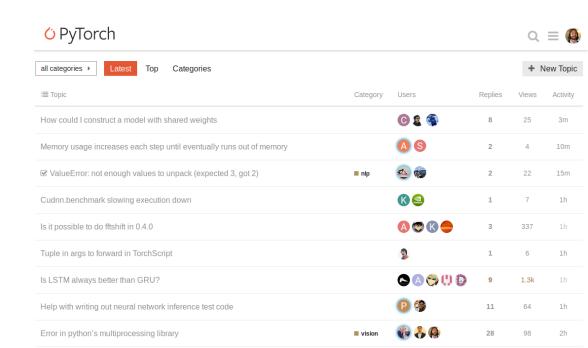
# Semantic Search on PyTorch discussions

#### PyTorch in Munich at Microsoft

- https://discuss.pytorch.org/
- Discussions on deep learning with PyTorch
  - Help with debugging
  - Performance issues
  - Model/training support
  - Feedback
  - **–** ...
- Welcoming community!

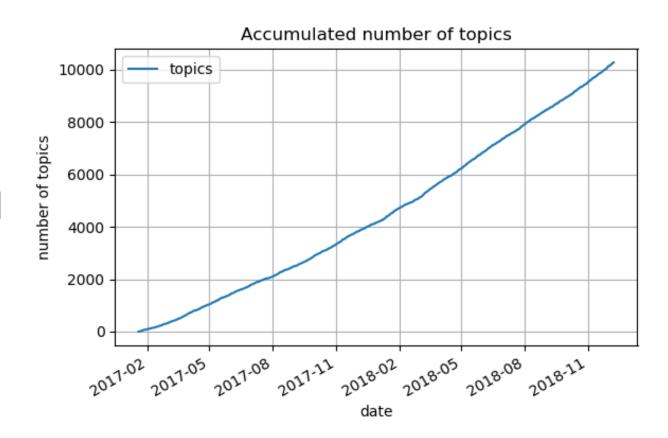


... and I like to hangout in the board!

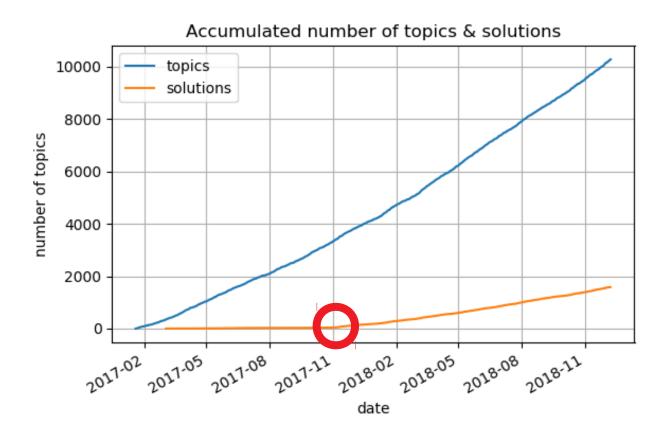
	Received	Given	Topics	Replies 🗸	Viewed	Read	Visits	Time Read
ptrblck	1.4k	286	1	4.0k	13.1k	38.2k	425	20d

So you might have seen me there.

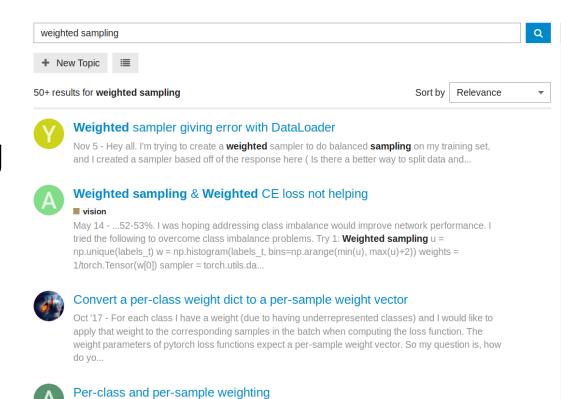
- Some stats
  - ~13,000 topics
  - ~52,000 posts
  - ~1,700 marked solutions



- Some stats
  - ~13,000 topics
  - ~52,000 posts
  - ~1,700 marked solutions



- Search works with keywords (lexical search)
- Fine in a lot of cases
- Semantic search: search with meaning



## Semantic search @github

- Hamel Husain & Ho-Hsiang Wu created semantic search demo for Github code search using deep learning
- Nice blog post: https://towardsdatascience.com/semantic-code-search-3cd6d244a39c
- Used function docstring pairs
- This work is highly inspired by Husain and Wu (thanks a lot for the great blog post and explanations)

## Semantic search @github

api app = ExampleAPI(host='localhost', port=5372)

api\_app.run()

Search

Result

```
Live Semantic Search of Code (Searching Holdout Set Only)
**search
start flask app
WARNING:root:Processing 1 rows
cosine dist:0.1288 url: https://github.com/Fire-Proof/cue-csqo/blob/master/cue csqo/csqo.py#
L97
def start webserver(self):
   app = Flask( name )
   app = self. setup routes(app)
    app.run(port=43555)
cosine dist:0.1294 url: https://github.com/sunary/ank/blob/master/examples/api app/processo
r.py#L13
def start(self):
```

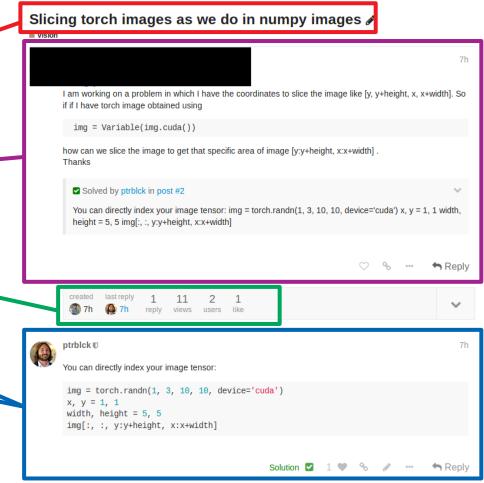
## Semantic search @github

```
Would this also work for our discussion board?
```

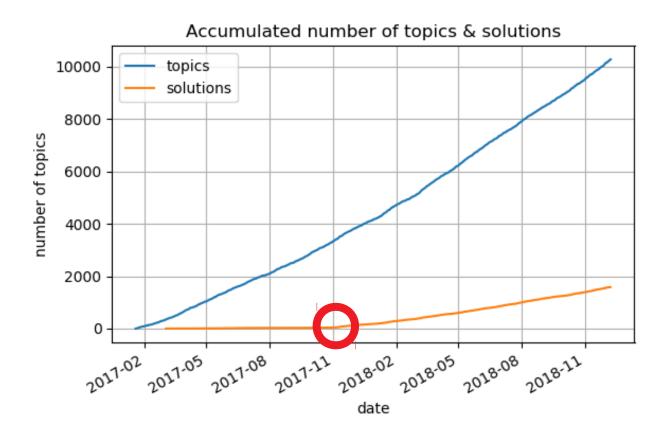
- Threads/topics:
  - Title (name)
  - Question (start post)
  - Stats
  - Solution?
  - Posts



- Threads/topics:
  - Title (name)
  - Question (start post)
  - Stats
  - Solution?
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- Some stats
  - ~**13,000** topics
  - ~52,000 posts
  - ~1,700 marked solutions



• 1,700 marked solutions might not be enough data to train deep learning model

- 1,700 marked solutions might not be enough data to train deep learning model
- But we have 13,000 topics!
- Workflow:
  - Use solution if available
  - Else: take post with highest score (not start post)

## Discourse post score

- Uses
  - Reply count
  - Likes
  - Links
  - Bookmark count
  - Reading time?
  - Number of reads

```
class ScoreCalculator
  def self.default_score_weights
      reply_count: 5,
      like_score: 15,
      incoming_link_count: 5,
      bookmark_count: 2,
      avg_time: 0.05,
      reads: 0.2
  end
```

#### Overview

- Data
- Model
- Loss function
- Training
- Testing

#### Get the data

- All pulled information is public (indexed by Google)
- Use discourse REST API to get all posts
- Save title, question, solution (or highest scored post)
- Create two datasets
  - small dataset (only solutions)
  - Bigger dataset (solutions or best post)

```
id:
                                  80969
 name:
                                  "ptrblck"
 username:
▼avatar template:
                                  "/user avatar/discuss.pytorch.org/ptrblck/{size}/1823 1.png"
                                  "2018-12-10T13:15:41.217Z"
 created at:
                                  "You can directly index your image tensor:\n<code</pre>

▼ cooked:
 post number:
                                  2
 post type:
 updated at:
                                  "2018-12-10T14:11:02.491Z"
 reply count:
 reply to post number:
                                  null
 quote count:
 avg time:
                                  18
 incoming link count:
 reads:
                                  16.7
 score:
```

#### Get the data

- REST API was quite easy to use (although the docs could get some more examples)
- Saved datasets:
  - Small dataset: 1582 threads
  - Bigger dataset: 10,280 threads

#### Get data

- Example of "raw" markdown data:
- "You can directly index your image
  tensor:\n```python\nimg = torch.randn(1, 3, 10, 10,
  device='cuda')\nx, y = 1, 1\nwidth, height = 5,
  5\nimg[:, :, y:y+height, x:x+width]\n```"
- How to clean and preprocess this kind of data?

- Basic approach:
  - Tokenize the raw text
  - Lower all words
  - ...
  - Create language (dictionary)
  - Done!

- Basic approach:
  - Tokenize the raw text
  - Lower all words
  - ...
  - Create language (dictionary)
  - Done!
  - Maybe not :(
- Tokenization of code seems to fail

```
['you', 'can', 'directly',
'index', 'your', 'image',
'tensor', ':', '``', '`',
'python', 'imq', '=',
'torch.randn(1', ',', '3',
',', '10', ',', '10', ','
"device='cuda", "'", ')', 'x',
'width', ',', 'height', '=',
'5', ',', '5', 'img', '[',
'y+height', ',', 'x',
'x+width', 'l',
```

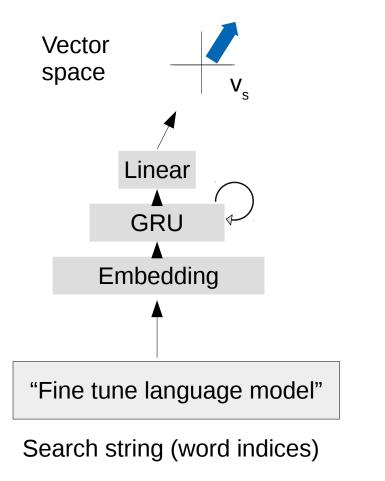
- New approach:
  - Tokenize text and code separately
    - Use regex to get markdown code
    - re\_code = r'(?:(?<!\\)((?:\\{2})+)(?=`+)|(?<!\\)
      (`+)(.+?)(?<!`)\2(?!`))'</pre>
    - (Taken from https://github.com/Python-Markdown/markdown)
    - Also, remove all links (+ image links)
  - Use tokens to create language

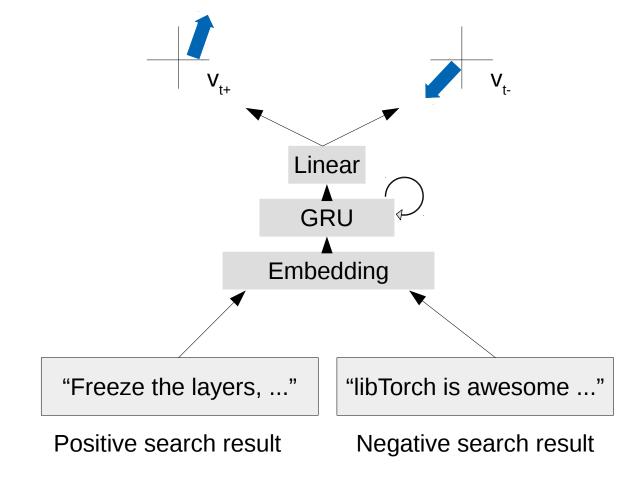
- New approach:
  - Tokenize text and code separately
  - Use tokens to create language (dictionary)
    - Represent each word as one-hot encoded vector
    - Create lookup table for word index
    - See PyTorch Seq2Seq Tutorial (class Lang)
       https://pytorch.org/tutorials/intermediate/seq2seq\_translation\_tutorial.html

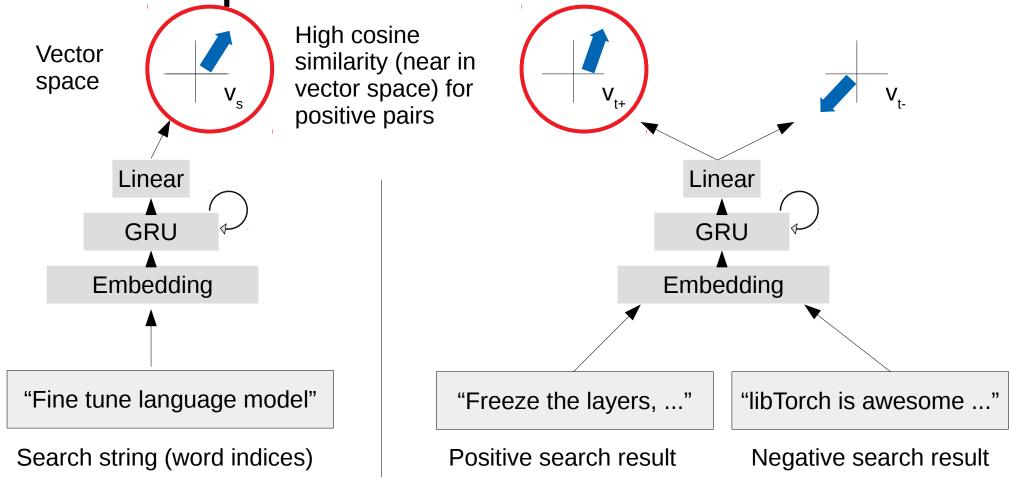
### Overview

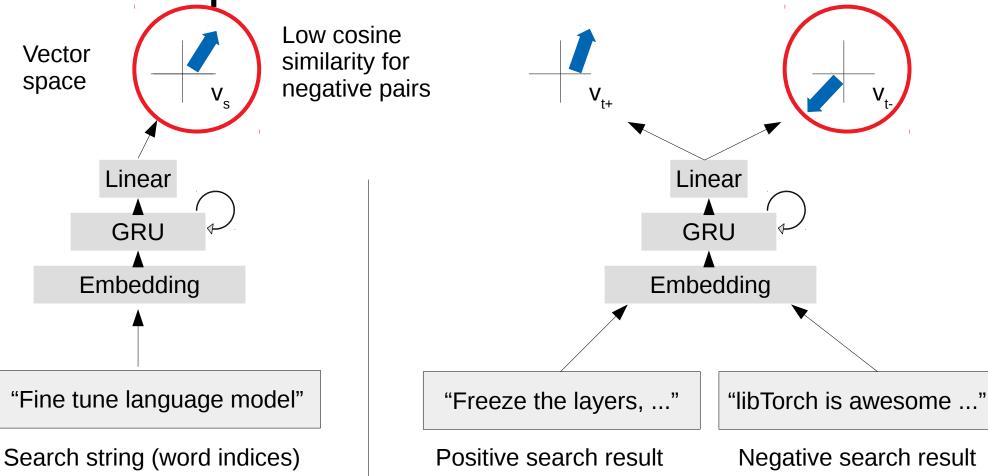
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- Basic idea
  - Represent search string and target (answer/thread) in a shared vector space
  - Use two neural networks for mapping
  - Search strings and targets with same "meaning" should be close in vector space
  - Different meaning → far apart in vector space
  - Use cosine similarity to measure distance









- Use model (encoder) from PyTorch Seq2Seq tutorial
- Add linear layer(s) to learn vector space

```
class EncoderRNN(nn.Module):
   def __init__(self, input_size, hidden_size):
        super(EncoderRNN, self).__init__()
        self.hidden size = hidden size
        self.embedding = nn.Embedding(input size, hidden size)
        self.gru = nn.GRU(hidden size, hidden size)
   def forward(self, input, hidden):
        embedded = self.embedding(input).view(1, 1, -1)
        output = embedded
        output, hidden = self.gru(output, hidden)
        return output, hidden
   def initHidden(self):
        return torch.zeros(1, 1, self.hidden_size, device=device)
```

#### Overview

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#### Loss function

- Start with simple cosine similarity [-1, 1]
  - Should be high for positive pairs (sim+)
  - Low for negative samples (sim-)
  - Shift by 1 to get zero loss instead of negative values loss = (1 sim+) + (1 + sim-)
  - Sum both similarities together
  - nn.CosineSimilarity()

#### Overview

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- Use standard setup (SGD, Ir=1e-3, batch\_size=64, ...)
- Start with small dataset
- ...

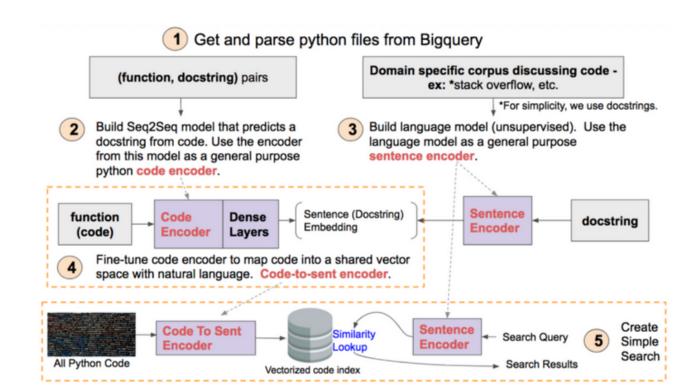
- Use standard setup (SGD, Ir=1e-3, batch\_size=64, ...)
- Start with small dataset
- ...
- Fail: Training+Validation loss hardly moving

Tune model hyperparameters (layer size)

- Tune model hyperparameters (layer size)
- Fail

- Things that have failed:
  - Model hyperparameter tuning
  - Tuning of optimization hyperparams (Ir, weight decay, different optimizer)
  - Adding some regularization (BatchNorm, Dropout)
  - Change GRU (bidirectional, more layers)
  - Use the bigger dataset
  - Use shorter sequences (cut or remove longer sequences)
- Nothing seems to be working!

- What have Husain & Wu done?
- Steps 2 and 3
   create a
   "language model"
   for both networks
- Step 4 learn the shared vector space



- What have Husain & Wu done?
- Two different approaches for pretraining
  - Seq2Seq model (use only encoder)
  - Try to learn to predict next word

- What have Husain & Wu done?
- Two different approaches for pretraining
  - Seq2Seq model (use only encoder)
    - Failed: probably too little data?
  - Try to learn to predict next word
    - Failed: No natural language (mixture of text + code)?

## Overview

- Data
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Back to Step1!

### Review the data

- Data consists of
  - A lot of numbers
    - Tensor/model shapes, random values etc.
  - A lot of single letter words
    - Variable names, etc.

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- Data consists of
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#### Review the data

- Data consists of
  - A lot of numbers
    - Tensor/model shapes, random values etc.
  - A lot of single letter words
    - Variable names, etc.
- Remove these and try training again with small dataset
- (Half) Fail: Model trades sim+ for sim- (at least moving at all!)

- Pretrain models using just search strings
  - Maybe this way the "language" will be learned?
- Then add targets to datasets

- Pretrain models using just search strings
  - Maybe this way the "language" will be learned?
- Then add targets to datasets
- Works OK! First success!
- Validation loss is still high
  - but it's a first step;)

- Pretrain models using just search strings
- Then add targets to datasets
- Change hyperparameters around
  - Add or remove capacity from models
  - Observe the losses
- Switch back to bigger dataset
- Change loss function to log(1 + exp(-1.0\*((sim+)-(sim-))))
  - Taken from Geo et. al, "An Introduction to Deep Learning for Natural Language Processing", Microsoft Research

- Works alright!
- Training and validation losses going down
  - Not as I would have wished, but anyway

### Overview

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- Question: 'tensor is not contiguous'
- Top10 Answers:
- how to compile pytorch from source without cuda default location
- how to merge by avg multiple inputs to layer
- how to specify gpu usage
- why does this assignment operation of variable not work

- apply part of tensor on function to avoid out of memory
- shuffle elements of tensor
- method object does not support item assignment
- tensor slicing on 3 dim tensors
- how to extend tensors inside variable
- tensor and variable are the same now

- Question: 'how to broadcast tensor'
- Top10 Answers:
- how to flip a tensor along a specified axis
- how to get the current value of a variable
- how to broadcast a 1d tensor with a 4d tensor
- how to transfer an existing tensor to another device based on other tensor
- how to get all registerd buffer by self register buffer

- how can I use the pre trained resnet to extract feautres from my own dataset
- how to convert a normal variable into a regular variable that can be inputted to a loss function
- how to merge tensor with weights
- how to choose a suitable weight decay
- how to keey the weight of conv layer unchanged

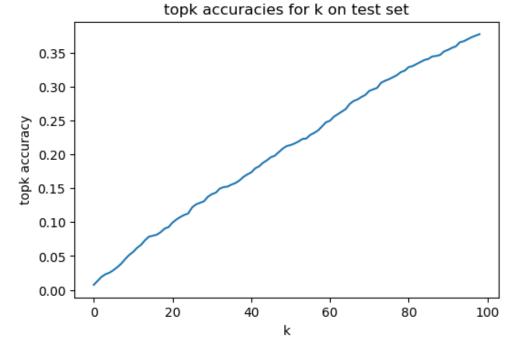
- Question: 'model is performing bad'
- Top10 Answers:
- unusual large memory for con2d with batch size 1
- gpu high memory usage low gpu volatile util
- pytorch example with cnn based object detection
- error loading bidirectional Istm model

- question about thstorage
- too many resources requested for launch
- what is pytorch
- what is nn embedding exactly doing
- will conda install pytorch torchvision c pytorch also install cuda and cudnn
- loading pytorch checkpoint in tf keras

Based on these results let's rather call this talk
 "First steps towards semantic search on PyTorch discussions"

Top10 accuracy:
~7%

Top10 random:
 10/1337=~0.7%





Thanks a lot to **all of you** for being such a great community!

Make sure to create an account at https://discuss.pytorch.org;)

Now let's have some beers and pizza, and hang out together!

Semantic Search on PyTorch discussions

PyTorch in Munich at Microsoft

Munich Applied Deep Learning Meetup Dec 11<sup>th</sup>, 2018 Piotr Bialecki