

The SFU logo consists of the letters 'SFU' in white, bold, sans-serif font, set against a solid red rectangular background.

SFU

SIMON FRASER UNIVERSITY
ENGAGING THE WORLD

Linking data across university silos to support data-driven research development strategies

Kyle Demes, Director Institutional Strategic Awards
Morgan Mameni, Director Research Intelligence

Talk at a glance

- Background
- Case studies
- Challenges and opportunities



Data Science at Universities

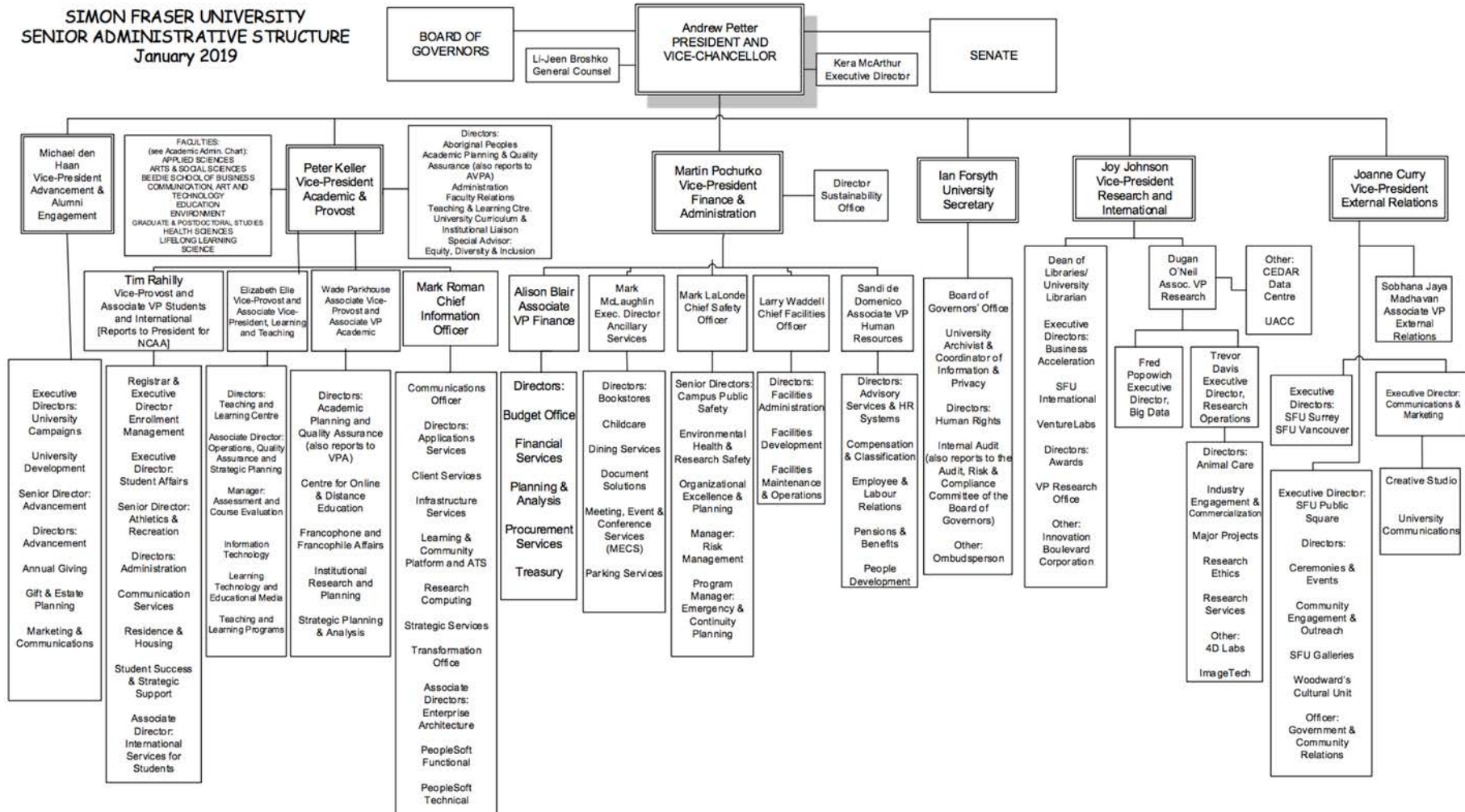
- Advances in data science are transforming research at unprecedented rates
- Universities are leaders in this field
- But University administrative structures are lagging considerably

Why the gap?

- The transformative potential is clear
- The talent and capacity already exist
- Historical momentum?
- Institutional silos?



Data follows organizational structure



A few case studies

- Identifying competitive award candidates
- Targeted communications
- Equity, diversity, and inclusion initiatives
 - empirically assessing bias
- Research metrics for applications

1- Identifying strong award candidates

- Research awards bring prestige and opportunities to Universities and faculty
- Awards are associated with achievements
- Universities' submissions are limited, so they only want to put forward nominees that are likely to succeed

1- Identifying strong award candidates

Can previous award data inform likelihood of applicant success and/or identify strong nominees?

Previous award
success data (binary)



Multi-dimension
research metrics



Model that best
predicts success

1- Identifying strong award candidates

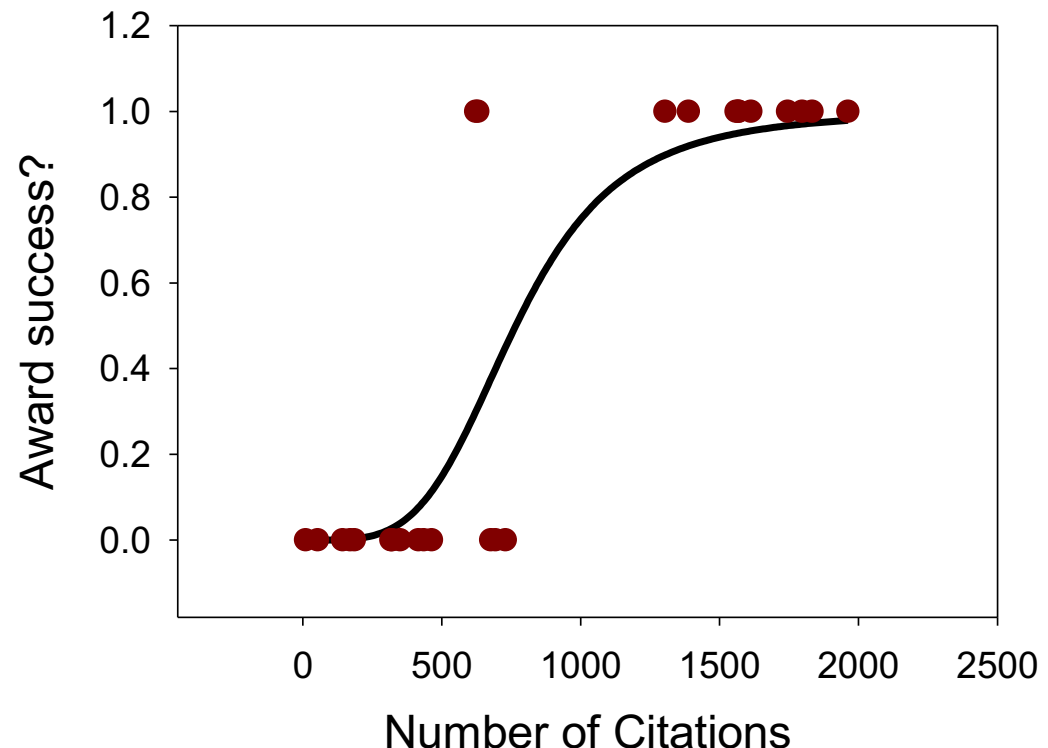
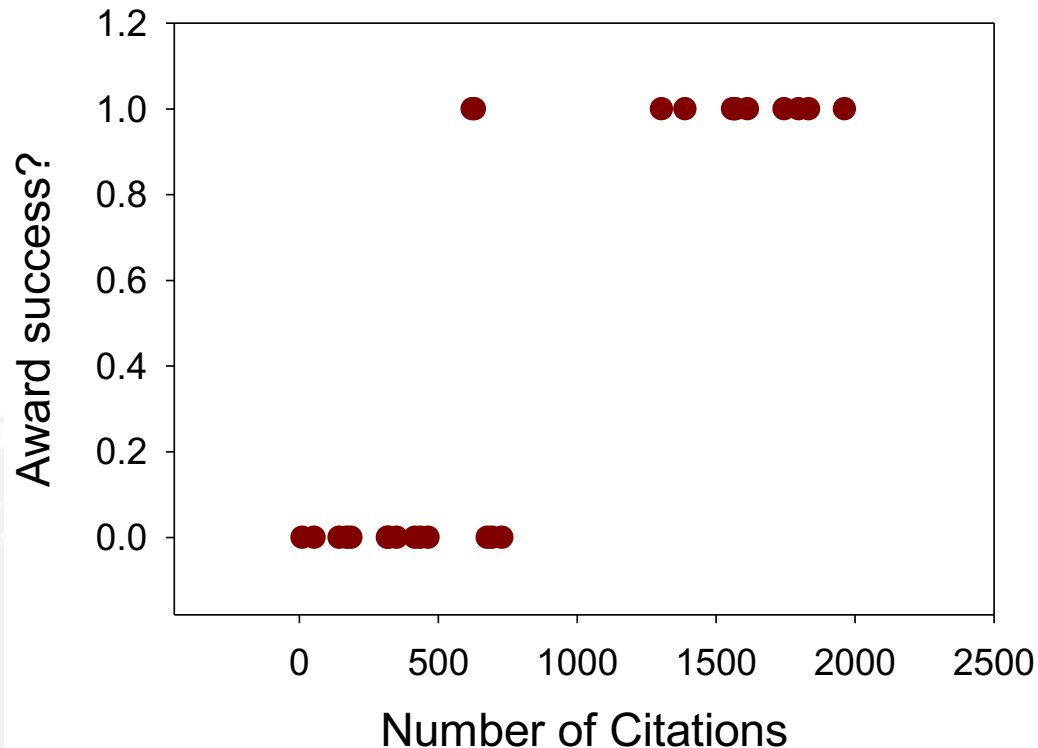
Previous award
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Multi-dimension
research metrics



Model that best
predicts success



1- Identifying strong award candidates

No universal predictor; useful models require dozens of factors from all dimensions

Name	Successful?	Year of PhD	Gender	Research \$	Industry \$	# Publications	FWCI	Significant Awards	Media mentions	HQP trained	Patents
Abel Andrzejewski	0	2008	Male	\$ 13,154,756	\$ 54,368	46	1	1	9	61	8
Kizzie Krawiec	0	2006	Female	\$ 5,576,573	\$ 90,268	250	2	3	29	73	15
Aleisha Alcala	0	2004	Female	\$ 4,195,727	\$ 129,630	45	1	6	30	36	13
Gayle Gaunt	1	2007	Male	\$ 8,795,831	\$ 137,561	34	1	2	41	23	8
Israel Infante	1	1991	Male	\$ 13,724,166	\$ 49,363	200	2	0	42	77	12
Bea Bracy	0	2008	Female	\$ 9,205,192	\$ 42,844	60	1	1	49	64	14
Daron Dieguez	1	2008	Male	\$ 1,480,339	\$ 80,085	46	2	4	31	76	0
Pierre Preusser	0	2009	Male	\$ 3,032,928	\$ 30,138	138	1	6	16	85	7
Scarlet Seagraves	0	1993	Female	\$ 6,826,748	\$ 52,641	369	2	6	40	9	15
Alysia Acheson	1	1993	Female	\$ 9,647,920	\$ 60,365	125	1	0	33	17	5

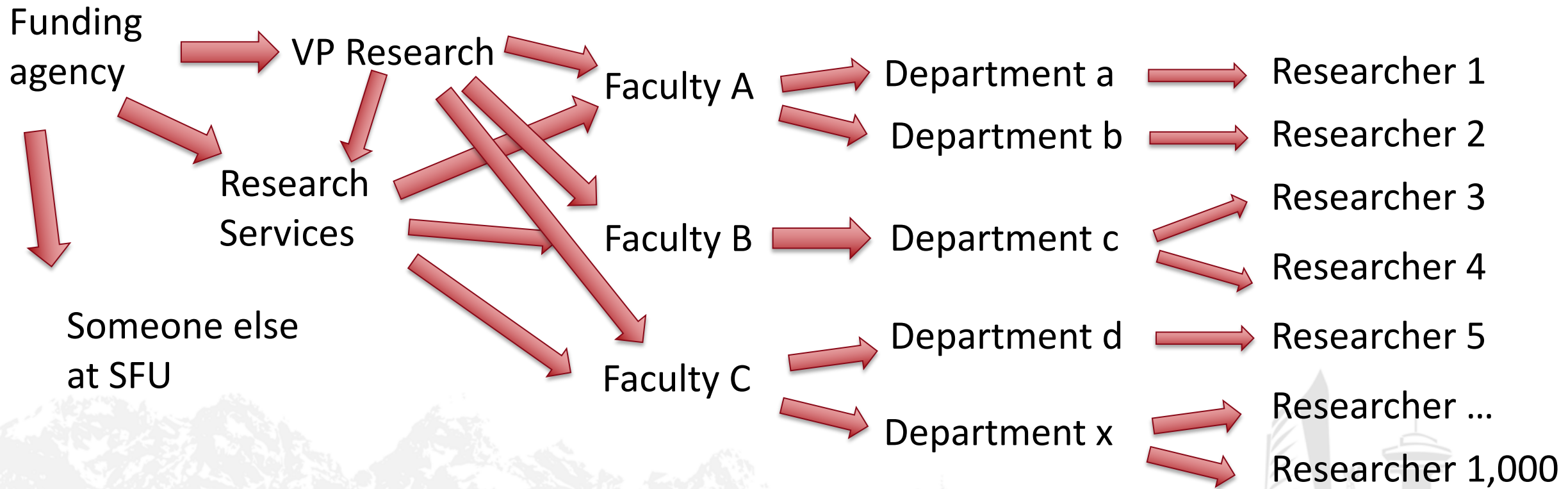
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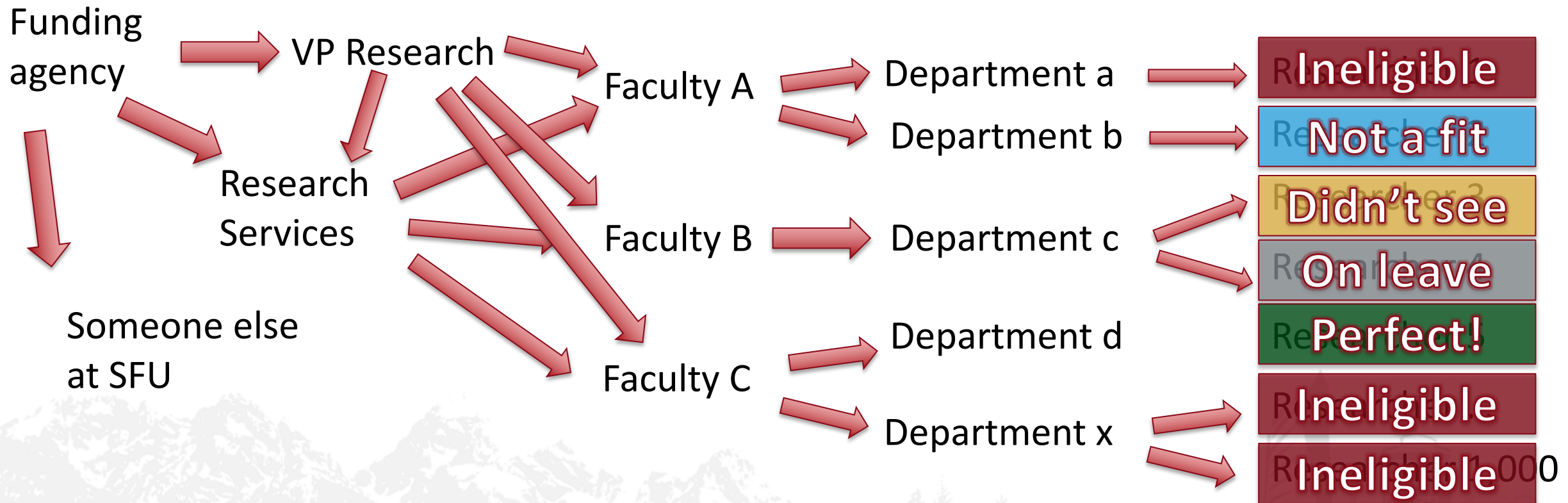
2- Targeted communications

- Matching funding opportunities with researchers is a perennial challenge
- Researchers don't know about grants:
 - programs change and pop-up sporadically
 - we don't know what they need at any point

2- Targeted communications



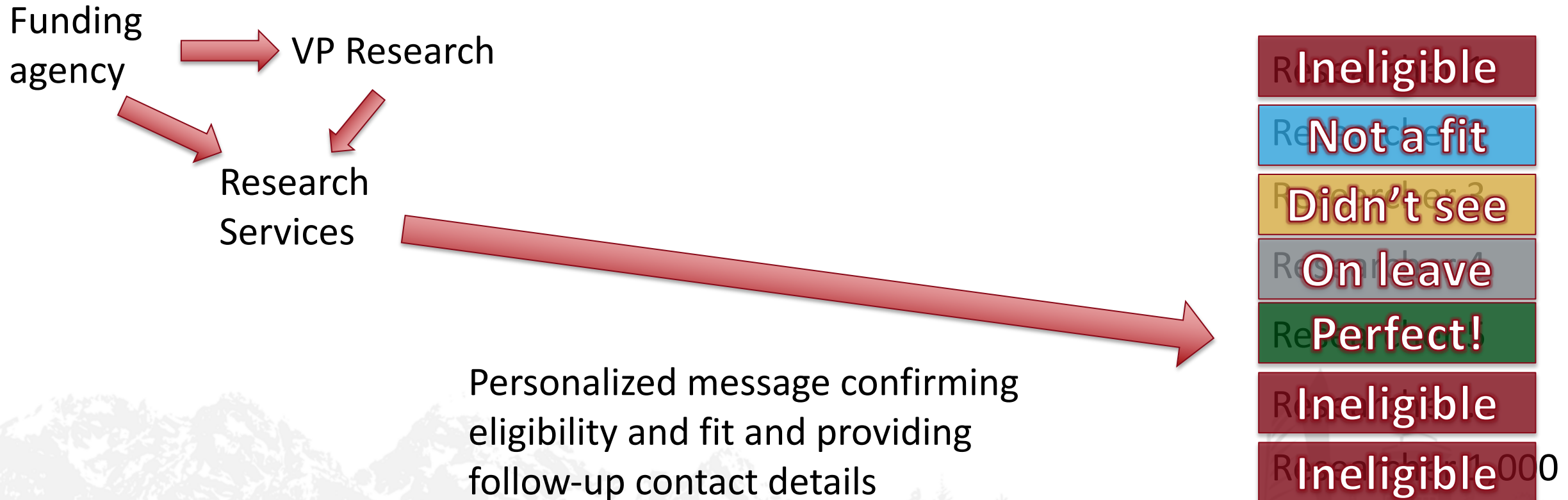
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2- Targeted communications

- Blast e-mails for every competition not effective
- Administrative data can help us filter by:
 - eligibility (rank, career stage, funding, etc.)
 - research expertise/ interest

2- Targeted communications

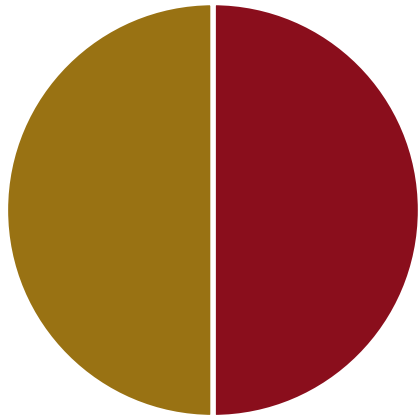


3- Equity, diversity, and inclusion

- EDI is an increasingly prominent priority
- Overarching principle is to address, evaluate, and correct systematic biases
 - this requires access to novel data
 - often data collection is designed *ad hoc*

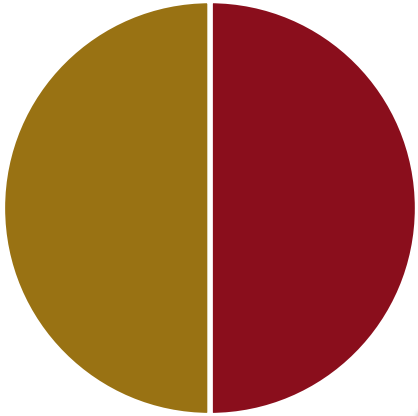
3- EDI in hiring: bias or chance?

Possible applicants

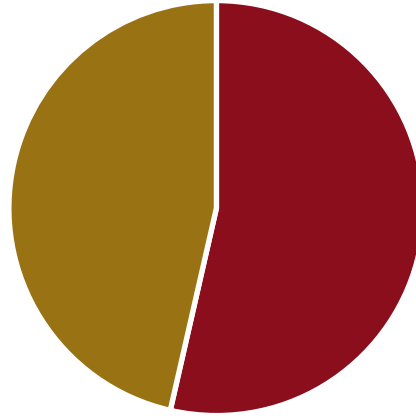


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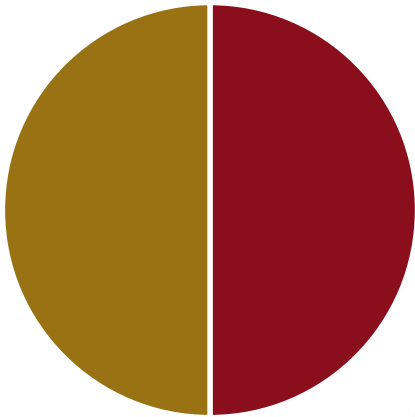
Applicant pool



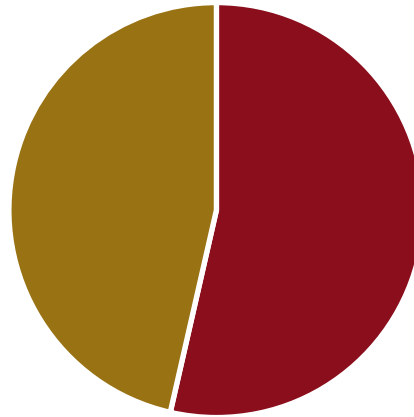
Chance?
 $p \gg 0.05$

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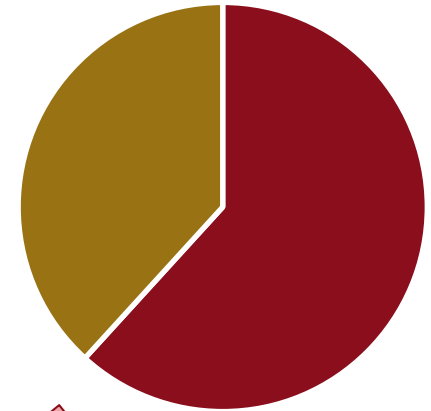
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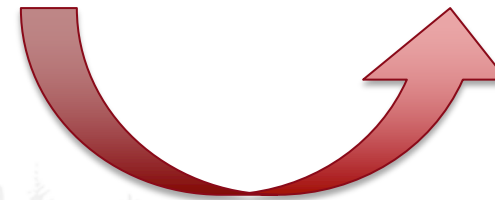
Applicant pool



Short list



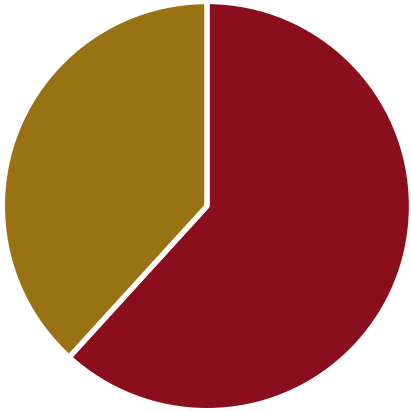
Chance?
 $p \gg 0.05$



Bias
 $p < 0.05$

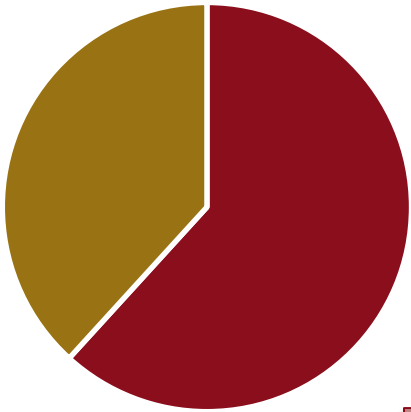
3- EDI in hiring: evaluating strategies

Possible applicants

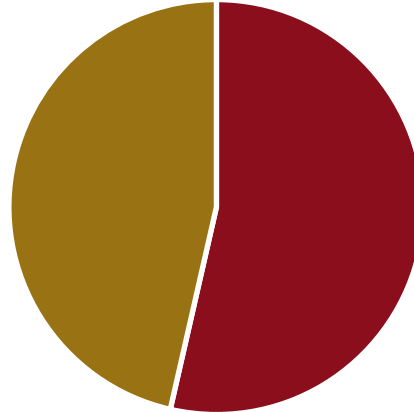


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Possible applicants



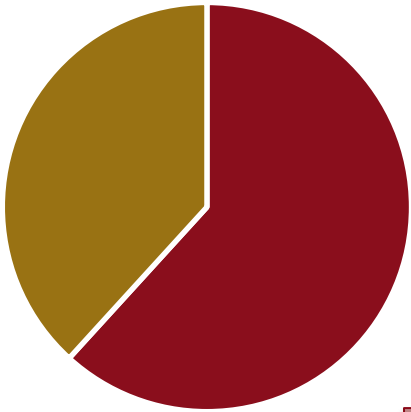
Applicant Pool



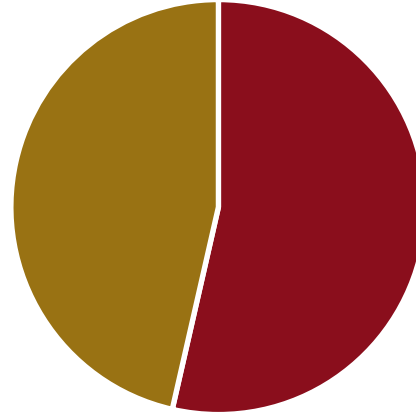
Strategic communication works
 $p < 0.05$

3- EDI in hiring: evaluating strategies

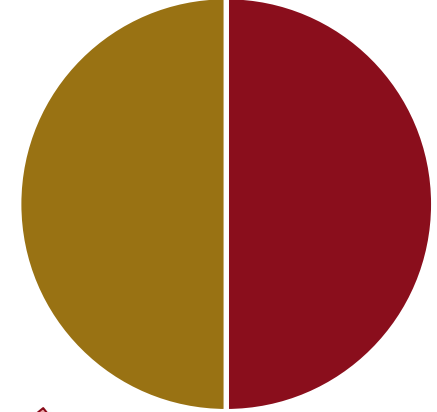
Possible applicants



Applicant Pool



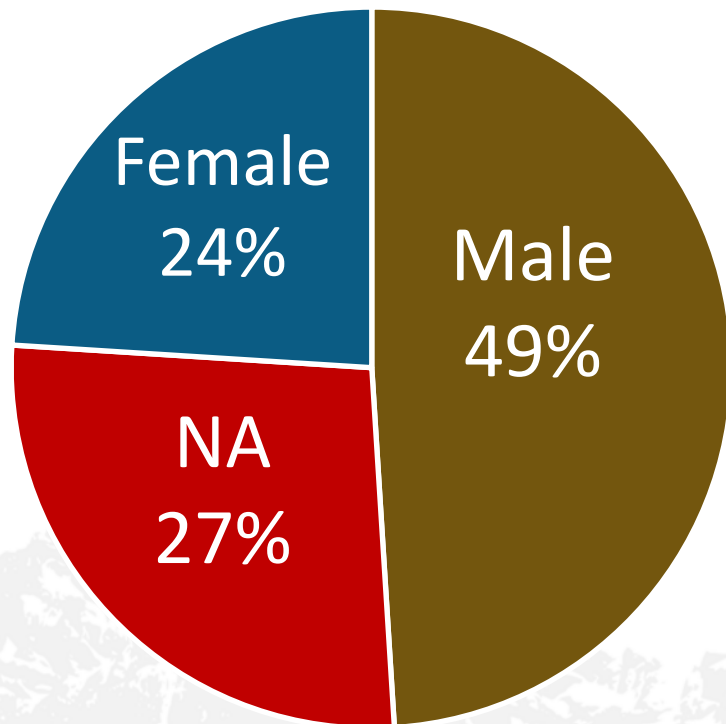
Short List



Strategic communication works
 $p < 0.05$

Process favours equity-seeking
group
 $p < 0.05$

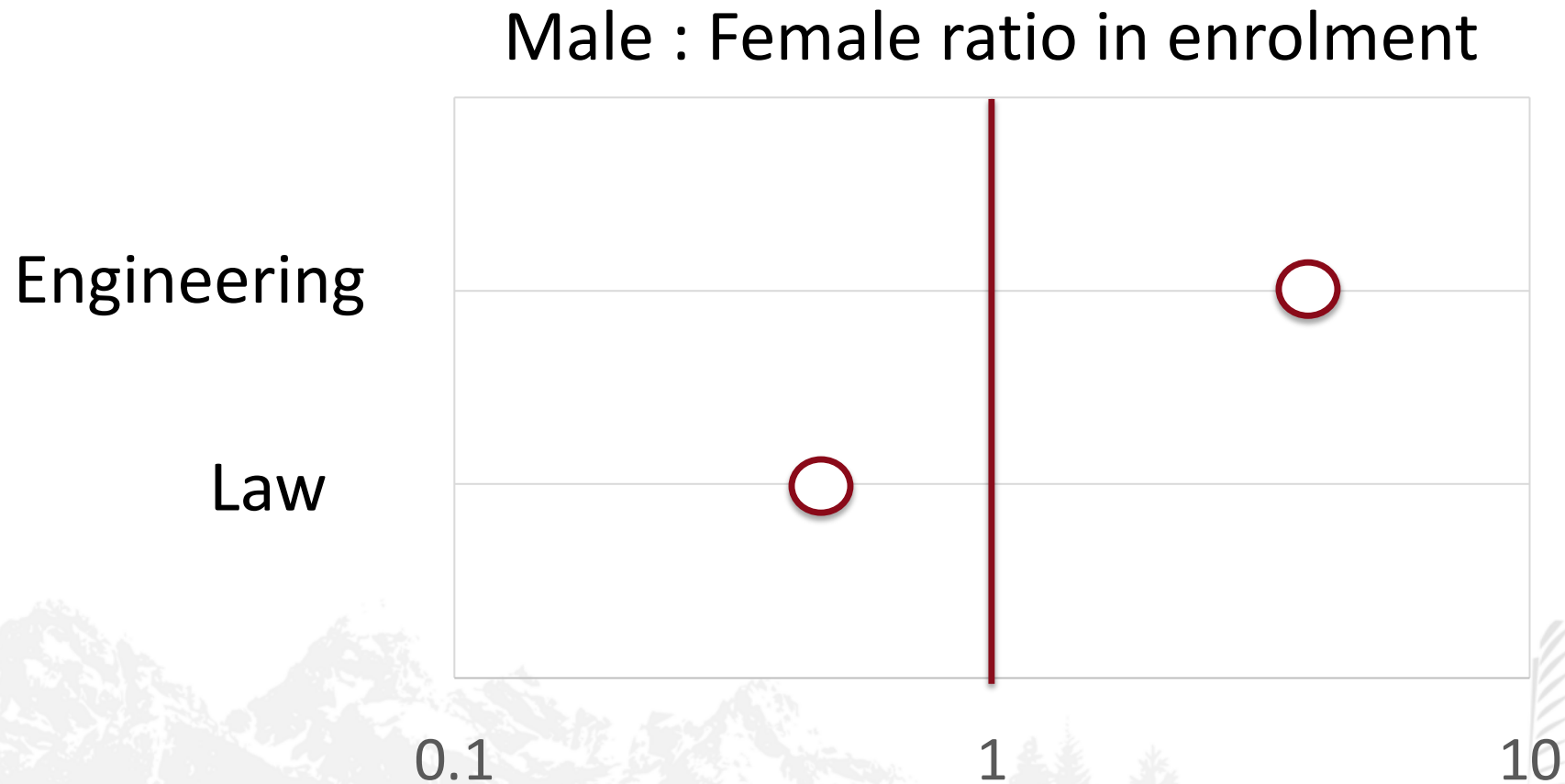
3- Bias in fellowship program?



- Overall?
- Bias in applications?
- Bias in process?

Simulated data

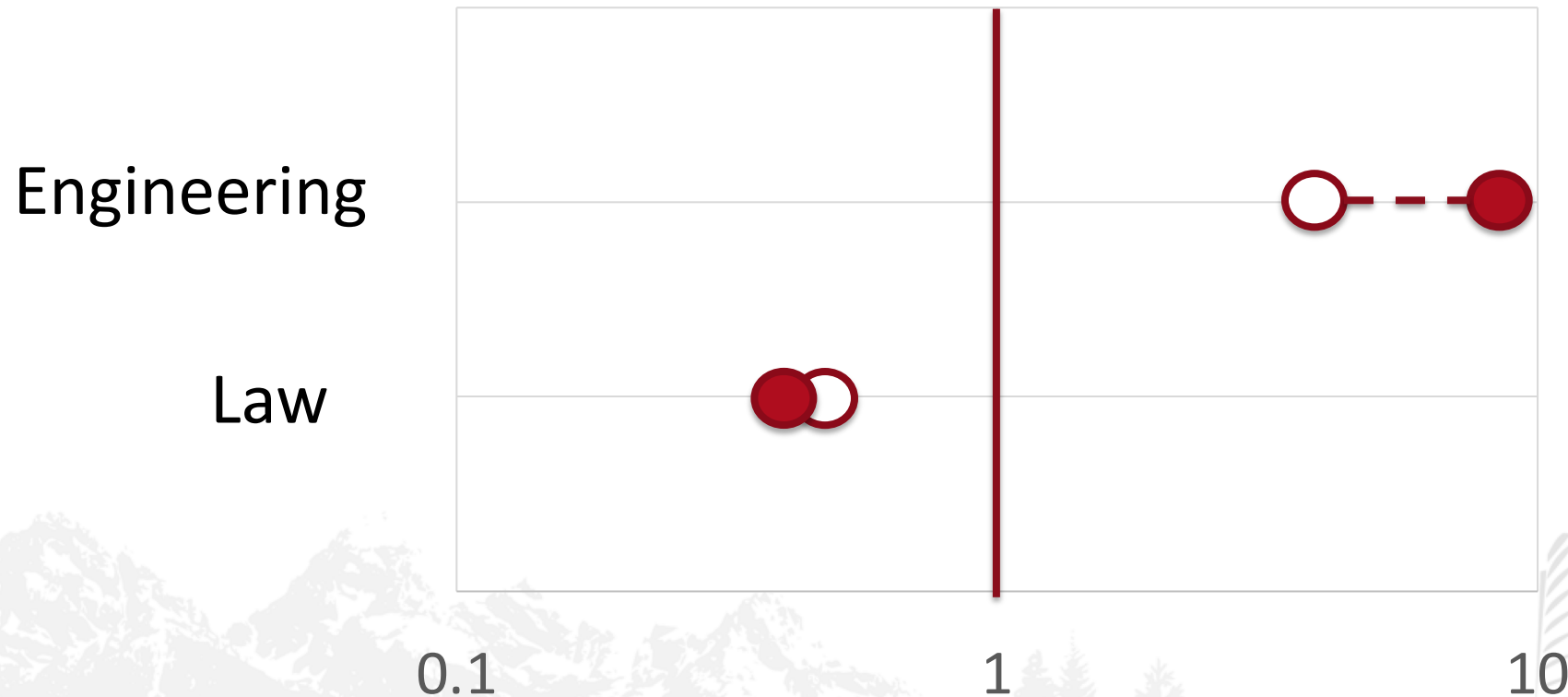
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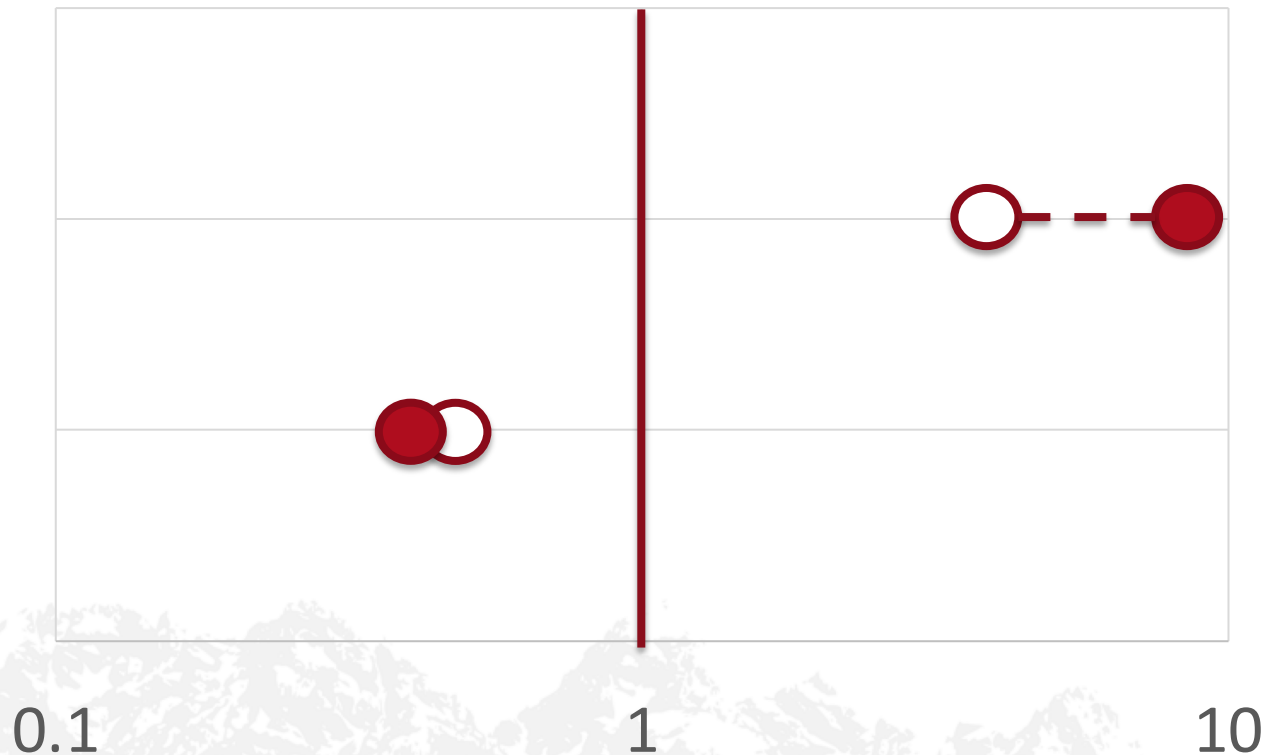
3- Bias in fellowship program?

Male : Female ratio in enrolment vs internships



Simulated data

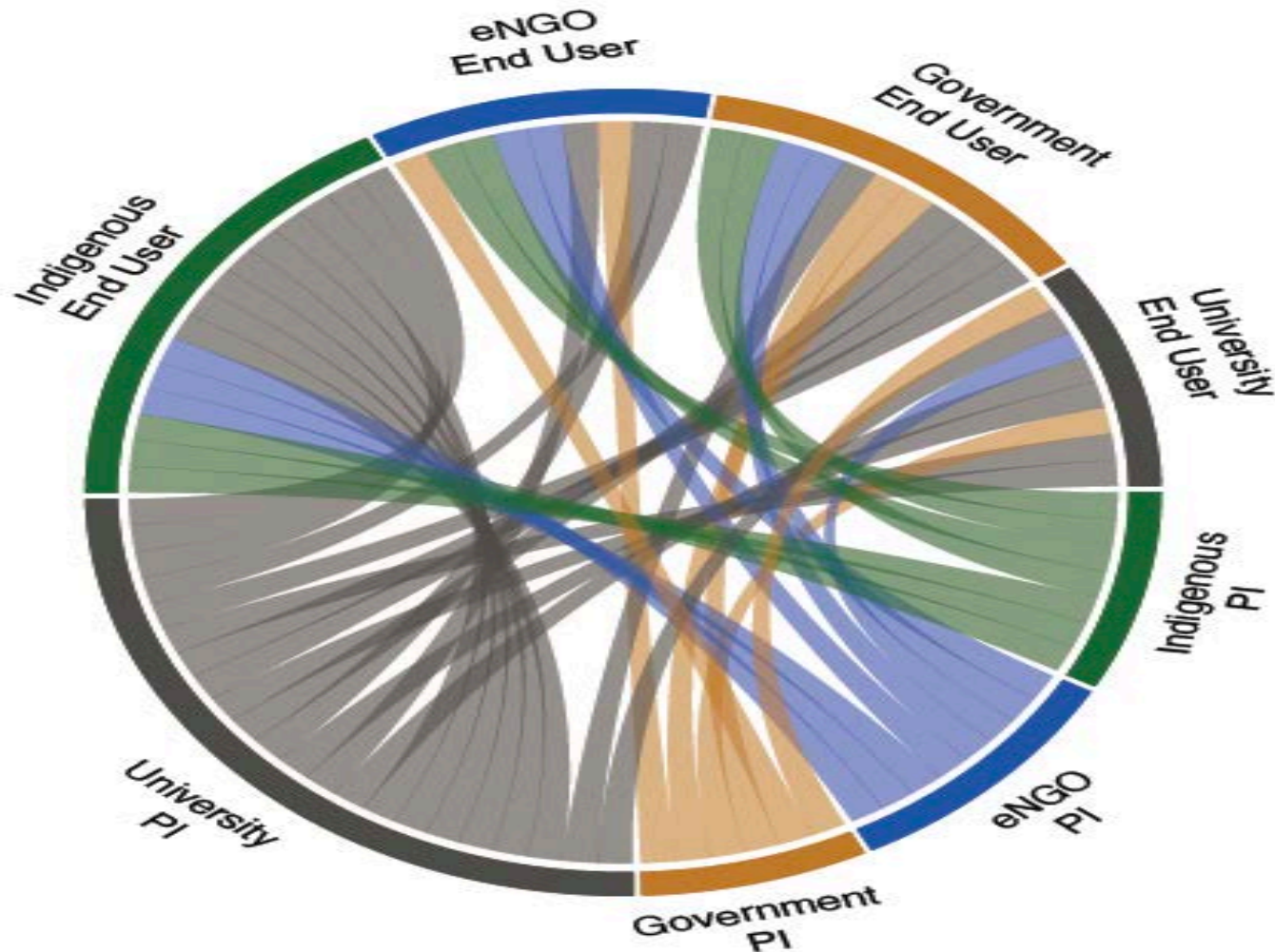
3- Bias in fellowship program?



Required data from:

- Fellowship agency
- Research services
- 'Creative' sources
- Graduate studies

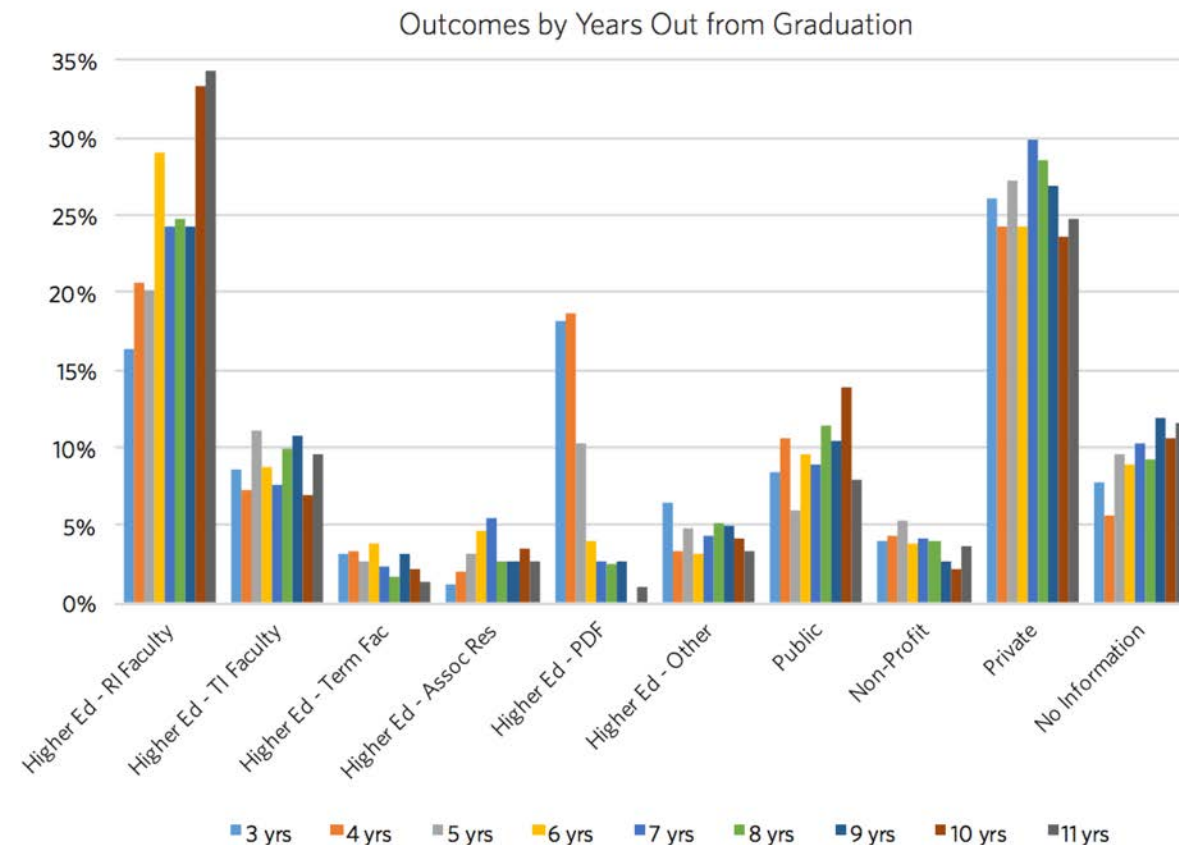
4- Research metrics for applications



‘Team-i-ness’

- co-publications
- team grants
- co-taught classes
- co-supervision

4- Research metrics for applications



‘excellence in training’

- # students
- time to completion
- career outcomes

4- Research metrics for applications

- ‘best in the world’
- ‘most innovative’
- ‘committed to community and public engagement’
- etc.



Big Data Big Challenges

- Advantages of data integration:
 - Attract competitive funding
 - Aid researcher recruitment
 - Improve internal services & administration
 - Enhance University reputation

Big Data Big Challenges

- Data integration not without challenges:
 - Community resistance to change
 - Technical & cultural
 - Suspicion regarding (irresponsible) use of data
 - Concerns over privacy

Community Resistance



Technical:

requires innovative approaches to business operations + (re-)training

Cultural:

requires communication, demonstration of advantages, institutional support + champions

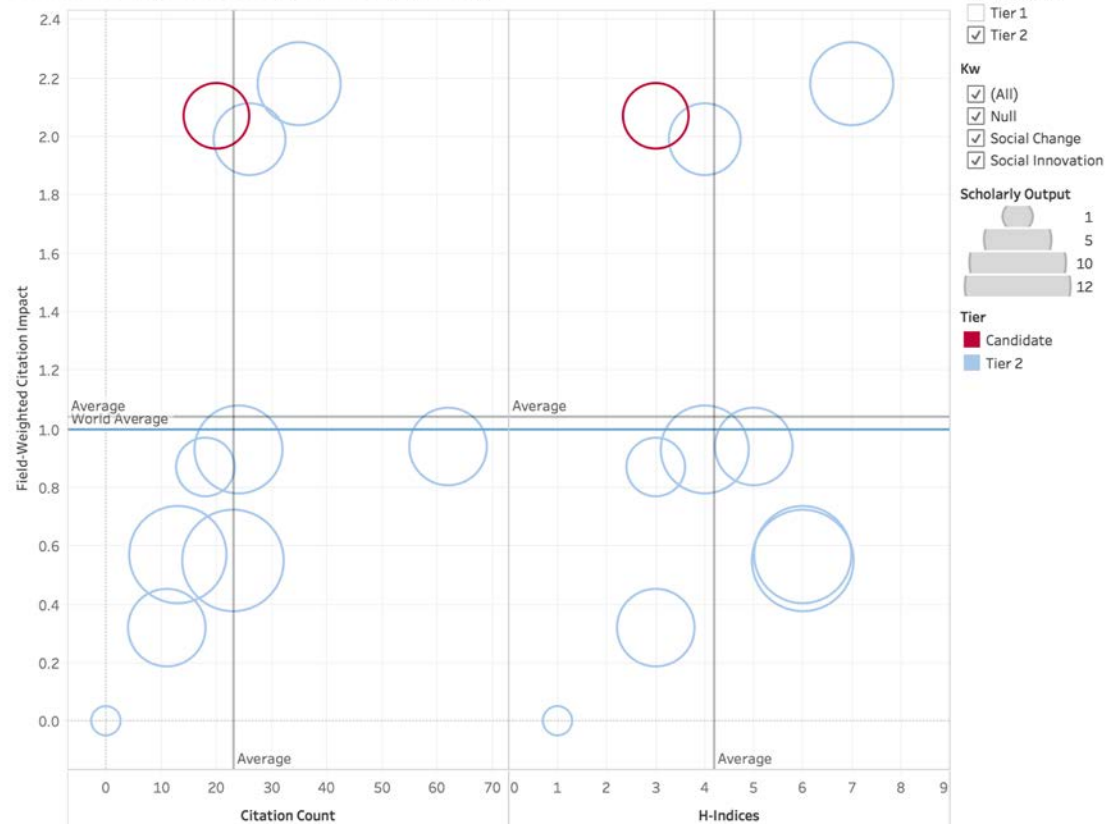
Data abuse

- Irresponsible use of metrics
 - Disregarding disciplinary diversity
 - Superficial analysis



Research metrics: use & abuse

Canada Research Chairs with keywords including *Social Innovation* or *Social Change*
Publications range: 2013-2018 (Source: Scopus/SciVal)



Blinkered approach? impact factors are being used too often to evaluate research

At least one in three research-intensive universities in North America examined by a study leaned on the journal impact factor of periodicals that academics had published in when making decisions on promotion and tenure, but the true proportion may be much higher.

The study, believed to be the first to examine the use of the journal impact factor in academic performance reviews, warns that there is an “undue reliance” on the controversial metric, which represents the number of times articles in particular publications have been cited in recent years.



Universities 'must read applicants' work to defeat impact factor'

[READ MORE](#)

Journal impact factors are being used too often “to evaluate the quality and significance of research, despite the numerous warnings against such use”, says the study, published on [PeerJ Preprints](#).

For the study, researchers from four countries collected and analysed review, promotion and tenure policies from 129 universities in the US and Canada.

Privacy



- Collection notice
 - *“I have the data but can I use it for this purpose?”*
- Personal information (perceived vs real)
 - *“Is this data personal? Is it current? Might out-of-date data reveal personal history?”*
- Data access
 - *Data governance models, managing access*

Where to go from here

- Data in action roadshows
- AI at the executive board
 - Data-informed decisions: good!
 - Robotic/formulaic decisions: bad!
- Accountability, reliability, transparency



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Thank you

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