

CS231n Project Design

Chris Waites

Agenda

1. Project expectations
 - a. Does my project meet expectations?
 - b. FAQs
2. Picking a project idea
 - a. Sources of inspiration
 - b. Reading papers efficiently
3. Proposal, milestone, and final report
 - a. Due dates, expectations, logistics
 - b. Support

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Project expectations

The course project is a (fun) way to explore concepts taught in the course on a topic of your choice!

- Fairly open-ended, anything related to vision ([link to project page](#))

Completed in groups of 1, 2, or 3 people

- Project expectations are higher for groups with more people

Generally, two tracks of work:

- **Applications:** If you have a specific background or interest (e.g. biology, engineering, physics), we'd love to see you apply ConvNets to problems related to your particular domain of interest.
- **Models:** You can build a new model (algorithm) and apply it to tackle vision tasks. This track might be more challenging, and could lead to a piece of publishable work.

Project expectations

The final report has the following structure:

- Title, Author(s)
- Abstract
- Related Work
- Data Description
- Methods
- Experiments
- Conclusion
- Supplementary Material (optional)

FAQ: Does my project meet expectations?

Rule of thumb:

- ***How much effort are you putting into your project?***

Strong projects might...

- Propose a novel variant of a technique (which takes a lot of effort)
- Adapts an existing technique to a totally new problem (which takes a lot of effort)

Weaker projects might...

- Spend several weeks collecting/cleaning data rather than testing hypotheses
- Clone an existing repo and do minimal stitching to make it work for a Kaggle competition

FAQ: Does my project meet expectations?

So, this **doesn't** mean:

- Your project has to be strictly novel to get a good grade (although, we encourage this!)
- You have to beat the state-of-the-art performance to get a good grade (you don't have to come up with the next best object detector to test an interesting hypothesis)

This **does** mean:

- You need to put a significant effort into your investigation, and you may have to try many different approaches

In your **analysis**, ask yourself:

- Are you *interpreting* and *understanding* your results, or merely stating them?
- Are you just plotting a loss curve, or are you evaluating the results of your approach from many different angles?

Project FAQs

Q: Can I apply convolutional networks to a purely NLP / audio / stock price problem?

- **A:** This is a computer vision course, so ***you must incorporate visual data in some form.***

Q: Can I change my project after the proposal, before the milestone?

- **A:** Yes - the proposal is to make sure you have a plausible project direction. If you need to change project directions, we understand.

Q: Can I change my project ***after*** the milestone?

- **A:** In general, we do not encourage this. At this point in the course, there will be little time to put together a sufficient project.

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Picking a project idea

First and foremost:

Do what is important or interesting to you, *not* what seems easiest.

- You will be far more motivated if you're invested in what you're doing
- What do you *really* care about? Healthcare? Self-driving cars? Surveillance? Sports? Ethics? You can probably find its intersection with computer vision

Practical considerations:

1. **Data:** Is there existing data for this problem? Will I need to spend weeks collecting it myself?
2. **Code & framework:** Will I have to implement this myself, or is there an existing implementation?

Picking a project idea

Conferences:

CVPR: IEEE Conference on Computer Vision and Pattern Recognition

ICCV: International Conference on Computer Vision

ECCV: European Conference on Computer Vision

NeurIPS: Neural Information Processing Systems

ICLR: International Conference on Learning Representations

ICML: International Conference on Machine Learning

*Note: Do **not** even begin to try to read through all of these papers, or even their titles. There are far too many. Use CMD+F to find papers with relevant keywords.*

Picking a project idea

Additional resources:

- [Stanford Vision Lab Publications](#)
- [Awesome Deep Vision](#)
- [Papers With Code](#)
- [Kaggle](#)
- [Previous CS229 Projects](#)

Reading papers

Do **not** read a paper linearly on your first pass

- First, read the abstract (word for word) as well as the figures & captions
- Does the paper still seem relevant? If so, read the methods as well as the results
- Finally, read the entire paper linearly (if the additional detail seems useful)

Papers are not always the most efficient way to digest an idea. Also try looking around for:

- Talks, videos, or blog posts on the topics
- Github repos, containing actual code for the idea

Reading papers

Example:

You Only Look Once: Unified, Real-Time Object Detection

Joseph Redmon*, Santosh Divvala*[†], Ross Girshick[¶], Ali Farhadi*[†]

University of Washington*, Allen Institute for AI[†], Facebook AI Research[¶]

<http://pjreddie.com/yolo/>

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Deliverables

Due dates:

- Proposal (4/18) - Monday!
- Milestone (5/10)
- Final report (6/3)
- Final presentation (6/4) - Video submitted to gradescope, async.

Milestone (Due 5/10)

1-2 page progress report, more or less containing:

- 1. Literature review (3+ sources)**
- 2. Indication that code is up and running**
- Data source explained correctly
- What Github repo or other code you're basing your work off of
- 5. Ran baseline model have results**
 - a. Yes, points are taken off for no model running & no preliminary results
- Data pipeline should be in place
- Brief discussion of your preliminary results

Support: CA areas of specialty

Day	Staff		
Mon	Kevin <i>Robotics, self supervised learning</i>	Haofeng <i>Detection, segmentation, multiple object tracking, video understanding, interactive annotation</i>	
Tue	Rachel <i>NLP (vision+language), Amazon Mechanical Turk, creating custom datasets, RL, robotics, 3D vision, AI for medical imaging.</i>	Geet <i>Performance (e.g. pruning, quantization), recommendation/embedding, sparse networks, transfer Learning, videos, RNNs</i>	Mandy <i>Action recognition, videos, medical imaging, optimization/learning methods</i>
Wed	Chris <i>Generative models (GANs), privacy, fairness & interpretability</i>	Sam <i>Video understanding, action recognition, speech recognition, bias, multi-task learning, medical imaging</i>	Nishant <i>Videos, unsupervised learning, self-supervised learning, 3D vision</i>
Thu	Guanzhi <i>RL, robotics, videos, self-supervised learning</i>	Russell <i>Videos, action recognition, graph neural networks, domain adaptation, image segmentation / denoising</i>	Sean <i>Graphics, image editing, virtual reality, 360 degree videos</i>
Fri	Yichen <i>3D vision, transfer learning, domain adaptation</i>	JQ <i>Image classification, image augmentation, medical imaging (classification)</i>	Lin <i>Robotics, RL, 3D vision</i>

Questions?