```
In [ ]: # This mounts your Google Drive to the Colab VM.
        from google.colab import drive
        drive.mount('/content/drive')
        # TODO: Enter the foldername in your Drive where you have saved the unzipped
        # assignment folder, e.g. 'cs6353/assignments/assignment2/'
        FOLDERNAME = 'CS6353/Assignments/assignment2/'
        assert FOLDERNAME is not None, "[!] Enter the foldername."
        # Now that we've mounted your Drive, this ensures that
        # the Python interpreter of the Colab VM can load
        # python files from within it.
        import sys
        sys.path.append('/content/drive/My Drive/{}'.format(FOLDERNAME))
        # This downloads the CIFAR-10 dataset to your Drive
        # if it doesn't already exist.
        %cd /content/drive/My\ Drive/$FOLDERNAME/cs6353/datasets/
        !bash get_datasets.sh
        %cd /content/drive/My\ Drive/$FOLDERNAME
        # Install requirements from colab_requirements.txt
        # TODO: Please change your path below to the colab_requirements.txt file
        ! python -m pip install -r /content/drive/My\ Drive/$FOLDERNAME/colab_requirements.txt
```

```
Mounted at /content/drive
/content/drive/My Drive/CS6353/Assignments/assignment2/assignment2/cs6353/datasets
--2024-09-29 20:36:13-- http://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz
Resolving www.cs.toronto.edu (www.cs.toronto.edu)... 128.100.3.30
Connecting to www.cs.toronto.edu (www.cs.toronto.edu) | 128.100.3.30 | :80... connected.
HTTP request sent, awaiting response... 200 OK
Length: 170498071 (163M) [application/x-gzip]
Saving to: 'cifar-10-python.tar.gz'
cifar-10-python.tar 100%[========>] 162.60M 16.0MB/s
2024-09-29 20:36:25 (14.3 MB/s) - 'cifar-10-python.tar.gz' saved [170498071/17049807
1]
cifar-10-batches-py/
cifar-10-batches-py/data_batch_4
cifar-10-batches-py/readme.html
cifar-10-batches-py/test_batch
cifar-10-batches-py/data batch 3
cifar-10-batches-py/batches.meta
cifar-10-batches-py/data_batch_2
cifar-10-batches-py/data_batch_5
cifar-10-batches-py/data_batch_1
/content/drive/My Drive/CS6353/Assignments/assignment2/assignment2
Requirement already satisfied: anyio==3.7.1 in /usr/local/lib/python3.10/dist-package
s (from -r /content/drive/My Drive/CS6353/Assignments/assignment2/assignment2//colab_
requirements.txt (line 1)) (3.7.1)
Collecting appnope==0.1.3 (from -r /content/drive/My Drive/CS6353/Assignments/assignm
ent2/assignment2//colab_requirements.txt (line 2))
  Downloading appnope-0.1.3-py2.py3-none-any.whl.metadata (1.2 kB)
Requirement already satisfied: argon2-cffi==23.1.0 in /usr/local/lib/python3.10/dist-
packages (from -r /content/drive/My Drive/CS6353/Assignments/assignment2/assignment
2//colab_requirements.txt (line 3)) (23.1.0)
Requirement already satisfied: argon2-cffi-bindings==21.2.0 in /usr/local/lib/python
3.10/dist-packages (from -r /content/drive/My Drive/CS6353/Assignments/assignment2/as
signment2//colab_requirements.txt (line 4)) (21.2.0)
Collecting arrow==1.2.3 (from -r /content/drive/My Drive/CS6353/Assignments/assignmen
t2/assignment2//colab requirements.txt (line 5))
  Downloading arrow-1.2.3-py3-none-any.whl.metadata (6.9 kB)
Collecting asttokens==2.2.1 (from -r /content/drive/My Drive/CS6353/Assignments/assig
nment2/assignment2//colab_requirements.txt (line 6))
  Downloading asttokens-2.2.1-py2.py3-none-any.whl.metadata (4.8 kB)
Collecting async-lru==2.0.4 (from -r /content/drive/My Drive/CS6353/Assignments/assig
nment2/assignment2//colab_requirements.txt (line 7))
  Downloading async_lru-2.0.4-py3-none-any.whl.metadata (4.5 kB)
Collecting attrs==23.1.0 (from -r /content/drive/My Drive/CS6353/Assignments/assignme
nt2/assignment2//colab requirements.txt (line 8))
  Downloading attrs-23.1.0-py3-none-any.whl.metadata (11 kB)
Collecting Babel==2.12.1 (from -r /content/drive/My Drive/CS6353/Assignments/assignme
nt2/assignment2//colab_requirements.txt (line 9))
  Downloading Babel-2.12.1-py3-none-any.whl.metadata (1.3 kB)
Requirement already satisfied: backcall==0.2.0 in /usr/local/lib/python3.10/dist-pack
ages (from -r /content/drive/My Drive/CS6353/Assignments/assignment2/assignment2//col
ab_requirements.txt (line 10)) (0.2.0)
Collecting beautifulsoup4==4.12.2 (from -r /content/drive/My Drive/CS6353/Assignment
s/assignment2/assignment2//colab_requirements.txt (line 11))
  Downloading beautifulsoup4-4.12.2-py3-none-any.whl.metadata (3.6 kB)
Collecting bleach==6.0.0 (from -r /content/drive/My Drive/CS6353/Assignments/assignme
nt2/assignment2//colab requirements.txt (line 12))
  Downloading bleach-6.0.0-py3-none-any.whl.metadata (29 kB)
```

```
Collecting certifi==2023.7.22 (from -r /content/drive/My Drive/CS6353/Assignments/ass
ignment2/assignment2//colab_requirements.txt (line 13))
  Downloading certifi-2023.7.22-py3-none-any.whl.metadata (2.2 kB)
Collecting cffi==1.15.1 (from -r /content/drive/My Drive/CS6353/Assignments/assignmen
t2/assignment2//colab_requirements.txt (line 14))
  Downloading cffi-1.15.1-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86 64.whl.
metadata (1.1 kB)
Collecting charset-normalizer==3.2.0 (from -r /content/drive/My Drive/CS6353/Assignme
nts/assignment2/assignment2//colab_requirements.txt (line 15))
  Downloading charset_normalizer-3.2.0-cp310-cp310-manylinux_2_17_x86_64.manylinux201
4 x86 64.whl.metadata (31 kB)
Collecting comm==0.1.4 (from -r /content/drive/My Drive/CS6353/Assignments/assignment
2/assignment2//colab_requirements.txt (line 16))
  Downloading comm-0.1.4-py3-none-any.whl.metadata (4.2 kB)
Collecting contourpy==1.1.0 (from -r /content/drive/My Drive/CS6353/Assignments/assig
nment2/assignment2//colab_requirements.txt (line 17))
  Downloading contourpy-1.1.0-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.
whl.metadata (5.7 kB)
Collecting cycler==0.11.0 (from -r /content/drive/My Drive/CS6353/Assignments/assignm
ent2/assignment2//colab requirements.txt (line 18))
  Downloading cycler-0.11.0-py3-none-any.whl.metadata (785 bytes)
Collecting debugpy==1.6.7.post1 (from -r /content/drive/My Drive/CS6353/Assignments/a
ssignment2/assignment2//colab_requirements.txt (line 19))
  Downloading debugpy-1.6.7.post1-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86
64.whl.metadata (1.1 kB)
Requirement already satisfied: decorator<=5.0 in /usr/local/lib/python3.10/dist-packa
ges (from -r /content/drive/My Drive/CS6353/Assignments/assignment2/assignment2//cola
b_requirements.txt (line 20)) (4.4.2)
Requirement already satisfied: defusedxml==0.7.1 in /usr/local/lib/python3.10/dist-pa
ckages (from -r /content/drive/My Drive/CS6353/Assignments/assignment2//c
olab_requirements.txt (line 21)) (0.7.1)
Collecting executing==1.2.0 (from -r /content/drive/My Drive/CS6353/Assignments/assig
nment2/assignment2//colab requirements.txt (line 22))
  Downloading executing-1.2.0-py2.py3-none-any.whl.metadata (8.9 kB)
Collecting fastjsonschema==2.18.0 (from -r /content/drive/My Drive/CS6353/Assignment
s/assignment2/assignment2//colab_requirements.txt (line 23))
  Downloading fastjsonschema-2.18.0-py3-none-any.whl.metadata (2.0 kB)
Collecting fonttools==4.42.1 (from -r /content/drive/My Drive/CS6353/Assignments/assi
gnment2/assignment2//colab requirements.txt (line 24))
  Downloading fonttools-4.42.1-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_6
4.whl.metadata (150 kB)
                                         ---- 151.0/151.0 kB 5.1 MB/s eta 0:00:00
Collecting fqdn==1.5.1 (from -r /content/drive/My Drive/CS6353/Assignments/assignment
2/assignment2//colab_requirements.txt (line 25))
  Downloading fqdn-1.5.1-py3-none-any.whl.metadata (1.4 kB)
Collecting idna==3.4 (from -r /content/drive/My Drive/CS6353/Assignments/assignment2/
assignment2//colab requirements.txt (line 26))
  Downloading idna-3.4-py3-none-any.whl.metadata (9.8 kB)
Collecting imageio==2.31.1 (from -r /content/drive/My Drive/CS6353/Assignments/assign
ment2/assignment2//colab_requirements.txt (line 27))
  Downloading imageio-2.31.1-py3-none-any.whl.metadata (4.7 kB)
Requirement already satisfied: ipykernel<=5.5.6 in /usr/local/lib/python3.10/dist-pac
kages (from -r /content/drive/My Drive/CS6353/Assignments/assignment2/assignment2//co
lab_requirements.txt (line 28)) (5.5.6)
Requirement already satisfied: ipython<=7.34.0 in /usr/local/lib/python3.10/dist-pack
ages (from -r /content/drive/My Drive/CS6353/Assignments/assignment2/assignment2//col
ab_requirements.txt (line 29)) (7.34.0)
Collecting isoduration==20.11.0 (from -r /content/drive/My Drive/CS6353/Assignments/a
ssignment2/assignment2//colab requirements.txt (line 30))
  Downloading isoduration-20.11.0-py3-none-any.whl.metadata (5.7 kB)
```

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Collecting jedi==0.19.0 (from -r /content/drive/My Drive/CS6353/Assignments/assignmen
t2/assignment2//colab_requirements.txt (line 31))
  Downloading jedi-0.19.0-py2.py3-none-any.whl.metadata (22 kB)
Collecting Jinja2==3.1.2 (from -r /content/drive/My Drive/CS6353/Assignments/assignme
nt2/assignment2//colab_requirements.txt (line 32))
  Downloading Jinja2-3.1.2-py3-none-any.whl.metadata (3.5 kB)
Collecting json5==0.9.14 (from -r /content/drive/My Drive/CS6353/Assignments/assignme
nt2/assignment2//colab_requirements.txt (line 33))
  Downloading json5-0.9.14-py2.py3-none-any.whl.metadata (10 kB)
Collecting jsonpointer==2.4 (from -r /content/drive/My Drive/CS6353/Assignments/assig
nment2/assignment2//colab requirements.txt (line 34))
  Downloading jsonpointer-2.4-py2.py3-none-any.whl.metadata (2.5 kB)
Collecting jsonschema==4.19.0 (from -r /content/drive/My Drive/CS6353/Assignments/ass
ignment2/assignment2//colab_requirements.txt (line 35))
  Downloading jsonschema-4.19.0-py3-none-any.whl.metadata (8.2 kB)
Collecting jsonschema-specifications==2023.7.1 (from -r /content/drive/My Drive/CS635
3/Assignments/assignment2/assignment2//colab requirements.txt (line 36))
  Downloading jsonschema_specifications-2023.7.1-py3-none-any.whl.metadata (2.8 kB)
Collecting jupyter-events==0.7.0 (from -r /content/drive/My Drive/CS6353/Assignments/
assignment2/assignment2//colab requirements.txt (line 37))
  Downloading jupyter_events-0.7.0-py3-none-any.whl.metadata (5.5 kB)
Collecting jupyter-lsp==2.2.0 (from -r /content/drive/My Drive/CS6353/Assignments/ass
ignment2/assignment2//colab_requirements.txt (line 38))
  Downloading jupyter_lsp-2.2.0-py3-none-any.whl.metadata (1.8 kB)
Requirement already satisfied: jupyter_client<8.0 in /usr/local/lib/python3.10/dist-p
ackages (from -r /content/drive/My Drive/CS6353/Assignments/assignment2/assignment2/
colab_requirements.txt (line 39)) (6.1.12)
Collecting jupyter_core==5.3.1 (from -r /content/drive/My Drive/CS6353/Assignments/as
signment2/assignment2//colab_requirements.txt (line 40))
  Downloading jupyter_core-5.3.1-py3-none-any.whl.metadata (3.4 kB)
Collecting jupyter_server==2.7.2 (from -r /content/drive/My Drive/CS6353/Assignments/
assignment2/assignment2//colab_requirements.txt (line 41))
  Downloading jupyter_server-2.7.2-py3-none-any.whl.metadata (8.6 kB)
Collecting jupyter_server_terminals==0.4.4 (from -r /content/drive/My Drive/CS6353/As
signments/assignment2//colab requirements.txt (line 42))
  Downloading jupyter_server_terminals-0.4.4-py3-none-any.whl.metadata (6.3 kB)
Collecting jupyterlab==4.0.5 (from -r /content/drive/My Drive/CS6353/Assignments/assi
gnment2/assignment2//colab requirements.txt (line 43))
  Downloading jupyterlab-4.0.5-py3-none-any.whl.metadata (15 kB)
Collecting jupyterlab-pygments==0.2.2 (from -r /content/drive/My Drive/CS6353/Assignm
ents/assignment2/assignment2//colab_requirements.txt (line 44))
  Downloading jupyterlab_pygments-0.2.2-py2.py3-none-any.whl.metadata (1.9 kB)
Collecting jupyterlab server==2.24.0 (from -r /content/drive/My Drive/CS6353/Assignme
nts/assignment2/assignment2//colab_requirements.txt (line 45))
  Downloading jupyterlab_server-2.24.0-py3-none-any.whl.metadata (5.8 kB)
Collecting kiwisolver==1.4.5 (from -r /content/drive/My Drive/CS6353/Assignments/assi
gnment2/assignment2//colab requirements.txt (line 46))
  Downloading kiwisolver-1.4.5-cp310-cp310-manylinux_2_12_x86_64.manylinux2010_x86_6
4.whl.metadata (6.4 kB)
Collecting MarkupSafe==2.1.3 (from -r /content/drive/My Drive/CS6353/Assignments/assi
gnment2/assignment2//colab_requirements.txt (line 47))
 Downloading MarkupSafe-2.1.3-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_6
4.whl.metadata (3.0 kB)
Collecting matplotlib==3.7.2 (from -r /content/drive/My Drive/CS6353/Assignments/assi
gnment2/assignment2//colab_requirements.txt (line 48))
 Downloading matplotlib-3.7.2-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_6
4.whl.metadata (5.6 kB)
Collecting matplotlib-inline==0.1.6 (from -r /content/drive/My Drive/CS6353/Assignmen
ts/assignment2/assignment2//colab_requirements.txt (line 49))
  Downloading matplotlib_inline-0.1.6-py3-none-any.whl.metadata (2.8 kB)
```

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Collecting mistune==3.0.1 (from -r /content/drive/My Drive/CS6353/Assignments/assignm
ent2/assignment2//colab_requirements.txt (line 50))
  Downloading mistune-3.0.1-py3-none-any.whl.metadata (1.7 kB)
Collecting nbclient==0.8.0 (from -r /content/drive/My Drive/CS6353/Assignments/assign
ment2/assignment2//colab_requirements.txt (line 51))
  Downloading nbclient-0.8.0-py3-none-any.whl.metadata (7.8 kB)
Collecting nbconvert==7.7.4 (from -r /content/drive/My Drive/CS6353/Assignments/assig
nment2/assignment2//colab_requirements.txt (line 52))
  Downloading nbconvert-7.7.4-py3-none-any.whl.metadata (8.0 kB)
Collecting nbformat==5.9.2 (from -r /content/drive/My Drive/CS6353/Assignments/assign
ment2/assignment2//colab_requirements.txt (line 53))
  Downloading nbformat-5.9.2-py3-none-any.whl.metadata (3.4 kB)
Collecting nest-asyncio==1.5.7 (from -r /content/drive/My Drive/CS6353/Assignments/as
signment2/assignment2//colab_requirements.txt (line 54))
  Downloading nest asyncio-1.5.7-py3-none-any.whl.metadata (2.7 kB)
Collecting notebook_shim==0.2.3 (from -r /content/drive/My Drive/CS6353/Assignments/a
ssignment2/assignment2//colab requirements.txt (line 55))
  Downloading notebook_shim-0.2.3-py3-none-any.whl.metadata (4.0 kB)
Collecting numpy<1.24,>=1.22 (from -r /content/drive/My Drive/CS6353/Assignments/assi
gnment2/assignment2//colab requirements.txt (line 56))
  Downloading numpy-1.23.5-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.wh
1.metadata (2.3 kB)
Collecting overrides==7.4.0 (from -r /content/drive/My Drive/CS6353/Assignments/assig
nment2/assignment2//colab_requirements.txt (line 57))
  Downloading overrides-7.4.0-py3-none-any.whl.metadata (5.7 kB)
Collecting packaging==23.1 (from -r /content/drive/My Drive/CS6353/Assignments/assign
ment2/assignment2//colab_requirements.txt (line 58))
  Downloading packaging-23.1-py3-none-any.whl.metadata (3.1 kB)
Collecting pandas<=1.5.3 (from -r /content/drive/My Drive/CS6353/Assignments/assignme
nt2/assignment2//colab_requirements.txt (line 59))
  Downloading pandas-1.5.3-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.wh
1.metadata (11 kB)
Collecting pandocfilters==1.5.0 (from -r /content/drive/My Drive/CS6353/Assignments/a
ssignment2/assignment2//colab_requirements.txt (line 60))
  Downloading pandocfilters-1.5.0-py2.py3-none-any.whl.metadata (9.0 kB)
Collecting parso==0.8.3 (from -r /content/drive/My Drive/CS6353/Assignments/assignmen
t2/assignment2//colab_requirements.txt (line 61))
  Downloading parso-0.8.3-py2.py3-none-any.whl.metadata (7.5 kB)
Collecting pexpect==4.8.0 (from -r /content/drive/My Drive/CS6353/Assignments/assignm
ent2/assignment2//colab_requirements.txt (line 62))
  Downloading pexpect-4.8.0-py2.py3-none-any.whl.metadata (2.2 kB)
Requirement already satisfied: pickleshare==0.7.5 in /usr/local/lib/python3.10/dist-p
ackages (from -r /content/drive/My Drive/CS6353/Assignments/assignment2/assignment2/
colab_requirements.txt (line 63)) (0.7.5)
Collecting Pillow==10.0.0 (from -r /content/drive/My Drive/CS6353/Assignments/assignm
ent2/assignment2//colab_requirements.txt (line 64))
  Downloading Pillow-10.0.0-cp310-cp310-manylinux_2_28_x86_64.whl.metadata (9.5 kB)
Collecting platformdirs==3.10.0 (from -r /content/drive/My Drive/CS6353/Assignments/a
ssignment2/assignment2//colab_requirements.txt (line 65))
  Downloading platformdirs-3.10.0-py3-none-any.whl.metadata (11 kB)
Collecting prometheus-client==0.17.1 (from -r /content/drive/My Drive/CS6353/Assignme
nts/assignment2/assignment2//colab requirements.txt (line 66))
  Downloading prometheus_client-0.17.1-py3-none-any.whl.metadata (24 kB)
Collecting prompt-toolkit==3.0.39 (from -r /content/drive/My Drive/CS6353/Assignment
s/assignment2/assignment2//colab_requirements.txt (line 67))
  Downloading prompt toolkit-3.0.39-py3-none-any.whl.metadata (6.4 kB)
Requirement already satisfied: psutil==5.9.5 in /usr/local/lib/python3.10/dist-packag
es (from -r /content/drive/My Drive/CS6353/Assignments/assignment2/assignment2//colab
requirements.txt (line 68)) (5.9.5)
Requirement already satisfied: ptyprocess==0.7.0 in /usr/local/lib/python3.10/dist-pa
```

olab_requirements.txt (line 69)) (0.7.0)

nment2/assignment2//colab_requirements.txt (line 70))

Downloading pure_eval-0.2.2-py3-none-any.whl.metadata (6.2 kB)

ckages (from -r /content/drive/My Drive/CS6353/Assignments/assignment2//csignment2//c

Collecting pure-eval==0.2.2 (from -r /content/drive/My Drive/CS6353/Assignments/assig

Collecting pycparser==2.21 (from -r /content/drive/My Drive/CS6353/Assignments/assign

```
ment2/assignment2//colab requirements.txt (line 71))
  Downloading pycparser-2.21-py2.py3-none-any.whl.metadata (1.1 kB)
Collecting Pygments==2.16.1 (from -r /content/drive/My Drive/CS6353/Assignments/assig
nment2/assignment2//colab_requirements.txt (line 72))
  Downloading Pygments-2.16.1-py3-none-any.whl.metadata (2.5 kB)
Collecting pyparsing==3.0.9 (from -r /content/drive/My Drive/CS6353/Assignments/assig
nment2/assignment2//colab_requirements.txt (line 73))
  Downloading pyparsing-3.0.9-py3-none-any.whl.metadata (4.2 kB)
Requirement already satisfied: python-dateutil==2.8.2 in /usr/local/lib/python3.10/di
st-packages (from -r /content/drive/My Drive/CS6353/Assignments/assignment2/assignmen
t2//colab requirements.txt (line 74)) (2.8.2)
Collecting python-json-logger==2.0.7 (from -r /content/drive/My Drive/CS6353/Assignme
nts/assignment2/assignment2//colab requirements.txt (line 75))
  Downloading python json logger-2.0.7-py3-none-any.whl.metadata (6.5 kB)
Collecting pytz==2023.3 (from -r /content/drive/My Drive/CS6353/Assignments/assignmen
t2/assignment2//colab_requirements.txt (line 76))
  Downloading pytz-2023.3-py2.py3-none-any.whl.metadata (22 kB)
Collecting PyYAML==6.0.1 (from -r /content/drive/My Drive/CS6353/Assignments/assignme
nt2/assignment2//colab requirements.txt (line 77))
 Downloading PyYAML-6.0.1-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.wh
1.metadata (2.1 kB)
Requirement already satisfied: pyzmq<25 in /usr/local/lib/python3.10/dist-packages (f
rom -r /content/drive/My Drive/CS6353/Assignments/assignment2/assignment2//colab_requ
irements.txt (line 78)) (24.0.1)
Collecting referencing==0.30.2 (from -r /content/drive/My Drive/CS6353/Assignments/as
signment2/assignment2//colab_requirements.txt (line 79))
  Downloading referencing-0.30.2-py3-none-any.whl.metadata (2.6 kB)
Collecting requests==2.31.0 (from -r /content/drive/My Drive/CS6353/Assignments/assig
nment2/assignment2//colab requirements.txt (line 80))
  Downloading requests-2.31.0-py3-none-any.whl.metadata (4.6 kB)
Collecting rfc3339-validator==0.1.4 (from -r /content/drive/My Drive/CS6353/Assignmen
ts/assignment2/assignment2//colab requirements.txt (line 81))
  Downloading rfc3339_validator-0.1.4-py2.py3-none-any.whl.metadata (1.5 kB)
Collecting rfc3986-validator==0.1.1 (from -r /content/drive/My Drive/CS6353/Assignmen
ts/assignment2/assignment2//colab_requirements.txt (line 82))
  Downloading rfc3986_validator-0.1.1-py2.py3-none-any.whl.metadata (1.7 kB)
Collecting rpds-py==0.9.2 (from -r /content/drive/My Drive/CS6353/Assignments/assignm
ent2/assignment2//colab_requirements.txt (line 83))
  Downloading rpds_py-0.9.2-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.wh
1.metadata (3.7 kB)
Collecting scipy==1.11.2 (from -r /content/drive/My Drive/CS6353/Assignments/assignme
nt2/assignment2//colab requirements.txt (line 84))
  Downloading scipy-1.11.2-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.wh
1.metadata (59 kB)
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Collecting seaborn==0.12.2 (from -r /content/drive/My Drive/CS6353/Assignments/assign
ment2/assignment2//colab_requirements.txt (line 85))
  Downloading seaborn-0.12.2-py3-none-any.whl.metadata (5.4 kB)
Collecting Send2Trash==1.8.2 (from -r /content/drive/My Drive/CS6353/Assignments/assi
gnment2/assignment2//colab_requirements.txt (line 86))
  Downloading Send2Trash-1.8.2-py3-none-any.whl.metadata (4.0 kB)
Requirement already satisfied: six==1.16.0 in /usr/local/lib/python3.10/dist-packages
(from -r /content/drive/My Drive/CS6353/Assignments/assignment2/assignment2//colab_re
quirements.txt (line 87)) (1.16.0)
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Collecting sniffio==1.3.0 (from -r /content/drive/My Drive/CS6353/Assignments/assignm
ent2/assignment2//colab_requirements.txt (line 88))
  Downloading sniffio-1.3.0-py3-none-any.whl.metadata (3.6 kB)
Collecting soupsieve==2.4.1 (from -r /content/drive/My Drive/CS6353/Assignments/assig
nment2/assignment2//colab_requirements.txt (line 89))
  Downloading soupsieve-2.4.1-py3-none-any.whl.metadata (4.7 kB)
Collecting stack-data==0.6.2 (from -r /content/drive/My Drive/CS6353/Assignments/assi
gnment2/assignment2//colab_requirements.txt (line 90))
  Downloading stack_data-0.6.2-py3-none-any.whl.metadata (18 kB)
Collecting terminado==0.17.1 (from -r /content/drive/My Drive/CS6353/Assignments/assi
gnment2/assignment2//colab_requirements.txt (line 91))
  Downloading terminado-0.17.1-py3-none-any.whl.metadata (5.9 kB)
Collecting tinycss2==1.2.1 (from -r /content/drive/My Drive/CS6353/Assignments/assign
ment2/assignment2//colab_requirements.txt (line 92))
  Downloading tinycss2-1.2.1-py3-none-any.whl.metadata (3.0 kB)
Collecting tornado<=6.3.2 (from -r /content/drive/My Drive/CS6353/Assignments/assignm
ent2/assignment2//colab requirements.txt (line 93))
  Downloading tornado-6.3.2-cp38-abi3-manylinux_2_5_x86_64.manylinux1_x86_64.manylinu
x_2_17_x86_64.manylinux2014_x86_64.whl.metadata (2.5 kB)
Collecting traitlets==5.9.0 (from -r /content/drive/My Drive/CS6353/Assignments/assig
nment2/assignment2//colab_requirements.txt (line 94))
  Downloading traitlets-5.9.0-py3-none-any.whl.metadata (10 kB)
Collecting tzdata==2023.3 (from -r /content/drive/My Drive/CS6353/Assignments/assignm
ent2/assignment2//colab_requirements.txt (line 95))
  Downloading tzdata-2023.3-py2.py3-none-any.whl.metadata (1.4 kB)
Collecting uri-template==1.3.0 (from -r /content/drive/My Drive/CS6353/Assignments/as
signment2/assignment2//colab_requirements.txt (line 96))
  Downloading uri_template-1.3.0-py3-none-any.whl.metadata (8.8 kB)
Collecting urllib3==2.0.4 (from -r /content/drive/My Drive/CS6353/Assignments/assignm
ent2/assignment2//colab_requirements.txt (line 97))
  Downloading urllib3-2.0.4-py3-none-any.whl.metadata (6.6 kB)
Collecting wcwidth==0.2.6 (from -r /content/drive/My Drive/CS6353/Assignments/assignm
ent2/assignment2//colab requirements.txt (line 98))
  Downloading wcwidth-0.2.6-py2.py3-none-any.whl.metadata (11 kB)
Collecting webcolors==1.13 (from -r /content/drive/My Drive/CS6353/Assignments/assign
ment2/assignment2//colab_requirements.txt (line 99))
  Downloading webcolors-1.13-py3-none-any.whl.metadata (2.6 kB)
Requirement already satisfied: webencodings==0.5.1 in /usr/local/lib/python3.10/dist-
packages (from -r /content/drive/My Drive/CS6353/Assignments/assignment2/assignment
2//colab_requirements.txt (line 100)) (0.5.1)
Collecting websocket-client==1.6.2 (from -r /content/drive/My Drive/CS6353/Assignment
s/assignment2/assignment2//colab_requirements.txt (line 101))
  Downloading websocket client-1.6.2-py3-none-any.whl.metadata (7.5 kB)
Requirement already satisfied: exceptiongroup in /usr/local/lib/python3.10/dist-packa
ges (from anyio==3.7.1->-r /content/drive/My Drive/CS6353/Assignments/assignment2/ass
ignment2//colab_requirements.txt (line 1)) (1.2.2)
Requirement already satisfied: typing-extensions>=4.0.0 in /usr/local/lib/python3.10/
dist-packages (from async-lru==2.0.4->-r /content/drive/My Drive/CS6353/Assignments/a
ssignment2/assignment2//colab_requirements.txt (line 7)) (4.12.2)
Collecting jupyter_client<8.0 (from -r /content/drive/My Drive/CS6353/Assignments/ass
ignment2/assignment2//colab_requirements.txt (line 39))
 Downloading jupyter client-7.4.9-py3-none-any.whl.metadata (8.5 kB)
Requirement already satisfied: tomli in /usr/local/lib/python3.10/dist-packages (from
jupyterlab==4.0.5->-r /content/drive/My Drive/CS6353/Assignments/assignment2/assignme
nt2//colab_requirements.txt (line 43)) (2.0.1)
Requirement already satisfied: ipython-genutils in /usr/local/lib/python3.10/dist-pac
kages (from ipykernel<=5.5.6->-r /content/drive/My Drive/CS6353/Assignments/assignmen
t2/assignment2//colab_requirements.txt (line 28)) (0.2.0)
Requirement already satisfied: setuptools>=18.5 in /usr/local/lib/python3.10/dist-pac
kages (from ipython<=7.34.0->-r /content/drive/My Drive/CS6353/Assignments/assignment
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2/assignment2//colab requirements.txt (line 29)) (71.0.4)
Requirement already satisfied: entrypoints in /usr/local/lib/python3.10/dist-packages
(from jupyter_client<8.0->-r /content/drive/My Drive/CS6353/Assignments/assignment2/a
ssignment2//colab requirements.txt (line 39)) (0.4)
Downloading appnope-0.1.3-py2.py3-none-any.whl (4.4 kB)
Downloading arrow-1.2.3-py3-none-any.whl (66 kB)
                                          - 66.4/66.4 kB 5.0 MB/s eta 0:00:00
Downloading asttokens-2.2.1-py2.py3-none-any.whl (26 kB)
Downloading async_lru-2.0.4-py3-none-any.whl (6.1 kB)
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Downloading Babel-2.12.1-py3-none-any.whl (10.1 MB)
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Downloading beautifulsoup4-4.12.2-py3-none-any.whl (142 kB)
                                          - 143.0/143.0 kB 10.3 MB/s eta 0:00:00
Downloading bleach-6.0.0-py3-none-any.whl (162 kB)
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Downloading certifi-2023.7.22-py3-none-any.whl (158 kB)
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Downloading cffi-1.15.1-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (4
41 kB)
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Downloading charset_normalizer-3.2.0-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_
x86_64.whl (201 kB)
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Downloading comm-0.1.4-py3-none-any.whl (6.6 kB)
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1 (300 kB)
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Downloading cycler-0.11.0-py3-none-any.whl (6.4 kB)
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4.whl (3.0 MB)
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hl (4.5 MB)
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Downloading fqdn-1.5.1-py3-none-any.whl (9.1 kB)
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Downloading json5-0.9.14-py2.py3-none-any.whl (19 kB)
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Downloading jsonschema specifications-2023.7.1-py3-none-any.whl (17 kB)
Downloading jupyter_events-0.7.0-py3-none-any.whl (18 kB)
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Downloading jupyter_core-5.3.1-py3-none-any.whl (93 kB)
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Downloading jupyter server terminals-0.4.4-py3-none-any.whl (13 kB)
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Downloading MarkupSafe-2.1.3-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.w
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Downloading overrides-7.4.0-py3-none-any.whl (17 kB)
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Downloading pexpect-4.8.0-py2.py3-none-any.whl (59 kB)
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Downloading Pillow-10.0.0-cp310-cp310-manylinux 2 28 x86 64.whl (3.4 MB)
                                        --- 3.4/3.4 MB 84.7 MB/s eta 0:00:00
Downloading platformdirs-3.10.0-py3-none-any.whl (17 kB)
Downloading prometheus_client-0.17.1-py3-none-any.whl (60 kB)
                                         - 60.6/60.6 kB 540.0 kB/s eta 0:00:00
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Downloading pure_eval-0.2.2-py3-none-any.whl (11 kB)
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Downloading python json logger-2.0.7-py3-none-any.whl (8.1 kB)
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Downloading tornado-6.3.2-cp38-abi3-manylinux_2_5_x86_64.manylinux1_x86_64.manylinux_
2 17 x86 64.manylinux2014 x86 64.whl (426 kB)
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Installing collected packages: wcwidth, pytz, pure-eval, json5, fastjsonschema, execu
ting, appnope, websocket-client, webcolors, urllib3, uri-template, tzdata, traitlets,
tornado, tinycss2, soupsieve, sniffio, Send2Trash, rpds-py, rfc3986-validator, rfc333
9-validator, PyYAML, python-json-logger, pyparsing, Pygments, pycparser, prompt-toolk
it, prometheus-client, platformdirs, Pillow, pexpect, parso, pandocfilters, packagin
g, overrides, numpy, nest-asyncio, mistune, MarkupSafe, kiwisolver, jupyterlab-pygmen
ts, jsonpointer, idna, fqdn, fonttools, debugpy, cycler, charset-normalizer, certifi,
bleach, Babel, attrs, async-lru, asttokens, terminado, stack-data, scipy, requests, r
eferencing, pandas, matplotlib-inline, jupyter core, Jinja2, jedi, imageio, contourp
y, comm, cffi, beautifulsoup4, arrow, matplotlib, jupyter_server_terminals, jupyter_c
lient, jsonschema-specifications, isoduration, seaborn, jsonschema, nbformat, nbclien
t, jupyter-events, nbconvert, jupyter_server, notebook_shim, jupyterlab_server, jupyt
er-lsp, jupyterlab
 Attempting uninstall: wcwidth
    Found existing installation: wcwidth 0.2.13
    Uninstalling wcwidth-0.2.13:
      Successfully uninstalled wcwidth-0.2.13
 Attempting uninstall: pytz
    Found existing installation: pytz 2024.2
   Uninstalling pytz-2024.2:
      Successfully uninstalled pytz-2024.2
 Attempting uninstall: fastjsonschema
    Found existing installation: fastjsonschema 2.20.0
   Uninstalling fastjsonschema-2.20.0:
      Successfully uninstalled fastjsonschema-2.20.0
 Attempting uninstall: websocket-client
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ERROR: pip's dependency resolver does not currently take into account all the package
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cts.
albucore 0.0.16 requires numpy>=1.24, but you have numpy 1.23.5 which is incompatibl
albumentations 1.4.15 requires numpy>=1.24.4, but you have numpy 1.23.5 which is inco
mpatible.
bigframes 1.18.0 requires numpy>=1.24.0, but you have numpy 1.23.5 which is incompati
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chex 0.1.86 requires numpy>=1.24.1, but you have numpy 1.23.5 which is incompatible.
cudf-cu12 24.4.1 requires pandas<2.2.2dev0,>=2.0, but you have pandas 1.5.3 which is
incompatible.
google-colab 1.0.0 requires pandas==2.1.4, but you have pandas 1.5.3 which is incompa
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pandas-stubs 2.1.4.231227 requires numpy>=1.26.0; python_version < "3.13", but you ha
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wcwidth-0.2.6 webcolors-1.13 websocket-client-1.6.2
```

s that are installed. This behaviour is the source of the following dependency confli

Multiclass Support Vector Machine exercise

Complete and hand in this completed worksheet (including its outputs and any supporting code outside of the worksheet) with your assignment submission. For more details see the assignments page on the course website.

In this exercise you will:

- implement a fully-vectorized loss function for the SVM
- implement the fully-vectorized expression for its **analytic gradient**
- check your implementation using numerical gradient
- use a validation set to tune the learning rate and regularization strength
- optimize the loss function with SGD
- visualize the final learned weights

```
In [ ]: # Run some setup code for this notebook.
        from __future__ import print_function
        import random
        import numpy as np
        from cs6353.data_utils import load_CIFAR10
        import matplotlib.pyplot as plt
        # This is a bit of magic to make matplotlib figures appear inline in the
        # notebook rather than in a new window.
        %matplotlib inline
        plt.rcParams['figure.figsize'] = (10.0, 8.0) # set default size of plots
        plt.rcParams['image.interpolation'] = 'nearest'
        plt.rcParams['image.cmap'] = 'gray'
        # Some more magic so that the notebook will reload external python modules;
        # see http://stackoverflow.com/questions/1907993/autoreload-of-modules-in-ipython
        %load ext autoreload
        %autoreload 2
```

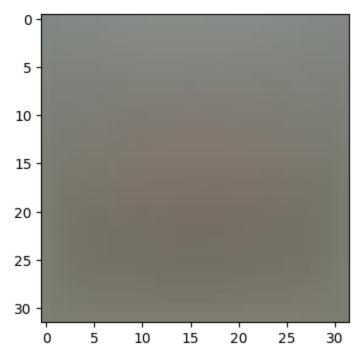
CIFAR-10 Data Loading and Preprocessing

```
In [ ]: # Load the raw CIFAR-10 data.
        cifar10_dir = 'cs6353/datasets/cifar-10-batches-py'
        # Cleaning up variables to prevent loading data multiple times (which may cause memory
        try:
           del X_train, y_train
           del X test, y test
           print('Clear previously loaded data.')
        except:
           pass
        X_train, y_train, X_test, y_test = load_CIFAR10(cifar10_dir)
        # As a sanity check, we print out the size of the training and test data.
        print('Training data shape: ', X_train.shape)
        print('Training labels shape: ', y_train.shape)
        print('Test data shape: ', X_test.shape)
        print('Test labels shape: ', y_test.shape)
        Training data shape: (50000, 32, 32, 3)
        Training labels shape: (50000,)
        Test data shape: (10000, 32, 32, 3)
        Test labels shape: (10000,)
```

```
In [ ]: # Visualize some examples from the dataset.
         # We show a few examples of training images from each class.
         classes = ['plane', 'car', 'bird', 'cat', 'deer', 'dog', 'frog', 'horse', 'ship', 'tru
         num classes = len(classes)
         samples_per_class = 7
         for y, cls in enumerate(classes):
             idxs = np.flatnonzero(y_train == y)
             idxs = np.random.choice(idxs, samples_per_class, replace=False)
             for i, idx in enumerate(idxs):
                 plt_idx = i * num_classes + y + 1
                 plt.subplot(samples_per_class, num_classes, plt_idx)
                 plt.imshow(X_train[idx].astype('uint8'))
                 plt.axis('off')
                 if i == 0:
                     plt.title(cls)
         plt.show()
          plane
                            bird
                                      cat
                                              deer
                                                       doa
                                                                froa
                                                                        horse
                                                                                 ship
                                                                                         truck
In [ ]: # Split the data into train, val, and test sets. In addition we will
         # create a small development set as a subset of the training data;
         # we can use this for development so our code runs faster.
         num_training = 49000
         num validation = 1000
         num_test = 1000
         num_dev = 500
         # Our validation set will be num_validation points from the original
         # training set.
```

```
mask = range(num training, num training + num validation)
        X_val = X_train[mask]
        y_val = y_train[mask]
        # Our training set will be the first num_train points from the original
        # training set.
        mask = range(num training)
        X_train = X_train[mask]
        y_train = y_train[mask]
        # We will also make a development set, which is a small subset of
        # the training set.
        mask = np.random.choice(num_training, num_dev, replace=False)
        X_{dev} = X_{train[mask]}
        y dev = y train[mask]
        # We use the first num_test points of the original test set as our
        # test set.
        mask = range(num test)
        X test = X test[mask]
        y_{\text{test}} = y_{\text{test}}[mask]
        print('Train data shape: ', X_train.shape)
        print('Train labels shape: ', y_train.shape)
                                        ', X_val.shape)
        print('Validation data shape:
        print('Validation labels shape: ', y_val.shape)
        print('Test data shape: ', X_test.shape)
        print('Test labels shape: ', y_test.shape)
        Train data shape: (49000, 32, 32, 3)
        Train labels shape: (49000,)
        Validation data shape: (1000, 32, 32, 3)
        Validation labels shape: (1000,)
        Test data shape: (1000, 32, 32, 3)
        Test labels shape: (1000,)
In [ ]: # Preprocessing: reshape the image data into rows
        X_train = np.reshape(X_train, (X_train.shape[0], -1))
        X_val = np.reshape(X_val, (X_val.shape[0], -1))
        X_test = np.reshape(X_test, (X_test.shape[0], -1))
        X_{dev} = np.reshape(X_{dev}, (X_{dev.shape}[0], -1))
        # As a sanity check, print out the shapes of the data
        print('Training data shape: ', X_train.shape)
        print('Validation data shape: ', X_val.shape)
        print('Test data shape: ', X_test.shape)
        print('dev data shape: ', X_dev.shape)
        Training data shape: (49000, 3072)
        Validation data shape: (1000, 3072)
        Test data shape: (1000, 3072)
        dev data shape: (500, 3072)
In [ ]: # Preprocessing: subtract the mean image
        # first: compute the image mean based on the training data
        mean_image = np.mean(X_train, axis=0)
        print(mean_image[:10]) # print a few of the elements
        plt.figure(figsize=(4,4))
        plt.imshow(mean image.reshape((32,32,3)).astype('uint8')) # visualize the mean image
        plt.show()
```

[130.64189796 135.98173469 132.47391837 130.05569388 135.34804082 131.75402041 130.96055102 136.14328571 132.47636735 131.48467347]



```
In []: # second: subtract the mean image from train and test data
    X_train -= mean_image
    X_val -= mean_image
    X_test -= mean_image
    X_dev -= mean_image

In []: # third: append the bias dimension of ones (i.e. bias trick) so that our SVM
    # only has to worry about optimizing a single weight matrix W.
    X_train = np.hstack([X_train, np.ones((X_train.shape[0], 1))])
    X_val = np.hstack([X_val, np.ones((X_val.shape[0], 1))])
    X_test = np.hstack([X_test, np.ones((X_test.shape[0], 1))])
    X_dev = np.hstack([X_dev, np.ones((X_dev.shape[0], 1))])
```

(49000, 3073) (1000, 3073) (1000, 3073) (500, 3073)

print(X_train.shape, X_val.shape, X_test.shape, X_dev.shape)

SVM Classifier

Your code for this section will all be written inside cs6353/classifiers/linear_svm.py.

As you can see, we have prefilled the function svm_loss_naive which uses for loops to evaluate the multiclass SVM loss function.

```
In [ ]: # Evaluate the naive implementation of the loss we provided for you:
    from cs6353.classifiers.linear_svm import svm_loss_naive
    import time

# generate a random SVM weight matrix of small numbers
W = np.random.randn(3073, 10) * 0.0001

loss, grad = svm_loss_naive(W, X_dev, y_dev, 0.000005)
```

The grad returned from the function above is right now all zero. Derive and implement the gradient for the SVM cost function and implement it inline inside the function svm_loss_naive. You will find it helpful to interleave your new code inside the existing function.

To check that you have correctly implemented the gradient correctly, you can numerically estimate the gradient of the loss function and compare the numeric estimate to the gradient that you computed. We have provided code that does this for you:

```
In []: # Once you've implemented the gradient, recompute it with the code below
# and gradient check it with the function we provided for you

# Compute the Loss and its gradient at W.
loss, grad = svm_loss_naive(W, X_dev, y_dev, 0.0)

# Numerically compute the gradient along several randomly chosen dimensions, and
# compare them with your analytically computed gradient. The numbers should match
# almost exactly along all dimensions.
from cs6353.gradient_check import grad_check_sparse
f = lambda w: svm_loss_naive(w, X_dev, y_dev, 0.0)[0]
grad_numerical = grad_check_sparse(f, W, grad)

# do the gradient check once again with regularization turned on
# you didn't forget the regularization gradient did you?
loss, grad = svm_loss_naive(W, X_dev, y_dev, 5e1)
f = lambda w: svm_loss_naive(w, X_dev, y_dev, 5e1)[0]
grad_numerical = grad_check_sparse(f, W, grad)
```

```
numerical: -1.836874 analytic: -1.836874, relative error: 4.933138e-11
numerical: 7.346344 analytic: 7.346344, relative error: 1.374043e-11
numerical: -1.583480 analytic: -1.572928, relative error: 3.343141e-03
numerical: 2.739409 analytic: 2.739409, relative error: 2.377995e-11
numerical: 9.991404 analytic: 9.991404, relative error: 5.010525e-12
numerical: -33.077543 analytic: -33.077543, relative error: 5.846195e-12
numerical: 8.632547 analytic: 8.669070, relative error: 2.110998e-03
numerical: 0.766813 analytic: 0.766813, relative error: 1.443983e-10
numerical: -38.545572 analytic: -38.627434, relative error: 1.060759e-03
numerical: -8.611617 analytic: -8.611617, relative error: 1.130704e-11
numerical: 9.394738 analytic: 9.394738, relative error: 7.961307e-12
numerical: -3.066593 analytic: -3.066593, relative error: 1.101293e-10
numerical: -13.585245 analytic: -13.585245, relative error: 5.498565e-12
numerical: 1.604586 analytic: 1.604586, relative error: 9.601810e-11
numerical: 6.820640 analytic: 6.820640, relative error: 6.083431e-11
numerical: 15.238316 analytic: 15.238316, relative error: 1.963482e-11
numerical: 10.441361 analytic: 10.441361, relative error: 3.038765e-11
numerical: 10.989256 analytic: 11.041811, relative error: 2.385489e-03
numerical: -28.241544 analytic: -28.289356, relative error: 8.457706e-04
numerical: -8.099178 analytic: -8.099178, relative error: 1.960946e-11
```

Inline Question 1:

It is possible that once in a while a dimension in the gradient check will not match exactly. What could such a discrepancy be caused by? Is it a reason for concern? What is a simple example in one dimension where a gradient check could fail? How would change the margin affect of the frequency of this happening? *Hint: the SVM loss function is not strictly speaking differentiable*

Your Answer: The mismatch in the dimension of the gradient check can occur due to the SVM loss function not being continously differentiable. The SVM loss function is defined as max(0, x), where x is the difference between the scores of incorrect and correct classes plus delta. If x < 0 the loss gets clipped to 0. Hence, at the Hinge x = 0, the loss function is undefined and hence we cannot differentiate. This can cause the discrepancy.

We can skip the gradient update step when this error occurs. Since the occurrence of such an error is rare, it is not an cause of concern.

A simple example where the gradient check will fail is: Modulus function |x|, at x = 0 the gradient check for this function will fail.

Consider the function max(0, x). If we increase the magin delta, the possibility of x being negative reduces, hence the possibility of loss being 0 reduces and hence the frequency of gradient check failing reduces.

```
In []: # Next implement the function svm_loss_vectorized; for now only compute the loss;
# we will implement the gradient in a moment.
tic = time.time()
loss_naive, grad_naive = svm_loss_naive(W, X_dev, y_dev, 0.000005)
toc = time.time()
print('Naive loss: %e computed in %fs' % (loss_naive, toc - tic))

from cs6353.classifiers.linear_svm import svm_loss_vectorized
tic = time.time()
```

```
loss_vectorized, _ = svm_loss_vectorized(W, X_dev, y_dev, 0.000005)
toc = time.time()
print('Vectorized loss: %e computed in %fs' % (loss_vectorized, toc - tic))

# The losses should match but your vectorized implementation should be much faster.
print('difference: %f' % (loss_naive - loss_vectorized))
Naive loss: % $207420100 computed in 0.0002365
```

Naive loss: 8.830742e+00 computed in 0.099236s Vectorized loss: 8.830742e+00 computed in 0.011614s difference: 0.000000

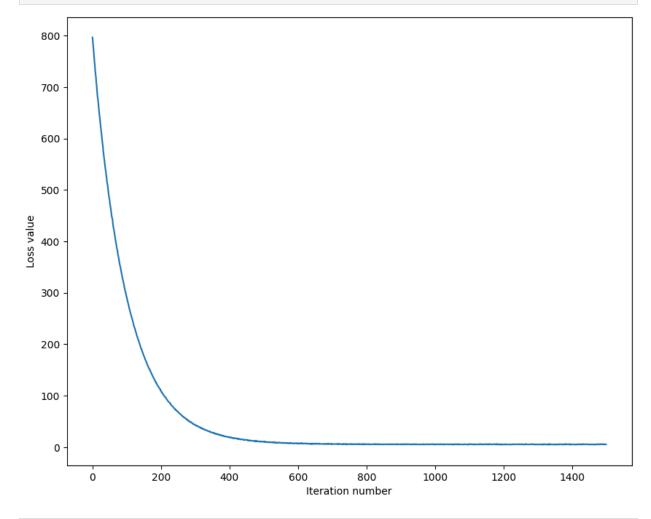
```
In [ ]: # Complete the implementation of svm_loss_vectorized, and compute the gradient
        # of the loss function in a vectorized way.
        # The naive implementation and the vectorized implementation should match, but
        # the vectorized version should still be much faster.
        tic = time.time()
        _, grad_naive = svm_loss_naive(W, X_dev, y_dev, 0.000005)
        toc = time.time()
        print('Naive loss and gradient: computed in %fs' % (toc - tic))
        tic = time.time()
        _, grad_vectorized = svm_loss_vectorized(W, X_dev, y dev, 0.000005)
        toc = time.time()
        print('Vectorized loss and gradient: computed in %fs' % (toc - tic))
        # The loss is a single number, so it is easy to compare the values computed
        # by the two implementations. The gradient on the other hand is a matrix, so
        # we use the Frobenius norm to compare them.
        difference = np.linalg.norm(grad_naive - grad_vectorized, ord='fro')
        print('difference: %f' % difference)
```

Naive loss and gradient: computed in 0.091855s Vectorized loss and gradient: computed in 0.009065s difference: 0.000000

Stochastic Gradient Descent

We now have vectorized and efficient expressions for the loss, the gradient and our gradient matches the numerical gradient. We are therefore ready to do SGD to minimize the loss. Your code for this part will be written inside cs6353/classifiers/linear_classifier.py.

```
iteration 0 / 1500: loss 796.412384
iteration 100 / 1500: loss 290.687019
iteration 200 / 1500: loss 108.979117
iteration 300 / 1500: loss 42.586333
iteration 400 / 1500: loss 18.861781
iteration 500 / 1500: loss 10.370017
iteration 600 / 1500: loss 6.483616
iteration 700 / 1500: loss 5.515384
iteration 800 / 1500: loss 5.363760
iteration 900 / 1500: loss 5.640452
iteration 1000 / 1500: loss 4.680338
iteration 1100 / 1500: loss 5.723385
iteration 1200 / 1500: loss 5.408608
iteration 1300 / 1500: loss 4.616177
iteration 1400 / 1500: loss 4.930543
That took 9.644955s
```



9/29/24, 9:41 PM sv y val pred = svm.predict(X val)

```
print('validation accuracy: %f' % (np.mean(y_val == y_val_pred), ))
       training accuracy: 0.371286
       validation accuracy: 0.376000
In [ ]: # Use the validation set to tune hyperparameters (regularization strength and
       # learning rate). You should experiment with different ranges for the learning
       # rates and regularization strengths; if you are careful you should be able to
       # get a classification accuracy of about 0.4 on the validation set.
       learning_rates = [1e-7, 2e-7, 5e-6, 5e-5]
       regularization_strengths = [5e3, 2.5e4, 5e4, 1e5]
       # results is dictionary mapping tuples of the form
       # (learning_rate, regularization_strength) to tuples of the form
       # (training accuracy, validation accuracy). The accuracy is simply the fraction
       # of data points that are correctly classified.
       results = {}
       best val = -1 # The highest validation accuracy that we have seen so far.
       best_svm = None # The LinearSVM object that achieved the highest validation rate.
       # TODO:
       # Write code that chooses the best hyperparameters by tuning on the validation #
       # set. For each combination of hyperparameters, train a linear SVM on the
       # training set, compute its accuracy on the training and validation sets, and #
       # store these numbers in the results dictionary. In addition, store the best
       # validation accuracy in best val and the LinearSVM object that achieves this #
       # accuracy in best svm.
                                                                            #
       # Hint: You should use a small value for num_iters as you develop your
       # validation code so that the SVMs don't take much time to train; once you are #
       # confident that your validation code works, you should rerun the validation
                                                                            #
       # code with a larger value for num iters.
       for lr in learning_rates:
           for reg in regularization strengths:
              svm = LinearSVM()
              svm.train(X_train, y_train, learning_rate=lr, reg=reg, num_iters=1500, verbose
              y_train_pred = svm.predict(X_train)
              y_val_pred = svm.predict(X_val)
              train_results = y_train_pred == y_train
              y_train_accuracy = np.mean(train_results)
              val_results = y_val_pred == y_val
              y_val_accuracy = np.mean(val_results)
              results[(lr, reg)] = (y_train_accuracy, y_val_accuracy)
              if y_val_accuracy > best_val:
                  best_val = y_val_accuracy
                  best svm = svm
       END OF YOUR CODE
```

iteration 0 / 1500: loss 178.244227 iteration 100 / 1500: loss 134.383499 iteration 200 / 1500: loss 108.965742 iteration 300 / 1500: loss 89.827161 iteration 400 / 1500: loss 73.827566 iteration 500 / 1500: loss 60.892688 iteration 600 / 1500: loss 50.312921 iteration 700 / 1500: loss 41.541130 iteration 800 / 1500: loss 35.354586 iteration 900 / 1500: loss 28.906567 iteration 1000 / 1500: loss 24.374831 iteration 1100 / 1500: loss 21.776227 iteration 1200 / 1500: loss 17.810968 iteration 1300 / 1500: loss 15.919390 iteration 1400 / 1500: loss 13.630636 iteration 0 / 1500: loss 791.273070 iteration 100 / 1500: loss 289.199177 iteration 200 / 1500: loss 108.022782 iteration 300 / 1500: loss 42.459306 iteration 400 / 1500: loss 18.652796 iteration 500 / 1500: loss 10.216656 iteration 600 / 1500: loss 7.226874 iteration 700 / 1500: loss 6.247032 iteration 800 / 1500: loss 5.329904 iteration 900 / 1500: loss 6.004492 iteration 1000 / 1500: loss 5.448198 iteration 1100 / 1500: loss 5.389966 iteration 1200 / 1500: loss 5.219112 iteration 1300 / 1500: loss 5.644886 iteration 1400 / 1500: loss 4.886158 iteration 0 / 1500: loss 1563.660630 iteration 100 / 1500: loss 210.325976 iteration 200 / 1500: loss 32.064180 iteration 300 / 1500: loss 8.797855 iteration 400 / 1500: loss 6.379050 iteration 500 / 1500: loss 5.811982 iteration 600 / 1500: loss 5.793851 iteration 700 / 1500: loss 5.642848 iteration 800 / 1500: loss 5.202007 iteration 900 / 1500: loss 4.934217 iteration 1000 / 1500: loss 5.627553 iteration 1100 / 1500: loss 5.560569 iteration 1200 / 1500: loss 5.297981 iteration 1300 / 1500: loss 5.244115 iteration 1400 / 1500: loss 5.524269 iteration 0 / 1500: loss 3073.123079 iteration 100 / 1500: loss 58.998161 iteration 200 / 1500: loss 7.132229 iteration 300 / 1500: loss 5.966786 iteration 400 / 1500: loss 6.069582 iteration 500 / 1500: loss 6.156074 iteration 600 / 1500: loss 6.090922 iteration 700 / 1500: loss 5.830623 iteration 800 / 1500: loss 6.569839 iteration 900 / 1500: loss 6.384720 iteration 1000 / 1500: loss 6.130254 iteration 1100 / 1500: loss 5.692147 iteration 1200 / 1500: loss 6.943978 iteration 1300 / 1500: loss 5.897260 iteration 1400 / 1500: loss 6.606903

iteration 0 / 1500: loss 172.598569 iteration 100 / 1500: loss 110.170134 iteration 200 / 1500: loss 75.924439 iteration 300 / 1500: loss 50.918345 iteration 400 / 1500: loss 35.379410 iteration 500 / 1500: loss 25.040118 iteration 600 / 1500: loss 19.052712 iteration 700 / 1500: loss 14.101669 iteration 800 / 1500: loss 10.694302 iteration 900 / 1500: loss 9.092919 iteration 1000 / 1500: loss 7.707675 iteration 1100 / 1500: loss 6.722602 iteration 1200 / 1500: loss 6.305849 iteration 1300 / 1500: loss 5.275060 iteration 1400 / 1500: loss 5.666545 iteration 0 / 1500: loss 783.232515 iteration 100 / 1500: loss 106.819093 iteration 200 / 1500: loss 18.907263 iteration 300 / 1500: loss 7.052838 iteration 400 / 1500: loss 5.505619 iteration 500 / 1500: loss 5.517656 iteration 600 / 1500: loss 5.581646 iteration 700 / 1500: loss 4.754594 iteration 800 / 1500: loss 5.220277 iteration 900 / 1500: loss 5.582272 iteration 1000 / 1500: loss 5.370475 iteration 1100 / 1500: loss 5.067590 iteration 1200 / 1500: loss 5.661360 iteration 1300 / 1500: loss 5.349802 iteration 1400 / 1500: loss 5.402219 iteration 0 / 1500: loss 1562.243201 iteration 100 / 1500: loss 32.191925 iteration 200 / 1500: loss 6.885046 iteration 300 / 1500: loss 5.602807 iteration 400 / 1500: loss 6.131675 iteration 500 / 1500: loss 5.640953 iteration 600 / 1500: loss 5.735482 iteration 700 / 1500: loss 5.928857 iteration 800 / 1500: loss 5.743055 iteration 900 / 1500: loss 5.531828 iteration 1000 / 1500: loss 5.203421 iteration 1100 / 1500: loss 6.086186 iteration 1200 / 1500: loss 5.902380 iteration 1300 / 1500: loss 5.302349 iteration 1400 / 1500: loss 5.934050 iteration 0 / 1500: loss 3098.534357 iteration 100 / 1500: loss 7.130491 iteration 200 / 1500: loss 6.094228 iteration 300 / 1500: loss 6.065124 iteration 400 / 1500: loss 6.282524 iteration 500 / 1500: loss 6.191154 iteration 600 / 1500: loss 5.976392 iteration 700 / 1500: loss 5.975025 iteration 800 / 1500: loss 5.645989 iteration 900 / 1500: loss 5.807639 iteration 1000 / 1500: loss 5.936774 iteration 1100 / 1500: loss 5.835507 iteration 1200 / 1500: loss 5.958050 iteration 1300 / 1500: loss 6.819106 iteration 1400 / 1500: loss 5.669680

iteration 0 / 1500: loss 173.198183 iteration 100 / 1500: loss 20.853773 iteration 200 / 1500: loss 18.109174 iteration 300 / 1500: loss 15.733888 iteration 400 / 1500: loss 13.055410 iteration 500 / 1500: loss 13.854221 iteration 600 / 1500: loss 10.420108 iteration 700 / 1500: loss 14.229794 iteration 800 / 1500: loss 13.924428 iteration 900 / 1500: loss 15.281499 iteration 1000 / 1500: loss 16.875358 iteration 1100 / 1500: loss 13.741402 iteration 1200 / 1500: loss 23.626028 iteration 1300 / 1500: loss 13.205194 iteration 1400 / 1500: loss 18.871571 iteration 0 / 1500: loss 779.975295 iteration 100 / 1500: loss 23.130887 iteration 200 / 1500: loss 33.690605 iteration 300 / 1500: loss 28.841234 iteration 400 / 1500: loss 23.207378 iteration 500 / 1500: loss 26.508271 iteration 600 / 1500: loss 21.876520 iteration 700 / 1500: loss 21.939206 iteration 800 / 1500: loss 25.317959 iteration 900 / 1500: loss 20.630530 iteration 1000 / 1500: loss 17.127213 iteration 1100 / 1500: loss 24.405165 iteration 1200 / 1500: loss 22.283576 iteration 1300 / 1500: loss 19.566886 iteration 1400 / 1500: loss 23.296308 iteration 0 / 1500: loss 1559.950278 iteration 100 / 1500: loss 35.393304 iteration 200 / 1500: loss 30.696443 iteration 300 / 1500: loss 33.169852 iteration 400 / 1500: loss 33.120913 iteration 500 / 1500: loss 33.478191 iteration 600 / 1500: loss 30.326305 iteration 700 / 1500: loss 33.692889 iteration 800 / 1500: loss 21.271244 iteration 900 / 1500: loss 28.265025 iteration 1000 / 1500: loss 33.888470 iteration 1100 / 1500: loss 32.588642 iteration 1200 / 1500: loss 29.961906 iteration 1300 / 1500: loss 28.233321 iteration 1400 / 1500: loss 32.649200 iteration 0 / 1500: loss 3106.209593 iteration 100 / 1500: loss 57.929791 iteration 200 / 1500: loss 74.851623 iteration 300 / 1500: loss 74.540410 iteration 400 / 1500: loss 68.043354 iteration 500 / 1500: loss 69.429617 iteration 600 / 1500: loss 63.305850 iteration 700 / 1500: loss 69.456483 iteration 800 / 1500: loss 57.876450 iteration 900 / 1500: loss 80.029685 iteration 1000 / 1500: loss 79.673372 iteration 1100 / 1500: loss 62.802481 iteration 1200 / 1500: loss 50.718919 iteration 1300 / 1500: loss 84.401221 iteration 1400 / 1500: loss 71.439129

iteration 0 / 1500: loss 175.468803 iteration 100 / 1500: loss 357.794190

```
iteration 200 / 1500: loss 261.610384
iteration 300 / 1500: loss 323.961122
iteration 400 / 1500: loss 236.804339
iteration 500 / 1500: loss 292.767227
iteration 600 / 1500: loss 338.246575
iteration 700 / 1500: loss 226.032447
iteration 800 / 1500: loss 299.913841
iteration 900 / 1500: loss 254.660160
iteration 1000 / 1500: loss 319.197245
iteration 1100 / 1500: loss 337.491635
iteration 1200 / 1500: loss 316.946059
iteration 1300 / 1500: loss 255.065476
iteration 1400 / 1500: loss 273.845585
iteration 0 / 1500: loss 794.020799
iteration 100 / 1500: loss 394760373444917569637606845452629573632.000000
iteration 200 / 1500: loss 6525072812573031292100673454664614464879456863556933790485
4949036761808896.000000
iteration 300 / 1500: loss 1078542277124495292016946587597207966718228388067316388335
3989456808185081499303733577127478046744740840341504.000000
iteration 400 / 1500: loss 1782744004485959901551228896455347888968974849861193021152
1987021846815679408570247155771735372325175882804807919132831712186340592330991720202
24.000000
iteration 500 / 1500: loss 2946733060853192242890543862612089291321853746624695376106
2412609312620855504400711067216409436130763183441465276364765271565317015346896412717
6997496598581885987809344378900578304.000000
iteration 600 / 1500: loss 4870713748062200503837272613858407099279604761114810011338
8087691493879638662815440760993692773055000442162218708269072306470479108164989868720
498729444972260726834281796386995680700894182750610075422519243279695872.000000
iteration 700 / 1500: loss 8050899733921997218766457257436512787846768561744804495639
8125075733144238517868823952847209756578607289500952999710954857546938719016050737394
8915980234832608473572482149694748882434308222412089708792544831739724783236841706512
9093766265386819387392.000000
iteration 800 / 1500: loss 1330749246995936717245578739746150383060986894711340920334
2789863200164643954264316571324759578047980950140023201099722823605441998790483201086
2589685600589912131414892180764058785897937613362917978810498622127545132469845246026
4737245018530225430950569430595262634345460622832523280384.000000
/content/drive/MyDrive/CS6353/Assignments/assignment2/assignment2/cs6353/classifiers/
linear svm.py:100: RuntimeWarning: overflow encountered in scalar multiply
 loss += reg * np.sum(W * W)
/usr/local/lib/python3.10/dist-packages/numpy/core/fromnumeric.py:88: RuntimeWarning:
overflow encountered in reduce
/content/drive/MyDrive/CS6353/Assignments/assignment2/assignment2/cs6353/classifiers/
linear_svm.py:100: RuntimeWarning: overflow encountered in multiply
  loss += reg * np.sum(W * W)
```

```
iteration 900 / 1500: loss inf
iteration 1000 / 1500: loss inf
iteration 1100 / 1500: loss inf
iteration 1200 / 1500: loss inf
iteration 1300 / 1500: loss inf
iteration 1400 / 1500: loss inf
iteration 0 / 1500: loss 1558.822219
iteration 100 / 1500: loss 4207798039246850533175537350026849985891957311066635763731
421068684681551426265889564213148587107320794663181729815999807488.000000
iteration 200 / 1500: loss 1086558597385951279432465208308215137742869207865414633594
6470231097929239192116128043930123319357484907821011382111736380297078929685288281354\\
4440572810648554778182559683387244449590057481321689025821234523991051138958670226822
38881250555199488.000000
iteration 300 / 1500: loss inf
iteration 400 / 1500: loss inf
iteration 500 / 1500: loss inf
/content/drive/MyDrive/CS6353/Assignments/assignment2/assignment2/cs6353/classifiers/
linear svm.py:125: RuntimeWarning: overflow encountered in multiply
  dW += 2 * reg * W
/content/drive/MyDrive/CS6353/Assignments/assignment2/assignment2/cs6353/classifiers/
linear classifier.py:70: RuntimeWarning: invalid value encountered in subtract
  self.W -= learning_rate * grad
iteration 600 / 1500: loss nan
iteration 700 / 1500: loss nan
iteration 800 / 1500: loss nan
iteration 900 / 1500: loss nan
iteration 1000 / 1500: loss nan
iteration 1100 / 1500: loss nan
iteration 1200 / 1500: loss nan
iteration 1300 / 1500: loss nan
iteration 1400 / 1500: loss nan
iteration 0 / 1500: loss 3071.075695
iteration 100 / 1500: loss 2182133309946366599253484069681885190785452080302820461479
7202861682346491760762762456234455162260005124946570008000177952981352557009769576134
2850588122872372114047619043797859518559655959724032.000000
iteration 200 / 1500: loss inf
iteration 300 / 1500: loss inf
/content/drive/MyDrive/CS6353/Assignments/assignment2/assignment2/cs6353/classifiers/
linear_svm.py:94: RuntimeWarning: invalid value encountered in subtract
 margins = np.maximum(0, scores - correct_class_scores + 1)
```

9/29/24, 9:41 PM svm iteration 400 / 1500: loss nan

iteration 500 / 1500: loss nan

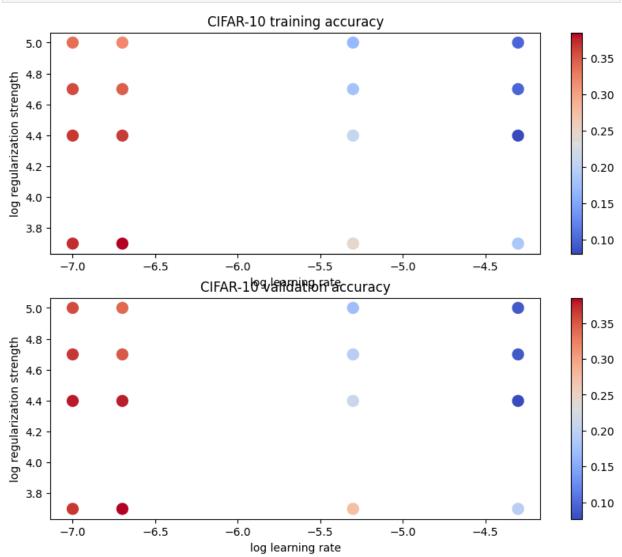
```
iteration 600 / 1500: loss nan
        iteration 700 / 1500: loss nan
        iteration 800 / 1500: loss nan
        iteration 900 / 1500: loss nan
        iteration 1000 / 1500: loss nan
        iteration 1100 / 1500: loss nan
        iteration 1200 / 1500: loss nan
        iteration 1300 / 1500: loss nan
        iteration 1400 / 1500: loss nan
        lr 1.000000e-07 reg 5.000000e+03 train accuracy: 0.372122 val accuracy: 0.367000
        lr 1.000000e-07 reg 2.500000e+04 train accuracy: 0.367551 val accuracy: 0.378000
        lr 1.000000e-07 reg 5.000000e+04 train accuracy: 0.356694 val accuracy: 0.366000
        lr 1.000000e-07 reg 1.000000e+05 train accuracy: 0.335776 val accuracy: 0.357000
        lr 2.000000e-07 reg 5.000000e+03 train accuracy: 0.384673 val accuracy: 0.385000
        lr 2.000000e-07 reg 2.500000e+04 train accuracy: 0.364837 val accuracy: 0.377000
        lr 2.000000e-07 reg 5.000000e+04 train accuracy: 0.344490 val accuracy: 0.350000
        lr 2.000000e-07 reg 1.000000e+05 train accuracy: 0.318592 val accuracy: 0.341000
        lr 5.000000e-06 reg 5.000000e+03 train accuracy: 0.245184 val accuracy: 0.274000
        lr 5.000000e-06 reg 2.500000e+04 train accuracy: 0.209959 val accuracy: 0.211000
        lr 5.000000e-06 reg 5.000000e+04 train accuracy: 0.177551 val accuracy: 0.193000
        lr 5.000000e-06 reg 1.000000e+05 train accuracy: 0.167510 val accuracy: 0.174000
        lr 5.000000e-05 reg 5.000000e+03 train accuracy: 0.183265 val accuracy: 0.194000
        lr 5.000000e-05 reg 2.500000e+04 train accuracy: 0.081184 val accuracy: 0.077000
        lr 5.000000e-05 reg 5.000000e+04 train accuracy: 0.100265 val accuracy: 0.087000
        lr 5.000000e-05 reg 1.000000e+05 train accuracy: 0.100265 val accuracy: 0.087000
        best validation accuracy achieved during cross-validation: 0.385000
In [ ]: # Visualize the cross-validation results
        import math
        x_scatter = [math.log10(x[0]) for x in results]
        y scatter = [math.log10(x[1]) for x in results]
        # plot training accuracy
        marker size = 100
        colors = [results[x][0] for x in results]
        plt.subplot(2, 1, 1)
        # plt.scatter(x_scatter, y_scatter, marker_size, c=colors)
        plt.scatter(x_scatter, y_scatter, marker_size, c=colors, cmap=plt.cm.coolwarm)
        plt.colorbar()
        plt.xlabel('log learning rate')
        plt.ylabel('log regularization strength')
        plt.title('CIFAR-10 training accuracy')
        # plot validation accuracy
        colors = [results[x][1] for x in results] # default size of markers is 20
        plt.subplot(2, 1, 2)
        # plt.scatter(x_scatter, y_scatter, marker_size, c=colors)
        plt.scatter(x_scatter, y_scatter, marker_size, c=colors, cmap=plt.cm.coolwarm)
```

plt.colorbar()

plt.xlabel('log learning rate')

plt.ylabel('log regularization strength')

```
plt.title('CIFAR-10 validation accuracy')
plt.show()
```

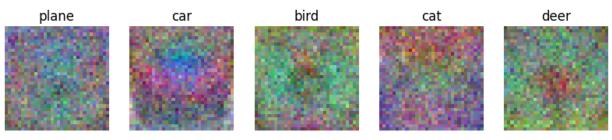


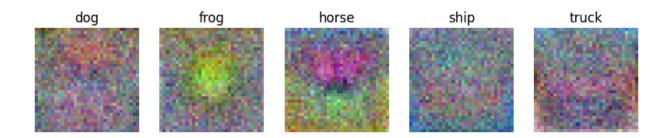
```
In [ ]: # Evaluate the best svm on test set
    y_test_pred = best_svm.predict(X_test)
    test_accuracy = np.mean(y_test == y_test_pred)
    print('linear SVM on raw pixels final test set accuracy: %f' % test_accuracy)
```

linear SVM on raw pixels final test set accuracy: 0.379000

```
In []: # Visualize the learned weights for each class.
# Depending on your choice of learning rate and regularization strength, these may
# or may not be nice to look at.
w = best_svm.W[:-1,:] # strip out the bias
w = w.reshape(32, 32, 3, 10)
w_min, w_max = np.min(w), np.max(w)
classes = ['plane', 'car', 'bird', 'cat', 'deer', 'dog', 'frog', 'horse', 'ship', 'tru
for i in range(10):
    plt.subplot(2, 5, i + 1)

# Rescale the weights to be between 0 and 255
wimg = 255.0 * (w[:, :, :, i].squeeze() - w_min) / (w_max - w_min)
plt.imshow(wimg.astype('uint8'))
plt.axis('off')
plt.title(classes[i])
```





Inline question 2:

Describe what your visualized SVM weights look like, and offer a brief explanation for why they look the way that they do.

Your answer: The visualized SVM weights act as templates for each class, learned from the patterns in the training data. These weights represent the core features that define the images belonging to each class. When classifying a test image, SVM computes the inner product between the test image and each class's template. This inner product measures how closely the test image aligns with the learned features of each class, helping determine which class the image most likely belongs to. For instance, an frog is usually green in colour, hence the image shows a big green circle in the centre.