

# Managing Research



# Outline

How to...

- Come up with research ideas
- Choose research topic
- MSc vs PhD
- Stay updated with new research
- Read papers efficiently
- Work with your supervisors
- Decide to start writing
- Connect with your research community



# How to come up with research ideas?

Subjective thoughts for this point in time

1. Narrow down
2. Specialize
3. Document into your tool-box
4. Read more broadly
5. Find a new idea



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# Narrow Down

Choose a *specific* topic

1. Is it **important**?



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*If you do not work on an important problem, it's unlikely you'll do important work. It's perfectly obvious.  
Richard Hamming, ["You and Your Research"](#)*



# Narrow Down

How to know what's important? Listen to leading researchers talks

## The Biggest Open Problems in NLP



Sebastian  
Ruder



Jade  
Abbott



Stephan  
Gouws



Omoju  
Miller

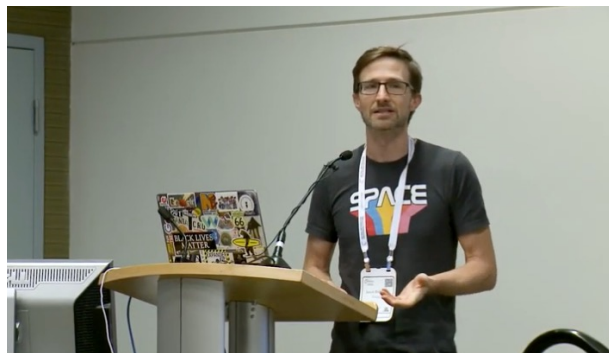


Bernardt  
Duvenhage



HUGGING FACE

**The Future of  
Natural Language Processing**



# Narrow Down

Be aware of red & blue oceans

- *“Blue Ocean” problems are safe, in “Red Ocean” problems you have more competition*



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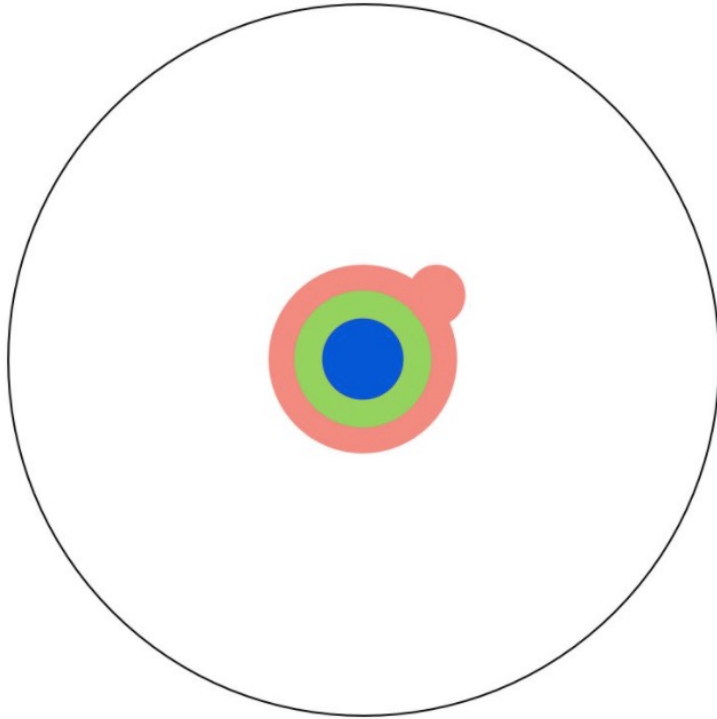
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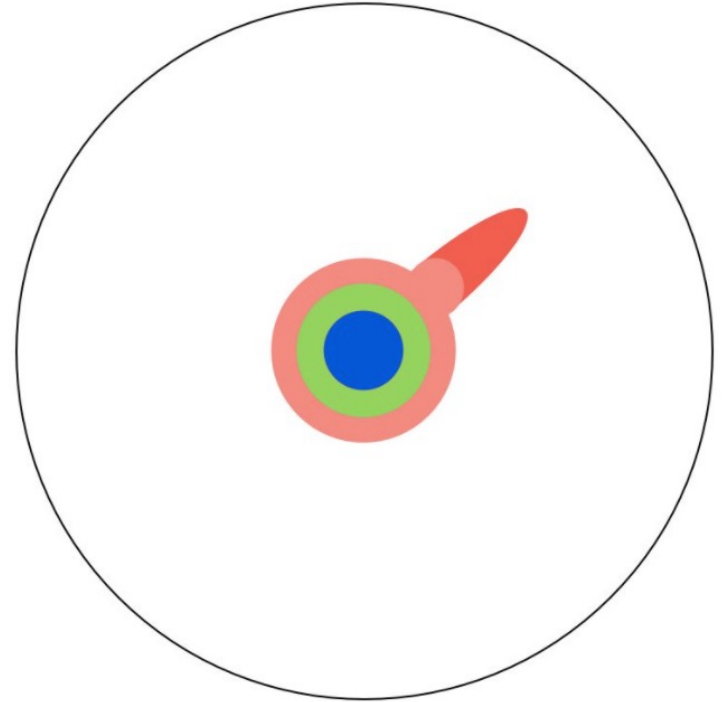
# Specialize

Read a lot about your topic

With a bachelor's degree, you gain a specialty:



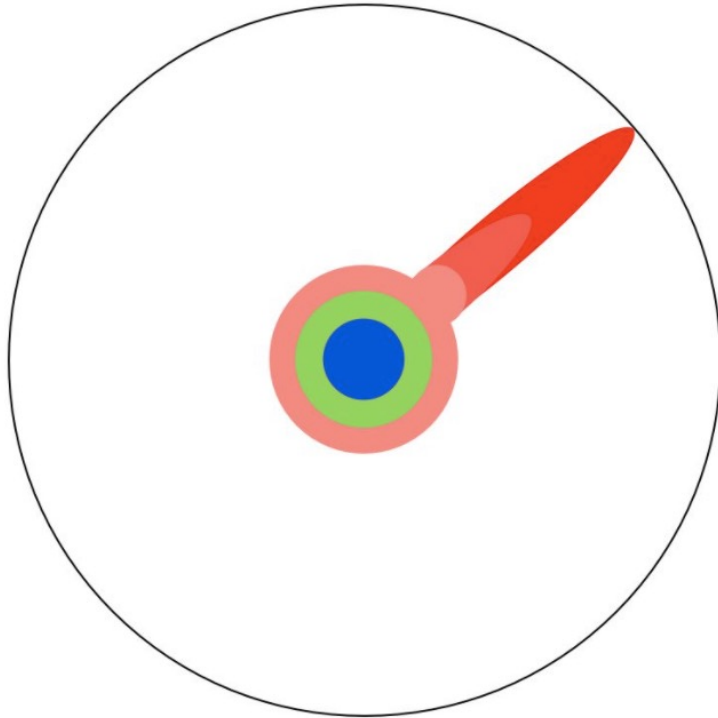
A master's degree deepens that specialty:



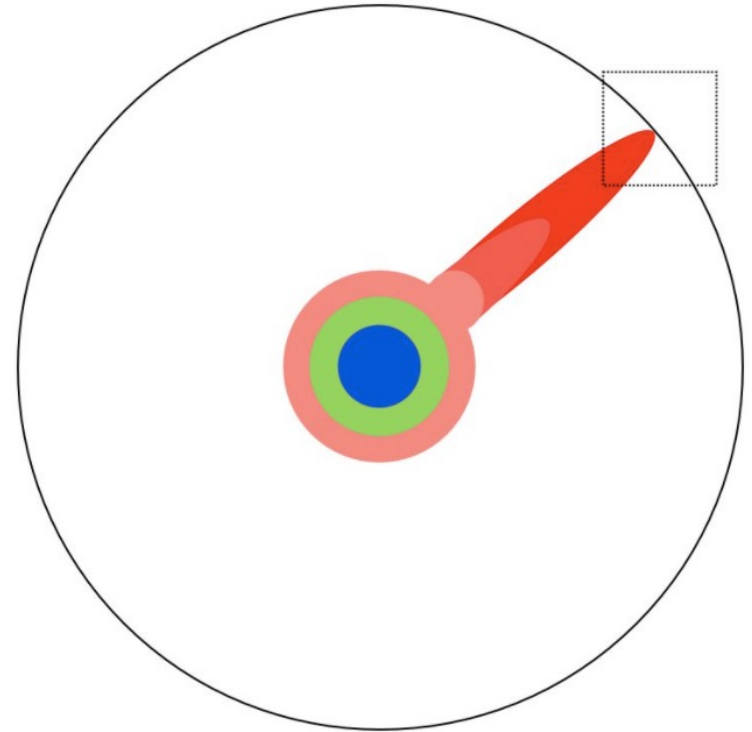
# Specialize

Read a lot about your topic

Reading research papers takes you to the edge of human knowledge:



Once you're at the boundary, you focus:



# Detour - PhD vs. MSc

	MSc	PhD
<b>Time</b>	2 years	4 years
<b>Focus</b>	Implementation (?) Single project <b>How</b> to research	Innovation Multiple projects <b>What</b> to research
<b>Projects risks</b>	Preferably low risk	It's ok to have high risk
<b>Who to work with?</b>	Mainly your advisor	Grow your network of collaborators, go to internships, etc
<b>Extra requirements</b>	~24 Points in courses Overhead of starting research	~12 points in courses
<b>Future Work</b>	Machine Learning Engineer, Research Engineer, Research Scientist	Academia (very competitive), Industry Research Labs



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1. Understand what was done in the field so far (Possible to read reviews). <sup>1</sup>
2. *Start from a seed of papers (or people!)*

<sup>1</sup> Uri Berger



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3. Then, repeat:
  - *Track new citations*
  - *Skim new papers, add to "papers pool" if interesting*
  - *Do it fast. Distinguish exploration time and times working on current projects*

6 New Citations for Papers You Follow Inbox x

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### **New Papers Matching Multiple Alerts**

Cites LXMERT: Learning Cross-Modality Encoder Representations from Transformers · Cites GQA: A New Dataset for Real-World Visual Reasoning and Compositional Question Answering

#### **Logically at the Factify 2022: Multimodal Fact Verification**

**Jie Gao, Hella-Franziska Hoffmann, ... Anil Bandhakavi**

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  - *Do it fast. Distinguish exploration time and times working on current projects*
4. Each conference (NAACL, ACL, EMNLP), read (or watch!) the papers most related to your work (< 10)

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“A paper is not a random collection of some experiments you ran that you report on. The paper sells a single thing that was not obvious or present before.”, Andrej Karpathy <sup>1</sup>

<sup>1</sup> [Andrej Karpathy \(Director of AI in Tesla, previously OpenAI, CS231n, PhD @ Stanford\): A Survival Guide to a PhD](#)

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- *What is the main idea of the method they use?*

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- *Persist, you will improve over time*
- *Can you **TLDR** it in 2-3 sentences?*
- *What is the problem this paper deals with?*
- *What is the main idea of the method they use?*
- *What are their main findings? (Not results)*
  - *Can you relate their work to your project? How?*

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- *(Continue reading if it's useful for your project)*

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## 1. Figure 1

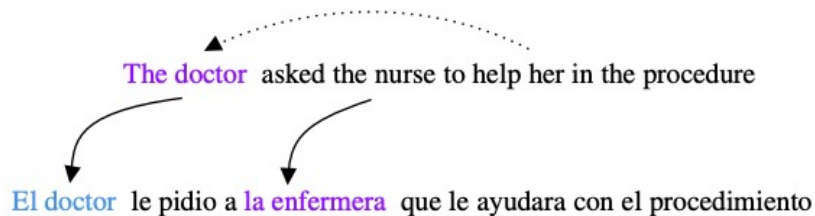


Figure 1: An example of gender bias in machine translation from English (top) to Spanish (bottom). In the English source sentence, the nurse's gender is unknown, while the coreference link with "her" identifies the "doctor" as a female. On the other hand, the Spanish target sentence uses morphological features for gender: "el doctor" (male), versus "la enfermera" (female). Aligning between source and target sentences reveals that a stereotypical assignment of gender roles changed the meaning of the translated sentence by changing the doctor's gender.

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2. Other figures if informative enough

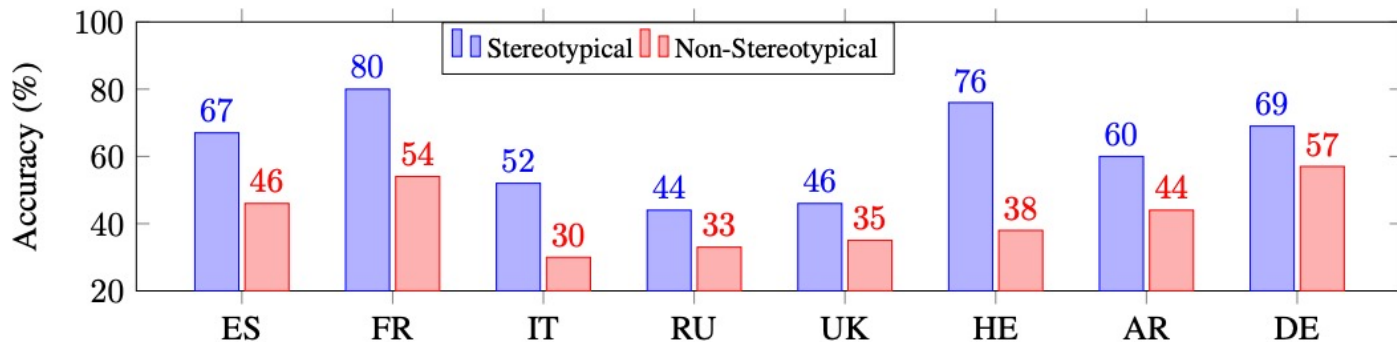


Figure 2: Google Translate's performance on gender translation on our tested languages. The performance on the stereotypical portion of WinoMT is consistently better than that on the non-stereotypical portion. The other MT systems we tested display similar trends.

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  3. Abstract

### Abstract

We present the first challenge set and evaluation protocol for the analysis of gender bias in machine translation (MT). Our approach uses two recent coreference resolution datasets composed of English sentences which cast participants into non-stereotypical gender roles (e.g., “The doctor asked the nurse to help *her* in the operation”). We devise an automatic gender bias evaluation method for eight target languages with grammatical gender, based on morphological analysis (e.g., the use of female inflection for the word “doctor”). Our analyses show that four popular industrial MT systems and two recent state-of-the-art academic MT models are significantly prone to gender-biased translation errors for all tested target languages. Our data and code are publicly available at [shorturl.at/dimuD](https://shorturl.at/dimuD).

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  4. Introduction "In this paper" part

## 1 Introduction

Learned models exhibit social bias when their training data encode stereotypes not relevant for the task, but the correlations are picked up anyway. Notable examples include gender biases in visual SRL (cooking is stereotypically done by women, construction workers are stereotypically men; [Zhao et al., 2017](#)), lexical semantics ("man is to computer programmer as woman is to homemaker"; [Bolukbasi et al., 2016](#)), and natural language inference (associating women with gossiping and men with guitars; [Rudinger et al., 2017](#)).

**In this work**, we conduct the first large-scale multilingual evaluation of gender-bias in machine translation (MT), following recent small-scale qualitative studies which observed that online MT services, such as Google Translate or Microsoft Translator, also exhibit biases, e.g., translating nurses as females and programmers as males, regardless of context ([Alvarez-Melis and Jaakkola, 2017](#); [Font and Costa-Jussà, 2019](#)). Google Trans-



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  6. Result tables

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# Document into your toolbox

TLDR stuff that you may want to use in the future

1. It should be *searchable*
2. The more probability you'll use it - the longer it should be

bias

0/33



NAACL 21: Measuring Social **Bias**es in Grounded Vision and Language Embeddings. <https://www.aclweb.org/anthology/2021.naacl-main.78.pdf>.

TLDR: Measuring social **bias**es in V+L word embeddings.

Contributions:

1. Adapting WEAT and SEAT metrics to V+L, ("Grounded" versions).
2. A new dataset for testing **bias**es in grounded systems (a small collection of men/women V+L pairs with stereotypes).
3. Findings ->.

Findings about social **bias**:

1. Grounded word embeddings have social **bias**es.
2. Grounded evidence has little impact on social **bias**es. (In other words - currently the image doesn't assist enough overcoming the **bias**).
3. **Bias**es mostly comes from the language modality.

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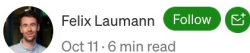
## Exploration

1. Many ideas in deep learning take inspiration from other fields
2. Diverse interests → rich repertoire
3. Follow important trends in the field

## NeurIPS 2021—10 papers you shouldn't miss

2334 papers, 60 workshops, 8 keynote speakers, 15k+ attendees. A dense landscape that's hard to navigate without a good guide and map, so here are some of our ideas!

### Top NLP Trends and Predictions 2022: is NeuralSpace set up for the future of NLP?



Oct 11 · 6 min read





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- Once you've mastered the foundational abilities
- “The Adjacent Possible”
  - The next big ideas in any field are found right beyond the current cutting edge, in the adjacent space that contains the possible new combinations of existing ideas”.<sup>1</sup>



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- Use your tool-box



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## Examples

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- From Jia-Bin Haung:
  - Find a different dimension, Ex: Text / audio / image / video / graph
  - Combine two ideas/problems
  - Add an adjective

Given an existing idea X, add an adjective to make it

- slow → fast
- batch → online
- sensitive → robust
- centralized → distributed
- single-step → progressive
- single-level → hierarchical
- fixed → adaptive, sth-aware
- data-hungry → data-efficient

and so on

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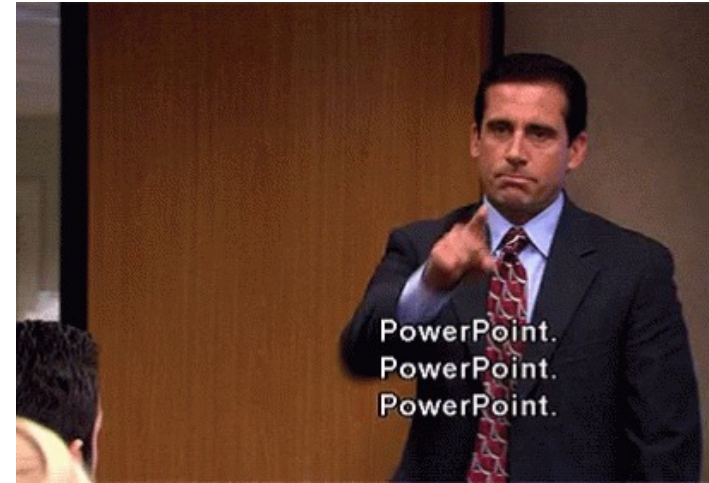
- Your future employers actually wants your supervisor. He is busy... they hope you are the most similar version they can afford<sup>1</sup>
- Repeating (weekly?) meetings
- The meetings are for you <sup>2</sup>
  - It's not a test. Receive feedback.
  - **Leshem**: If you don't need / not ready for the meeting - cancel it
  - **Gabi**: Never cancel a meeting

<sup>1</sup> Cal Newport

<sup>2</sup> Leshem Choshen

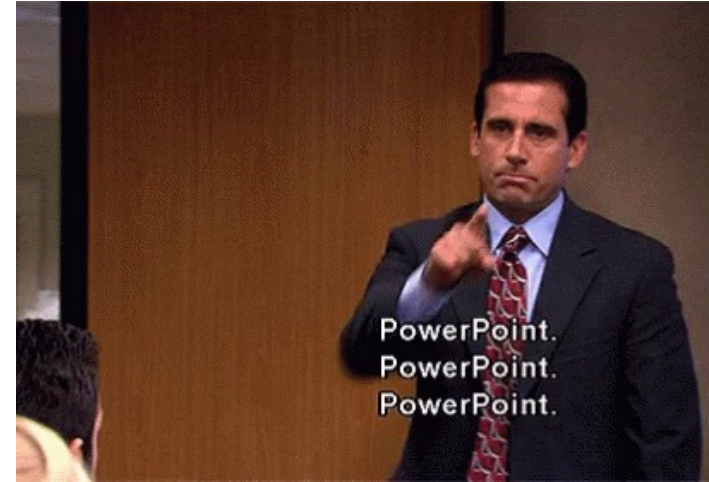
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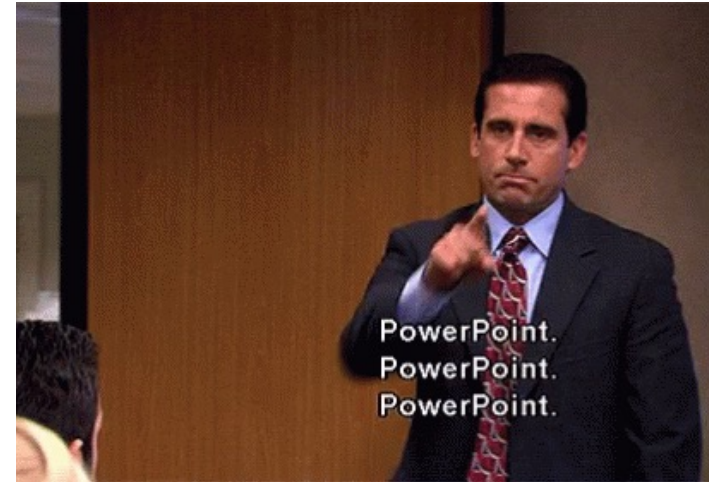
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- Write action items for next meeting
- Try their suggestions (all of them)

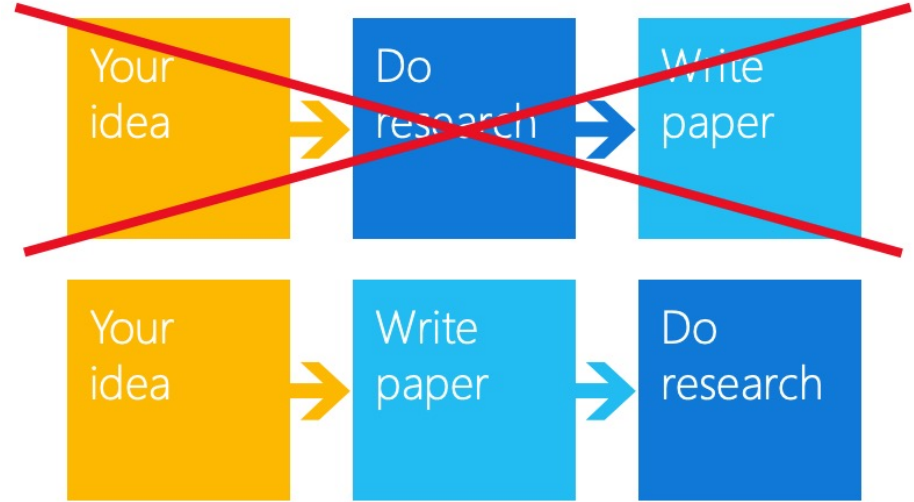


# Planning Research Tips



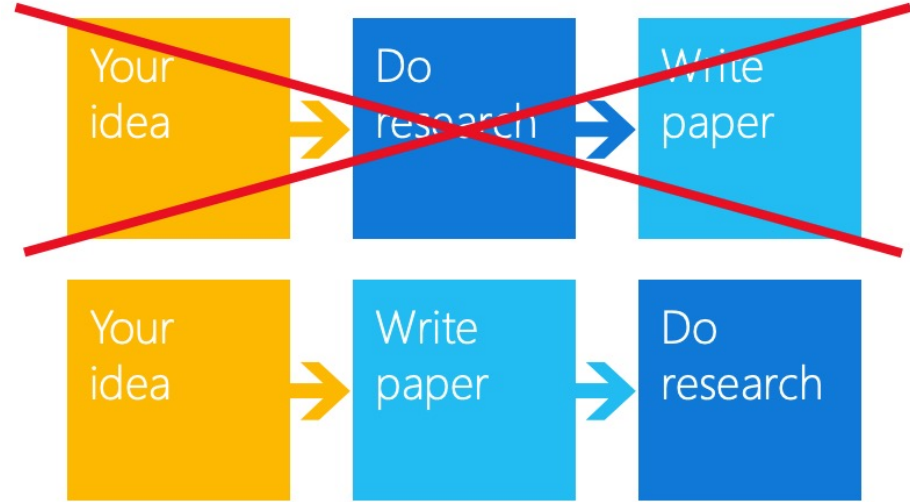
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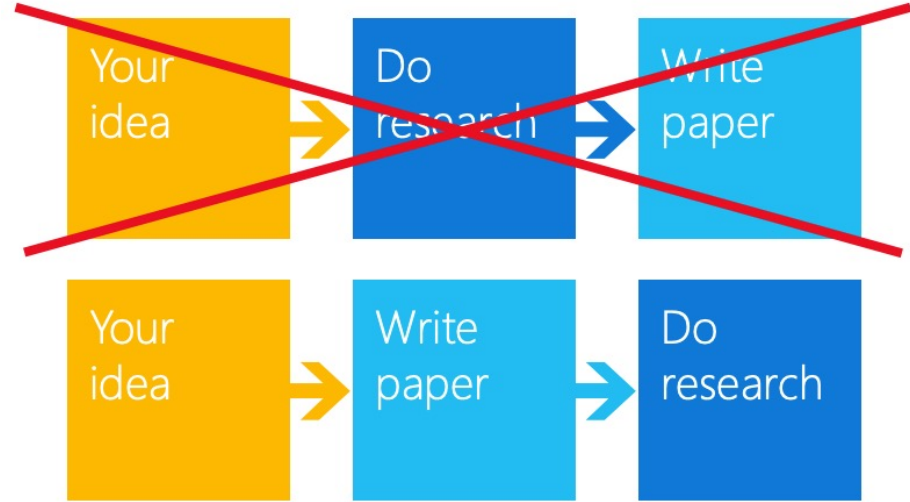
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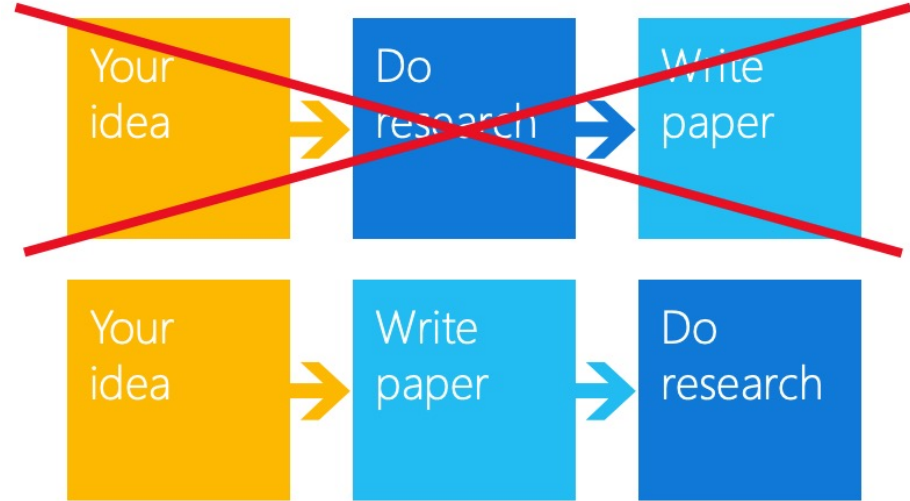
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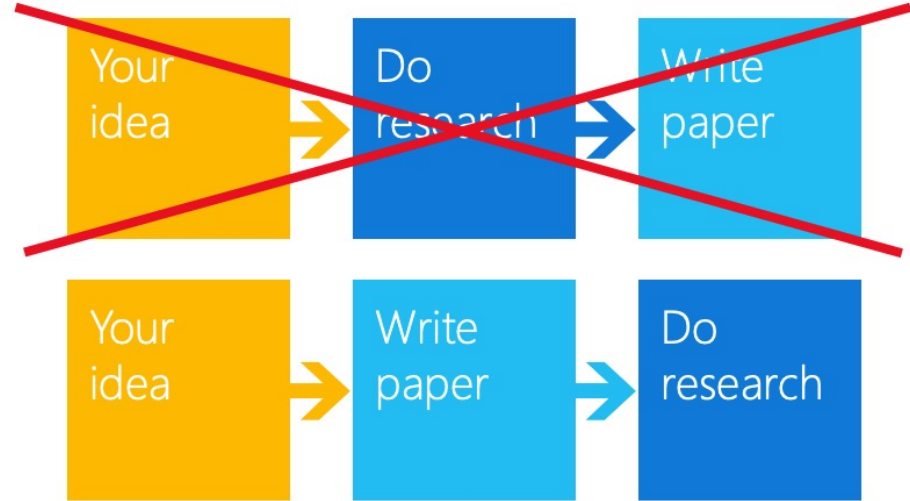
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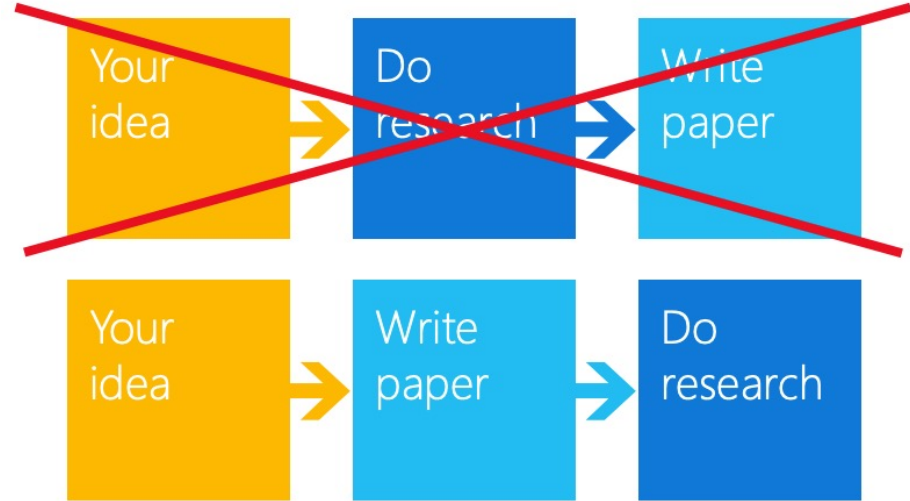
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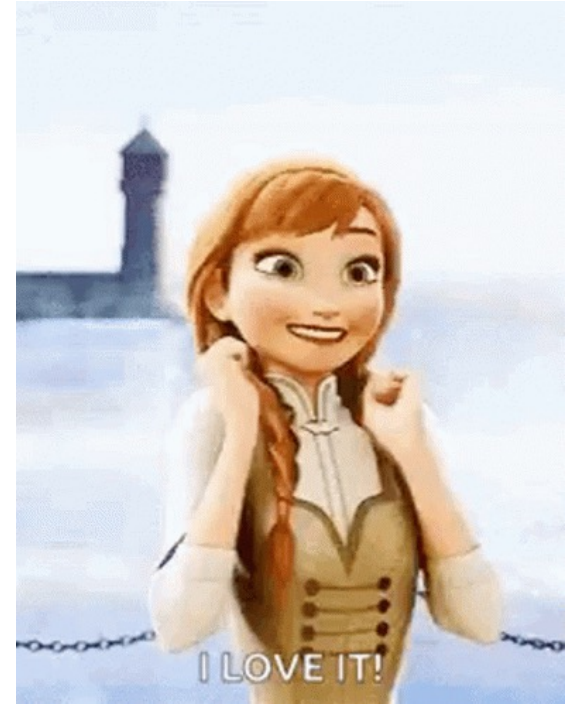


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- Can you “sell” your abstract?
- Are the experiments convincing?
- Leave the **Introduction** for last

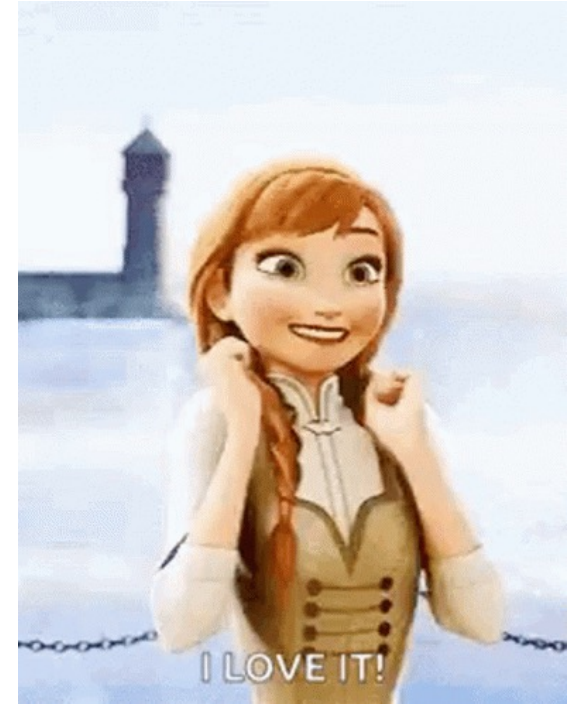


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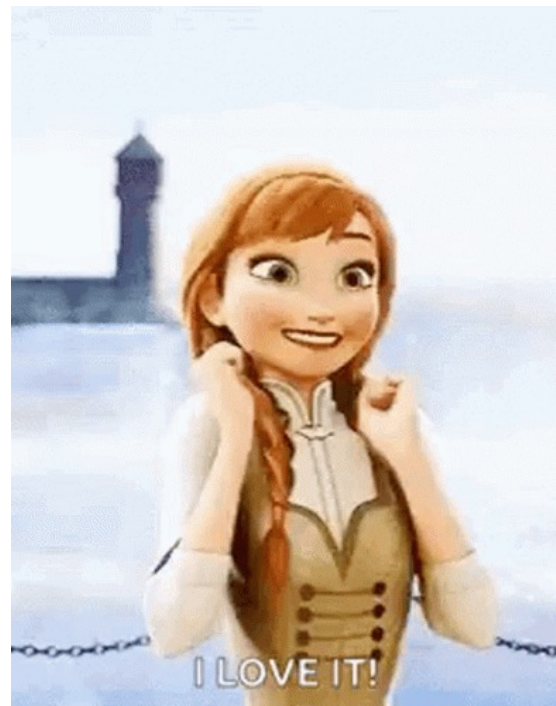
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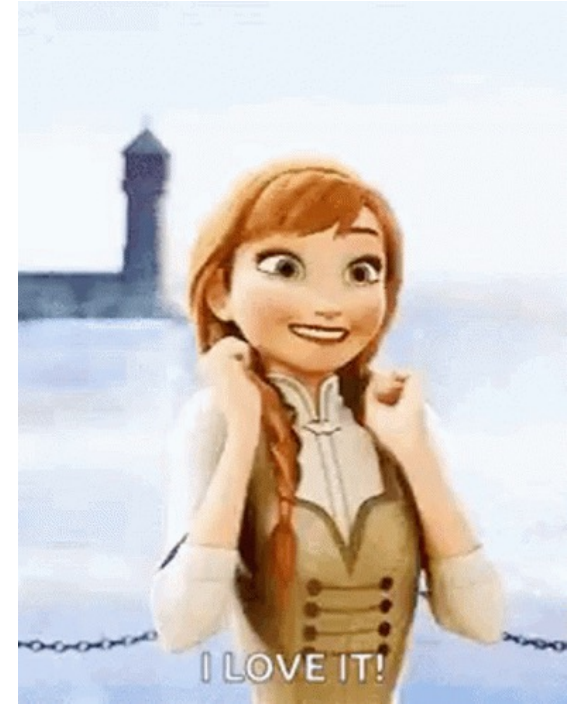
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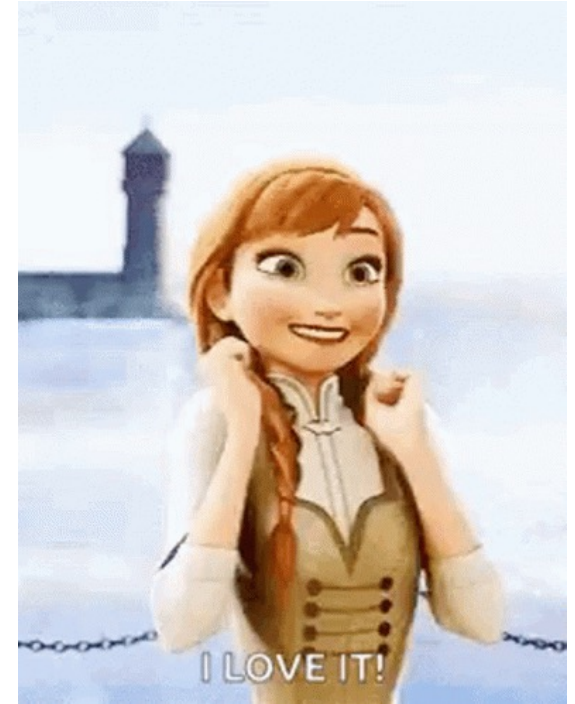
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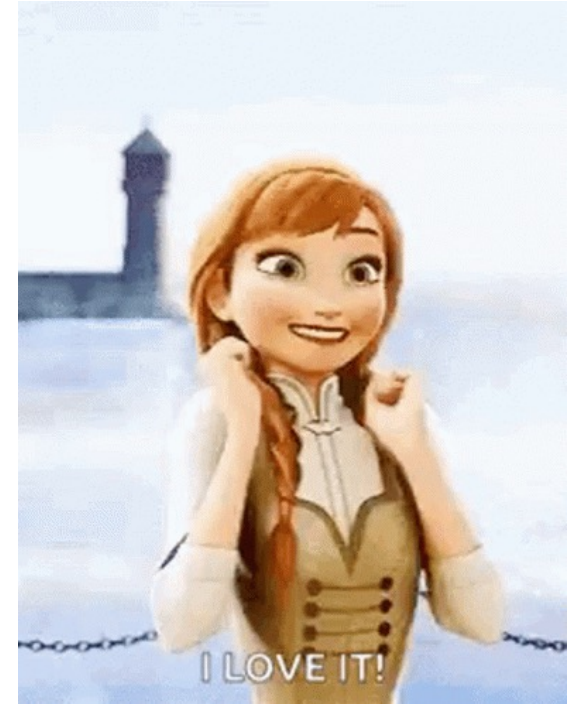
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- There will always be criticism
- The sole purpose is to help make the work better
- Seek criticism and be open



# Connect with your research community





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“If I have seen further, it is by standing on the shoulders of giants.” – Isaac Newton

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  - Schedule a coffee



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“The idea is like grass. It craves light, likes crowds, thrives on crossbreeding, grows better for being stepped on.” – Ursula K. Le Guin





Himan Abdollahpouri

@himan\_abd

Best advice I ever received from one of my professors 5 years ago:

And MSc

“The output of your PhD is not your thesis. The output is YOU”

#Phd

Thank You! 🙏

Take a look on Ronen Tamari’s [“Dark Research” talk](#):

“Dark Research”

