

How to find innovative ideas for your project

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Frequent questions

- Is my idea a good one?
- How to find a good idea?

Criteria of “good work” (in reviewers’ eyes)

1. Novel

2. Convincing

Criteria of “good work” (in reviewers’ eyes)

1. Novel

- *Bored reviewers want surprise!*
- *Has anyone done similar things before you?*

2. Convincing

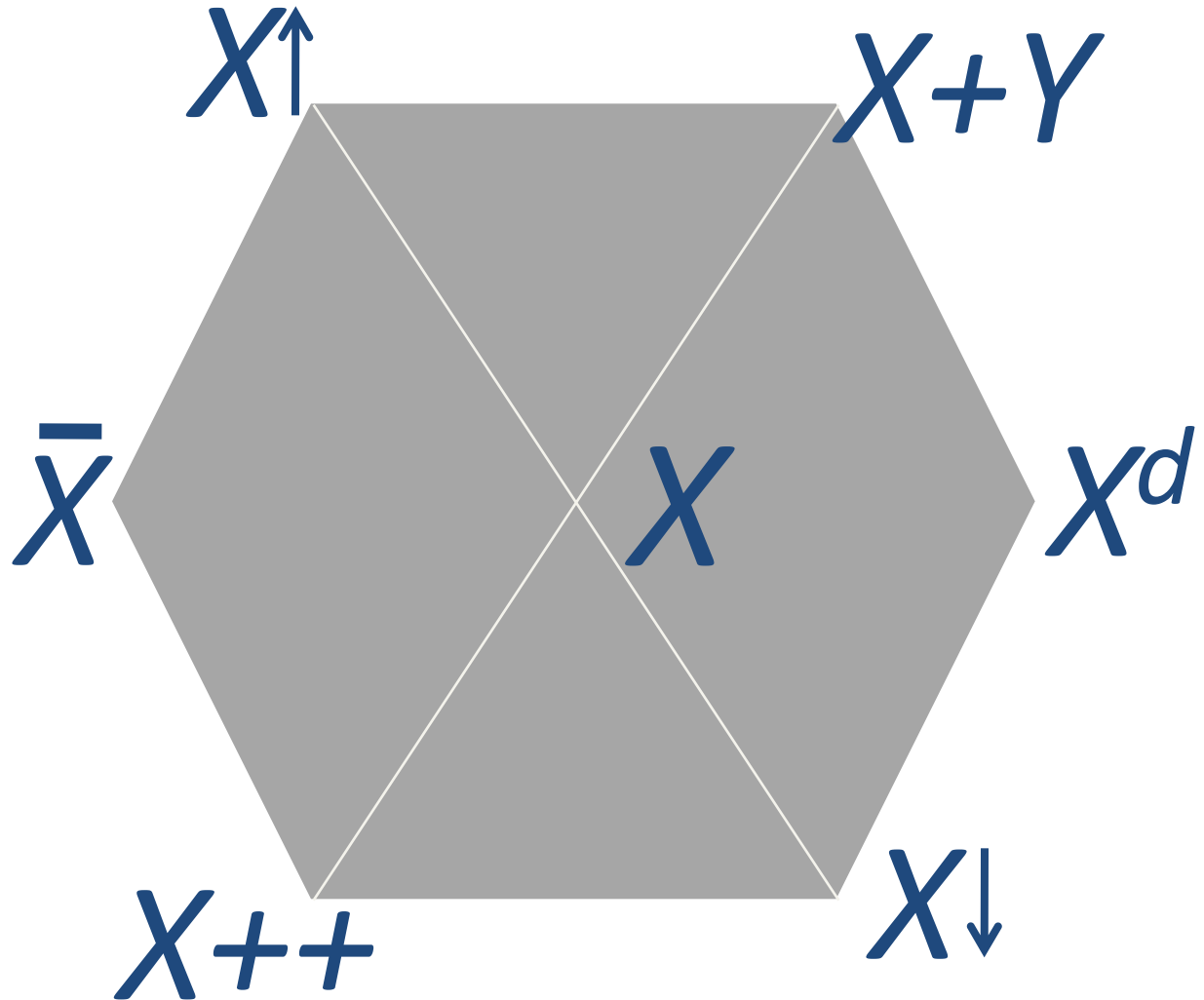
- *Reviewers are suspicious.*
- *Reviewers want convincing insights and solid experiments
(or solid math proof).*

Smell of risk ideas

- Do something you are not excited or interested
 - *Quickly you will find your time is wasted.*
- Do it because it can be done
 - *Where is the “impossible”?*
- Overlook the previous literature
 - *Where is the innovation?*
- Very complicated pipeline at your first try
 - *Some pieces may easily fall down.*

Ramesh Raskar's Innovation Hexagon

*Six ways of
coming up with
new ideas based
on an idea 'X'.*



What are good candidates of “X”s

- Successful applications or algorithms
- (Very) recently developed
or
- Has been there for long but overlooked

Brainstorm list of good “X”s

Try to name 3+ by discussing with the students next to you

- Successful applications
 - Translation
 - Speech2text
 - Face recognition
 - Question answering
 - ...
- Recent cool algorithm
 - Batch (re)normalization
 - Residual network ...
 - Generative Adversarial Networks (GAN) and Wasserstein GAN
 - Fast text
 - SqueezeNet
 - Distilling
 - Curriculum learning

My list is strongly biased but hopefully could give you some idea

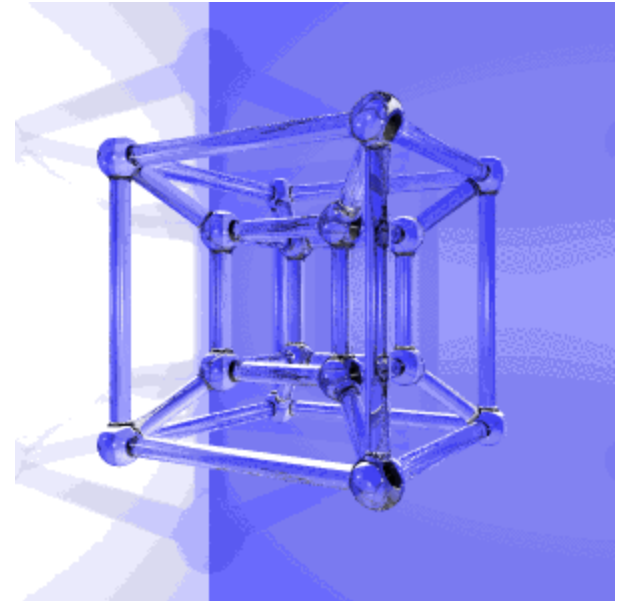
Examples of innovation hexagon

Note a lot of following slides are courtesy to Ramesh Raskar

I only add some examples of deep learning research.

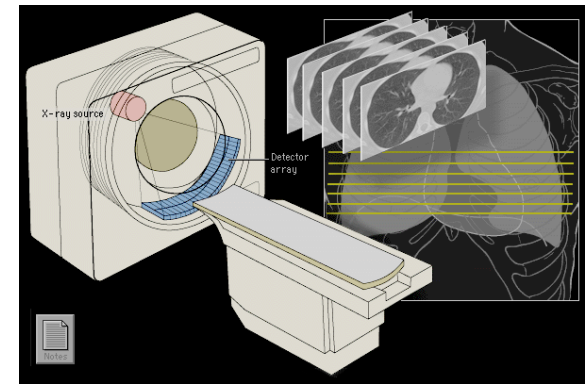
Strategy #1: X^d

- Extend it to next dimension (or some other) dimension
 - Flickr to Youtube
 - Images to infrared, sound, ultrasound to EM spectrum
- Example in deep learning
 - Image based question answering
 - => Album/video based QA



Strategy #2: X+Y

- Fusion of the dissimilar
 - More dissimilar, more spectacular the output
- Example
 - Scientific imaging + Photography
 - Coded aperture
 - Tomography
- Keys
 - Deep insights on why $X + Y$
 - The connection should be a surprise!
- Example of surprise:
 - Visually indicated sound



Strategy #3: \bar{X}

Do exactly the opposite



Bundesarchiv, Bild 183-50305-0030
Foto: Mittelstadt, Rainer | 5. März 1977

Straddle Method for High Jump

Replacement of landing surfaces with foam rubber



http://en.wikipedia.org/wiki/Dick_Fosbury



Strategy #3: \bar{X}

Do exactly the opposite

- From LSTM to word2vec
 - LSTM: long context and deep model
 - Word2vec: small window and shallow network

Word2vec works well for word embedding!
- From [Faster R-CNN](#) to [YOLO](#)/[SSD](#)
 - [Faster R-CNN](#): object proposal as regions and then detect
 - [YOLO](#)/[SSD](#): predict the bounding box simultaneously with detecting

YOLO/SSD are faster!

Strategy #4:



- Given a Hammer ..
 - Find all the nails
 - Sometimes even screws and bolts
- Given a cool solution/technique/Opportunity
 - Find other problems
- Examples
 - CNN was first developed for vision
 - Now it has been used for speech and text classification

Strategy #5: X ↓

- Given a nail,
 - Find all hammers
 - Sometimes even screwdrivers and pliers may work
- Discover a problem,
 - Find possible solutions
- Examples
 - [AlphaGo](#)

Strategy #6: X++

- Pick an adjective:

$$\text{neXt} = \text{adjective} + \text{X}$$

Examples:

- Faster: e.g., [R-CNN](#) -> [Fast R-CNN](#) -> [Faster R-CNN](#)
- Cheaper: e.g., [AlexNet](#) -> [Squeeze Net](#)
- More efficient: e.g., [Distilled learning](#)
- ...

Pitfalls

- These six ways are only a start
- They are good for projects which help you start your research career
- But
 - Significant innovations may not share a pattern
 - The risk of following patterns exists by creating a problem which does not exist in real
- The important thing is to develop your own deep insights
 - What is the true challenge?
 - How are you going to solve it?