

What are the computational and data sciences?

Computation



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Modes of science



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Defining computational science



Defining data science



Computational and data sciences



Big data [refers to] data sets that are so big and complex that traditional dataprocessing application software [is] inadequate to deal with them. Big data challenges include capturing data, data storage, data analysis, search, sharing, transfer, visualization, querying, updating, [and] information privacy [...] There are a number of concepts associated with big data: originally there were 3 concepts volume, variety, velocity. Other concepts later attributed with big data are veracity (i.e., how much noise is in the data) and value.

— Wikipedia

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- Automated workflows with error-checking are very important when working with large datasets
- Researchers in the computational and data sciences often need to communicate results to non-experts, which requires effective visualizations and developing the ability to write and present a clear and compelling story

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- Data science methods applied to other fields: medicine, the humanities, political science, law, and the list goes on

Focus of this course

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 - If you also want to be introduced to the *computational science* side of things, consider taking CDS 130!

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- Special topic: basics of web scraping

Credits

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