

Your first machine learning project

Sayak Paul | Deep Learning Associate at [PyImageSearch](#)

January 04, 2020

Mehsana, Gujarat, India



Agenda

- Your first ML project
 - Thinking about a problem statement
 - Things to consider for the project
 - Executing the project
 - Presenting the project and having feedback
 - ...
- Next steps

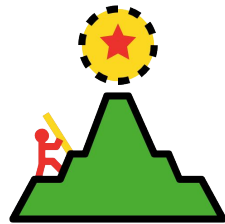
Motivation



[Source](#)

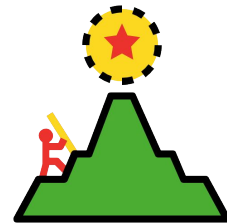
Your first ML project - Challenges

- How much should I know before I start my ML project?

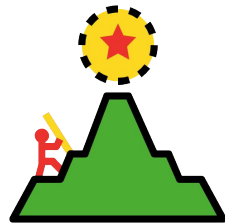


Your first ML project - Challenges

- How much should I know before I start my ML project?
- ML is **interdisciplinary**. How to not get overwhelmed?

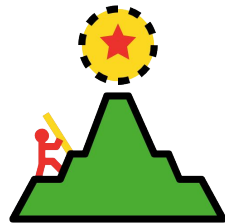


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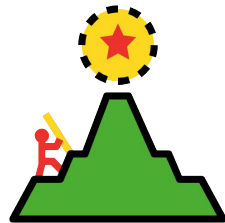


- How much should I know before I start my ML project?
- ML is **interdisciplinary**. How to not get overwhelmed?
- What problem statement do I choose for the project?
- Lots of moving parts in the project. Where do I start?

**Let's talk about
solutions now!**

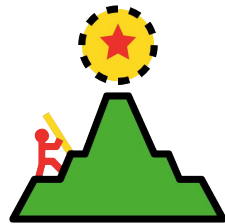


Your first ML project - Solutions



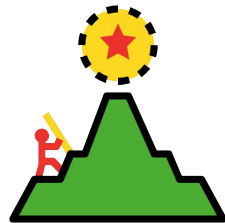
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Your first ML project - Solutions



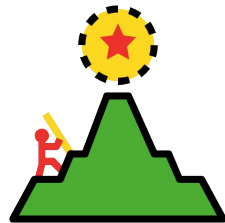
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 - You ***don't need to know everything*** before you start the project. Just basic concepts should be enough to start off.
 - I typically follow a **60:40** ratio (**things I know** : **learning opportunities**).
Make your own ratios and adjust them over time.

Your first ML project - Solutions



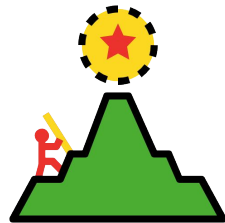
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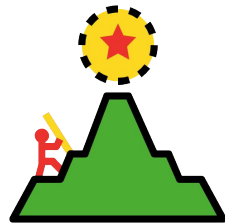
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Your first ML project - Solutions



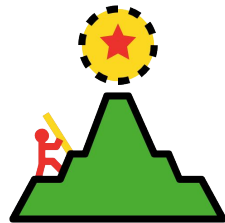
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 - Don't rush it out!

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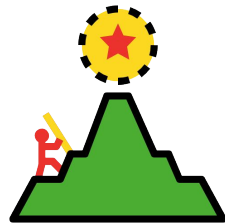
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 - Don't rush it out!
 - **So, how much understanding is needed?**

Your first ML project - Solutions



- How much should I know before I start my ML project?
- ML is **interdisciplinary**. How to not get overwhelmed?
 - [...]
 - Don't rush it out!
 - **So, how much understanding is needed?**
 - **Just enough to convince yourself!** Understanding can always be iterated!

Your first ML project - Solutions



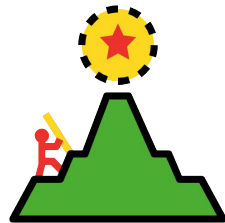
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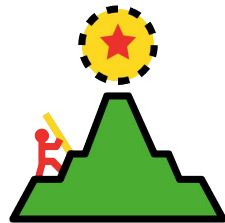
sayak.dev

Your first ML project - Solutions



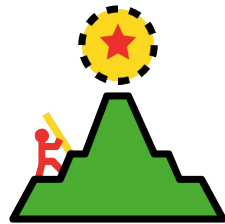
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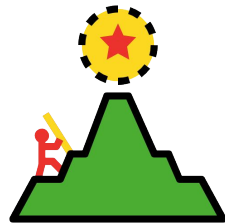
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 - Decide if the problem is machine learning friendly ([problem framing](#)).

Your first ML project - Solutions



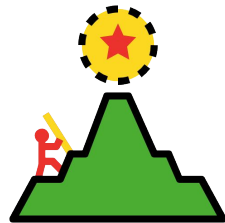
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 - Try identifying the problems in your surroundings - places you visit, platforms you use and so on.
 - Decide if the problem is machine learning friendly (problem framing).
 - Participate in **Kaggle competitions**, but ...

Your first ML project - Solutions



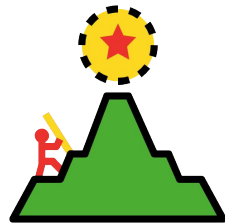
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 - Don't chase the leaderboard!

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- ML is **interdisciplinary**. How to not get overwhelmed?
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 - [...]
 - Participate in **Kaggle competitions**, but ...
 - Don't chase the leaderboard!
 - Always think of the bigger picture.

Your first ML project - Solutions

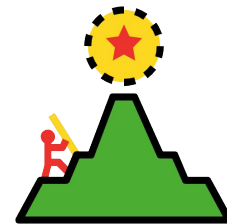


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- What problem statement do I choose for the project?
 - [...]
 - Participate in **Kaggle competitions**, but ...
 - [...]
 - Here's an amazing resource to learn creative ML:
<https://mlwave.com/>

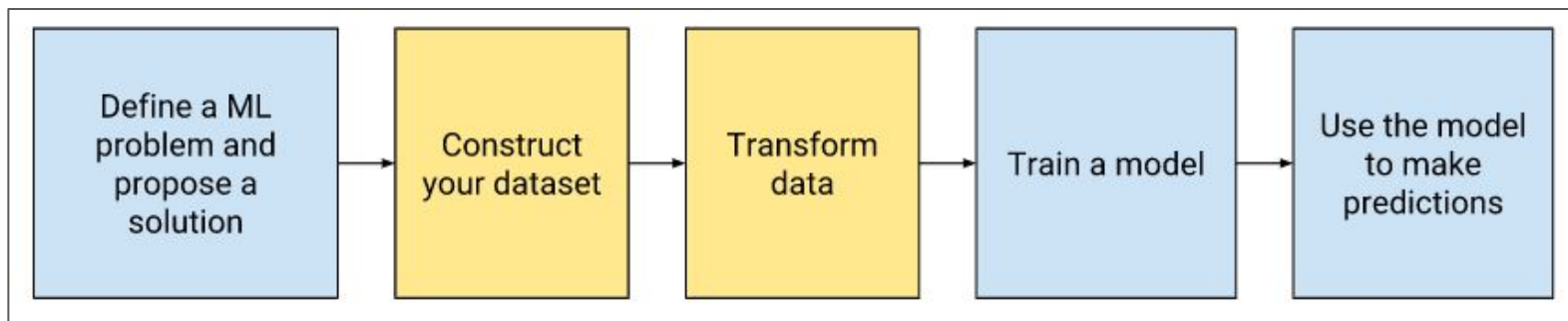
Executing an ML project: Some thoughts



Some things to consider



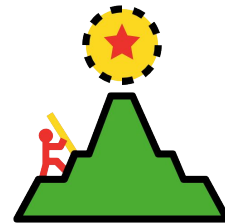
- I find it consistent to start with a **sequential** flow:



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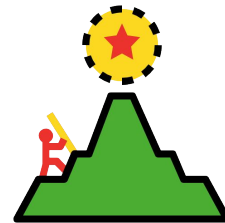
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- I find it consistent to start with a **sequential** flow.
- **Data!**

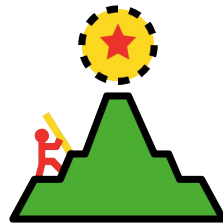


Some things to consider

- I find it consistent to start with a **sequential** flow.
- **Data!**
 - How do I collect data?

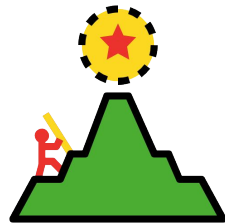


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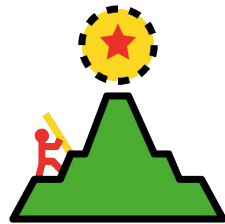
- I find it consistent to start with a **sequential** flow.
- **Data!**
 - How do I collect data? (Kaggle and other online resources, web scraping, manually collect data)

Some things to consider



- I find it consistent to start with a **sequential** flow.
- **Data!**
 - How do I collect data?
 - Is there a similar dataset available?

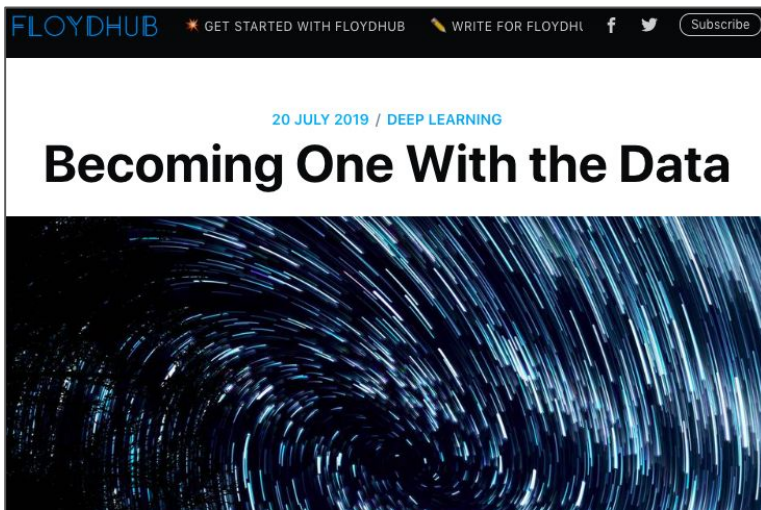
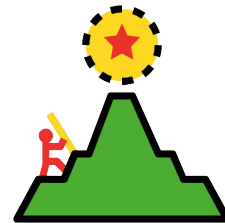
Some things to consider



- I find it consistent to start with a **sequential** flow.
- **Data!**
 - How do I collect data?
 - Is there a similar dataset available?
 - How do I become one with the data?

Some things to consider

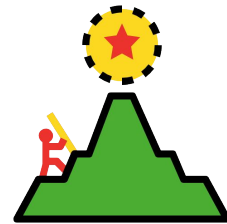
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Article [link](#)

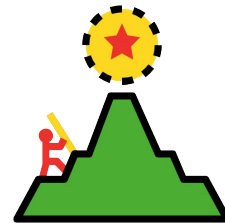
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- **Modeling!**

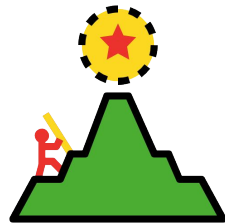


Some things to consider

- I find it consistent to start with a **sequential** flow.
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 - Which model should I consider?

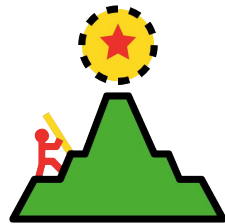


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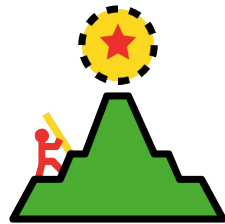
- I find it consistent to start with a **sequential** flow.
- **Data!**
- **Modeling!**
 - Which model should I consider?
 - How do I train a model?

Some things to consider



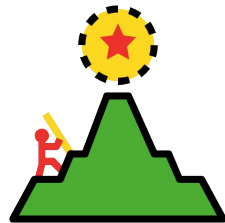
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 - How can I train my model faster?

Some things to consider



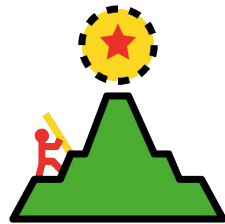
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 - How can I train it better?

Some things to consider



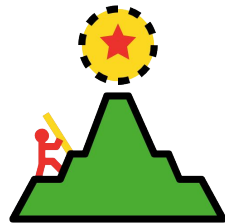
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Some things to consider



- I find it consistent to start with a **sequential** flow.
- **Data!**
- **Modeling!**
 - Which model should I consider?
 - How do I train a model?
 - How do I validate a model?
 - How do I debug a model?

Some things to consider



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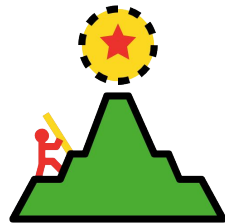
- **Modeling!**

- [...]
- How do I debug a model?

- Check out this course:

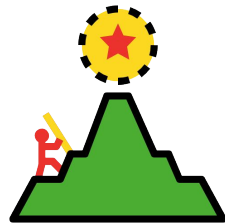
<https://developers.google.com/machine-learning/testing-debugging>

Some things to consider



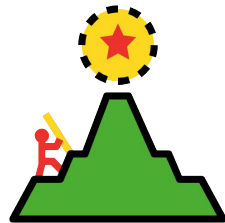
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Some things to consider



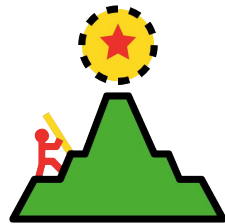
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 - A web application

Some things to consider



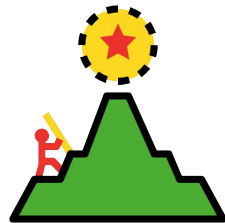
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 - A mobile application

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- **Modeling!**
- Does my ML model integrate well with other systems?
 - A web application
 - A mobile application
 - A Raspberry Pi

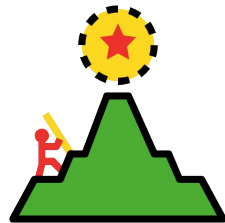
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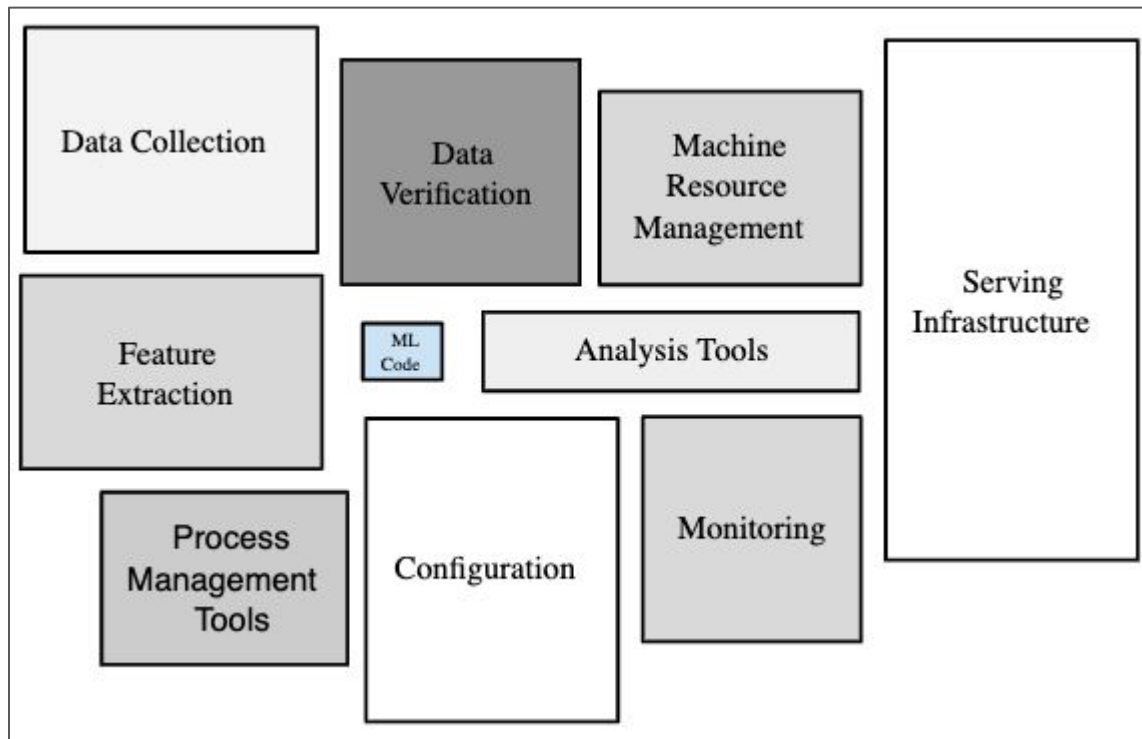
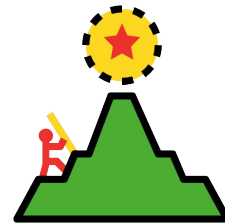
Possibilities are endless here!

Some things to consider



- [...]
- Does my ML model integrate well with other systems?
- Think about the other components of your project too, ***not just ML!***

Some things to consider



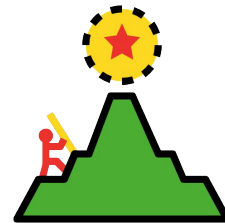
[Source](#)

Presenting your project

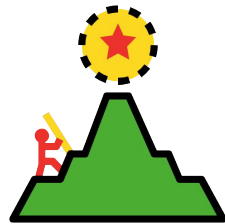


Project out in the wild

- Nothing like a structured GitHub repository

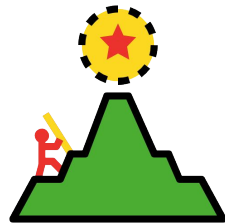


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 - Be sure to add a proper **README** including a demo of your project

Project out in the wild



- Nothing like a structured GitHub repository
 - Be sure to add a proper **README** including a demo of your project
 - A polished directory structure

Project out in the wild

```

apparel_classifier/          # Package that can be deployed as a self-contained prediction system
__init__.py

apparel_predictor.py        # Takes a raw image and obtains a prediction

datasets/                   # Code for loading datasets
__init__.py
dataset.py                  # Base class for datasets - logic for downloading data (if not available)
fmnist_dataset.py
fmnist_essentials.json
dataset_sequence.py

models/                     # Code for instantiating models, including data preprocessing and loss functions
__init__.py
base.py                    # Base class for models
image_model.py

networks/                   # Code for building neural networks (i.e., 'dumb' input->output mappings) used by models
__init__.py
mlp.py

tests/
support/                    # Raw data used by tests
test_apparel_predictor.py   # Test model on a few key examples

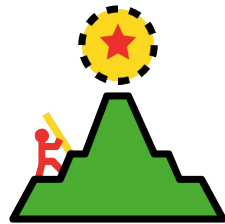
weights/                    # Weights for production model
Image_Model_FMNIST_Dataset_weights.h5

util.py

training/                   # Code for running training experiments and selecting the best model.
run_experiment.py           # Parse experiment config and launch training.
util.py                     # Logic for training a model with a given config
```

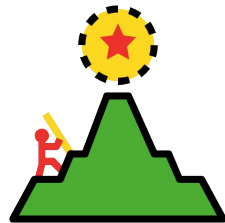
Click to [enlarge](#).

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- Write out a blog on the project and be as detailed as you can be!

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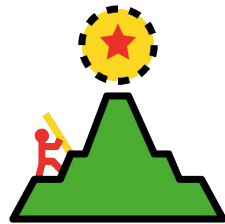


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- [FloydHub AI Writer Program](#)
- [Weights and Biases Content Developers](#)
- [Nanonets Writers](#)

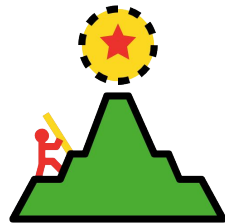
Paid writing opportunities.

Project out in the wild



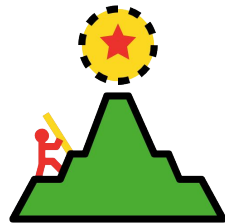
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- Share your project with the communities

Project out in the wild



- Nothing like a structured GitHub repository
- Write out a blog on the project and be as detailed as you can be!
- Share your project with the communities
 - Kaggle
 - [AIDL Facebook Group](#)
 - Twitter
 - [FastAI Forums](#)

Project out in the wild



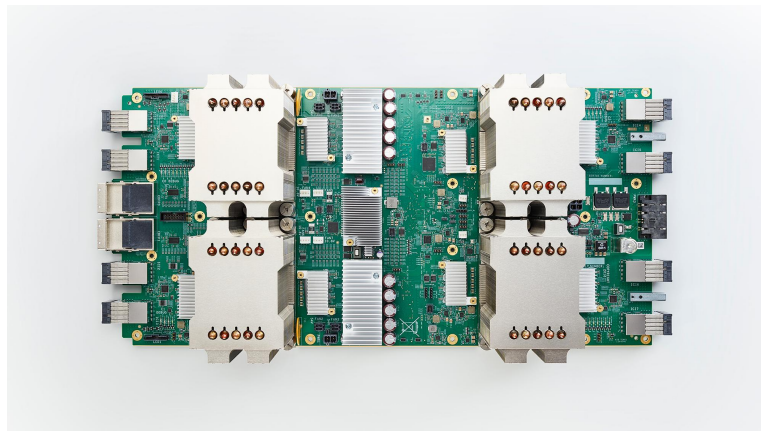
- Nothing like a structured GitHub repository
- Write out a blog on the project and be as detailed as you can be!
- Share your project with the communities
- Be open to constructive feedback.

Departing thoughts



- Figure out what interests you. Machine learning is a huuuuuge field!
 - Some good directions here: <https://www.sayak.dev/interviews>
- Discuss your work with like-minded people.
- Finally, [apply, learn, make mistakes and repeat!](#)

Free Cloud TPUs to support your ML research



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Simply apply here before Jan 7  bit.ly/bfdi-bootcamp



Are you ready to Solve for India?



[#BuildForDigitalIndia](#)

[Source](#)

Just to finish off in style



Slides available here: <http://bit.ly/dloop20>

See you next time



 [@RisingSayak](https://twitter.com/RisingSayak)

Thank you very much :)



Experts

