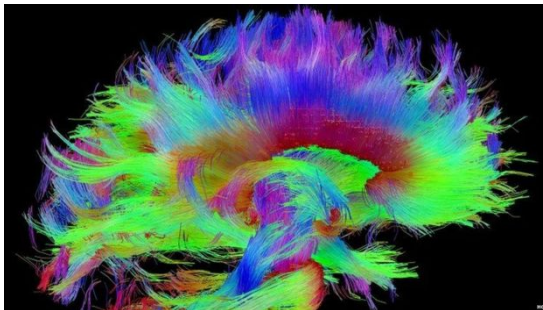


CS 4884: Brain Graphs

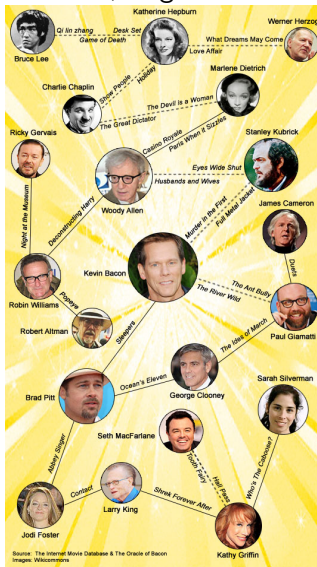
T. M. Murali

January 30, 2020



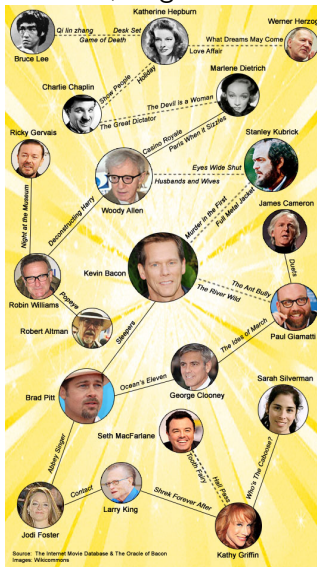
Creating Graphs

Node \equiv , Edge \equiv



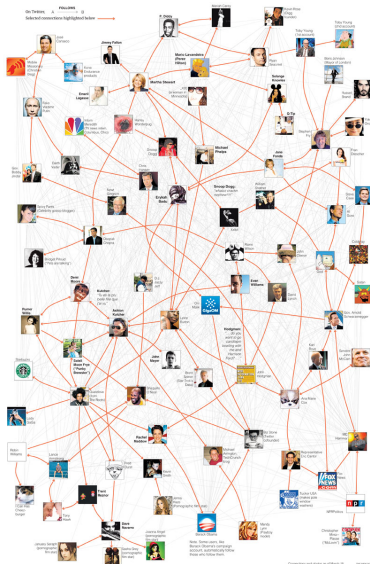
Creating Graphs

Node \equiv Person, Edge \equiv In same movie



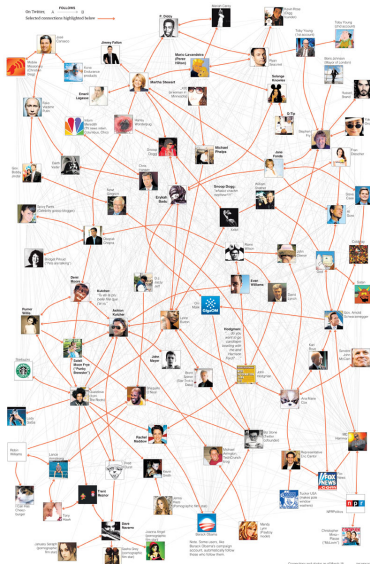
Creating Graphs

Node \equiv , Edge \equiv



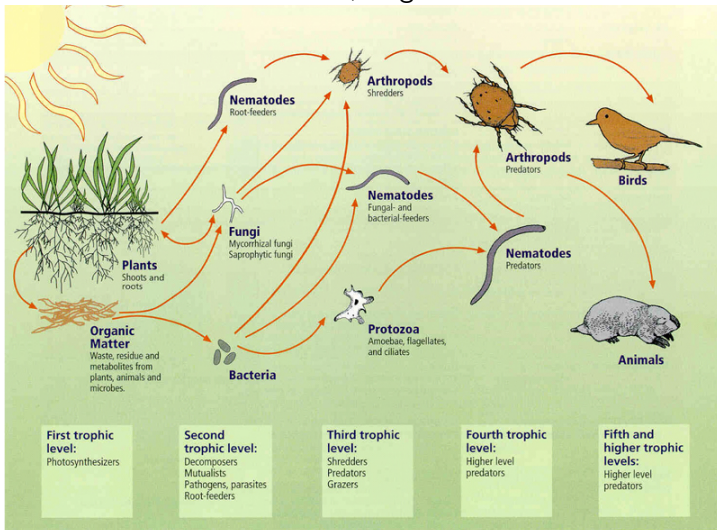
Creating Graphs

Node \equiv Person, Edge \equiv Follows on Twitter



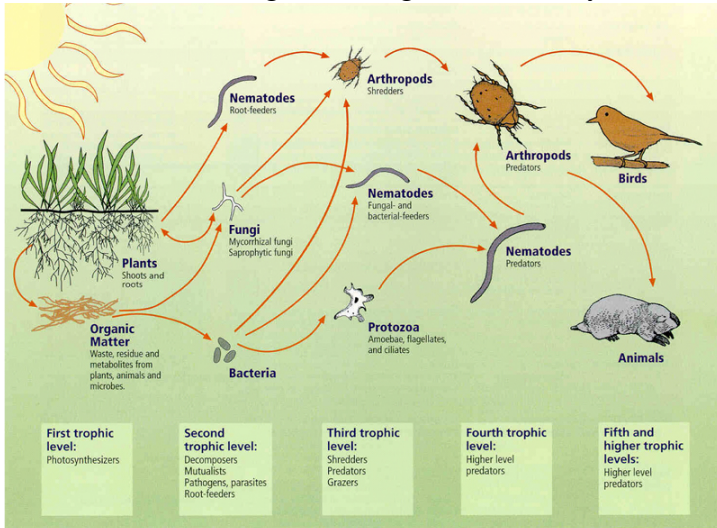
Creating Graphs

Node \equiv , Edge \equiv



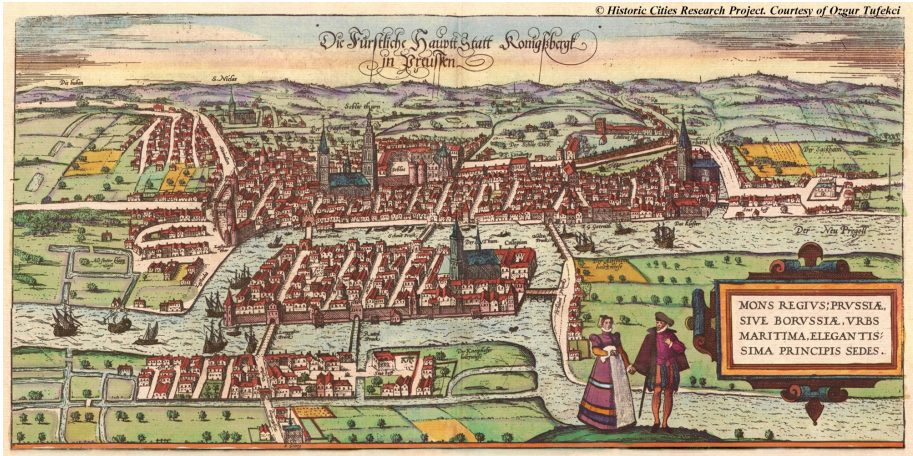
Creating Graphs

Node \equiv Organism, Edge \equiv Is eaten by



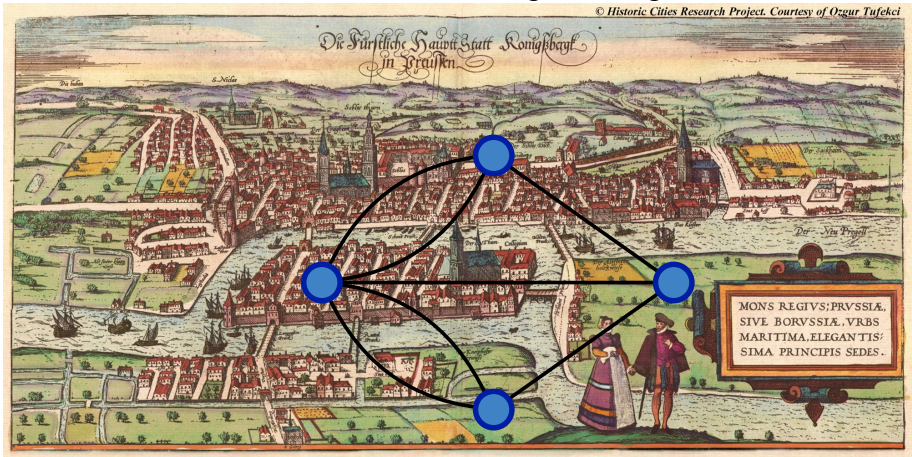
Creating Graphs

Node \equiv , Edge \equiv



Creating Graphs

Node \equiv Land mass, Edge \equiv Bridge



← from Torgersen Hall, 620 Drillfield Dr, Blacksburg, VA 24061
to Blacksburg Health & Fitness at The Weight Club, 801 University City Blvd, Blacksburg, VA 24061

7 min (1.3 miles)

via Stanger St and Prices Fork Rd
Fastest route, despite the usual traffic



Torgersen Hall

620 Drillfield Dr, Blacksburg, VA 24061

↑ Head northwest on Stanger St

174 ft

➔ Turn right to stay on Stanger St

0.4 mi

➔ Turn left onto Prices Fork Rd

0.5 mi

➔ Turn right onto Old Glade Rd

0.2 mi

➔ Turn right onto Glade Rd

0.1 mi

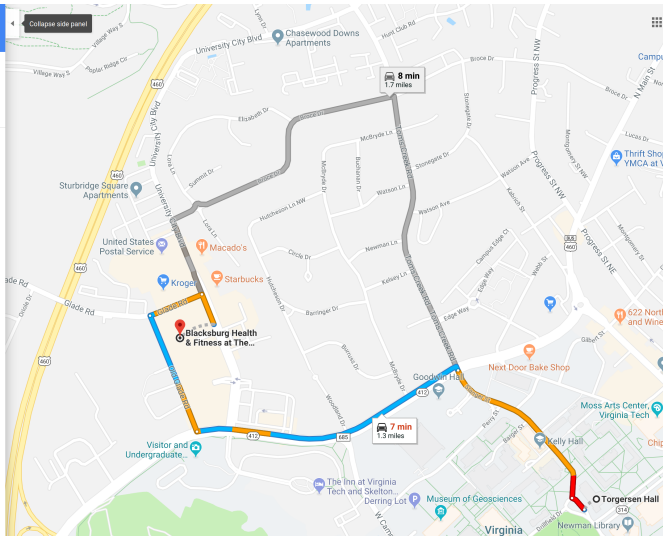
➔ Turn right onto University City Blvd

325 ft

Blacksburg Health & Fitness at The Weight Club

801 University City Blvd #6, Blacksburg, VA 24060

These directions are for planning purposes only. You may find that construction projects, traffic, weather, or other events may cause conditions to differ from the map results, and you should plan your route accordingly. You must obey all signs or notices regarding your route.



Node \equiv

, Edge \equiv

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0.1 mi

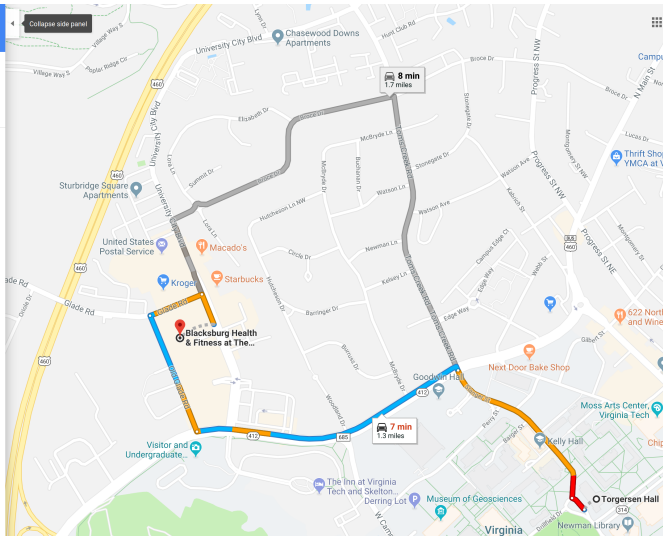
➤ Turn right onto University City Blvd

325 ft

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These directions are for planning purposes only. You may find that construction projects, traffic, weather, or other events may cause conditions to differ from the map results, and you should plan your route accordingly. You must obey all signs or notices regarding your route.



Node \equiv Intersection/Fork, Edge \equiv Street segment

← from Torgersen Hall, 620 Drillfield Dr, Blacksburg, VA 24061
to Blacksburg Health & Fitness at The Weight Club, 801 University City Blvd, Blacksburg, VA 24061

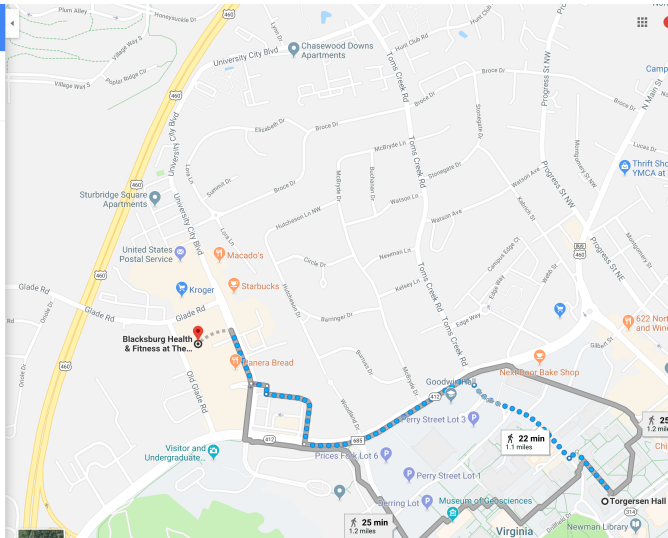
22 min (1.1 miles)
via Prices Fork Rd
Mostly flat

Use caution—walking directions may not always reflect real-world conditions

Torgersen Hall
620 Drillfield Dr, Blacksburg, VA 24061

- ↑ Head northwest
0.1 mi
- ↩ Turn left toward Stanger St
105 ft
- ↪ Turn right onto Stanger St
0.2 mi
- ↩ Turn left onto Prices Fork Rd
0.4 mi
- ↪ Turn right toward University City Blvd
0.2 mi
- ↪ Turn right toward University City Blvd
89 ft
- ↩ Turn left toward University City Blvd
167 ft
- ↪ Turn right onto University City Blvd
0.1 mi

Blacksburg Health & Fitness at The Weight Club
801 University City Blvd #6, Blacksburg, VA 24061



Node ≡

, Edge ≡

← from Torgersen Hall, 620 Drillfield Dr, Blacksburg, VA 24061
to Seattle's Best Coffee, Burke Johnson Student Cent...

7 min (0.4 mile)

via unnamed roads
Mostly flat



Torgersen Hall

620 Drillfield Dr, Blacksburg, VA 24061

↑ Head northwest

436 ft

↙ Turn left

138 ft

↙ Turn left

43 ft

↘ Turn right

0.1 mi

↘ Turn right

23 ft

↙ Turn left

266 ft

↘ Turn right

266 ft

↙ Turn left

23 ft

↘ Turn right

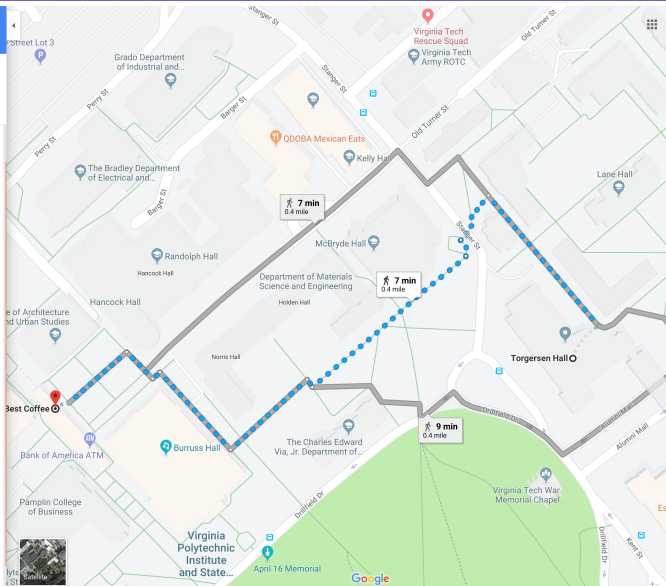
95 ft

↙ Turn left

194 ft

Seattle's Best Coffee

Burke Johnson Student Center, Blacksburg, VA 24060



Node ≡

, Edge ≡

← from Torgersen Hall, 620 Drillfield Dr, Blacksburg, VA 24061
to Seattle's Best Coffee, Burke Johnson Student Cent...

7 min (0.4 mile)

via unnamed roads
Mostly flat



Torgersen Hall

620 Drillfield Dr, Blacksburg, VA 24061

↑ Head northwest

436 ft

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138 ft

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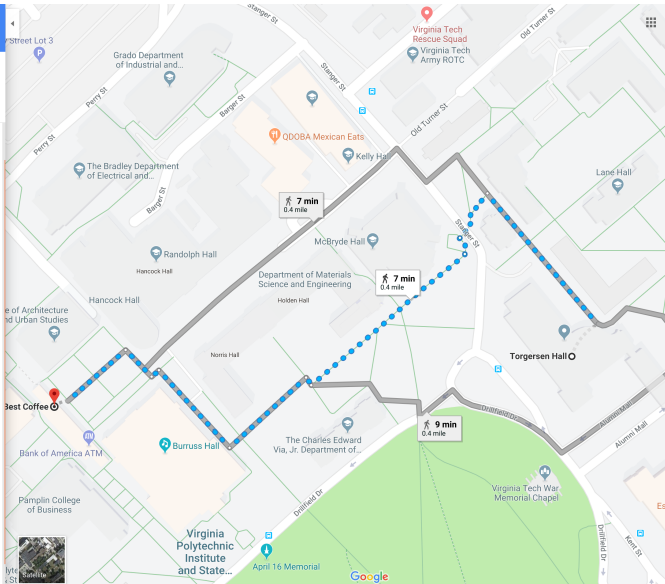
95 ft

↙ Turn left

194 ft

Seattle's Best Coffee

Burke Johnson Student Center, Blacksburg, VA 24060



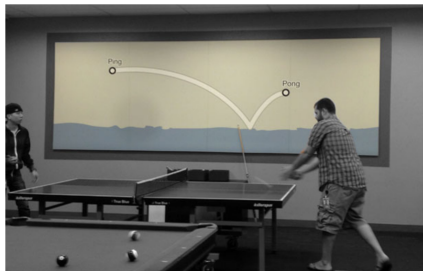
Node \equiv "Intersection", Edge \equiv Walkway/Unnamed road

How Do We Create Street Maps?

How Google Builds Its Maps—and What It Means for the Future of Everything

An exclusive look inside Ground Truth, the secretive program to build the world's best accurate maps

ALEXIS C. MADRIGAL | SEP 6, 2012 | TECHNOLOGY



How Do We Create Street Maps?





I was slated to meet with Gupta and the engineering ringleader on his team, former NASA engineer Michael Weiss-Malik, who'd spent his 20 percent time working on Google Mars, and Nick Volmar, an "operator" who actually massages map data.

"So you want to make a map," Weiss-Malik tells me as we sit down in front of a massive monitor. "There are a couple of steps. You acquire data through partners. You do a bunch of engineering on that data to get it into the right format and conflate it with other sources of data, and then you do a bunch of operations, which is what this tool is about, to hand massage the data. And out the other end pops something that is higher quality than the sum of its parts."


This is what they started out with, the [TIGER data from the US Census Bureau](#) (though the base layer could and does come from a variety of sources in different countries).



How Do We Create Street Maps?

Census.gov    

25th Anniversary of TIGER



TIGER is celebrating its 25th anniversary. The **Topologically Integrated Geographic Encoding and Referencing** database—the first nationwide digital map of roads, boundaries, and other features—was initially created for the 1990 Census to modernize the once-a-decade head count. However, its impact went well beyond its initial purpose by offering common map data in electronic form that powers the geographic information system (GIS) industry today. Through its TIGER/Line products, the Census Bureau has provided the common geospatial framework for use in linking statistical and other data in GIS.

The idea for TIGER developed within the Census Bureau. In the 1970s mathematicians, geographers, and software developers designed a spatial data handling system that resembled one big spreadsheet. Custom-built solutions were the norm for most GIS software companies in the two decades leading up to TIGER's release. TIGER was like a giant

How Do We Create Street Maps?



How Do We Create Street Maps?

OpenStreetMap powers map data on thousands of web sites, mobile apps, and hardware devices

OpenStreetMap is built by a community of mappers that contribute and maintain data about roads, trails, cafés, railway stations, and much more, all over the world.



Local Knowledge

OpenStreetMap emphasizes local knowledge. Contributors use aerial imagery, GPS devices, and low-tech field maps to verify that OSM is accurate and up to date.



Community Driven

OpenStreetMap's community is diverse, passionate, and growing every day. Our contributors include enthusiast mappers, GIS professionals, engineers running the OSM servers, humanitarians mapping disaster-affected areas, and many more. To learn more about the community, see the [OpenStreetMap Blog](#), [user diaries](#), [community blogs](#), and the [OSM Foundation](#) website.



Open Data


OpenStreetMap is *open data*: you are free to use it for any purpose as long as you credit OpenStreetMap and its contributors. If you alter or build upon the data in certain ways, you may distribute the result only under the same licence. See the [Copyright and License page](#) for details.

How Do We Correct Street Maps?

GREG MILLER SCIENCE 12.08.14 06:45 AM

THE HUGE, UNSEEN OPERATION BEHIND THE ACCURACY OF GOOGLE MAPS



Inside Atlas, Google's map-editing program, operators can see where Street View cameras have captured images (colored dots), and zoom in with a spyglass tool.  [GOOGLE MAPS](#)

How Do We Correct Street Maps?



APPS & SOFTWARE

Image Source: Tobamony Press Inc.

People are still driving into lakes because their GPS tells them to



Yoni Heisler [@edibleapple](#)
May 17th, 2016 at 1:12 PM

[Share](#)[Tweet](#)

We've seen the "GPS directions gone awry" scenario play out on TV shows any number of times, but perhaps the most well-known example comes from an old episode of *The Office* where Michael and Dwight both ignore their common-sense instincts and instead blindly drive a car into a lake.

Recently, and rather unfortunately, a Canadian woman on the road had her own Michael Scott moment when, in the midst of a rainstorm and accompanying fog with low visibility, she followed her GPS's driving directions – only to find herself in the middle of a freezing cold lake in Ontario.

How Do We Correct Street Maps?

Waze fixes app after police say it left drivers stranded on unpaved roads 45 miles away from the casino they were trying to reach



By [Alisha Ebrahimji](#), CNN
Updated 7:29 PM ET, Wed January 22, 2020



The Waze navigation app left drivers headed to an Atlantic City, NJ casino stranded and in the middle of Colliers Mills Wildlife Management Area

How Do We Correct Street Maps?

Report an error on the map

If we need to add roads to Google Maps, or something on the map isn't right, you can tell us. Let us know if you notice errors in Google Maps such as:



- Incorrect road names
- Wrong info about one-way and two-way roads
- Incorrectly drawn road
- Road closures
- A road on the map doesn't exist
- Missing roads
- Wrong addresses or marker locations

Notes:

- To edit info about a business or landmark, you can [suggest an edit](#).
- To update your business in Google Maps, you can [edit the business listing](#).
- Learn how to [set or change your address in Google Maps](#).
- You can only help us correct an error in Maps in some countries and regions.

[COMPUTER](#) [ANDROID](#) [IPHONE & IPAD](#)

Tell us about an error

1. On your computer, open [Google Maps](#)  . Make sure you're signed in.
2. In the top left, click Menu  > **Send feedback** > **Edit the map**.
3. Follow the instructions.
4. Click **Submit**.

Nodes and Edges

- Nodes and edges are elemental building blocks of networks.

What are the nodes and edges of brain graphs?

- Multiscale architecture of brain makes the answer challenging.
- There is no single, privileged scale for the analysis of brain networks.
- No single technology that can measure brain networks over all biologically relevant scales of space or time.

Multiscale Organisation of Brain Anatomy



(a) 10 cm

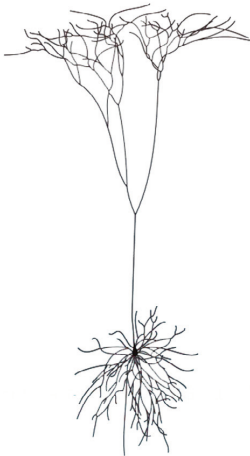
Broad divisions: cortical lobes, cytoarchitectural areas



(b) 1 cm

Neurons aggregate into columns, layers, and cell groups

Multiscale Organisation of Brain Anatomy



c)

1mm

Neuronal processes such as
dendritic trees and axons

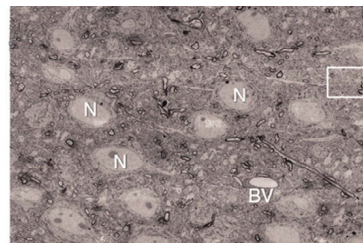


d)

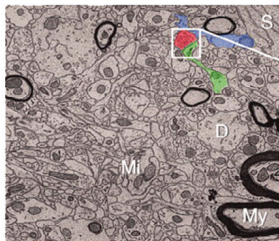
100 μm

Structure of individual fibers and
dendritic spines

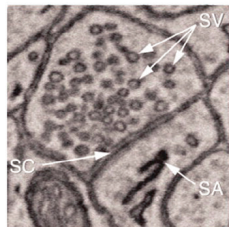
Multiscale Organisation of Brain Anatomy



(e)

100 μm 

(f)

10 μm 

(g)

1 μm

N: neuron, BV: blood vessel,

S: soma (cell body), Mi: mitochondria, My: myelinated axon, D: dendrite,

Blue: glial processes, Red: presynaptic terminal, Green: dendritic spines,

SV: synaptic vesicles, SC: synaptic cleft, SA: spine apparatus

Three Spatial Scales

- *Microscopic scale*: properties that are too small to resolve with the naked eye.
 - ▶ Require the use of microscopic techniques for visualization.
 - ▶ Scale that is synonymous with networks reconstructed at the level of individual neurons and synapses.

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 - ▶ Most commonly refers to analyses of structural and functional interactions between large-scale populations of neurons.
 - ▶ Characterized with MRI, MEG or EEG.

Three Spatial Scales

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- *Mesosopic scale* bridges the microscopic and macroscopic.
 - ▶ Analyses at this scale combine microscopic and macroscopic techniques.
 - ▶ Goal is to understand neuronal connectivity with high precision across the brain.

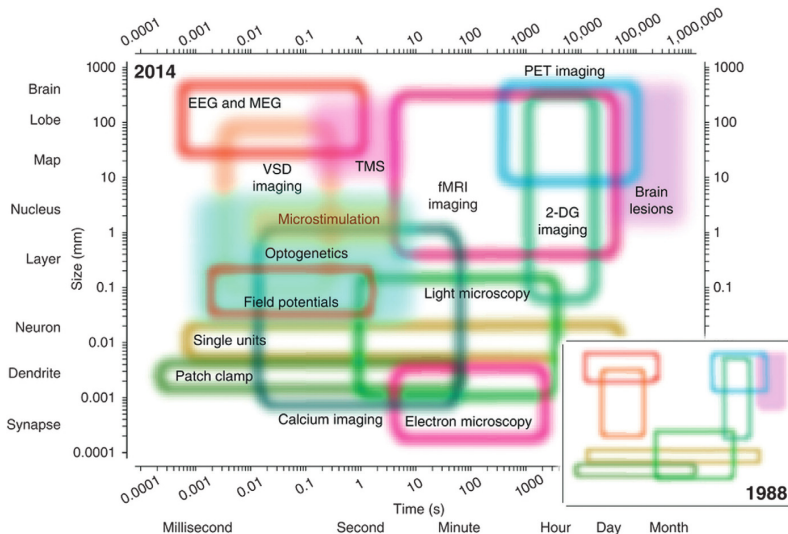
Three Spatial Scales

- *Microscopic scale*: properties that are too small to resolve with the naked eye.
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- *Mesosopic scale* bridges the microscopic and macroscopic.
 - ▶ Analyses at this scale combine microscopic and macroscopic techniques.
 - ▶ Goal is to understand neuronal connectivity with high precision across the brain.
- Techniques used at each scale constrain the way in which nodes and edges are defined.

Three Types of Connectivity

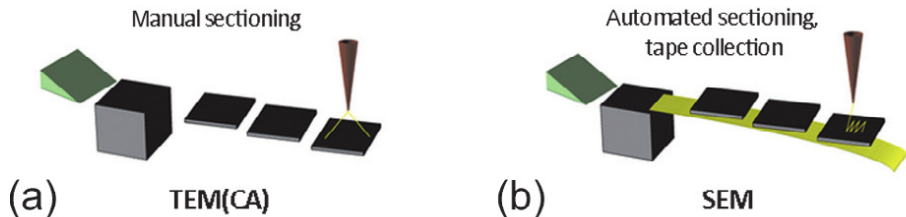
- *Structural*: anatomical connections between neural elements
 - ▶ Example: axons and synapses between neurons at the microscale.
 - ▶ Example: Large-scale fiber bundles that link cortical areas and subcortical nuclei at meso- and macroscales.
 - ▶ Measured using techniques such as electron microscopy (micro), axonal tract-tracing (meso), and diffusion MRI (macro).
- *Functional*: statistical dependence between physiological recordings that have been acquired from distinct neural elements.
 - ▶ Example: Correlation between spiking output of two neurons.
 - ▶ Measured by mathematical definitions of correlations.
- *Effective*: direct, causal influence that one neural element exerts on another's activity.

Spatiotemporal Resolution of Measurement Techniques



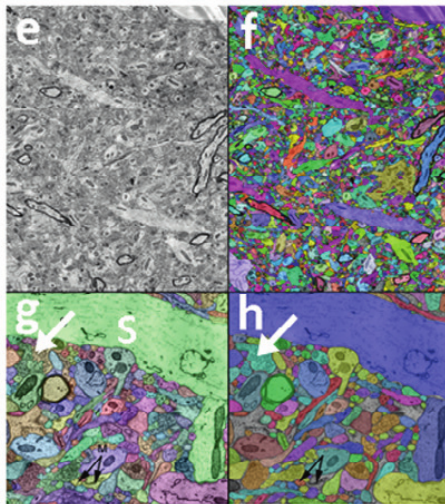
Open areas: measurement, filled areas: perturbation

Structural Connectivity at the Microscale



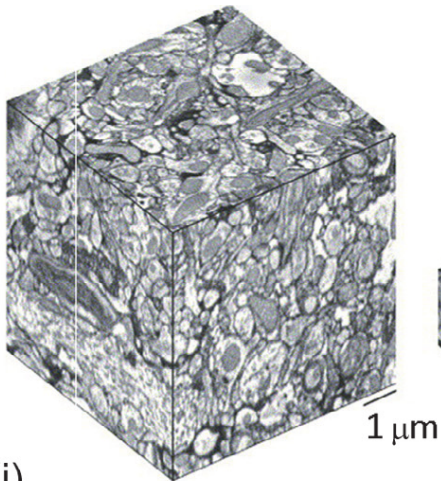
- Transmission electron microscopy (TEM): a beam of electrons is transmitted through a specimen to form an image.
- Scanning electron microscopy (SEM): produce images of a sample by scanning the surface with a focused beam of electrons.

Structural Connectivity at the Microscale

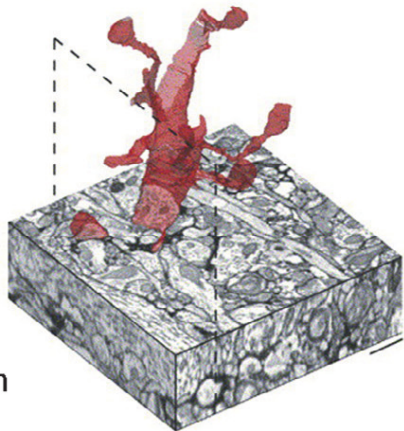


Mouse cortex section: $40 \mu\text{m} \times 20 \mu\text{m} \times 30 \text{nm}$.

Structural Connectivity at the Microscale



(i)



Reconstruction of the Fly Hemi-Brain

WIRED

BUSINESS

CULTURE

GEAR

IDEAS

SCIENCE

SECURITY

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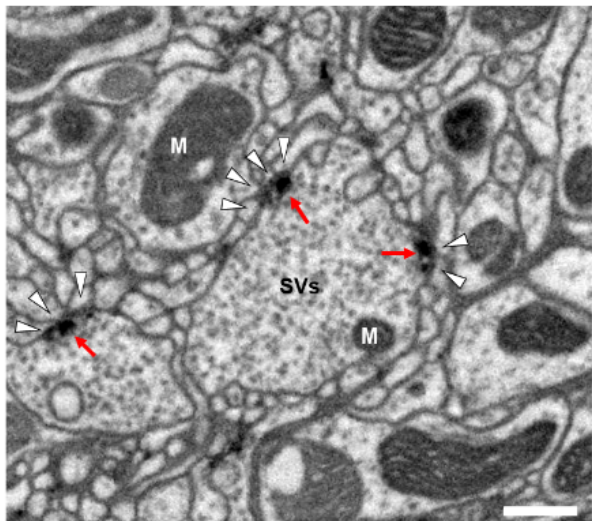
GREGORY BARBER

SCIENCE 01.22.2020 11:00 AM

The Most Complete Brain Map Ever Is Here: A Fly's 'Connectome'

It took 12 years and at least \$40 million to chart a region about 250 micrometers across—about the thickness of two strands of hair.

Reconstruction of the Fly Hemi-Brain



(a) *Drosophila* Synapse in EM


Reconstruction of the Fly Hemi-Brain

of 26 teravoxels of data, each with 8 bits of information. We applied numerous machine learning algorithms and over 50 person-years of proofreading effort over 2 calendar years to extract a variety of more compact and useful representations, such as neuron skeletons, synapse locations, and connectivity graphs. These are both more useful and much smaller than the raw grayscale data - for example, the connectivity can be reasonably summarized by a graph with 25,000 nodes and 3 million edges. Even with connections broken down by brain region, such a graph takes only 26 MB, roughly a million fold reduction in data size.

Reconstruction of the Fly Hemi-Brain

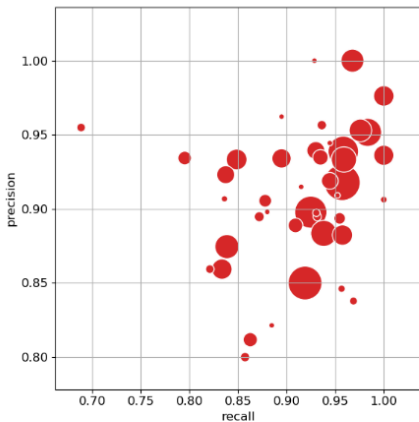
Article | Published: 16 July 2018

High-precision automated reconstruction of neurons with flood-filling networks

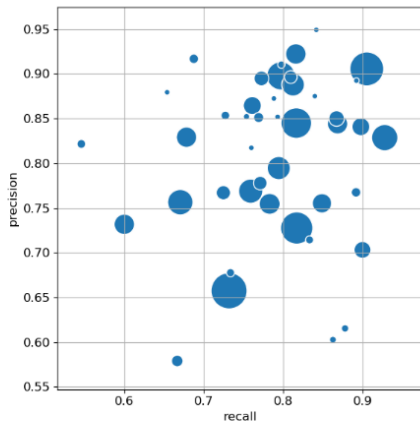
Michał Januszewski, Jürgen Kornfeld, Peter H. Li, Art Pope, Tim Blakely, Larry Lindsey, Jeremy Maitin-Shepard, Mike Tyka, Winfried Denk & Viren Jain 

Nature Methods **15**, 605–610(2018) | [Cite this article](#)

Reconstruction of the Fly Hemi-Brain



(a) T-bars



(b) PSDs

Reconstruction of the Fly Hemi-Brain

A Connectome of the Adult *Drosophila* Central Brain

 C. Shan Xu,  Michal Januszewski,  Zhiyuan Lu,  Shin-ya Takemura,  Kenneth J. Hayworth, Gary Huang, Kazunori Shinomiya,  Jeremy Maitin-Shepard, David Ackerman, Stuart Berg, Tim Blakely,  John Bogovic, Jody Clements, Tom Dolafi,  Phillip Hubbard,  Dagmar Kainmueller,  William Katz, Takashi Kawase, Khaled A. Khairy, Laramie Leavitt,  Peter H. Li, Larry Lindsey,  Nicole Neubarth, Donald J. Olbris, Hideo Otsuna, Eric T. Troutman, Lowell Umayam, Ting Zhao, Masayoshi Ito, Jens Goldammer, Tanya Wolff, Robert Svirskas,  Philipp Schlegel,  Erika R. Neace,  Christopher J. Knecht Jr.,  Chelsea X. Alvarado,  Dennis A. Bailey,  Samantha Ballinger,  Jolanta A Borycz,  Brandon S. Canino, Natasha Cheatham,  Michael Cook, Marisa Dreher,  Octave Duclos,  Bryon Eubanks, Kelli Fairbanks,  Samantha Finley,  Nora Forknall,  Audrey Francis,  Gary Patrick Hopkins,  Emily M. Joyce,  SungJin Kim,  Nicole A. Kirk,  Julie Kovalyak,  Shirley A. Lauchie,  Alanna Lohff,  Charli Maldonado,  Emily A. Manley,  Sari McLin,  Caroline Mooney,  Miatta Ndama,  Omotara Ogundeyi,  Nneoma Okeoma,  Christopher Ordish,  Nicholas Padilla,  Christopher Patrick, Tyler Paterson, Elliott E. Phillips,  Emily M. Phillips,  Neha Rampally,  Caitlin Ribeiro,  Madelaine K Robertson,  Jon Thomson Rymer,  Sean M. Ryan,  Megan Sammons, Anne K. Scott,  Ashley L. Scott,  Aya Shinomiya, Claire Smith, Kelsey Smith,  Natalie L. Smith,  Margaret A. Sobeski,  Alia Suleiman, Jackie Swift,  Satoko Takemura,  Iris Talebi,  Dorota Tarnogorska,  Emily Tenshaw,  Temour Tokhi,  John J. Walsh, Tansy Yang,  Jane Anne Horne, Feng Li,  Ruchi Parekh,  Patricia K. Rivlin,  Vivek Jayaraman, Kei Ito, Stephan Saalfeld, Reed George,  Ian Meinertzhagen,  Gerald M. Rubin, Harald F. Hess,  Louis K. Scheffer,  Viren Jain,  Stephen M. Plaza

doi: <https://doi.org/10.1101/2020.01.21.911859>

Reconstruction of the Fly Hemi-Brain

ZL developed sample preparation and fixed and stained the sample; CSX, KJH, HFH developed the imaging hardware and imaged the sample; LS, ETT, DK, KAK, DA and SS aligned, flattened and integrated the slabs into a coherent volume; WK wrote and managed the versioned data system; SB, TZ, PH, LU, TD, DRS, DJO, NN, SP, JC, LS, ET, PS, TK wrote proofreading and analysis software; MJ, JMS, PHL, VJain developed and applied segmentation approaches; TB, LL, JMS developed analysis, visualization and pipeline software; MJ, LL, VJain developed and applied tissue classification approaches; GH developed and applied methods to identify synapses; KS, ST, JG, MI, TW, FL, KI, RP, JAH defined brain regions and cell types; JB and HO did the EM-optical mapping; CXA, DAB, SB, JAB, BSC NC, MC, MD, OD, BE, KF, SF, NF, AF, GPH, EMJ, SK, NAK, JK, SAL, AL, CM, EAM, SM, CM, MN, OO, NO, Co, NP, CP, TP, EEP, EMP, NR, CR, MKR, JTR, SMR, MS, AKS, ALS, AS CS, KS, NLS, MAS, AS, JS, ST, IT, DT, ET, TT, JJW AND TY performed proofreading; EN and CJK provided proofreading analytics; VJay, IAM, ST, KS, PR, RP, FL, GMR did biological interpretation and analysis; LS, SP did connectivity analysis and wrote the paper; JAH, FL, PR, RP managed proofreading; RG, GMR, SP managed the overall effort.

Graphs from Microscale Structural Connectivity

- Node \equiv neuron, edge \equiv synapse.

Graphs from Microscale Structural Connectivity

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- Reconstructing connectomes in this manner is computationally demanding, time-consuming, and labor intensive.
 - ▶ One mm^3 of rat cortex imaged at nanometre resolution: 2PB of data.
 - ▶ A complete atlas of rat cortex (vol $\approx 500 \text{ mm}^3$): 1 EB (1000 PB).
 - ▶ A complete human cortex will require about 1000 EB.

Graphs from Microscale Structural Connectivity

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 - ▶ One mm^3 of rat cortex imaged at nanometre resolution: 2PB of data.
 - ▶ A complete atlas of rat cortex (vol $\approx 500 \text{ mm}^3$): 1 EB (1000 PB).
 - ▶ A complete human cortex will require about 1000 EB.
 - ▶ Reconstructing nervous system of *C. elegans* took over 10 years (White et al., 1986).

Functional Connectivity at the Microscale

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 - ▶ Pearson's correlation coefficient. Read Box 2.2 on pages 50–51.
 - ▶ Rank correlation coefficients.
 - ▶ Mutual information.

From Microscale to Mesoscale

- Microscale connectomics
 - ▶ Pro: offers unparalleled precision for resolving synaptic connectivity and spiking activity of individual neurons
 - ▶ Con: Techniques are not scalable to large-scale neural systems.
 - ▶ Con: High plasticity of synaptic connectivity makes it difficult to distinguish stable characteristics of neuronal networks from more transient features.

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- Mesoscale connectomics
 - ▶ Pro: Can smooth out some variability.
 - ▶ Pro: Offers a more robust means for characterizing time-invariant aspects of brain architecture.
 - ▶ Con: Not at the level of individual neurons.
 - ▶ Con: Depends on the parcellation.

Defining Nodes at the Mesoscale

- Goal is to map connectivity between neuronal populations or cell assemblies, rather than individual neurons.
- Exploit aggregation of neurons aggregate into populations that perform the same or related functions and are spatially proximal.
- Treat each volume as a node.
 - ▶ A volume may contain thousands or millions of cells.
 - ▶ Size of volume can which can range in size from cortical columns to larger cytoarchitectural areas and subcortical nuclei.

Defining Nodes at the Mesoscale

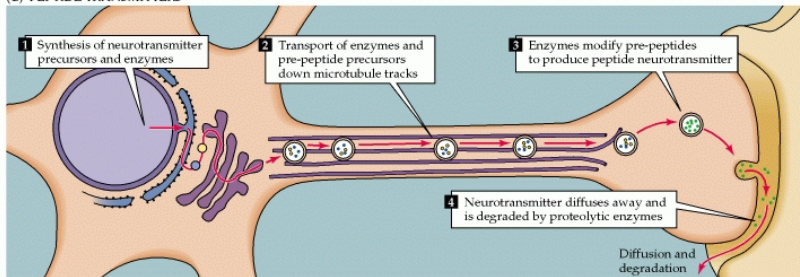
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- No gold standard for defining nodes; use approximations based on cytoarchitecture and anatomical landmarks.
- Coarse approach to defining nodes that results in the loss of information.
- Counterbalanced by an improved ability to map network structure over long distances.

Tracers for Structural Connectivity at the Mesoscale

- Invasive tract tracing is the main technique.
- A fluorescent dye or other tracer molecule injected into a specific part of the brain.
- Cellular membranes are permeable to these tracers.
- Once the tracer inside the cell, active axonal transport transfers it from the soma to peripheral axon terminals.
- After the tracer has had sufficient time to fill the entire extent, sacrifice the animal, dissect the brain, and determine sites of tracer uptake.

Types of Tracers

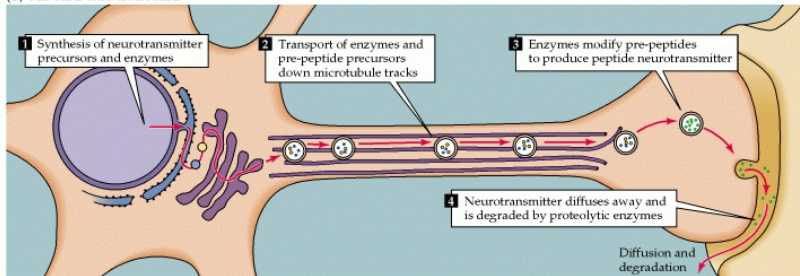
(C) PEPTIDE TRANSMITTERS



- Direction of transport distinguishes tracers.

Types of Tracers

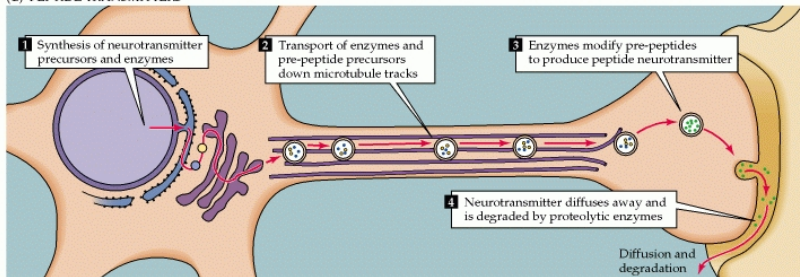
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- Direction of transport distinguishes tracers.
- *Anterograde tracers*: transported from the cell body to the axon terminal; used to map the efferent projection sites of an injected area.
- *Retrograde tracers*: transported from the cell periphery to the soma; used to map the upstream sources of afferent projections to the injection site.

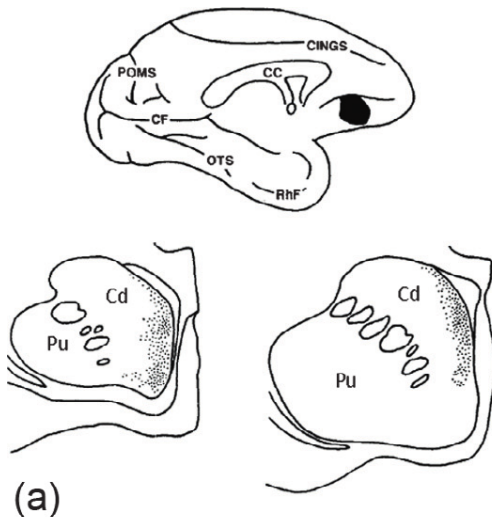
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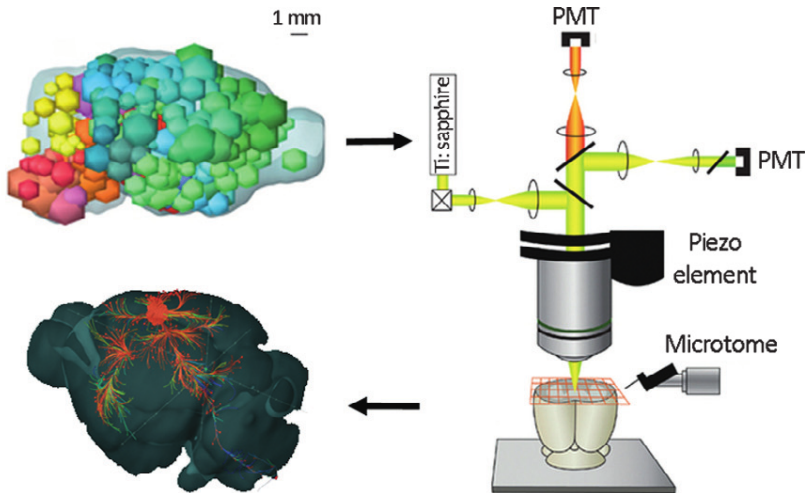
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- Viral tracers can cross synaptic junctions, allowing the mapping of polysynaptic pathways.

Structural Connectivity at the Mesoscale: Mouse



Results from a traditional tract tracing experiment

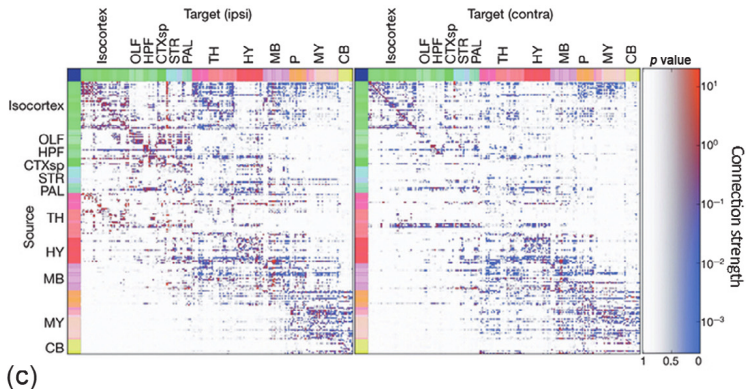
Structural Connectivity at the Mesoscale: Mouse



(b)

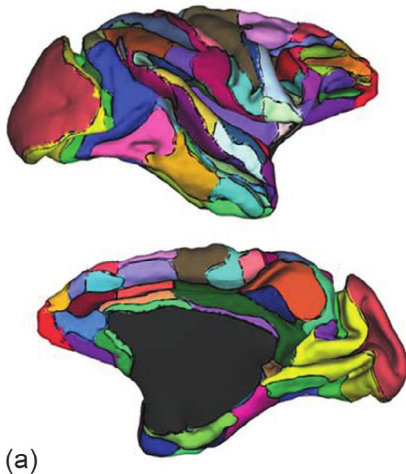
Modern experimental setup for whole-brain connectivity mapping
469 distinct tracer experiments

Structural Connectivity at the Mesoscale: Mouse



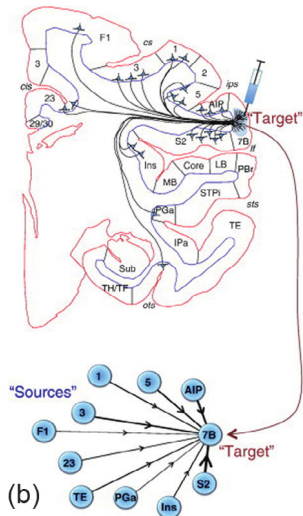
Example of connectivity matrix.
Edge weights range over four orders of magnitude.

Structural Connectivity at the Mesoscale: Macaque



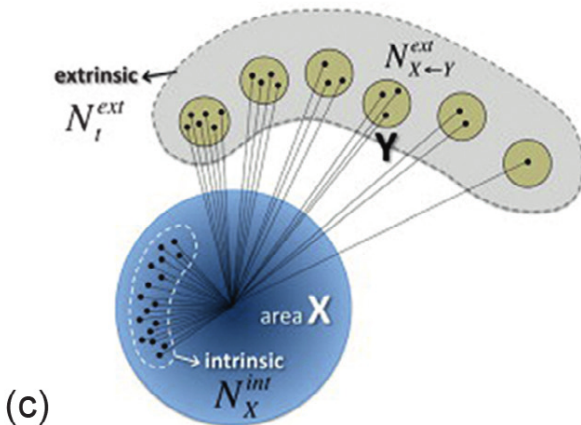
Parcellate the macaque cortex into 91 areas, defined according to cytoarchitecture and sulco-gyral landmarks.

Structural Connectivity at the Mesoscale: Macaque



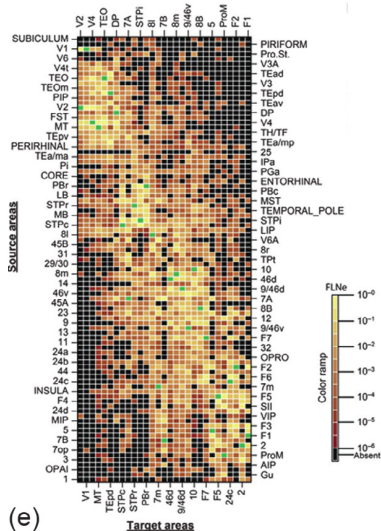
Use retrograde tract tracing. Determine edges coming into node representing area of injection from “labelled” nodes representing neurons that the tracer reaches.

Structural Connectivity at the Mesoscale: Macaque



Injection is at X: $w(Y, X) = \frac{\text{number of neurons labelled in } Y}{\text{total number of labelled neurons}}$

Structural Connectivity at the Mesoscale: Macaque



Example of connectivity matrix.

Edge weights range over six orders of magnitude.

Functional Connectivity at the Mesoscale

Please read Chapter 2.2.2 of the textbook.

From Mesoscale to Macroscale

- Experimental methods we have discussed so far for measuring brain connectivity are invasive.
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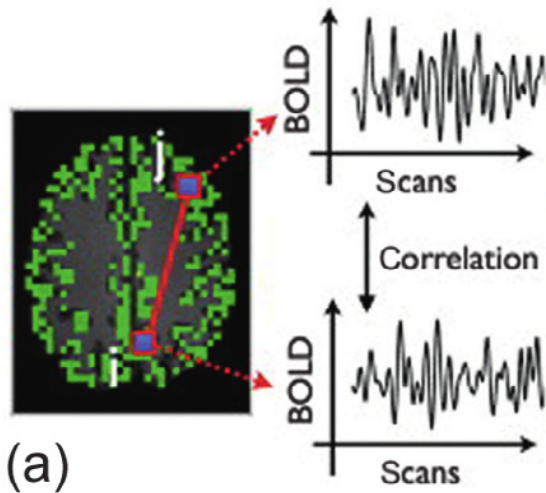
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 - ★ Typical MRI has a voxel resolution of 1 cubic mm, which contains an estimated 20,000–30,000 neurons and billions of synapses.
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From Mesoscale to Macroscale

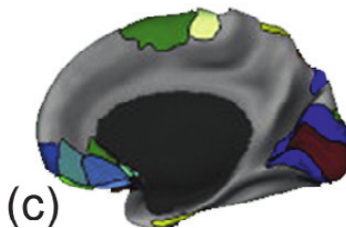
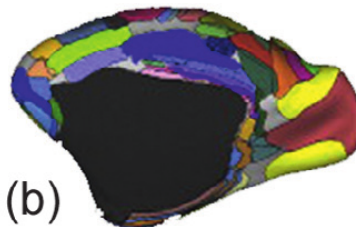
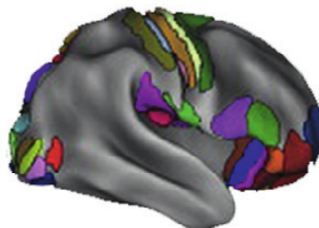
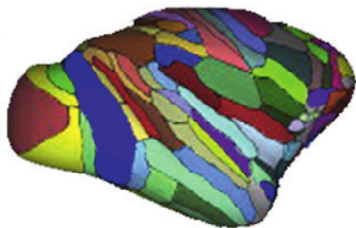
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 - ★ EEG and MEG have even lower resolution.
- Coarse spatial resolution means
 - ▶ We must aggregate measurements over ever-larger populations of neurons, axons, and synapses.
 - ▶ Reduces precision of node and edge definition.

Defining Nodes at Macroscale



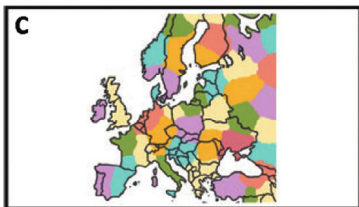
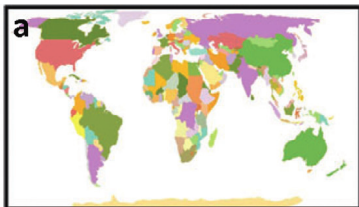
Each voxel is a node. Correlation between measurements for node pairs defines edges.

Defining Nodes at Macroscale

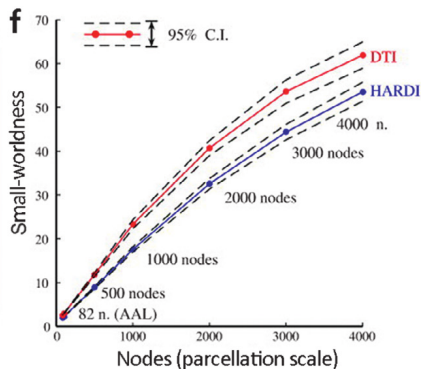
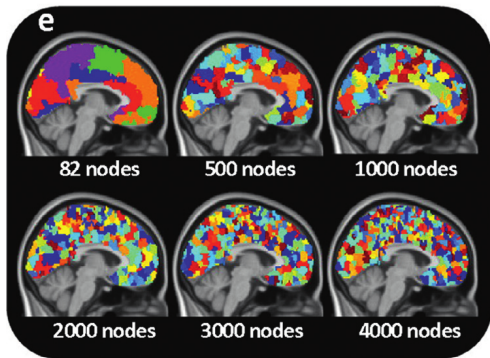


(b) Cytoarchitectural atlases mapped to standard stereotaxic space. (b) Macaque brain. (c) Human brain.

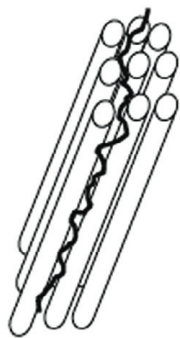
Parcellation Can Affect Network Properties



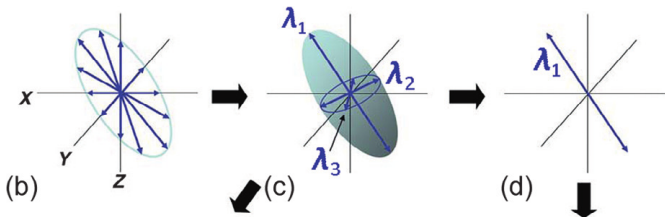
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Structural Connectivity at the Macroscale



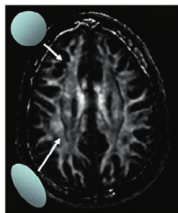
(a)



(b)

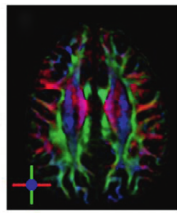
(c)

(d)



(g)

Fractional anisotropy map



(f)

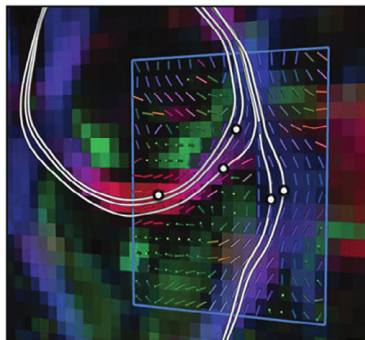
Color-coded orientation map



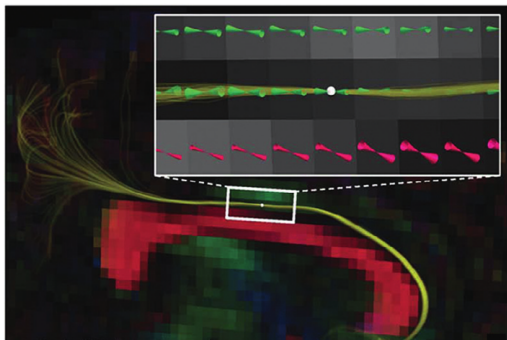
(e)

Orientation-color conversion

Structural Connectivity at the Macroscale

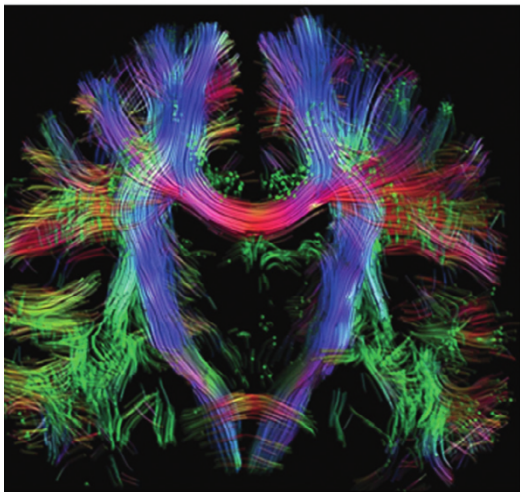


(h) Deterministic

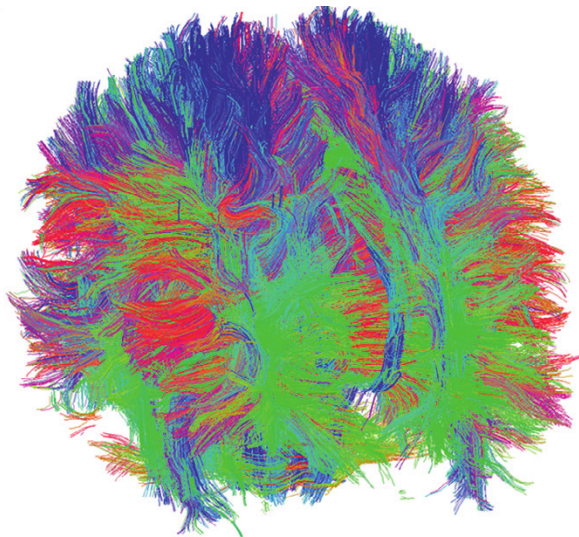


(i) Probabilistic

Structural Connectivity at the Macroscale



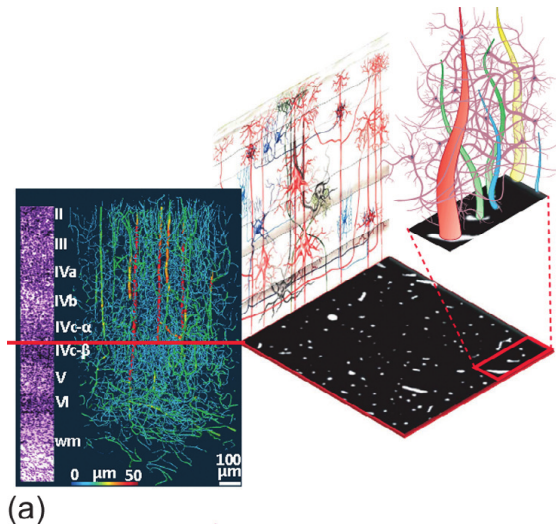
Structural Connectivity at the Macroscale



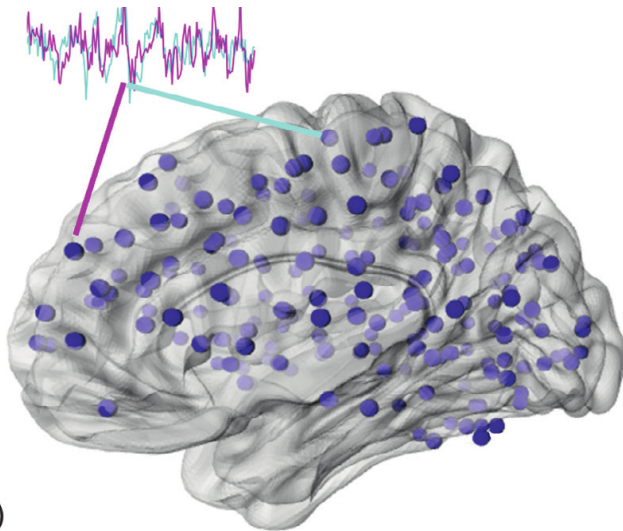
Functional Magnetic Resonance Imaging

fMRI - How it Works and What it's Good For, Video, 6:41"

Functional Connectivity at the Macroscale



Functional Connectivity at the Macroscale



Functional Connectivity at the Macroscale

