

# MSDS at New York University

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Deputy Director, DGS, NYU

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# Mission



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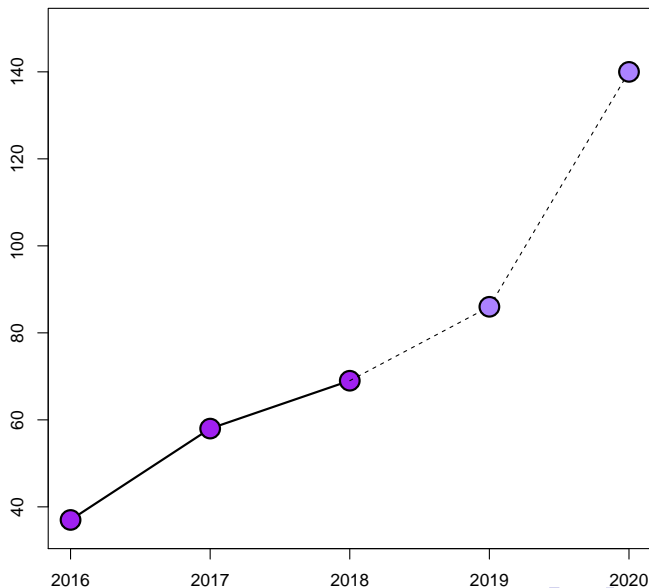
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Natural Language Processing with Representation Learning

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General electives

plus **ethics!**



# Ethics





Interest in ethics is **asymmetric**:

The New York Times



## *Tech's Ethical 'Dark Side': Harvard, Stanford and Others Want to Address It*



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NYU

DATA  
SCIENCE



The New York Times



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But sustainability is hard: ultimately need (joint) hire in ethics!

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Ph *Physics*

Bio *Biology*

Med *Biomedical* (from Fall 2019)







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# Professional Development



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**More general university resources:** alumni, faculty, NYU career services



# Outcomes I

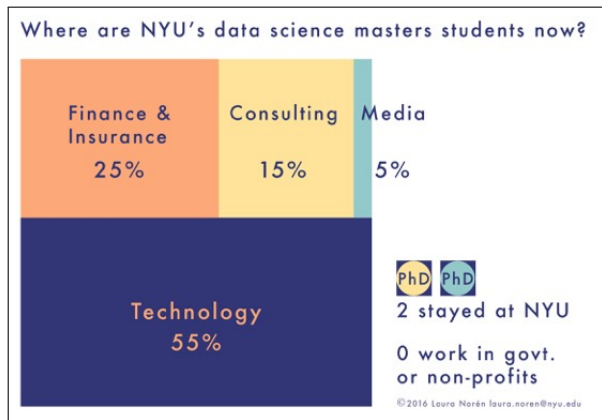


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'Narrative': understand program and what they want out of it



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We want to hear from you and your students!



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<https://cgs.nyu.edu/admissions/>

