

Deep Learning III

Programs: Matlab, Swift, Python
Data

MacOS, Clouds: eGPU, GPU array

Software: Metal, OpenGL, OpenCL
Apps

Index of /matconvnet/models

<u>Name</u>	<u>Last modified</u>	<u>Size</u>	<u>Description</u>
<u>Parent Directory</u>		-	
<u>beta10/</u>	2015-04-13 07:52	-	
<u>beta13/</u>	2015-09-19 12:10	-	
<u>beta16/</u>	2015-11-21 11:30	-	
<u>beta18/</u>	2016-02-21 02:32	-	
<u>beta22/</u>	2016-05-02 02:44	-	
<u>fast-rcnn-caffenet-p..></u>	2016-09-30 00:37	203M	
<u>fast-rcnn-caffenet-p..></u>	2016-09-30 09:50	67K	
<u>imagenet-matconvnet-..></u>	2016-09-30 00:33	366M	
<u>imagenet-matconvnet-..></u>	2016-09-30 09:50	56K	
<u>imagenet-matconvnet-..></u>	2016-09-30 00:34	366M	
<u>imagenet-matconvnet-..></u>	2016-09-30 09:50	56K	
<u>imagenet-matconvnet-..></u>	2016-09-30 00:34	492M	
<u>imagenet-matconvnet-..></u>	2016-09-30 09:50	110K	
<u>imagenet-resnet-101-..></u>	2016-09-30 00:33	159M	
<u>imagenet-resnet-101-..></u>	2016-09-30 09:50	1.3M	
<u>imagenet-resnet-152-..></u>	2016-09-30 00:33	215M	
<u>imagenet-resnet-152-..></u>	2016-09-30 09:50	1.9M	
<u>imagenet-resnet-50-d..></u>	2016-09-30 00:33	92M	
<u>imagenet-resnet-50-d..></u>	2016-09-30 09:50	664K	
<u>imagenet-vgg-f.mat</u>	2016-09-30 00:35	217M	
<u>imagenet-vgg-f.svg</u>	2016-09-30 09:50	60K	
<u>imagenet-vgg-m-1024.mat</u>	2016-09-30 00:35	310M	
<u>imagenet-vgg-m-1024.svg</u>	2016-09-30 09:50	60K	

The VLFeat [open source](#) library implements popular computer vision algorithms specializing in image understanding and local features extraction and matching. Algorithms include Fisher Vector, VLAD, SIFT, MSER, k-means, hierarchical k-means, agglomerative information bottleneck, SLIC superpixels, quick shift superpixels, large scale SVM training, and many others. It is written in C for efficiency and compatibility, with interfaces in MATLAB for ease of use, and detailed documentation throughout. It supports Windows, Mac OS X, and Linux. The latest version of VLFeat is **0.9.21**.

ACM OpenSource
Award

Download

- [VLFeat 0.9.21](#) (Windows, Mac, Linux)
- [Source code and installation](#)
- [git repository](#), [bug tracking](#).

Documentation

- [MATLAB commands](#)
- [C API](#) with algorithm descriptions
- [Command line tools](#)

Tutorials

- Features: [Covariant detectors](#), [HOG](#), [SIFT](#), [MSER](#), [Quick shift](#), [SLIC](#)
- Clustering: [IKM](#), [HIKM](#), [AIB](#)
- Matching: [Randomized kd-trees](#)
- [All tutorials](#)

Example applications

- [Caltech-101 classification](#)
- [SIFT matching for auto-stitching](#)
- [All example applications](#)

Citing

```
@misc{vedaldi08vlfeat,  
  Author = {A. Vedaldi and B. Fulkerson},  
  Title = {{VLFeat}: An Open and Portable Library  
          of Computer Vision Algorithms},  
  Year = {2008},  
  Howpublished = {\url{http://www.vlfeat.org/}}  
}
```

Acknowledgments

  [UCLA Vision Lab](#) [Oxford VGG](#).

News

8/1/2018 VLFeat 0.9.21 released

Maintenance release. Fixed **v1_argparse** to be compatible with MatConvNet. Fixed the binaries for recent versions of macOS.

14/1/2015 VLFeat 0.9.20 released

Maintenance release. Bugfixes.

12/9/2014 **MatConvNet**

ImageNet is an image database organized according to the **WordNet** hierarchy (currently only the nouns), in which each node of the hierarchy is depicted by hundreds and thousands of images. Currently we have an average of over five hundred images per node. We hope ImageNet will become a useful resource for researchers, educators, students and all of you who share our passion for pictures.

[Click here](#) to learn more about ImageNet, [Click here](#) to join the ImageNet mailing list.



What do these images have in common? *Find out!*

[Research updates on improving ImageNet data](#)



Simple Tutorial - Image Recognition -CNN

Python notebook using data from [Images Dataset](#) · 3,078 views · 4mo ago
deep learning, tutorial, cnn

40

Copy and Edit 55



Seven categories for testing

Version 73

73 commits

Notebook

Tutorial - Image Recognition

Data

Comments

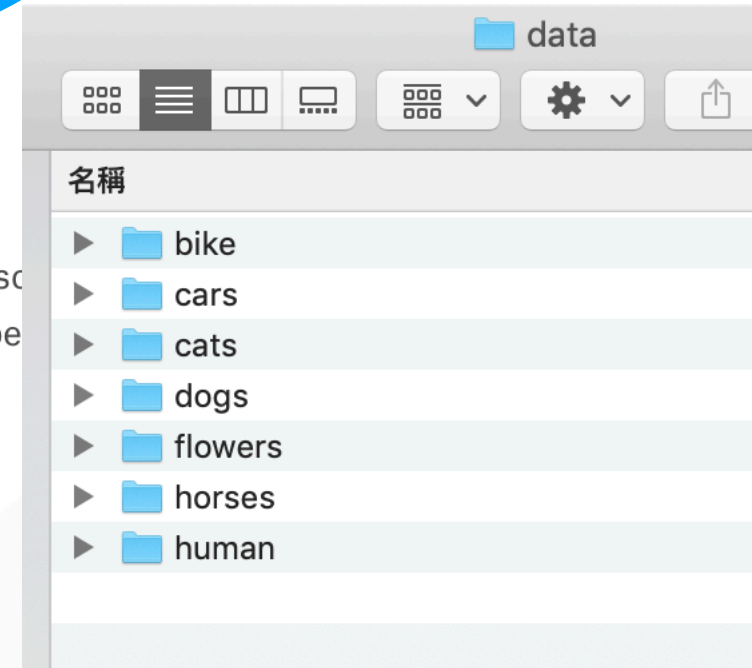
Tutorial - Image Recognition

This kernel covers the Image Recognition using CNN, Keras ,TensorFlow implementation using image dataset . The following topics will be

- Loading and preprocessing own dataset
- Designing and training a CNN model in Keras
- Plotting the Loss and Accuracy curve
- Evaluating the model & Predicting the output class of a test image
- Visualizing the intermediate layer output of CNN
- Plotting the confusion matrix for your result

Loading and preprocessing own dataset

The dataset that I am using for this kernel is my own accumulated dataset of 7 types of classes namely 'flowers', 'cars', 'cats', 'horses', 'human', 'b' image samples.



Anna from Kaggle

eGPU

- External GPU
- MacOS 10.13.4

The screenshot shows the NVIDIA website's navigation menu and footer. The navigation menu is a dark bar with white text and dropdown arrows for: 平台 (Platform), 開發者 (Developers), 社群 (Community), 驅動程式 (Drivers), 支援 (Support), and 關於 NVIDIA (About NVIDIA). Below the navigation menu is a white section with a green link "閱讀更多 >". The footer is a dark bar with a white grid of links categorized into four columns: 平台 (Platform), 產品 (Products), 開發者 (Developers), and 公司訊息 (Company Information).

平台	產品	開發者	公司訊息
CUDA-X	DGX 系統	NVIDIA 開發者	NVIDIA 合作夥伴網絡
智慧機器	DRIVE PX	開發者新聞	加入我們
資料中心	GeForce RTX 20 系列	開發者部落格	聯繫我們
深度學習與人工智慧	NVIDIA 虛擬 GPU	開發者論壇	產品安全
設計視覺化	Jetson	開源平台	Communities
醫療保健與生命科學	Quadro	訓練課程	NVIDIA 部落格
自動駕駛汽車	Tesla	GPU 技術大會	訂閱電子報
遊戲與娛樂	T4 企業伺服器	CUDA	隱私中心
NGC			

Data Parallel Problems

- Plenty of problems fall into this category (luckily 😊)
 - Graphics, image & video processing, physics, scientific computing, ...
- This type of parallelism is called *data parallelism*
- And GPUs are the perfect solution for them!
 - In fact the more the data, the more efficient GPUs become at these algorithms
 - Bonus: You can relatively easily add more processing cores to a GPU and increase the throughput

Parallelism in CPUs v. GPUs

- CPUs use *task parallelism*
 - Multiple tasks map to multiple threads
 - Tasks run different instructions
 - 10s of relatively heavyweight threads run on 10s of cores
 - Each thread managed and scheduled explicitly
 - Each thread has to be individually programmed
- GPUs use *data parallelism*
 - SIMD model (Single Instruction Multiple Data)
 - Same instruction on different data
 - 10,000s of lightweight threads on 100s of cores
 - Threads are managed and scheduled by hardware
 - Programming done for batches of threads (e.g. one pixel shader per group of pixels, or draw call)

Stream Processing

- **What we just described:**
 - Given a (typically large) set of data (“**stream**”)
 - Run the same series of operations (“**kernel**” or “**shader**”) on all of the data (SIMD)
- **GPUs use various optimizations to improve throughput:**
 - Some on-chip memory and local caches to reduce bandwidth to external memory
 - Batch groups of threads to minimize incoherent memory access
 - Bad access patterns will lead to higher latency and/or thread stalls.
 - Eliminate unnecessary operations by exiting or killing threads
 - Example: Z-Culling and Early-Z to kill pixels which will not be displayed



Blackmagic eGPU for MacBook Pro available exclusively from Apple

將外接式繪圖處理器搭配 Mac 使用

配備 Thunderbolt 3 並執行 macOS High Sierra 10.13.4 或以上版本的 Mac 可以連接至外接式繪圖處理器 (亦稱為 eGPU)，取得額外的繪圖效能。

eGPU 可提供 Mac 額外的繪圖效能，以執行專業 app、3D 遊戲、VR 內容創作等各種用途。

任何執行 macOS High Sierra 10.13.4 或以上版本並配備 Thunderbolt 3 的 Mac₁，皆支援 eGPU。瞭解[如何更新 Mac 上的軟體](#)。

您可以透過 eGPU 在 Mac 上執行以下操作：

- 加速用到 Metal、OpenGL 和 OpenCL 的 app
- 連接更多外接監視器和顯示器
- 使用插入 eGPU 的虛擬實境頭戴顯示器
- 在使用 eGPU 的同時為 MacBook Pro 充電
- 在闔上 MacBook Pro 的內建顯示器時，使用 eGPU 搭配 MacBook Pro
- 在使用者登入時連接 eGPU
- 使用 Mac 上的多個 Thunderbolt 3 (USB-C) 連接埠連接超過一個以上的 eGPU₂
- 使用選單列項目  安全地中斷連接 eGPU





[教學][TB3][eGPU] macOS 全版本 eGPU 外接式顯卡 NVIDIA / AMD 顯卡安 裝教學

POSTED ON 2019-01-13

#Thunderbolt3 擴充計畫



eGPU

2018 年 3 月，Apple 釋出 [macOS 10.13.4](#) 更新，正式支援 eGPU，正式進入 Thunderbolt 3 大 eGPU 時代。只要你多課金買了 Thunderbolt 3 的 eGPU Box，加上不錯的顯卡，你就可以大大改善 macOS 裝置顯卡都很低規的問題

來源：<https://support.apple.com/zh-tw/HT208544>

“ macOS High Sierra 10.13.4 和以上版本中的 eGPU 支援，是要藉由強大的 eGPU 效能來提升 Metal、OpenGL 和 OpenCL app 的速度。並非所有 app 都支援 eGPU 加速
– 新增 [外接繪圖處理器 \(eGPU\)](#) 支援。

任何執行 macOS High Sierra 10.13.4 或以上版本並 [配備 Thunderbolt 3 的](#)

Metal by Example

High-performance graphics and data-parallel programming for iOS and macOS

About the Author

The Book

Swift

Using Basis Universal Texture Compression with Metal

July 11, 2019 by Warren Moore

In this short article, we'll take a look at a relatively new compressed texture format called Basis. Basis is developed by [Binomial, LLC](#), a company founded by Rich Geldreich (of [crunch](#) fame) and Stephanie Hurlburt.

Basis is unique among compression formats in that it emphasizes efficient transcoding between compressed formats. This means that a single .basis file can be transformed, at runtime, into a format that's optimal for the target platform, without decompressing it in memory, saving space and bandwidth.

Training a Neural Network with Metal Performance Shaders

Use an MPS neural network graph to train a simple neural network digit classifier.

Download

SDKs

macOS 10.15+

Xcode 11.0+

Framework

Metal Performance
Shaders

On This Page

[Overview](#) ⌵

[See Also](#) ⌵

Overview

The sample code describes how to write a neural network using [MPSNNGraph](#) and how to train the network to recognize a digit in an image. The sample trains a network for 300 iterations on a batch size of 40 images. You'll see how to set up training of weights and biases using data sources, including how to initialize and update weights. You'll also see how to validate the network using a test dataset.

Note

This sample code project is associated with WWDC 2019 session [614: Metal for Machine Learning](#).

Forge: neural network toolkit for Metal

24 APRIL 2017 ⌚ 10 minutes

Today I am happy to announce [Forge](#), an open source library that makes it a bit easier to build neural networks with MPSCNN.



Convolutional neural networks on the iPhone with VGGNet

30 AUGUST 2016 © 34 minutes

I implemented the VGGNet architecture for image recognition on the iPhone, using the new convolutional neural network API from the Metal Performance Shaders framework.

In this post I explain how CNNs work and specifically how to get VGGNet running on your iPhone using Metal.

The [demo app](#) sends the video feed from the iPhone's camera through the neural network to get the top-5 classification scores for whatever you're looking at:

```
cd matconvnet-1.0-beta25
run matlab/vl_compilenn ;
```

Step 1

Quick Start

```
urlwrite(...
'http://www.vlfeat.org/matconvnet/models/imagenet-vgg-f.mat', ...
'imagenet-vgg-f.mat') ;

% Setup MatConvNet.
run matlab/vl_setupnn ;

% Load a model and upgrade it to MatConvNet current version.
net = load('imagenet-vgg-f.mat') ;
net = vl_simplenn_tidy(net) ;

% Obtain and preprocess an image.
im = imread('peppers.png') ;
im_ = single(im) ; % note: 255 range
im_ = imresize(im_, net.meta.normalization.imageSize(1:2)) ;
im_ = im_ - net.meta.normalization.averageImage ;

% Run the CNN.
res = vl_simplenn(net, im_) ;

% Show the classification result.
scores = squeeze(gather(res(end).x)) ;
[bestScore, best] = max(scores) ;
figure(1) ; clf ; imagesc(im) ;
title(sprintf('%s (%d), score %.3f', ...
net.meta.classes.description{best}, best, bestScore)) ;
```

Step 2

Save as
demo_imagenet.m

```
% Setup MatConvNet.  
run matlab/vl_setupnn ;  
  
% Load a model and upgrade it to MatConvNet current version.  
net = load('imagenet-vgg-f.mat') ;  
net = vl_simplenn_tidy(net) ;  
  
% Obtain and preprocess an image.  
im = imread('peppers.png') ;  
im_ = single(im) ; % note: 255 range  
im_ = imresize(im_, net.meta.normalization.imageSize(1:2)) ;  
im_ = im_ - net.meta.normalization.averageImage ;  
  
% Run the CNN.  
res = vl_simplenn(net, im_) ;  
  
% Show the classification result.  
scores = squeeze(gather(res(end).x)) ;  
[bestScore, best] = max(scores) ;  
figure(1) ; clf ; imagesc(im) ;  
title(sprintf('%s (%d), score %.3f', ...  
    net.meta.classes.description{best}, best, bestScore)) ;
```

demo_imagenet.m

The screenshot shows the MATLAB IDE interface. The left pane displays the file explorer for the current folder, listing various subfolders and files, including `demo_imagenet.m` and `imagenet-vgg-f.mat`. The main editor window shows the code for `demo_imagenet.m`, which includes comments and code for running a CNN and displaying classification results. The Command Window shows the output of the code, including a 1x1000 cell array of classification results.

```
12 % Run the CNN.
13 res = vl_simplenn(net, im_) ;
14
15 % Show the classification result.
16 scores = squeeze(gather(res(end).x)) ;
17 [bestScore, best] = max(scores) ;
18 figure(1) ; clf ; imagesc(im) ;
19 title(sprintf('%s (%d), score %.3f', ...
20             net.meta.classes.description{best}, best, bestScore))
```

Command Window

```
ans =
1x1000 cell array
Columns 1 through 2
{'tench, Tinca ti...'} {'goldfish, Caras...'}
Columns 3 through 4
```

Imagenet-vgg-f.mat

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FILE VARIABLE CODE SIMULINK ENVIRONMENT RESOURCES

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- matconvnet.vcxproj
- matconvnet.vcxproj.filters
- README.md

Details

Select a file to view details

```

1 func
2 %CNN
3 %

Command Window

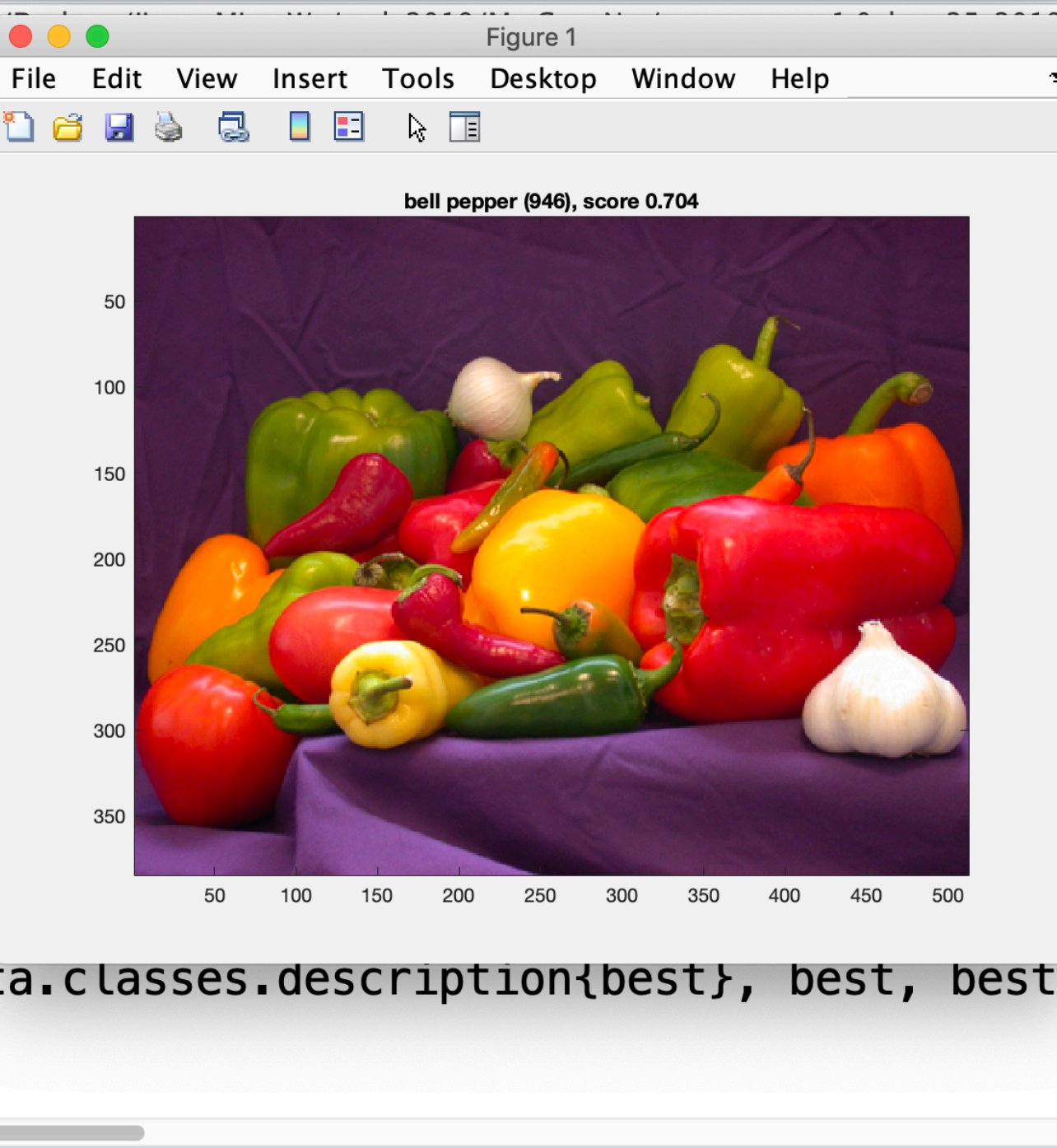
im_ = imre
im_ = im_

% Run the
= vl_s

show the
scores = s
[bestScore
figure(1)
title(spr
net.meta.classes.description{best}, best, bestScore)) ;

fx >>

```



```

getBatch
for trai
menting s

ize(1:2)) ;

```

EX-I

net.meta.classes.description{best}, best, bestScore)) ;

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PLOTS

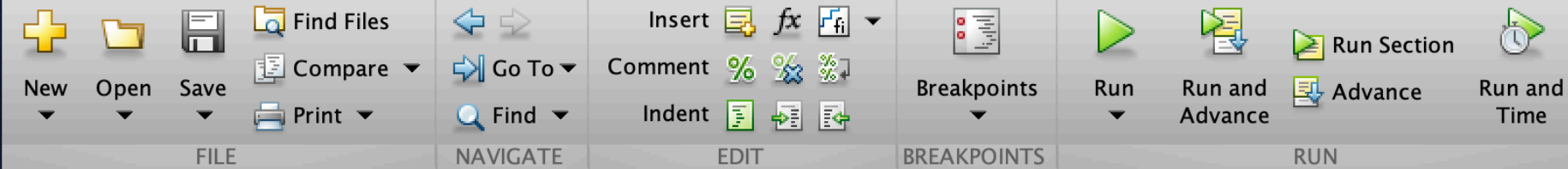
APPS

EDITOR

PUBLISH

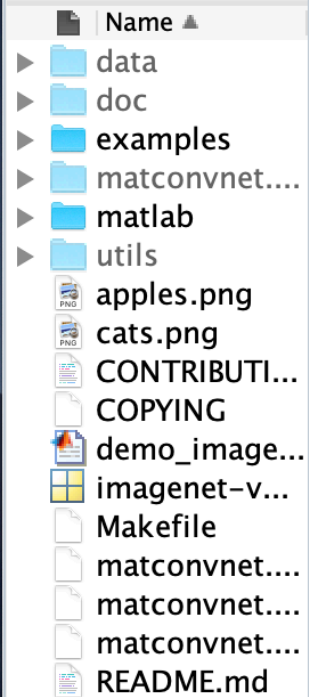
VIEW

Search Documentation



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Current Fol...

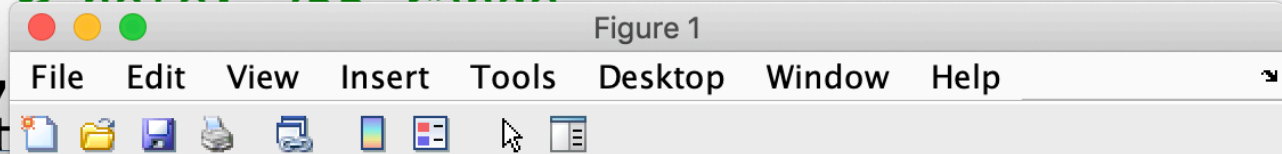


Editor - /Users/apple/Desktop/Jiann-Ming Wu/code2019/MatConvNet/matconvnet-1.0-beta25/demo_imagenet.m

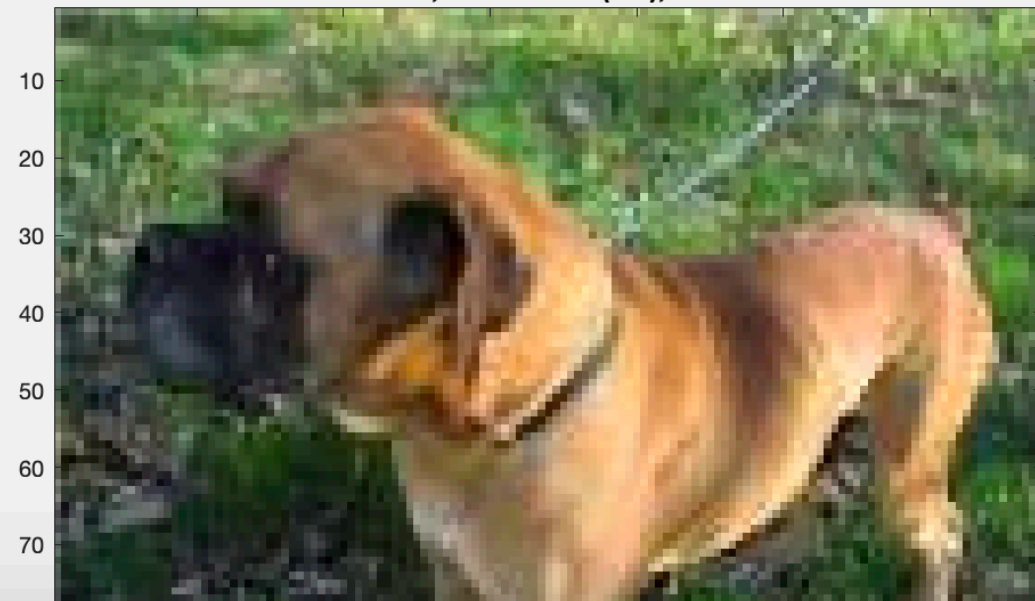
```
+16 interpolating_3Dface_elstic_nen.m x cnn_imagenet_camdemo.m x ToolboxDemo.m x demo_imagenet.m x vl_simplenn.m x +
2 % Load a model and upgrade it to MatConvNet current version.
3 net = load('imagenet-vgg-f.mat') ;
4 net = vl_simplenn_tidy(net) ;
5
6 % Obtain and preprocess an image.
7 % im = imread('peppers.png') ;
8 im = imread('../..../image_recognition_CNN/data/dogs/dog.11.jpg');
9 im_ = single(im) ; % note: 255 range
10 im_ = imresize(im_, [100 100]);
11 im = im - net.mean;
```

Command Window

```
>> demo_imagenet
fx >>
```



bloodhound, sleuthound (164), score 0.658



Details

Select a file to view

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Insert Comment Indent Breakpoints Run Run and Advance Run Section Advance Run Time

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- demo_image...
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- matconvnet...
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Details

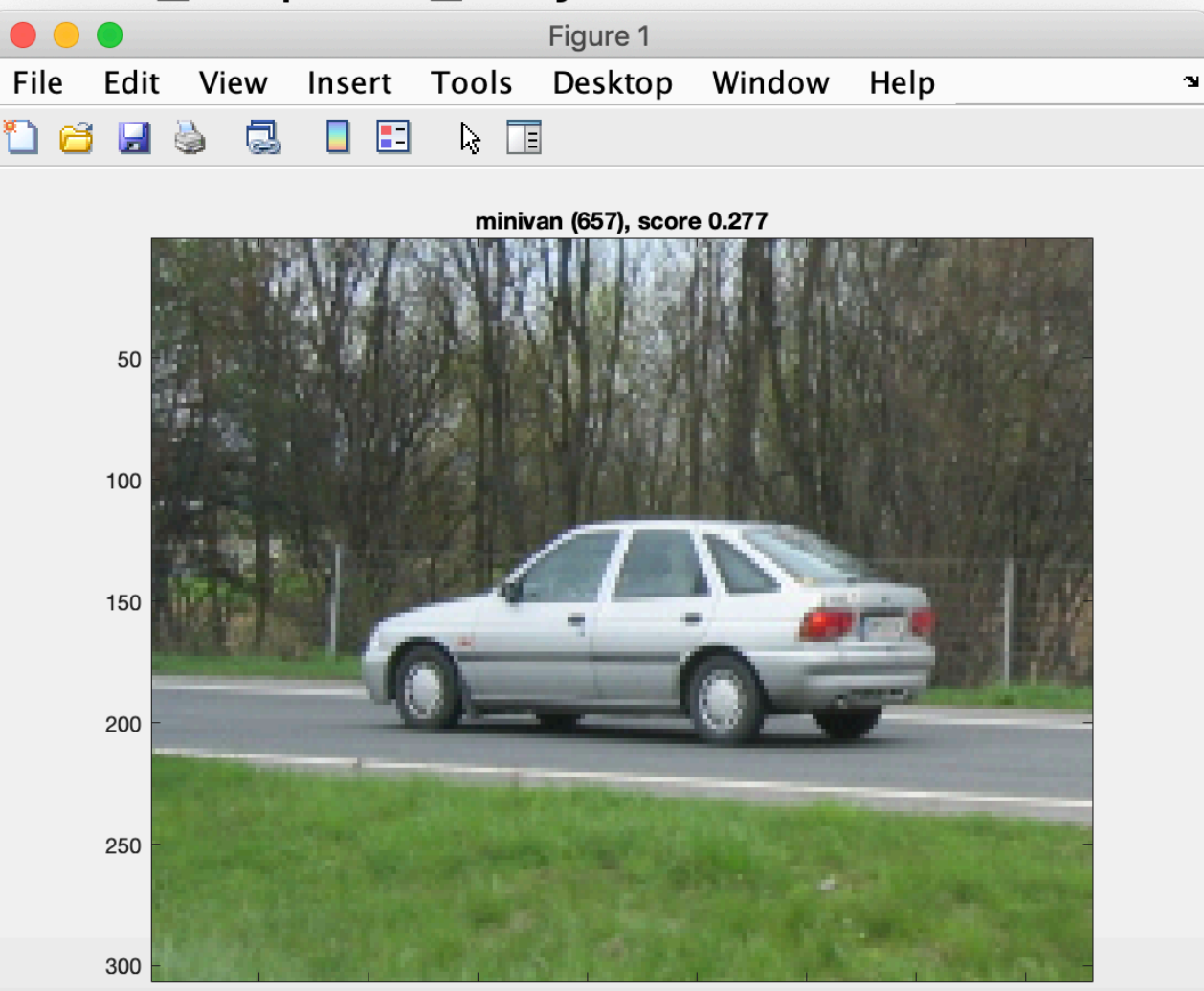
Select a file to view

```
1 run matlab/vl_setupnn ;
2 % Load a model and upgrade it to MatConvNet current version.
3 net = load('imagenet-vgg-f.mat') ;
4 net = vl_simplenn_tidy(net) ;
5
6 %
7 %
8 im
9 im_
10 im_
```

```
data/cars/mycar.png');
imageSize(1:2)) ;
```

Command Window

```
>> demo
fx >>
```



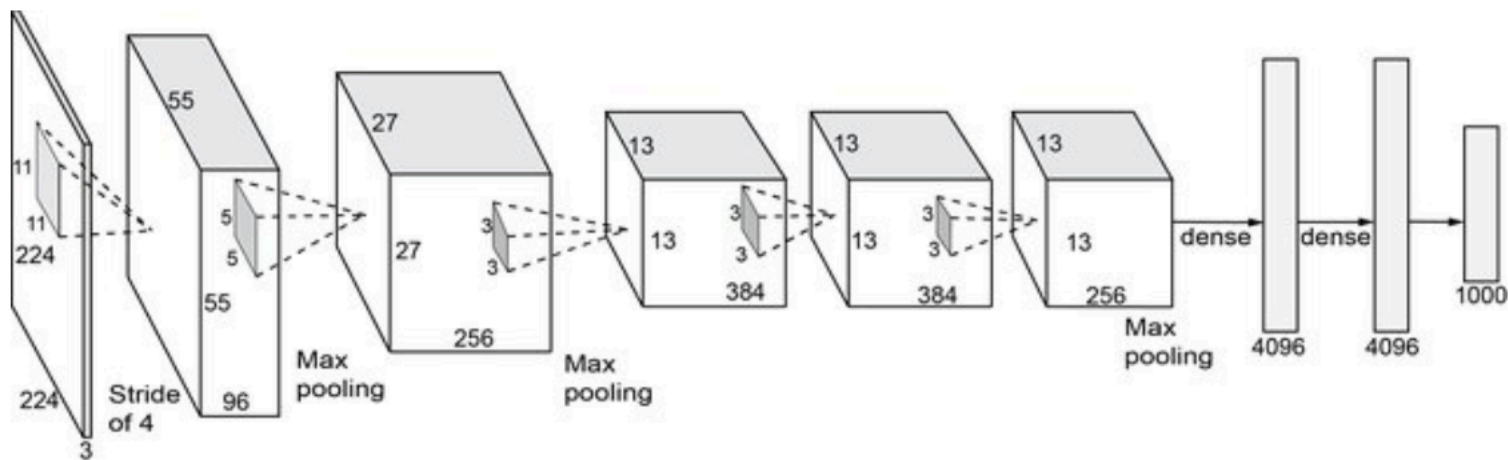
VERY DEEP CONVOLUTIONAL NETWORKS FOR LARGE-SCALE IMAGE RECOGNITION

Karen Simonyan* & Andrew Zisserman⁺

Visual Geometry Group, Department of Engineering Science, University of Oxford
{karen, az}@robots.ox.ac.uk

ABSTRACT

In this work we investigate the effect of the convolutional network depth on its accuracy in the large-scale image recognition setting. Our main contribution is a thorough evaluation of networks of increasing depth using an architecture with very small (3×3) convolution filters, which shows that a significant improvement on the prior-art configurations can be achieved by pushing the depth to 16–19 weight layers. These findings were the basis of our ImageNet Challenge 2014 submission, where our team secured the first and the second places in the localisation and classification tracks respectively. We also show that our representations generalise well to other datasets, where they achieve state-of-the-art results. We have made our two best-performing ConvNet models publicly available to facilitate further research on the use of deep visual representations in computer vision.



Figure

Caption

Fig. 1: Architecture of VGG-F.

```
>> net.layers{1}
ans =
  struct with fields:
    name: 'conv1'
    type: 'conv'
    weights: {[11x11x3x64 single] [64x1 single]}
    size: [11 11 3 64]
    pad: [0 0 0 0]
    stride: [4 4]
    precious: 0
    dilate: 1
    opts: {}
```

This figure was uploaded by [Bogdan Kwolek](#)

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demo_imagenet

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Details

Select a file to view details

```

+18 ToolboxDemo.m x demo_imagenet.m x vl_simplenn.m x cnn_cifar.m x +
9 - im_ = single(im) ; % note: 255 range
10 - im_ = imresize(im_, net.meta.normalization
11 - im_ = im_ - net.meta.normalization.average
12
13 % Run the CNN.
14 ● → res = vl_simplenn(net, im_) ;
15
16 % Show the classification result.
17 - scores = squeeze(gather(res(end).x)) ;
18 - [bestScore, best] = max(scores) ;
19 - figure(1) ; clf ; imagesc(im) ;
20 - title(sprintf('%s (%d), score %.3f', ...
21 - net.meta.classes.description{best}, bes

```

Command Window

```

>> demo_imagenet
14 res = vl_simplenn(net, im_) ;
fx K>>

```

Workspace - dem...

Name ▲	Value
im	306x4
im_	224x2
net	1x1 st

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Function Call Stack: vl_simplenn

Quit Debugging

FILE NAVIGATE EDIT BREAKPOINTS DEBUG

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- matconvnet.vcxproj.filters
- README.md

Details

Select a file to view details

```

289 %
290 -----
291 -----
292 -----
293 -----
294 for i=1:n
295     if opts.skipForward, break; end;
296     l = net.layers{i};
297     res(i).time = tic;
298     switch l.type
299     case 'conv'
300         res(i+1).x = vl_nnconv(res(i).x, l.
301             'pad', l.pad, ...

```

break

opts.skipForward: logical = 0

Workspace - vl_si...

Name	Value
backPropLim	1
bnormCudnn	1x1 c...
cudnn	1x1 c...
doder	0
gpuMode	0
n	21
net	1x1 s...
opts	1x1 s...
res	1x22...
testMode	0
varargin	0x0 c...
x	224x2...

Command Window

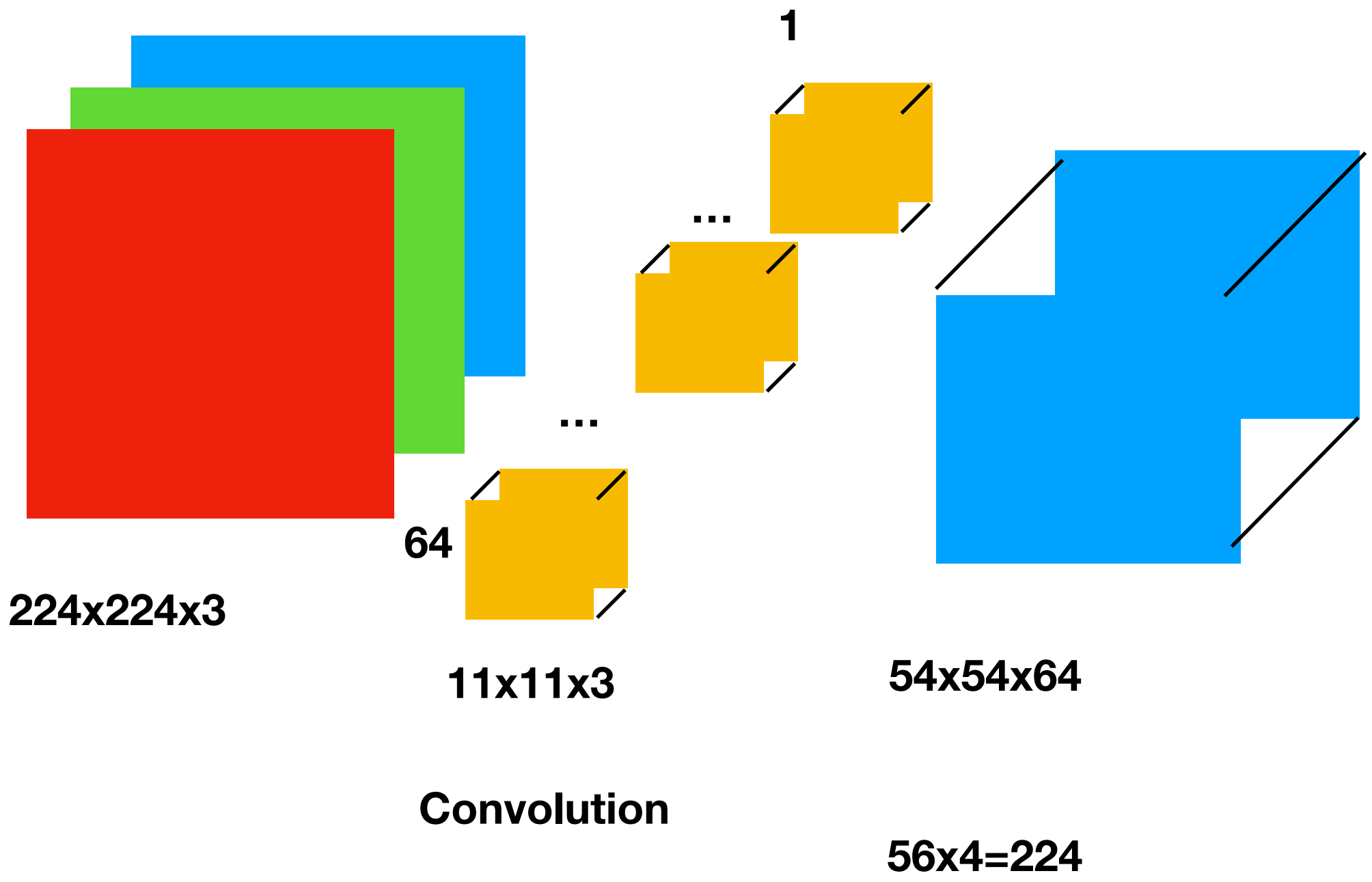
```

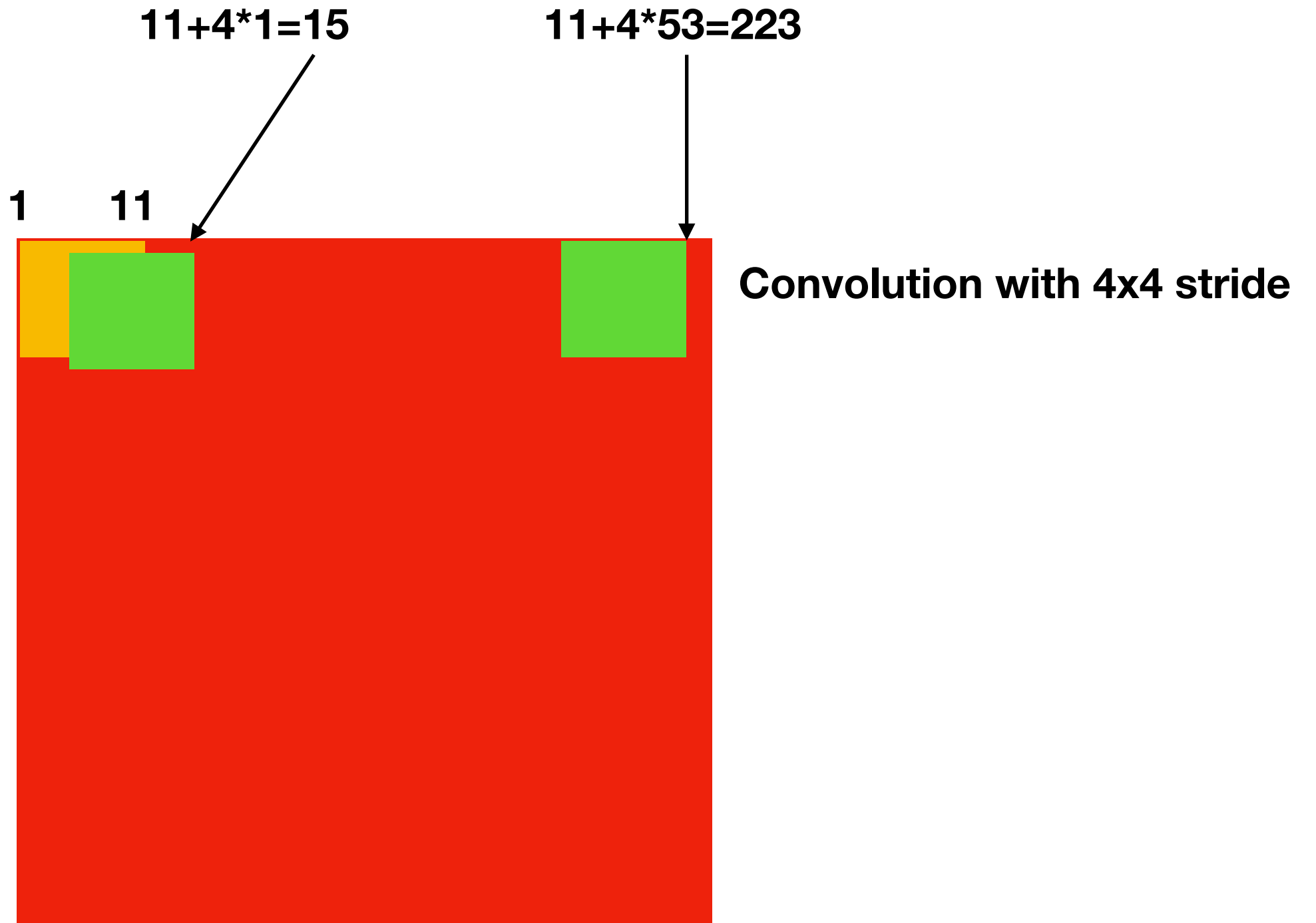
>> demo_imagenet
14 res = vl_simplenn(net, im_);
fx K>>

```

```
>> net.layers{1}
ans =
  struct with fields:
    name: 'conv1'
    type: 'conv'
    weights: {[11x11x3x64 single] [64x1 single]}
    size: [11 11 3 64]
    pad: [0 0 0 0]
    stride: [4 4]
    precious: 0
    dilate: 1
    opts: {}
```

stride





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Insert fx fi Comment % % Indent Breakpoints Continue Step Step In Step Out Run to Cursor Function Call Stack: vl_simplenn Quit Debugging

FILE NAVIGATE EDIT BREAKPOINTS DEBUG

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- Name
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- utils
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- cats.png
- CONTRIBUTING.md
- COPYING
- demo_imagenet.m
- imagenet-vgg-f.mat
- Makefile
- matconvnet.sln
- matconvnet.vcxproj
- matconvnet.vcxproj.filters
- README.md

Details

Select a file to view details

```

378 end
379
380 % optionally forget intermediate result
381 needsBProp = dodger && i >= backPropLim;
382 forget = opts.conserveMemory && ~needsB
383 if i > 1

```

Command Window

```

14 res = vl_simplenn(net, im) :
K>> size(res(i+1).x)

ans =

    54    54    64

K>> size(res(i).x)

ans =

    224    224     3

```

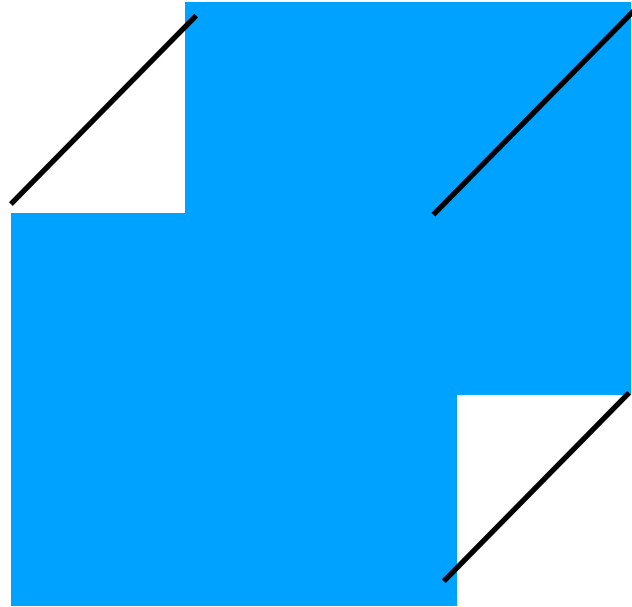
Output size of 1st hidden layer

Input size of 1st hidden layer

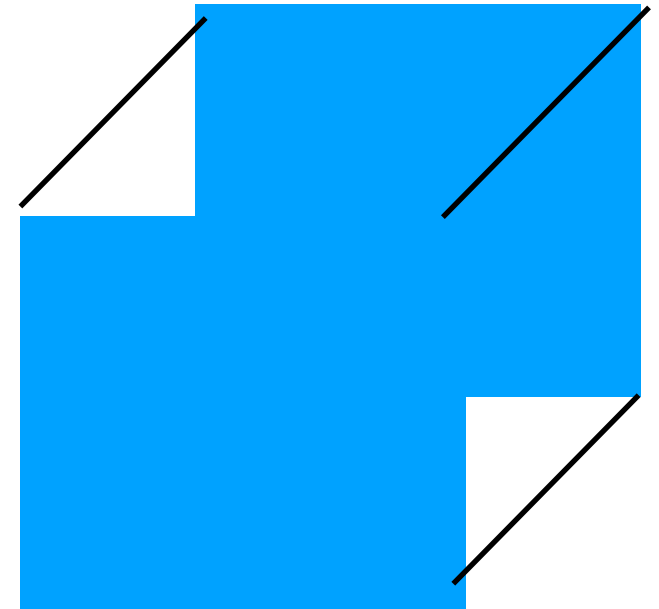
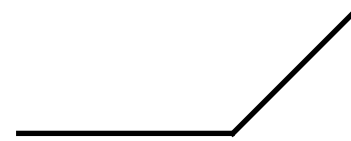
Workspace - vl_si...

Name	Value
ans	[224,2
backPropLim	1
bnormCudnn	1x1 ce
cudnn	1x1 ce
dodger	0
gpuMode	0
i	1
l	1x1 st
n	21
net	1x1 st
opts	1x1 st
res	1x22 s
testMode	0
varargin	0x0 ce
x	224x2

ReLU

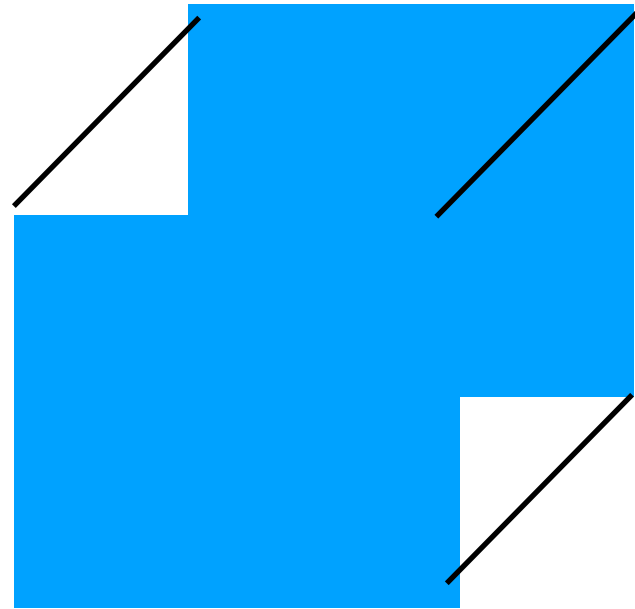


54x54x64



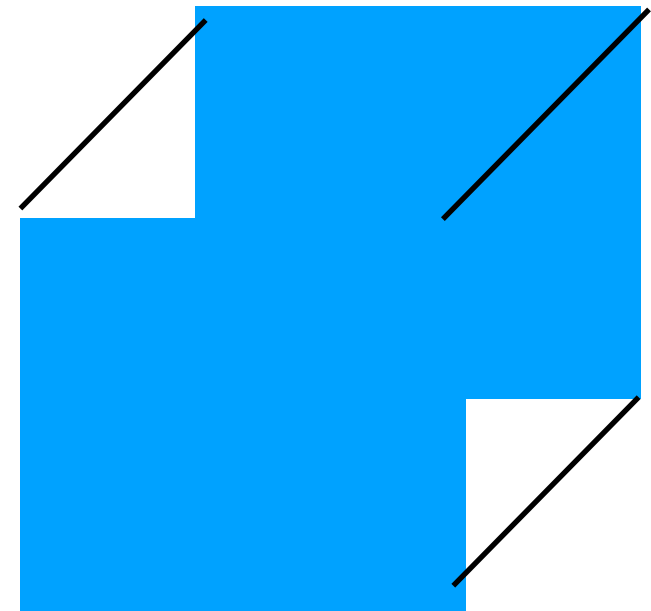
54x54x64

Norm



54x54x64

Normalization

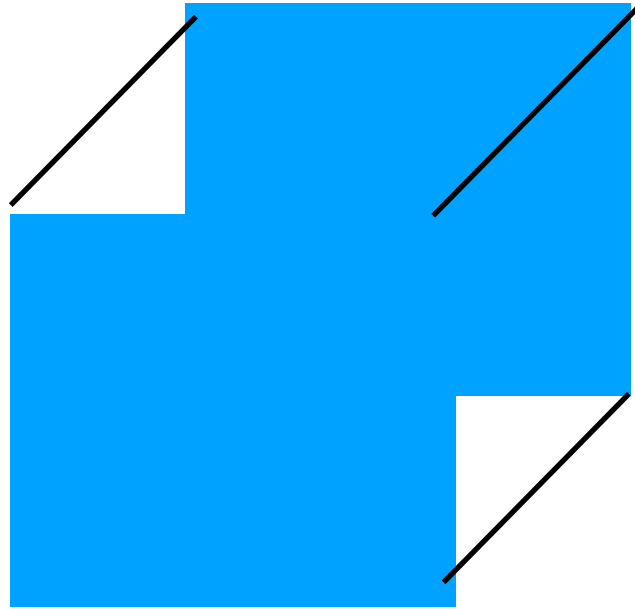


54x54x64

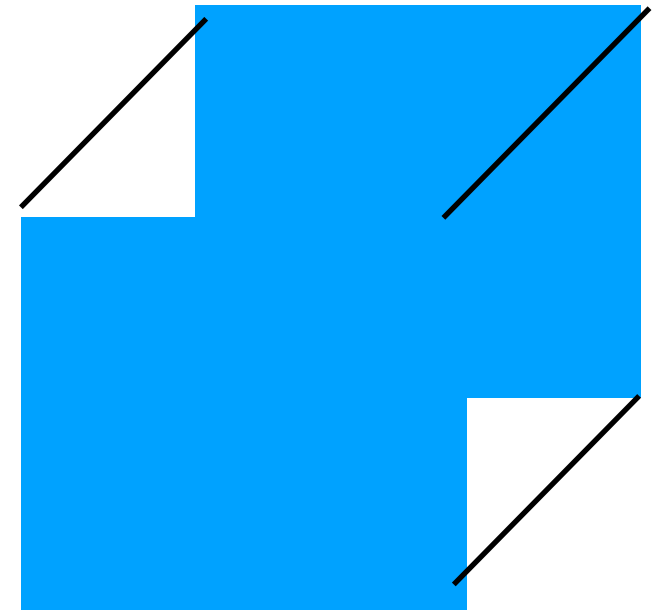
struct with fields:

```
name: 'pool1'  
type: 'pool'  
method: 'max'  
pool: [3 3]  
stride: [2 2]  
pad: [0 1 0 1]  
weights: {}  
precious: 0  
opts: {}
```

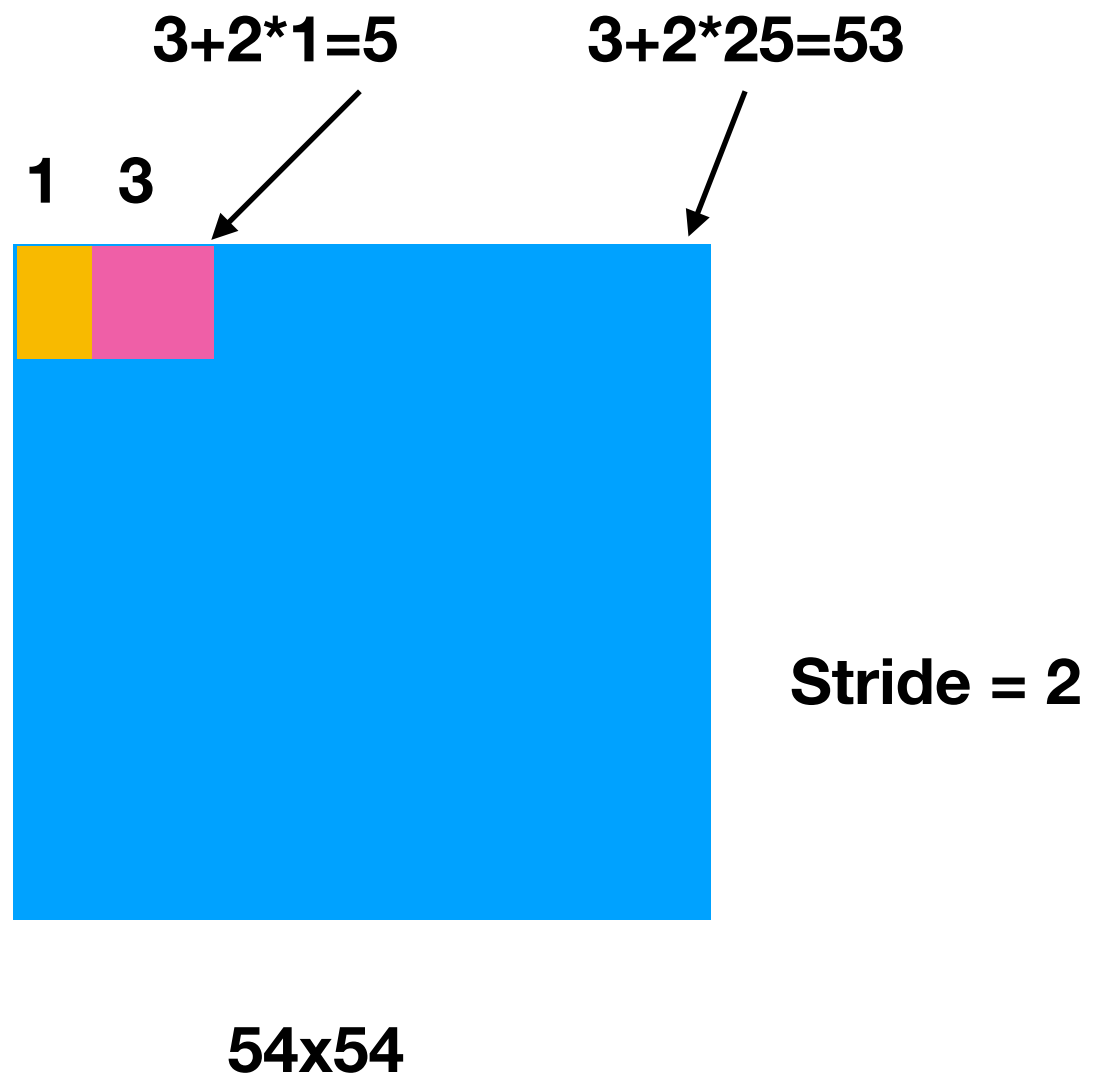
Max pooling



54x54x64



27x27x64



MATLAB R2019a - academic use

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FILE NAVIGATE EDIT BREAKPOINTS DEBUG

Function Call Stack: vl_simplenn

Quit Debugging

Users > apple > Desktop > Jiann-Ming Wu > code2019 > MatConvNet > matconvnet-1.0-beta25

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- imagenet-vgg-f.mat
- Makefile
- matconvnet.sln
- matconvnet.vcxproj
- matconvnet.vcxproj.filt...
- README.md
- test1.png

Editor - /Users/apple/Desktop/Jiann-Ming Wu/code2019/MatConvNet/matconvnet-1.0-beta25/matlab/simplenn/vl_simplenn.m

```
379  
380 % optionally forget intermediate results  
381 needsBProp = doder && i >= backPropLim;  
382 forget = opts.conserveMemory && ~needsBProp ;  
383 if i > 1
```

Command Window

```
K>> size(res(i+1).x)  
  
ans =  
  
    27    27    64  
  
K>> size(res(i).x)  
  
ans =  
  
    54    54    64  
  
fx K>>
```

3 usages of "needsBProp" found

MatConvNet

[< Back to Alex Krizhevsky's home page](#)

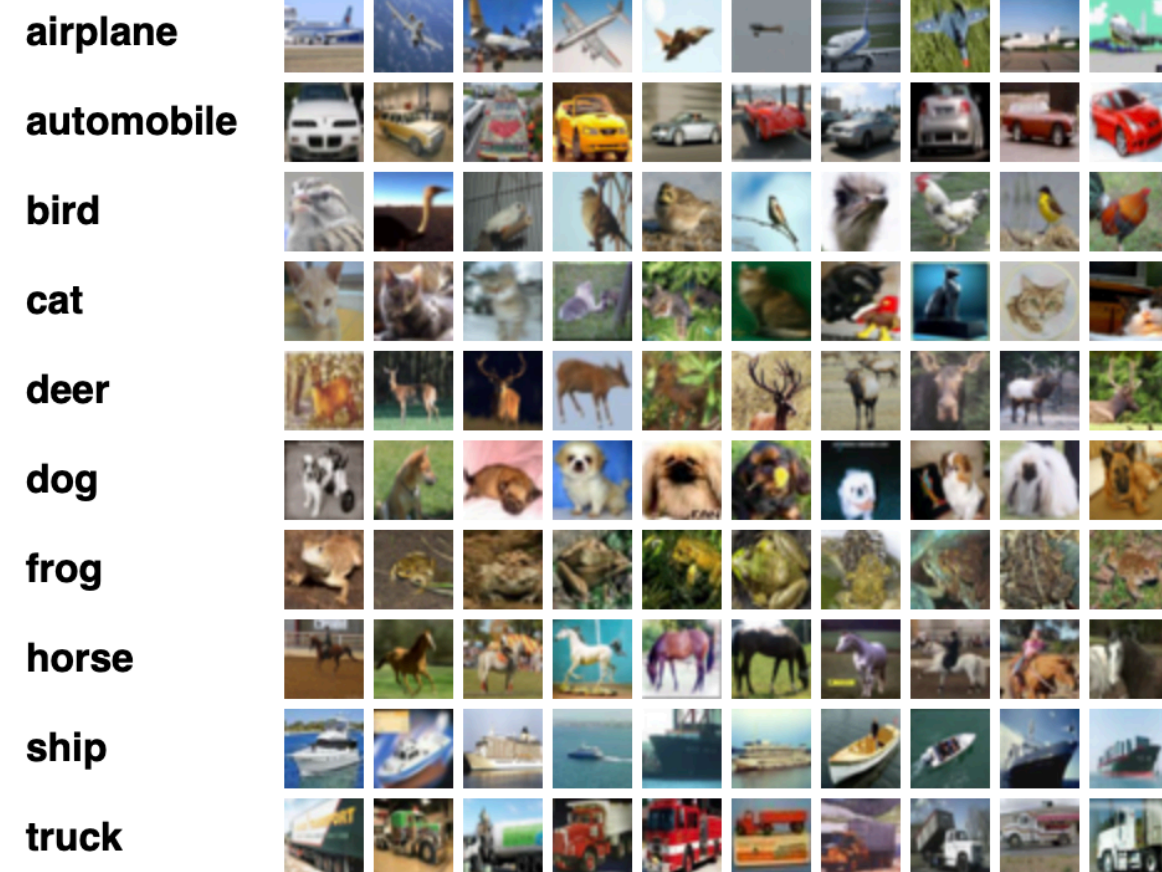
The CIFAR-10 and CIFAR-100 are labeled subsets of the [80 million tiny images](#) dataset. They were collected by Alex Krizhevsky, Vinod Nair, and Geoffrey Hinton.

The CIFAR-10 dataset

The CIFAR-10 dataset consists of 60000 32x32 colour images in 10 classes, with 6000 images per class. There are 50000 training images and 10000 test images.

The dataset is divided into five training batches and one test batch, each with 10000 images. The test batch contains exactly 1000 randomly-selected images from each class. The training batches contain the remaining images in random order, but some training batches may contain more images from one class than another. Between them, the training batches contain exactly 5000 images from each class.

Here are the classes in the dataset, as well as 10 random images from each:



The classes are completely mutually exclusive. There is no overlap between automobiles and trucks. "Automobile" includes sedans, SUVs, things of that sort. "Truck" includes only big trucks. Neither includes pickup trucks.

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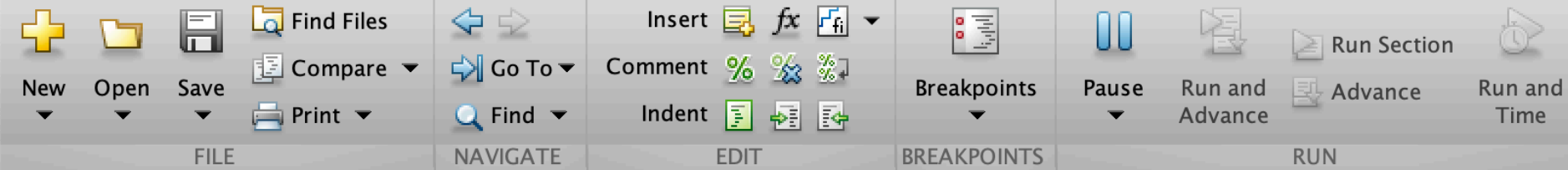
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Current Folder

Name ▲

- cnn_cifar.m
- cnn_cifar_init.m
- cnn_cifar_init_nin.m

Editor - /Users/apple/Desktop/Jiann-Ming Wu/code2019/MatConvNet/matconvnet-1.0-beta25_2019_10/exampl...

```

1  function [net, info] = cnn_cifar(varargin)
2  % CNN_CIFAR Demonstrates MatConvNet on CIFAR-10
3  % The demo includes two standard model: LeNet and
4  % Network (NIN). Use the 'modelType' option to cho
5
6  run(fullfile(fileparts(mfilename('fullpath')), ...
7  '..' , '..' , 'matlab' , 'vl_setupnn.m')) ;
8
9  opts.modelType = 'lenet' ;
10 [opts, varargin] = vl_argparse(opts, varargin) ;

```

cnn_cifar.m (Function)

Demonstrates MatConvNet on
CIFAR-10

- cnn_cifar(varargin)
- getBatch(opts)
- getSimpleNNBatch(imdb, batch)
- getDagNNBatch(opts, imdb, bat...
- getCifarImdb(opts)

Command Window

```

117/500: 617.5 (643.0) Hz objective: 0.808 top1err: 0.279 t
118/500: 617.7 (643.0) Hz objective: 0.809 top1err: 0.278 t
119/500: 617.8 (630.6) Hz objective: 0.809 top1err: 0.278 t
120/500: 617.8 (611.6) Hz objective: 0.807 top1err: 0.278 t
121/500: 617.7 (610.3) Hz objective: 0.807 top1err: 0.278 t
122/500: 617.9 (635.9) Hz objective: 0.808 top1err: 0.278 t
fx: 123/500:

```



```
Editor - /Users/apple/Desktop/Jiann-Ming Wu/code2019/MatConvNet/matconvnet-1.0-beta25_2019_10/examples/cifar/cnn_cifa...
script_load_scan_lattice_structured.m x script_render_fittings.m x script_load_scan.m x cnn_cifar.m x +
22 - opts = vl_argparse(opts, varargin) ;
23 - if ~isfield(opts.train, 'gpus'), opts.train.gpus = []; end;
24
25 % -----
26 %                                     Prepare r
27 % -----
28
29 switch opts.modelType
30     case 'lenet'
31         net = cnn_cifar_init('networkType', opts.networkType) ;
32     case 'nin'
```

Command Window

```
train: epoch 44: 50/500: 642.1 (639.2) Hz objective: 0.168 top1err:
train: epoch 44: 51/500: 642.5 (662.4) Hz objective: 0.169 top1err:
train: epoch 44: 52/500: 642.4 (637.2) Hz objective: 0.168 top1err:
train: epoch 44: 53/500: 642.6 (652.2) Hz objective: 0.169 top1err:
train: epoch 44: 54/500: 642.9 (659.1) Hz objective: 0.168 top1err:
fx train: epoch 44: 55/500:
```


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Current Folder: /Users/apple/Desktop/Jiann-Ming Wu/code2019/MatConvNet/matconvnet-1.0-beta25/examples/cifar

- cnncifar.m
- cnncifar_i...
- cnncifar_i...

```

22 - opts = vl_argparse(opts, varargin) ;
23 - if ~isfield(opts.train, 'gpus'), opts.train.gpus = []
24
25 -
    
```

Command Window

```

val: epoch 45: 97/100
val: epoch 45: 98/100
val: epoch 45: 99/100
val: epoch 45: 100/100
    
```

ans =

struct with fields:

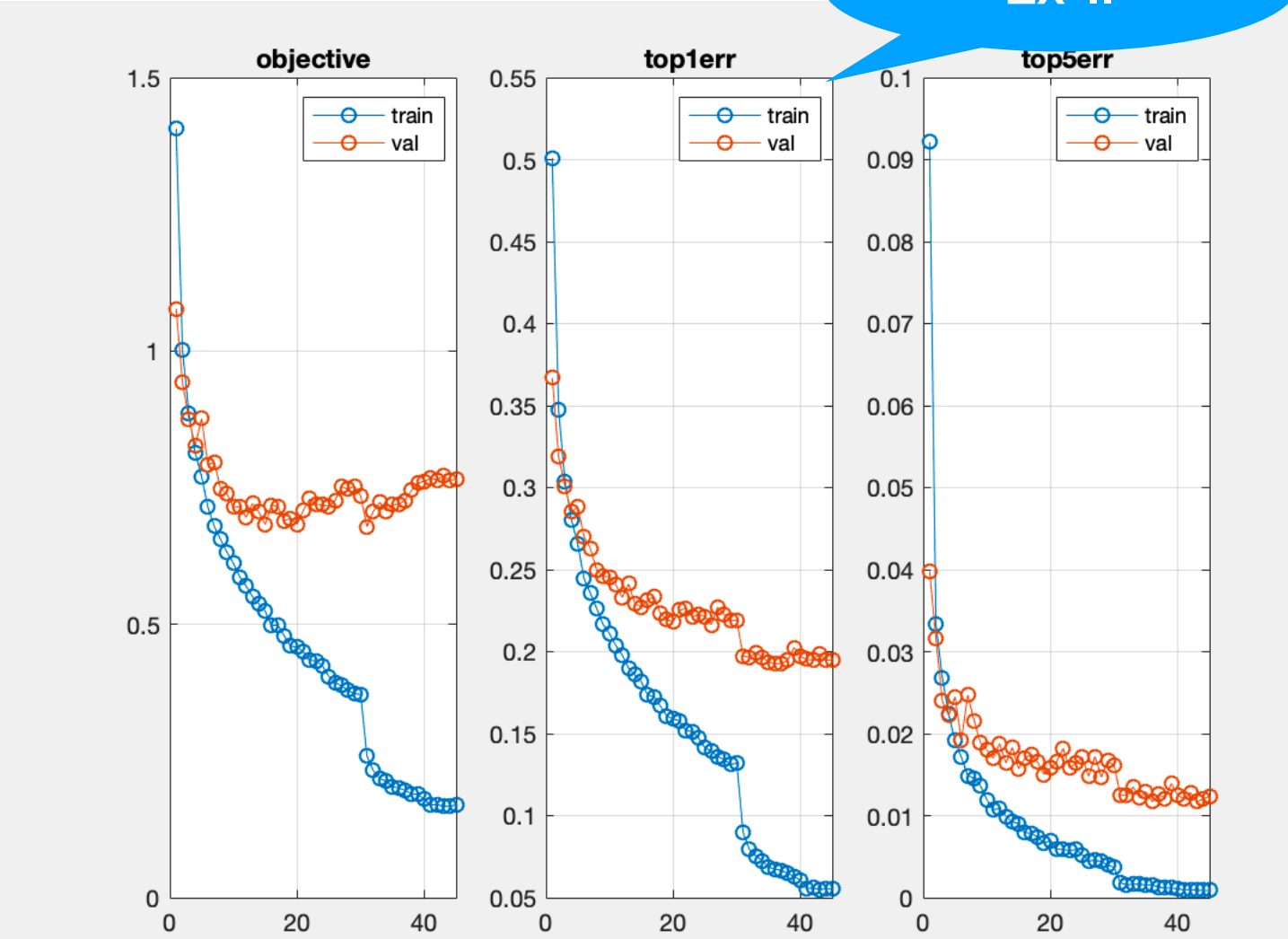
```

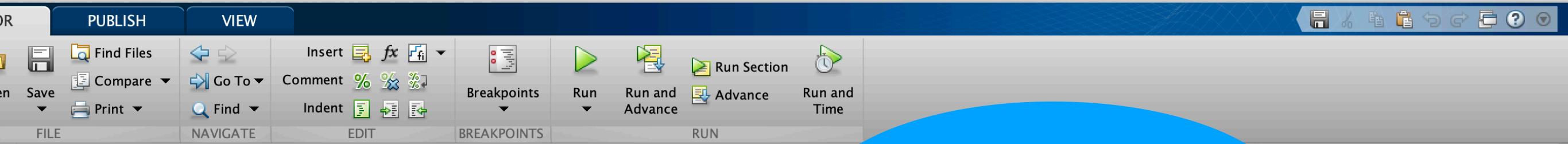
layers: {1x13 cell}
meta: [1x1 struct]
    
```

fx >>

Figure 1

File Edit View Insert Tools Desktop Window Help





Deep Convolution neural network

```
% Define network CIFAR10-quick
net.layers = {} ;

% Block 1
net.layers{end+1} = struct('type', 'conv', ...
    'weights', {{0.01*randn(5,5,3,32, 'single'), zeros(1, 32, 'single')}, ...
    'learningRate', lr, ...
    'stride', 1, ...
    'pad', 2) ;
net.layers{end+1} = struct('type', 'pool', ...
    'method', 'max', ...
    'pool', [3 3], ...
    'stride', 2, ...
    'pad', [0 1 0 1]) ;
net.layers{end+1} = struct('type', 'relu') ;

% Block 2
net.layers{end+1} = struct('type', 'conv', ...
    'weights', {{0.05*randn(5,5,32,32, 'single'), zeros(1,32, 'single')}, ...
    'learningRate', lr, ...
    'stride', 1, ...
```

Training set

The image shows a MATLAB IDE window with a file explorer on the left and a code editor on the right. The file explorer shows a folder named 'Current Folder' containing files: 'batches.meta.mat', 'data_batch_1.mat', 'data_batch_2.mat', 'data_batch_3.mat', 'data_batch_4.mat', 'data_batch_5.mat', 'readme.html', and 'test_batch.mat'. A blue callout bubble points from the 'Training set' text to the 'data_batch_1.mat' file. The code editor shows the following MATLAB code:

```
47 ... 'pad', [0 1 0 1]) ; % Emulate caffe
48
49 % Block 4
50 net.layers{end+1} = struct('type', 'conv', ...
51 ... 'weights', {0.05*randn(4,4,64,64, 'si
52 ... 'learningRate', lr, ...
53 ... 'stride', 1, ...
54 ... 'pad', 0) ;
55 net.layers{end+1} = struct('type', 'relu') ;
56
57 % Block 5
58 net.layers{end+1} = struct('type', 'conv', ...
59 ... 'weights', {0.05*randn(1,1,64,10, 'si
60 ... 'learningRate', .1*lr, ...
61 ... 'stride', 1, ...
```

A second blue callout bubble points from the 'Testing set' text to the 'test_batch.mat' file in the file explorer.

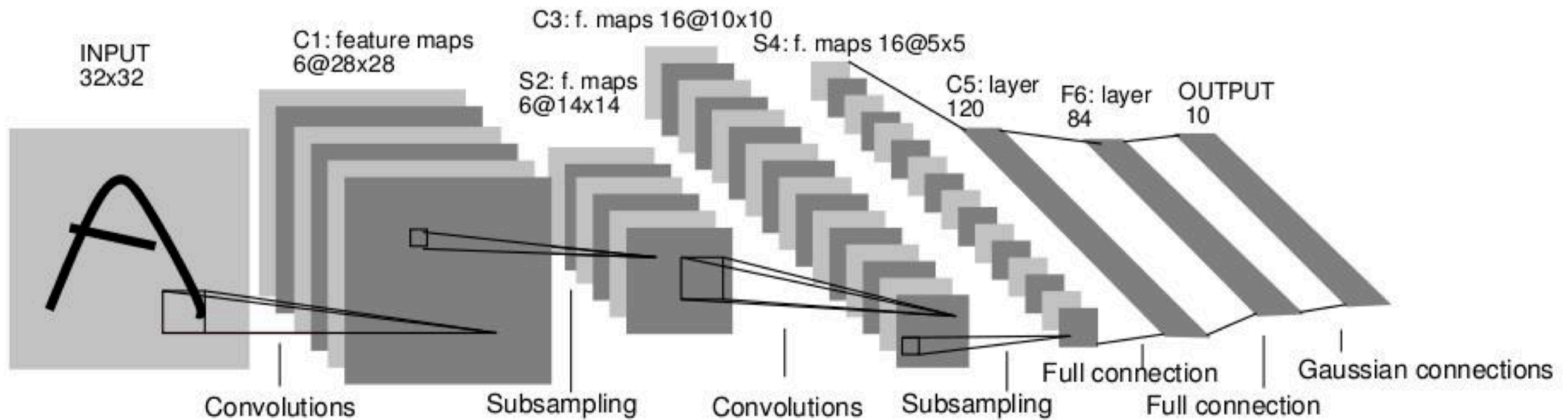
Testing set

Learning Multiple Layers of Features from Tiny Images

Alex Krizhevsky

April 8, 2009

LeNet 5



Y. LeCun, L. Bottou, Y. Bengio, and P. Haffner,

[Gradient-based learning applied to document recognition](#), Proc. IEEE 86(11): 2278–2324, 1998.

```

% Define network CIFAR10-quick
net.layers = {};

% Block 1
net.layers{end+1} = struct('type', 'conv', ...
    'weights', {{0.01*randn(5,5,3,32, 'single'), zeros(1, 32, 'single')}}}, ...
    'learningRate', lr, ...
    'stride', 1, ...
    'pad', 2);
net.layers{end+1} = struct('type', 'pool', ...
    'method', 'max', ...
    'pool', [3 3], ...
    'stride', 2, ...
    'pad', [0 1 0 1]);
net.layers{end+1} = struct('type', 'relu');

% Block 2
net.layers{end+1} = struct('type', 'conv', ...
    'weights', {{0.05*randn(5,5,32,32, 'single'), zeros(1,32,'single')}}}, ...
    'learningRate', lr, ...
    'stride', 1, ...
    'pad', 2);
net.layers{end+1} = struct('type', 'relu');
net.layers{end+1} = struct('type', 'pool', ...
    'method', 'avg', ...
    'pool', [3 3], ...
    'stride', 2, ...
    'pad', [0 1 0 1]); % Emulate caffe

% Block 3
net.layers{end+1} = struct('type', 'conv', ...
    'weights', {{0.05*randn(5,5,32,64, 'single'), zeros(1,64,'single')}}}, ...
    'learningRate', lr, ...
    'stride', 1, ...
    'pad', 2);
net.layers{end+1} = struct('type', 'relu');
net.layers{end+1} = struct('type', 'pool', ...
    'method', 'avg', ...
    'pool', [3 3], ...
    'stride', 2, ...
    'pad', [0 1 0 1]); % Emulate caffe

% Block 4
net.layers{end+1} = struct('type', 'conv', ...
    'weights', {{0.05*randn(4,4,64,64, 'single'), zeros(1,64,'single')}}}, ...
    'learningRate', lr, ...
    'stride', 1, ...
    'pad', 0);
net.layers{end+1} = struct('type', 'relu');

% Block 5
net.layers{end+1} = struct('type', 'conv', ...
    'weights', {{0.05*randn(1,1,64,10, 'single'), zeros(1,10,'single')}}}, ...
    'learningRate', .1*lr, ...
    'stride', 1, ...
    'pad', 0);

% Loss layer
net.layers{end+1} = struct('type', 'softmaxloss');

```


Caffe

Deep learning framework
by [BAIR](#)

Created by
[Yangqing Jia](#)
Lead Developer
[Evan Shelhamer](#)

 [View On GitHub](#)

Alex's CIFAR-10 tutorial, Caffe style

Alex Krizhevsky's [cuda-convnet](#) details the model definitions, parameters, and training procedure for good performance on CIFAR-10. This example reproduces his results in Caffe.

We will assume that you have Caffe successfully compiled. If not, please refer to the [Installation page](#). In this tutorial, we will assume that your caffe installation is located at `CAFFE_ROOT`.

We thank @chyojn for the pull request that defined the model schemas and solver configurations.

This example is a work-in-progress. It would be nice to further explain details of the network and training choices and benchmark the full training.

Prepare the Dataset

You will first need to download and convert the data format from the [CIFAR-10 website](#). To do this, simply run the following commands:

```
cd $CAFFE_ROOT
./data/cifar10/get_cifar10.sh
./examples/cifar10/create_cifar10.sh
```

If it complains that `wget` or `gunzip` are not installed, you need to install them respectively. After running the script there should be the dataset, `./cifar10-leveldb`, and the data set image mean `./mean.binaryproto`.

Architecture and Deep Learning

Example : network CIFAR10-quick

Current Folder: / Users apple Desktop Jiann-Ming Wu code2019 MatConvNet matconvnet-1.0-beta25 data cifar cifar-10-batches-mat

Files in Current Folder:
batches.meta.mat
data_batch_1.mat
data_batch_2.mat
data_batch_3.mat
data_batch_4.mat
data_batch_5.mat
readme.html
test_batch.mat

Editor - cnn_cifar.m
Variables - meta

Workspace:
batch_label
data
im
labels

airplane
automobile
bird
cat
deer
dog
frog
horse
ship
truck

Figure 1
File Edit View Insert Tools Desktop Window Help

```
>> load('test_batch.mat')  
>> im=reshape(data(1,:),32,32,3);  
>> imshow(im)  
fx >>
```

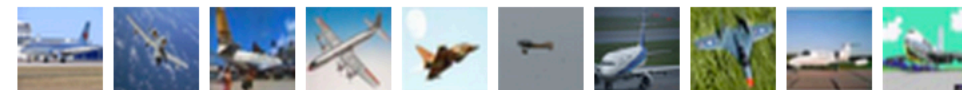
```
>> load('test_batch.mat')
>> im=reshape(data(1,:),32,32,3);
>> imshow(im)
>> labels(1)
```

ans =

uint8

3

airplane



automobile



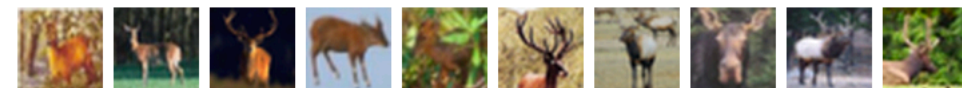
bird



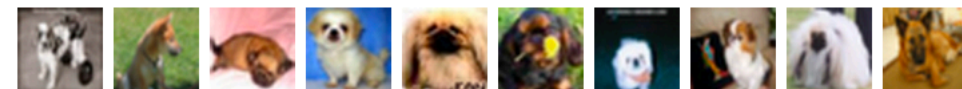
cat



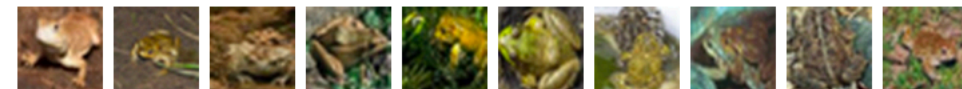
deer



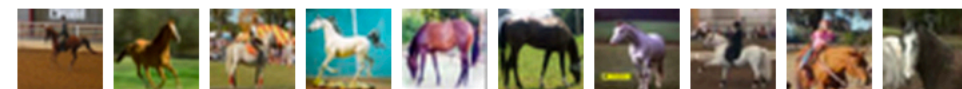
dog



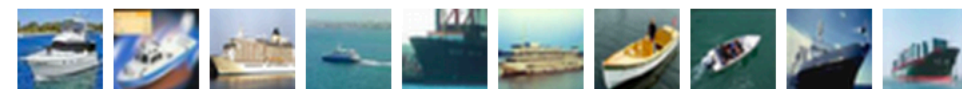
frog



horse



ship



truck



Mid-term project

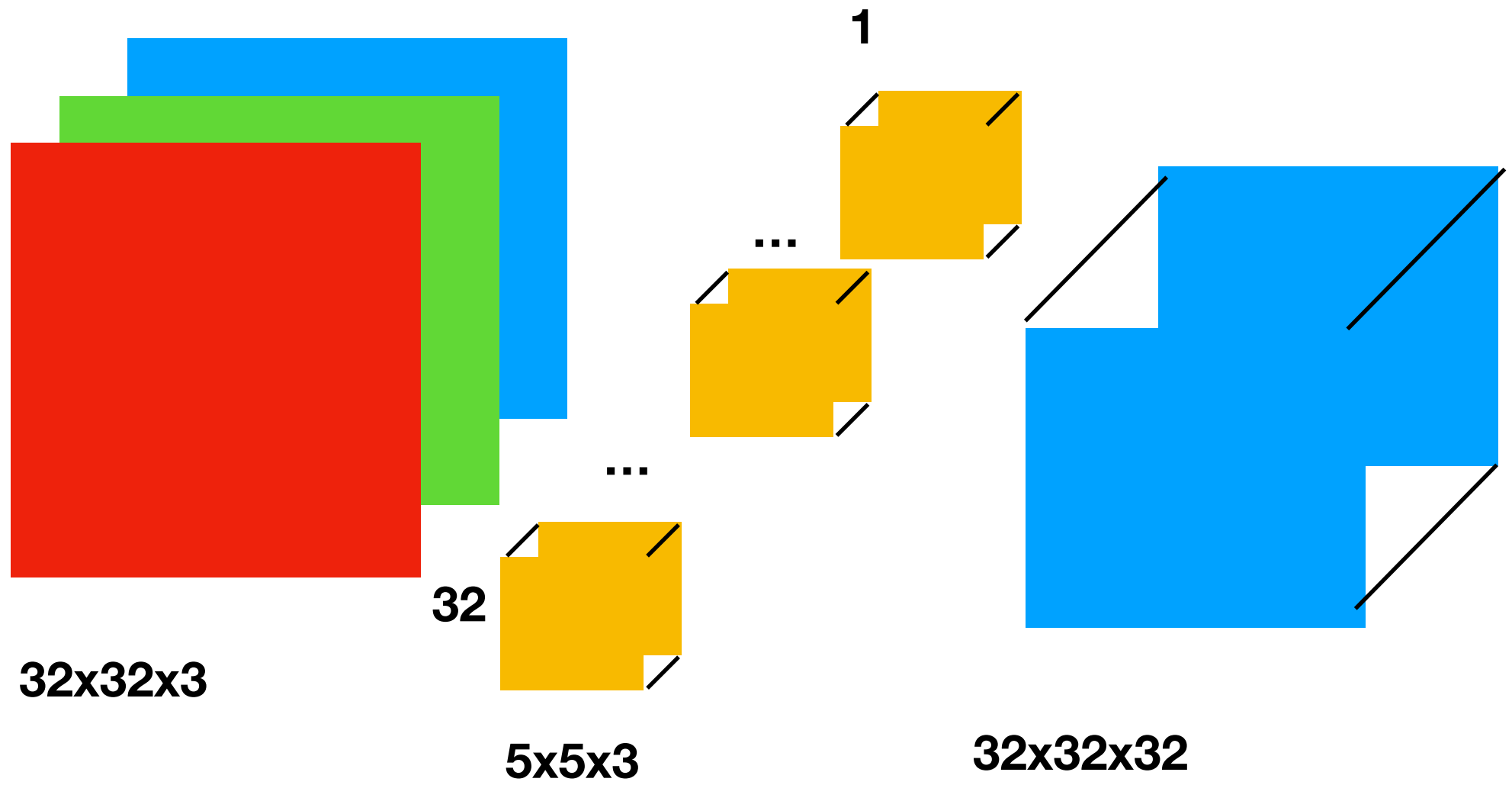
Exercise 1

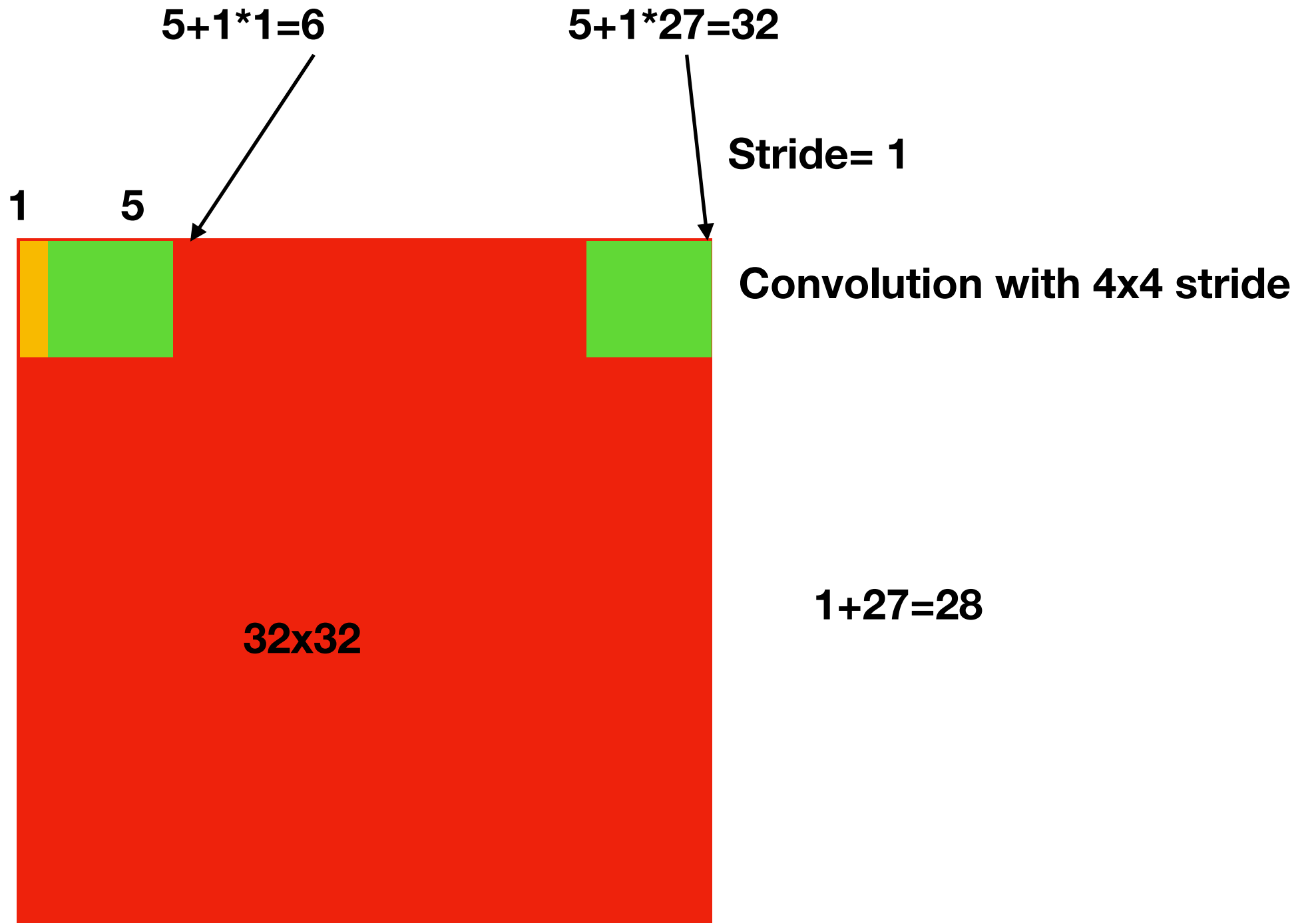
Execute `cnn_cifar.m` one epoch and trace `vl_simplenn.m`.

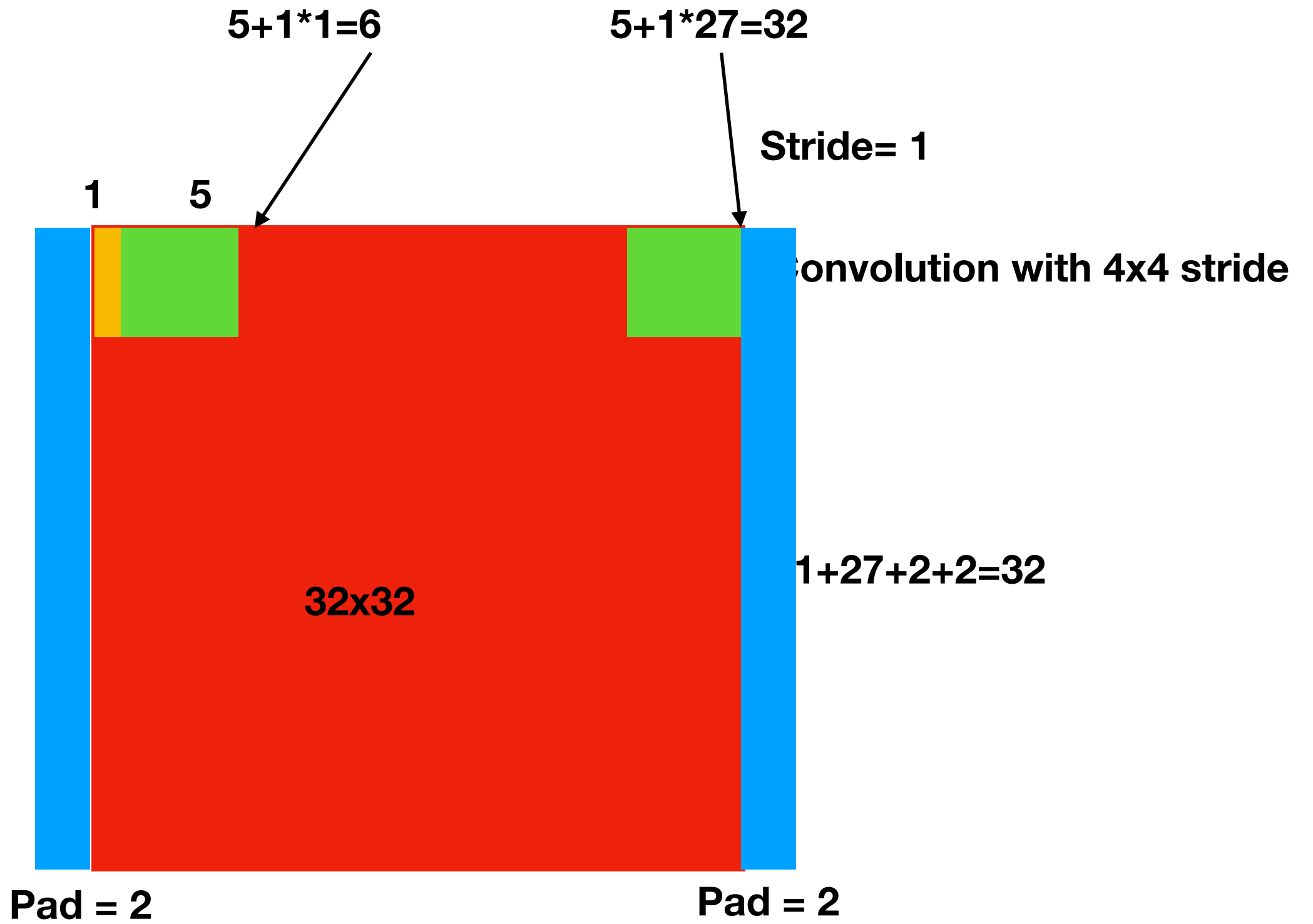
State the input size and output size of each layer and sketch the architecture

100 images per batch

Convolution







Current Folder

- net-epoch-25.mat
- net-epoch-26.mat
- net-epoch-27.mat
- net-epoch-28.mat
- net-epoch-29.mat
- net-epoch-30.mat
- net-epoch-31.mat
- net-epoch-32.mat
- net-epoch-33.mat
- net-epoch-34.mat
- net-epoch-35.mat
- net-epoch-36.mat
- net-epoch-37.mat
- net-epoch-38.mat
- net-epoch-39.mat
- net-epoch-40.mat
- net-epoch-41.mat
- net-epoch-42.mat
- net-epoch-43.mat
- net-epoch-44.mat
- net-train.pdf

Details

```

299 - case 'conv'
300 -> res(i+1).x = vl_nnconv(res(i).x, l.
301     'pad', l.pad, ...
302     'stride', l.stride, ...
303     'dilate', l.dilate, ...
304     l.opts{:}, ...
305     cudnn{:}) ;
306
307 - case 'convt'

```

Command Window

data mem	1MB	12MB	3MB	3MB	3MB
param mem	n/a	10KB	0B	0B	100KB

parameter memory | 569KB (1.5e+05 parameters) |
data memory | 31MB (for batch size 100) |

epoch 45: 1/500: 294 for i=1:n

J> K>>

delete net-epoch-44.mat

Select a file to

HOME

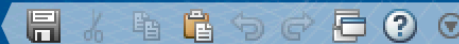
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 Go To
 Find

Insert
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 Indent
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 Continue
 Step
 Run to Cursor

Step In
 Step Out
 Quit Debugging

Function Call Stack:

vl_simplenn

Quit Debugging

FILE

NAVIGATE

EDIT

BREAKPOINTS

DEBUG

/ Users apple Desktop Jiann-Ming Wu code2019 MatConvNet matconvnet-1.0-beta25 examples cifar

Current Folder

- Name
- cnn_cifar.m
- cnn_cifar_init.m
- cnn_cifar_init_nin.m

Editor - /Users/apple/Desktop/Jiann-Ming Wu/code2019/MatConvNet/matconvnet-1.0-...

```

+18
290 % -----
291 %
292 % -----
293
294 for i=1:n
295     if opts.skipForward, break; end;
296     l = net.layers{i} ;
297     res(i).time = tic ;
298     switch l.type
  
```

Workspace - vl_si...

Name	Value
backPropLim	1
bnormCudnn	1x1 ce
cudnn	1x1 ce
doder	1
dzdy	1
gpuMode	0
n	13
net	1x1 st
opts	1x1 st
res	1x14 s
testMode	0
varargin	1x16 c
x	4-D s

Command Window

```

data mem| 1MB| 12MB| 3MB| 3MB|
param mem| n/a| 10KB| 0B|
  
```

```

parameter memory|569KB (1.5e+05 para
data memory| 31MB (for batch size
  
```

```

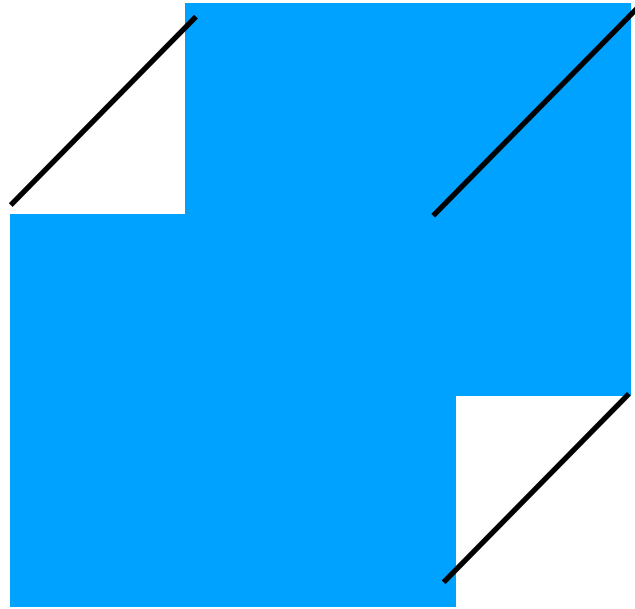
cnn_train: resuming by loading epoch 44
train: epoch 45: 1/500:294 for i=1:n
  
```

fx K>>

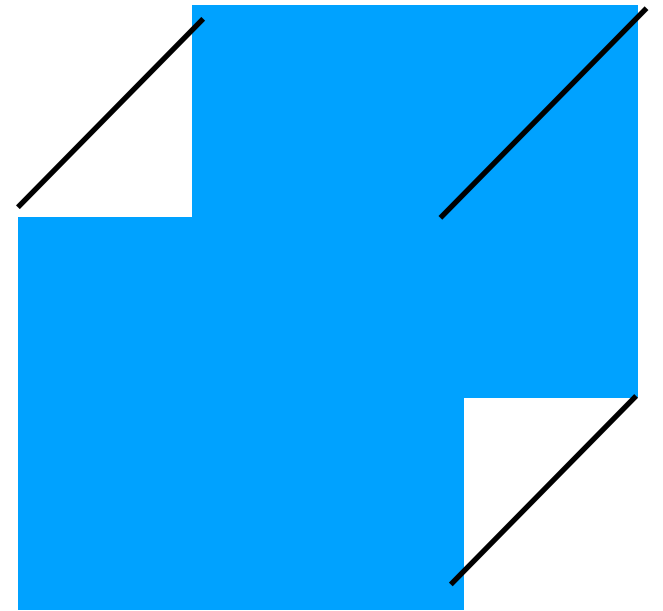
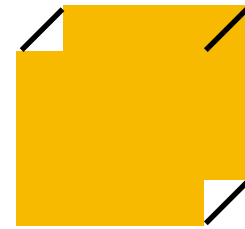
Execute cnn_cifar.m one epoch

Select a file to view details

Max pooling



??x??x??



??x??x??

Exercise II

- 1. Use deep network designer App to design cifar10-quick**
- 2. Train and test your deep network**

Exercise III





- 1. Use deep network designer App to design your deep network for improving cifar10-quick**
- 2. Train and test your deep network**

 New	 Import	 Duplicate	 Cut	 Copy	 Paste	 Fit to View	 Zoom In	 Zoom Out	 Auto Arrange	 Analyze	 Export
FILE		BUILD			NAVIGATE		LAYOUT		ANALYSIS		EXPORT







LAYER LIBRARY

Filter layers...



INPUT

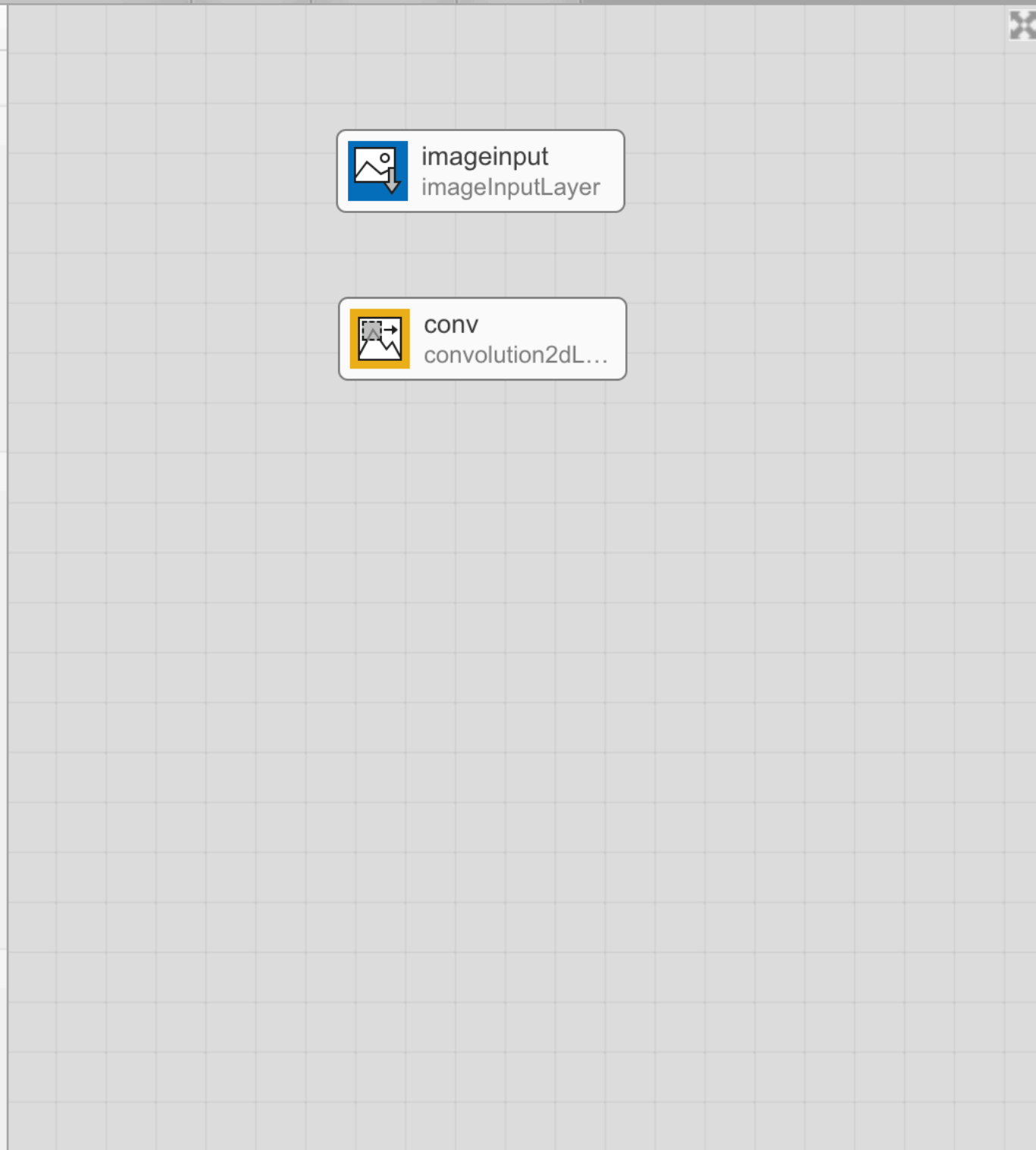
-  imageInputLayer
-  image3dInputLayer
-  sequenceInputLayer
-  roiInputLayer

CONVOLUTION AND FULLY CONNECTED

-  convolution2dLayer
-  convolution3dLayer
-  groupedConvolution2dLayer
-  transposedConv2dLayer
-  transposedConv3dLayer
-  fullyConnectedLayer

SEQUENCE

-  lstmLayer
-  bilstmLayer



PROPERTIES

Number of layers	2
Number of connections	0
Input type	Image
Output type	None