that 346 347 they rigorously and consistently conform to the domain they are designed to model [3,15–19]. A common pitfall when 348 developing symbolic models of anatomy is mixing spatial 349 and functional information in the same hierarchy [3,15,20]. 350 In order to avoid this problem, the scope of our ontology 351 352 is strictly limited to representing functional anatomy, specif-353 ically the functional anatomy of motor neural networks (abstraction level principle) [3]. Consequently, our design 354 plan conforms to a functional context (unified context princi-355

ple) [3], and the entities of the symbolic model, along with 356 their attributes and relationships are defined in respect to 357 their functional role, and not to spatial relationships or 358 embryologic origin criteria (definition principle, relationship *constraint principle*) [3]. 360

Defining the dominant concept (dominant concept principle) and the organizational unit (organizational unit principle) [3] are crucial steps in ontology design. The dominant concept serves as reference for defining all classes (concepts) of the ontology.

As discussed in the previous section, the neural network is the anatomo-functional unit of the nervous system. Hence, in order to represent brain function, we identify the "neural network" concept as the dominant concept of our ontology. We define this concept as follows:

Neural network is a biological entity consisting of neurons and their processes, connected by synapses in a specific, genetically determined pattern, which performs specific elementary signal processing operations, and constitutes the functional organization unit of the nervous system.

Microscopic neuronal networks are grouped together into functionally specialized, gray matter structures (subcortical nuclei, cortical areas). Gray matter structures are connected in a specific, genetically determined pattern, via long axonal processes (neural pathways, fiber tracts), into functionally specialized macroscopic neural networks (e.g. networks of subcortical nuclei).

The organizational unit of our symbolic model is the neuron, since all elements (classes) of our symbolic representation can be derived from neurons or parts of neurons. At its current stage of development, our ontology represents anatomic structures at macroscopic scale.

In accordance with the content constraint principle [3], 388 the largest entity that can be modeled is a functional system (collection of macroscopic neural networks performing elementary signal processing operations as part of a specific brain function, e.g. motor function), and the smallest entity that could be represented is a biological molecule con-393 cerned with synaptic transmission in the nervous system (neurotransmitter).

#### 2.3.3. High-level schema

The general structure (high-level schema) of our ontol-397 ogy consists of the following elements: anatomo-functional 398 abstraction, neural network taxonomy, and neural network 399 component taxonomies (nodes taxonomy, connections 400 taxonomy). 401

2.3.3.1. Anatomo-functional abstraction. The anatomofunctional abstraction describes the properties and relationships between the entities of the functional ontology.

There are two types of relationships that need to be represented in an ontology of functional anatomy: meronymic (partitive) and functional (physiologic) relationships.

(1) Meronymic ("is part of", "has parts") relationships. Each neural network is part of a functional system (e.g.

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411 motor system, sensory systems), and *has parts*, i.e. network
412 nodes (subcortical gray matter nuclei/cortical areas), and
413 network connections (white matter fiber tracts, neural
414 pathways), respectively.

415 Examples:

# 416 "The motor initiation neural network <u>is</u> 417 <u>part of</u> the motor system."

418 "The motor system <u>has parts</u>: the voluntary 419 motor neural network and motor feedback 420 neural networks."

421 "Globus pallidus *is part of* the motor initiation
422 network, and thus it *is part of* the motor system."

423 White matter structures (white matter fiber tracts, neural 424 pathways) are also parts of a neural network.

(2) Functional (physiologic) relationships. Neural net-425 works provide the anatomic basis for neurologic function. 426 Typically, the physiologic influence of one network node 427 on another is conveyed via neural pathways (neural net-428 work connections). A pathway can only have one origin 429 (cortical area/subcortical nucleus) and one target (cortical 430 431 area/subcortical nucleus), and it conveys one type of phys-432 iologic effect to the target (excitation, or inhibition). The signal propagation is unidirectional, from the origin to 433 434 the target (dynamic polarization).

435 Example:

436

437 "The internal pallidal segment exerts an
438 <u>inhibitory</u> influence on the ventral anterior thalamic
439 nucleus; this inhibitory influence is conveyed via the pal440 lido-thalamic pathway."

Since neural pathways convey the physiologic action of
the origin node to the target node, they can be represented
as verbs in the syntactic structure of the ontology, while the
origin node can be represented as subject, and the target
node as object (Fig. 4).

2.3.3.2. Neural network taxonomy and neural network
element taxonomies. Since neural network elements cannot
be represented in the form of "is a"—subclasses of the generic neural network class, we accomplished the goal of creating a comprehensive class-subsumption hierarchy in an
operational manner.

First, we created a template taxonomy of neural networks, with "neural network" as its root class (Fig. 5).

Specific neural networks belonging to particular func-454 tional systems (e.g. motor neural network, somatosensory 455 456 neural network) are represented as subclasses of the root class. We further elaborated the "motor neural network" 457 class, by adding the following subclasses: "voluntary move-458 ment neural network", and "motor feedback neural net-459 work". The latter subclass also has two subclasses: 460 "motor initiation neural network" and "motor modulation 461 462 neural network". The right and left motor initiation neural networks are represented as instances of the "motor initia-463 tion neural network" subclass. 464

Second, using the template neural network taxonomy, we created separate taxonomies for each network element type (nodes and connections, respectively).

The classes in the nodes taxonomy were given the following properties: "is part of" (neural network), "*output*", "*input*". The "input/output" slots can only be filled with instances of classes from the network connections taxonomy.

One important feature of neural networks participating in the same brain function is the fact that they share some of their nodes. For instance, the motor cortex is at the same time part of the voluntary motor network and of the motor initiation network. The basal ganglia send output back to the motor cortex, where the initial motor command originates (see Fig. 1). We accounted for this feature by creating a "shared node" subclass of the "motor neural network node" class.

The classes in the connections taxonomy were given the following properties: "origin", "target", and "physiologic effect" (excitationlinhibition). The "origin" and "target" slots can only be filled with class instances from the nodes taxonomy.

In the final step, we combined the two network element taxonomies into a comprehensive functional ontology. In this design, the classes of the nodes taxonomy fill slots of classes in the connections taxonomy (Figs. 6 and 7).

2.3.3.3. Representing somatotopy and pathway cross-over. 492 Specific sites of the motor system control the motor activity 493 of different body regions (head, neck, trunk, limbs). This is 494 called somatotopic representation. For example, the corti-495 cal area located on the medial aspect of the precentral 496 gyrus controls voluntary movement of the contralateral 497 leg, while voluntary movement of the face muscles is con-498 trolled by an area located in the inferior portion of the pre-499 central gyrus. In order to represent somatotopy in our 500 ontology, we added instances to the "neural network 501 node" and "neural network connection" classes, according 502 to the body regions they represent. For example, we added 503 instances to the "motor cortex" class corresponding to leg, 504 trunk, upper limb, lower limb, etc. 505 506

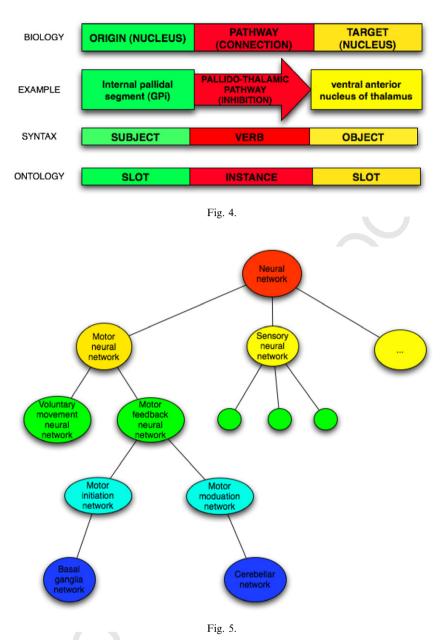
Due to the fact that the corticospinal tract crosses over to the contralateral side, voluntary movement of the two body halves is controlled by the contralateral motor cortex. Pathway cross-over is represented in our ontology in the following manner:

origin	node	( <u>right</u>	side) —	→pathway	512
(instanc	e of	neural	network	connec-	513
tion) $\rightarrow$ target node ( <u>left</u> side).					514

#### 3. Results

Using the approach presented in the previous sections, 516 we developed of a prototype ontology describing the basic 517

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functional organization of the motor system. There is a 518 one-to-one mapping from each instance in the ontology 519 and a gray matter structure in the brain involved with 520 motor function (e.g. right and left primary motor cortex, 521 right and left putamen, globus pallidus pars interna and 522 externa, etc.). For example, the internal pallidal segment 523 is represented as an instance of the NeuralNetworkNode 524 class. Furthermore, the connections (neural pathways) 525 between gray matter structures, each having the appropri-526 ate attributes to specify their functional action on the target 527 network node (gray matter structure), i.e. excitatory or 528 inhibitory influence, are also represented in our ontology, 529 as instances of the NeuralNetworkConnection class. The 530 531 ontological representation of the different neural networks that compose the motor system is accomplished in an oper-532 ational manner, i.e. the network is reconstructed from its 533 elements: class instances of the neural network nodes tax-534

onomy fill "target" slots of classes in the neural network connections taxonomy.

Our modeling approach is suitable not only for representing neural connectivity in the normal brain. It can be easily extended to represent impaired brain connectivity as well.

We used the prototype ontology described above to create a symbolic, computable model of neural connectiv-542 ity and function of the motor initiation network in the 543 normal brain (Fig. 8B). While this ontology-based sym-544 bolic model has a very similar appearance to the original 545 graphical representation from which it was derived (Figs. 546 1 and 8A), the entire model is a computable representa-547 tion. Each node and connection in the diagram represents 548 an object in the ontology. For example, the arc connect-549 ing thalamus and cortex is an instance in the ontology 550 in which the thalamus is linked with the motor cortex 551

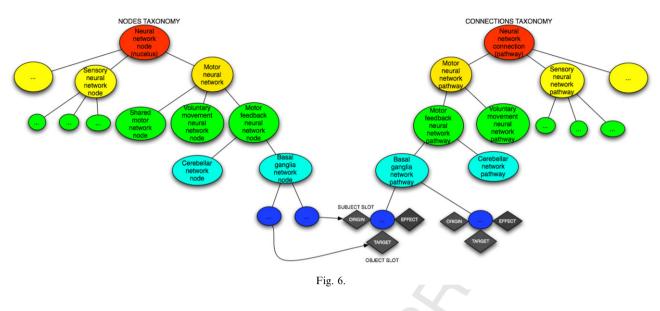
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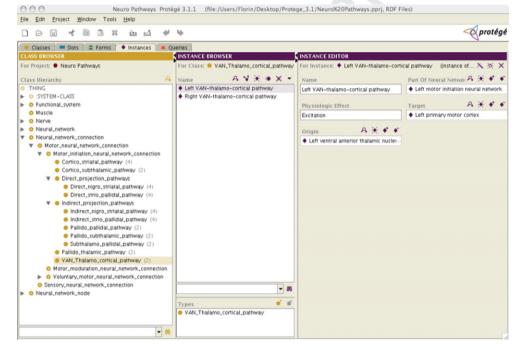
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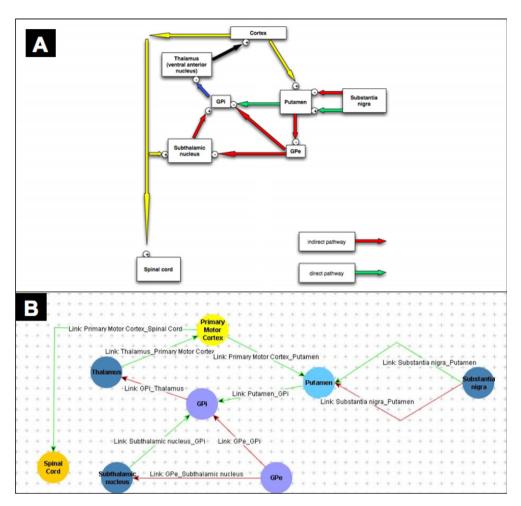


via an excitatory connection. This instance contains information about the structure from which it originates and the target structure to which it projects, as well as the type of physiologic influence it exerts on the target. Likewise, other connection objects are instances of an "inhibitory connection" class, meaning that they inhibit nodes to which they connect.

The entire model is computable, and by traversing all links, it is possible to compute the net excitation or inhibition of every anatomic structure.

To demonstrate the extensibility of our approach for representing pathologic conditions, we created a second model representing impaired neural connectivity in Parkinson's disease, by modifying the normal model. We replaced two arcs in the normal model with arcs that 566 are of type "impaired excitatory neural connection". 567 These arcs represent neural connections that produce less 568 excitatory output than normal. Accordingly, the ontolog-569 ical representation of the Parkinson's disease model per-570 mits us to assess the consequences of the impaired 571 connectivity-that there will be increased inhibitory 572 activity in the internal pallidal segment (GP<sub>i</sub>), and conse-573 quently increased inhibition of the motor cortex (Fig. 9). 574 In a similar manner, we could model the consequences of 575 impaired connectivity in hyperkinetic movement disor-576 ders, such as Chorea and Hemiballism by replacing links 577 with the appropriate functional types according to the 578 pathology of the disorder. 579

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The net inhibitory effect of the internal pallidal segment 580 on the motor cortex, via the anterior thalamic nucleus, is 581 determined by the ratio between activation levels of the 582 direct vs. indirect pathway  $(A_{DP}: A_{IDP})$  [21]. Voluntary 583 584 movement can only be initiated when the activation level of the direct pathway is greater than that of the indirect 585 pathway. In hypokinetic movement disorders, such as Par-586 kinson's disease, the  $A_{DP}$ :  $A_{IDP}$  ratio is lower than normal, 587 whereas in hyperkinetic movement disorders (Chorea, 588 589 Hemiballism), this ratio is higher than normal.

By attributing arbitrary strength values to the different 590 network connections (positive for excitatory connections, 591 592 negative for inhibitory connections), it is possible to create a computer reasoning application that computes the 593 net excitation levels of the motor cortex under normal 594 and pathologic circumstances, because our ontology-595 based model of neural connectivity is machine-process-596 able and net excitation of all nodes can thus be com-597 puted (Figs. 8 and 9). For example, in the Parkinson's 598 disease model, the ontological representation of the func-599 600 tional aspects of neural network connections would permit a computer reasoning service to evaluate the net 601 activation in the different nodes of the MotorInitia-602 tionNeuralNetwork and conclude that there is net 603

inhibition of the PrimaryMotorCortex node. Accordingly, the value of creating our ontological representation 605 of the functional organization of the motor system is to 606 make the anatomic and functional aspects of neural 607 structures accessible to intelligent computer reasoning 608 services. 609

Such reasoning services, combined with patient-specific 610 imaging-based brain atlases, may be used in creating deci-611 sion support applications to help surgical planning and 612 personalized patient care. Image-based, geometric models 613 of brain anatomy provide spatially accurate, implicit repre-614 sentations of brain structure. However, they lack explicit 615 knowledge about their contents, such as the functional role 616 of the anatomic structures they represent or the functional 617 consequences of pathologic changes. By combining the 618 explicit functional knowledge provided by ontology-based 619 models with image-based geometric models of brain anat-620 omy, it may be possible to develop surgical planning appli-621 cations designed to predict consequences of injuries to 622 brain structures resulting from particular surgical 623 approaches, or to support identification of appropriate tar-624 gets for stereotactic functional surgery in movement disor-625 ders, that can be highlighted on patient-specific image 626 datasets. 627



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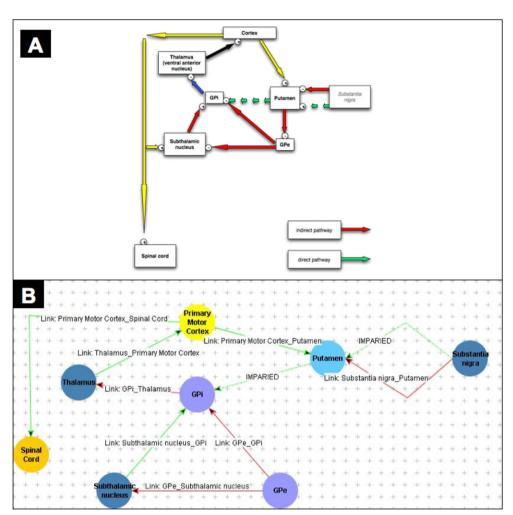
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#### 628 **4. Discussion and conclusion**

In this paper, we introduced an ontological modeling 629 630 approach of functional neuroanatomy, and presented a prototype ontology of canonical functional anatomy of 631 the motor system. Our ontology-based model is intended 632 to support symbolic lookup, logical inference and mathe-633 matical modeling by integrating descriptive, qualitative 634 635 and quantitative functional neuroanatomical knowledge. We have shown that our approach permits us to generate 636 symbolic models of impaired brain connectivity under 637 638 pathologic conditions, such as movement disorders. Our methods also provide a computational framework in which 639 to create applications that can reason about the functional 640 consequences of brain injuries. 641

Our functional ontology shares several important fea-642 643 tures with the Foundational Model of Anatomy (FMA) [3]: common vocabulary, common modeling principles, 644 645 and a common modeling platform (Protégé). Aside from 646 the limited coverage of our ontology compared to the FMA, one essential difference is the fact that these ontologies 647 provide two different, but complementary views of neuro-648 anatomy. While the FMA describes the spatial organization 649

of the nervous system, our ontology describes its *functional*650organization. The structural knowledge of the FMA enables651automated identification of anatomic structures that may be652affected by certain injuries with a given spatial distribution.653However, structural knowledge is not sufficient to provide654an understanding of how the neural structures work, or to655predict functional consequences of injuries.656

Recently, parameterized models have been designed for specific applications, such as simulating the effects of deep brain stimulation (DBS) on the activity of basal ganglia [22,23]. Such parameterized models incorporate actual clinical and experimental observations, and use systems of partial differential equations to describe temporal variations of physiologic signals in a quantitative manner. These parameterized models are highly specialized applications developed with the purpose of solving a particular clinical problem. They lack explicit declarative anatomic information about the components of which they are comprised.

To our knowledge, no ontology of functional neuroanatomy has been developed to date. As opposed to parametrized models, domain ontologies, are repositories of coherent, explicit knowledge, stored in a format understandable by both humans and computers. They are not

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intended as end-user applications or designed for solving a 673 specific problem. They represent generalizable and reusable 674 sources of knowledge about a particular domain. Ontolo-675 gies of functional neuroanatomy enable qualitative reason-676 ing about functional consequences of abnormalities and 677 can serve as basis for a multitude of specific applications, 678 679 including parameterized modeling of neurologic function. Furthermore, they are extensible, and can be mapped to 680 structural ontologies, such as the FMA, as well as to med-681 ical images. 682

683 One limitation of our ontology is that it currently covers 684 only large macroscopic neural components. We are cur-685 rently working on extending this representation to more 686 granular levels of anatomic and functional detail.

Another limitation of our prototype ontology is the fact 687 that it currently covers the narrow domain of functional 688 organization and abnormal neural connectivity of the 689 motor system. While our ontology is potentially useful to 690 enable different types of intelligent applications, it may 691 not be able to tackle a broader range of reasoning applica-692 tions beyond the scope of our focused domain. However, 693 since the principles of functional connectionism, that lay 694 695 at the foundation of our ontological representation, apply to the entire brain, it may be possible to extend the scope of 696 our ontological representation to incorporate other func-697 tional systems as well, with the final goal of creating a 698 domain ontology of functional neuroanatomy. 699

In summary, in this preliminary report, we have shown 700 that functional neuroanatomical knowledge about the 701 motor system can be represented using an ontology, which 702 can be exploited by computer reasoning applications. The 703 declarative knowledge encoded in our ontology can be per-704 used and maintained by domain experts, interpreted by 705 machines, and serve as basis for a multitude of decision sup-706 port, computerized modeling, and teaching applications. 707

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