NOTE: The bez2018model Python auditory nerve model wrapper must be installed within the Singularity environment: cd /packages/bez2018model python setup.py build ext --inplace 1. Evaluating pre-trained pitch network models on new stimuli In [1]: import os import sys import json import time import IPython.display as ipd import numpy as np import tensorflow as tf %matplotlib inline import matplotlib.pyplot as plt # Import Python wrapper for Bruce et al. (2018) auditory nerve model sys.path.append('/packages/bez2018model/') import bez2018model # Import functions for generating figures and audio stimuli sys.path.append('/packages/msutil/') import util\_figures import util\_stimuli # Import functions for deep neural network models sys.path.append('/packages/tfutil/') import functions\_brain\_network import functions\_graph\_assembly # Import F0 classification bin functions from dataset\_util sys.path.append('./assets datasets/') from dataset util import get f0 bins, f0 to label, label to f0 Generate sound waveform Generate audio waveform to test model(s) on. Default settings create a sine-phase harmonic complex tone with 200 Hz F0 (150 ms duration and 32 kHz sampling rate). The tone is embedded in modified uniform masking noise (0 dB SNR) and the stimulus is re-scaled to 60 dB SPL. In [2]: np.random.seed(0) signal fs = 32000signal = util stimuli.complex tone( f0=200, fs=signal\_fs, dur=0.150, harmonic\_numbers=np.arange(1, 31), frequencies=None, amplitudes=None, phase\_mode='sine', offset\_start=True, strict nyquist=True) noise = util\_stimuli.modified\_uniform\_masking\_noise( fs=signal\_fs, dur=0.150, dBHzSPL=15.0, attenuation\_start=600.0, attenuation slope=2.0) signal = util\_stimuli.combine\_signal\_and\_noise(signal, noise, snr=0.0) signal = util\_stimuli.set\_dBSPL(signal, 60.0) ipd.display(ipd.Audio(signal, rate=signal\_fs)) Generate simulated auditory nerve representation Default parameters correspond to the standard auditory nerve model (most human-like). In [3]: kwargs nervegram = { 'nervegram\_dur': 0.050, # auditory nerve representation is clipped to 50 ms 'nervegram fs': 20e3, # auditory nerve representation is sampled at 20 kHz 'buffer\_start\_dur': 0.070, 'buffer\_end\_dur': 0.010, 'pin fs': 100e3, pin\_dBSPL\_flag': 0, 'pin dBSPL': None, 'species': 2, 'bandwidth scale factor': 1.0, # scales bandwidths of cochlear filters 'cf list': None, 'num cf': 100, # simulate 100 auditory nerve fibers with CFs from 125 to 14000 Hz 'min cf': 125.0, 'max cf': 14e3, 'synapseMode': 0, 'max spikes per train': -1, 'num\_spike\_trains': 40, 'cohc': 1.0, 'cihc': 1.0, 'IhcLowPass\_cutoff': 3e3, # IHC lowpass filter cutoff in Hz (default 3000 Hz) 'IhcLowPass order': 7, 'spont': 70.0, 'noiseType': 1, 'implnt': 0, 'tabs': 6e-4, 'trel': 6e-4, 'random seed': None, 'return\_vihcs': False, 'return\_meanrates': True, # Pitch DNN models expect `meanrates` nervegram input 'return spike times': False, 'return spike tensor sparse': False, 'return spike tensor dense': False, 'nervegram spike tensor fs': 100e3, 'squeeze channel dim': True t0 = time.time() nervegram\_output\_dict = bez2018model.nervegram(signal, signal\_fs, \*\*kwargs\_nervegram) t1 = time.time() print("time to generate bez2018model nervegram: {:.2f} seconds".format(t1-t0)) print("bez2018model nervegram shape: {}".format(nervegram\_output\_dict['nervegram\_meanrates'].shape)) time to generate bez2018model nervegram: 3.31 seconds bez2018model nervegram shape: (100, 1000) Visualize simulated auditory nerve representation The sound waveform is plotted at the top. The stimulus power spectrum (with ERB-scaled frequency axis) is plotted to the left. The grey-scale image plots the array of instantaneous auditory nerve firing rates (i.e., the nervegram\_meanrates). Lighter colors indicate more spiking. The time-averaged firing rates (the excitation pattern) are plotted to the right. In [4]: fig, ax\_arr = plt.subplots( nrows=2, ncols=3, figsize=(6, 3.5),gridspec\_kw={'wspace': 0.15, 'hspace': 0.15, 'width\_ratios': [1,6,1], 'height\_ratios': [1,4]}) util\_figures.make\_stimulus\_summary\_plot( ax arr, ax\_idx\_waveform=1, ax\_idx\_spectrum=3, ax idx nervegram=4, ax\_idx\_excitation=5, waveform=nervegram output \_dict['signal'], nervegram=nervegram\_output\_dict['nervegram\_meanrates'], sr\_waveform=nervegram\_output\_dict['signal\_fs'], sr\_nervegram=nervegram\_output\_dict['nervegram\_fs'], cfs=nervegram output dict['cf list'], tmin=None, tmax=None, treset=True, vmin=None, vmax=None, erb freq axis=True, spines\_to\_hide\_waveform=[], spines\_to\_hide\_spectrum=[], spines\_to\_hide\_excitation=[], nxticks=6, nyticks=6, kwargs plot={}, limits\_buffer=0.2, ax arr clear leftover=True) plt.show() 14000 3072 -1361 537 140 20 200 55 0 10 30 50 0 Power Time Excitation (dB SPL) (ms) (spikes/s) Build pitch model deep neural network(s) and evaluate them on the stimulus In [5]: # Specify list of model directories to run (one directory per network) # NOTE: To evaluate only one network, simply comment out the other 9 list dir model = [ 'models/default/arch\_0083', 'models/default/arch 0154', 'models/default/arch 0190', 'models/default/arch\_0191', 'models/default/arch\_0286', 'models/default/arch 0288', 'models/default/arch 0302', 'models/default/arch\_0335', 'models/default/arch 0338', 'models/default/arch 0346', basename arch = 'brain arch.json' # Model architecture file within directory basename\_ckpt = 'brain\_model.ckpt' # Model checkpoint file within directory # Construct a tensorflow graph containing all specified networks tf.reset default graph() tf.get\_logger().setLevel('ERROR') # Suppress tensorflow deprecation warnings tf.random.set\_random\_seed(0) # Input nervegrams are provided via a placeholder (networks expect shape [100, 1000]) placeholder nervegram = tf.placeholder(tf.float32, shape=[100, 1000]) tensor\_nervegram = placeholder\_nervegram[tf.newaxis, : , :, tf.newaxis] # Iterate over model directories and build DNN graphs dict\_model\_vars = {} # Dictionary of variables for each network dict\_model\_softmax = {} # Dictionary of softmax tensors for each network for dir model in list dir model: tensor\_logits, tensor\_dict = functions\_graph\_assembly.build\_brain\_graph( tensor\_nervegram, N\_CLASSES\_DICT={'f0\_label': 700}, config=os.path.join(dir\_model, basename\_arch), batchnorm\_flag=False, dropout flag=True, save\_pckl\_path=None, save\_arch\_path=None, trainable=False, only\_include\_layers=None, var scope name=dir model, var scope reuse=False) # Collect variables for current network var\_list = tf.get\_collection(tf.GraphKeys.GLOBAL\_VARIABLES, scope=dir\_model) dict model vars[dir model] = { v.name[:-2].replace(dir\_model, 'brain\_network'): v for v in var\_list # Convert current network outputs to F0 class probabilities dict model softmax[dir model] = tf.nn.softmax(tensor logits['f0 label']) t0 = time.time() with tf.Session() as sess: sess.run(tf.global\_variables\_initializer()) # Iterate over model directories and load network weights from checkpoints for dir model in list dir model: functions graph assembly build saver ( sess, dict\_model\_vars[dir\_model], dir model, restore\_model\_path=None, ckpt prefix name=basename ckpt, attempt load=True) t1 = time.time() print('time to initialize and load model(s): {:.2f} seconds'.format(t1-t0)) # Evaluate the networks on `nervegram\_output\_dict['nervegram\_meanrates']` t0 = time.time() feed dict = {placeholder nervegram: nervegram output dict['nervegram meanrates']} dict\_model\_softmax\_output = sess.run(dict\_model\_softmax, feed\_dict=feed\_dict) t1 = time.time() print('time to run model(s): {:.2f} seconds'.format(t1-t0)) print('network output shapes (F0 class probabilities)') for key in sorted(dict\_model\_softmax\_output.keys()): print(' {}: {}'.format(key, dict model softmax output[key].shape)) Loading brain network config from models/default/arch 0083/brain arch.json Loading brain network config from models/default/arch\_0154/brain\_arch.json Loading brain network config from models/default/arch\_0190/brain\_arch.json Loading brain network config from models/default/arch 0191/brain arch.json Loading brain network config from models/default/arch\_0286/brain\_arch.json Loading brain network config from models/default/arch\_0288/brain\_arch.json Loading brain network config from models/default/arch 0302/brain arch.json Loading brain network config from models/default/arch\_0335/brain\_arch.json Loading brain network config from models/default/arch\_0338/brain\_arch.json Loading brain network config from models/default/arch 0346/brain arch.json ### Loading variables from latest checkpoint: models/default/arch\_0083/brain\_model.ckpt-70000 ### Loading variables from latest checkpoint: models/default/arch\_0154/brain\_model.ckpt-55000 ### Loading variables from latest checkpoint: models/default/arch 0190/brain model.ckpt-45000 ### Loading variables from latest checkpoint: models/default/arch\_0191/brain\_model.ckpt-60000 ### Loading variables from latest checkpoint: models/default/arch\_0286/brain\_model.ckpt-60000 ### Loading variables from latest checkpoint: models/default/arch 0288/brain model.ckpt-45000 ### Loading variables from latest checkpoint: models/default/arch\_0302/brain\_model.ckpt-50000 ### Loading variables from latest checkpoint: models/default/arch\_0335/brain\_model.ckpt-60000 ### Loading variables from latest checkpoint: models/default/arch 0338/brain model.ckpt-50000 ### Loading variables from latest checkpoint: models/default/arch\_0346/brain\_model.ckpt-55000 time to initialize and load model(s): 5.56 seconds time to run model(s): 5.70 seconds network output shapes (F0 class probabilities) models/default/arch\_0083: (1, 700) models/default/arch 0154: (1, 700) \_\_ models/default/arch\_0190: (1, 700) \_\_ models/default/arch\_0191: (1, 700) models/default/arch 0286: (1, 700) \_\_ models/default/arch\_0288: (1, 700) \_\_ models/default/arch\_0302: (1, 700) models/default/arch 0335: (1, 700) models/default/arch\_0338: (1, 700) models/default/arch\_0346: (1, 700) Convert model output probabilities to F0 estimates In [6]: print('network F0 predictions in Hz:') f0\_bins = get\_f0\_bins() for key in sorted(dict model softmax output.keys()): predicted\_labels = np.argmax(dict\_model\_softmax\_output[key], axis=1) print(' \_\_\_ {}: {}'.format(key, label\_to\_f0(predicted\_labels, f0\_bins))) network F0 predictions in Hz: models/default/arch\_0083: [199.41587152] \_\_models/default/arch\_0154: [199.41587152] models/default/arch 0190: [197.981218] \_\_ models/default/arch\_0191: [199.41587152] \_\_ models/default/arch\_0286: [200.13709211] \_\_ models/default/arch\_0288: [200.86092112] \_\_ models/default/arch\_0302: [200.13709211] \_\_ models/default/arch\_0335: [199.41587152] \_\_models/default/arch\_0338: [198.69724993] \_\_ models/default/arch\_0346: [200.13709211] 2. Example SLURM bash script for training new pitch networks NOTE: Model training requires simulated auditory nerve representations of the training stimuli. #!/bin/bash #SBATCH --job-name=pitchnet train #SBATCH --out="slurm-%A\_%a.out" #SBATCH --cpus-per-task=6 #SBATCH --mem=18000 #SBATCH --gres=gpu:1 #SBATCH --time=0-48:00:00 #SBATCH --array=0-9 #### Specify list of model directories to train and use SLURM\_ARRAY\_TASK\_ID to select one per job declare -a list\_model\_dir=( '/models/demo\_train/arch\_0191/', '/models/demo train/arch 0302/', '/models/demo\_train/arch\_0288/', '/models/demo\_train/arch\_0335/', '/models/demo train/arch 0346/', '/models/demo\_train/arch\_0286/', '/models/demo train/arch 0083/', '/models/demo\_train/arch\_0154/', '/models/demo train/arch 0190/', '/models/demo\_train/arch\_0338/', offset=0 job\_idx=\$((\$SLURM\_ARRAY\_TASK\_ID + \$offset)) model\_dir=\${list\_model\_dir[\$job\_idx]} #### Specify paths for training dataset and validation dataset tfrecords DATA\_TRAIN="dataset\_pitchnet\_train/sr20000\_cf100\_species002\_spont070\_BW10eN1\_IHC3000Hz\_IHC7order/bez2018meanrate s 0[0-7]\*.tfrecords" DATA VALID="dataset pitchnet train/sr20000 cf100 species002 spont070 BW10eN1 IHC3000Hz IHC7order/bez2018meanrate s 0[8-9]\*.tfrecords" #### Run training routine within Singularity environment (NOTE: customize paths for -B flags) singularity exec --nv \ -B /om2/user/\$USER/pitchnet/ \ -B /om2/user/\$USER/pitchnet/packages:/packages \ -B /om2/user/\$USER/pitchnet/models:/models \ tensorflow-1.13.1-pitchnet.simg \ python -u /packages/tfutil/run train or eval.py "\$model dir" \ -dt "\$DATA\_TRAIN" \ -de "\$DATA VALID" \ -t -e -f 3. Example SLURM bash script for evaluating pitch networks NOTE: Model evaluation requires simulated auditory nerve representations of the psychophysical stimuli. #!/bin/bash #SBATCH -- job-name=pitchnet eval #SBATCH --out="slurm-%A %a.out" #SBATCH --cpus-per-task=6 #SBATCH --mem=18000 #SBATCH --gres=gpu:1 #SBATCH --time=0-48:00:00 #SBATCH --array=0-9 #### Specify list of model directories to train and use SLURM ARRAY TASK ID to select one per job declare -a list\_model\_dir=( '/models/default/arch\_0191/', '/models/default/arch\_0302/', '/models/default/arch\_0288/', '/models/default/arch\_0335/'**,** '/models/default/arch 0346/', '/models/default/arch\_0286/', '/models/default/arch\_0083/', '/models/default/arch\_0154/', '/models/default/arch\_0190/', '/models/default/arch\_0338/', offset=0 job\_idx=\$((\$SLURM\_ARRAY\_TASK\_ID + \$offset)) model\_dir=\${list\_model\_dir[\$job\_idx]} #### Specify path to psychophysical stimulus dataset tfrecords (uncomment one experiment below) TFRECORDS PATH="sr20000 cf100 species002 spont070 BW10eN1 IHC3000Hz IHC7order/bez2018meanrates\*tfrecords" #### Experiment A (Bernstein and Oxenham, 2005) DATA EVAL="data psychophysics/bernox2005/lowharm v01/\$TFRECORDS PATH" EVAL\_OUTPUT\_FILENAME="EVAL\_SOFTMAX\_lowharm\_v01\_bestckpt.json" #### Experiment B (Shackleton and Carlyon, 1994) # DATA\_EVAL="data\_psychophysics/mooremoore2003/freqshifted\_v01/\$TFRECORDS\_PATH" # EVAL\_OUTPUT\_FILENAME="EVAL\_SOFTMAX\_freqshifted\_v01\_bestckpt.json" #### Experiment C (Moore and Moore, 2003) # DATA\_EVAL="data\_psychophysics/moore1985/mistunedharm\_v01/\$TFRECORDS\_PATH" # EVAL\_OUTPUT\_FILENAME="EVAL\_SOFTMAX\_mistunedharm\_v01\_bestckpt.json" #### Experiment D (Moore et al., 1985) # DATA EVAL="data psychophysics/oxenham2004/transposedtones v01/\$TFRECORDS PATH" # EVAL\_OUTPUT\_FILENAME="EVAL\_SOFTMAX\_transposedtones\_v01\_bestckpt.json" #### Experiment E (Oxenham et al., 2004) # DATA\_EVAL="data\_psychophysics/shackcarl1994/altphase\_v01/\$TFRECORDS\_PATH" # EVAL\_OUTPUT\_FILENAME="EVAL\_SOFTMAX\_altphase\_v01\_bestckpt.json" #### Run evaluation routine within Singularity environment (NOTE: customize paths for -B flags) singularity exec --nv \ -B /om2/user/\$USER/pitchnet/ \ -B /om2/user/\$USER/pitchnet/packages:/packages \ -B /om2/user/\$USER/pitchnet/models:/models \ tensorflow-1.13.1-pitchnet.simg \ python pitchnet\_evaluate\_best.py \ -de "\$DATA\_EVAL" \ -efn "\$EVAL OUTPUT FILENAME" \ -o "\$model\_dir" \ -wpo 1 4. Example SLURM bash script for simulating psychophysical experiments Once network predictions for all of the psychophysical stimulus datasets have been stored, this script can be used to simulate the actual psychophysical experiments (e.g., measure discrimination thresholds or shifts in the perceived pitch). #!/bin/bash #SBATCH --job-name=pitchnet analyze #SBATCH --out="slurm-%A\_%a.out" #SBATCH --cpus-per-task=6 #SBATCH --mem=36000 #SBATCH --time=0-6:00:00 #SBATCH --array=0-9 #### Specify regular expression that captures all model directories MODEL\_DIR\_REGEX="/models/default/arch\_????" #### Specify prior to impose when computing pitch estimates (+/- 0.5 octaves of experiment-specific value) PRIOR\_RANGE="0.5" #### Run each experiment in the Singularity environment (NOTE: customize paths for -B flags) #### Experiment A (Bernstein and Oxenham, 2005) singularity exec \ -B /om2/user/\$USER/pitchnet/ \ -B /om2/user/\$USER/pitchnet/packages:/packages \ -B /om2/user/\$USER/pitchnet/models:/models \ tensorflow-1.13.1-pitchnet.simg \ python assets\_psychophysics/f0dl\_bernox.py \ -r "\${MODEL\_DIR\_REGEX}/EVAL\_SOFTMAX\_lowharm\_v01\_bestckpt.json" \ -p \${PRIOR\_RANGE} \ -j \${job\_idx} #### Experiment B (Shackleton and Carlyon, 1994) singularity exec \ -B /om2/user/\$USER/pitchnet/ \ -B /om2/user/\$USER/pitchnet/packages:/packages \ -B /om2/user/\$USER/pitchnet/models:/models \ tensorflow-1.13.1-pitchnet.simg \ python assets\_psychophysics/f0experiment\_alt\_phase.py \ -r "\${MODEL\_DIR\_REGEX}/EVAL\_SOFTMAX\_altphase\_v01\_bestckpt.json" \ -p \${PRIOR\_RANGE} \ -j \${job\_idx} #### Experiment C (Moore and Moore, 2003) singularity exec \ -B /om2/user/\$USER/pitchnet/ \ -B /om2/user/\$USER/pitchnet/packages:/packages \ -B /om2/user/\$USER/pitchnet/models:/models \ tensorflow-1.13.1-pitchnet.simg \ python assets\_psychophysics/f0experiment\_freq\_shifted.py \ -r "\${MODEL\_DIR\_REGEX}/EVAL\_SOFTMAX\_freqshifted\_v01\_bestckpt.json" \ -p \${PRIOR\_RANGE} \ -j \${job\_idx} #### Experiment D (Moore et al., 1985) singularity exec \ -B /om2/user/\$USER/pitchnet/ \ -B /om2/user/\$USER/pitchnet/packages:/packages \ -B /om2/user/\$USER/pitchnet/models:/models \ tensorflow-1.13.1-pitchnet.simg \ python assets psychophysics/f0experiment mistuned harmonics.py \ -r "\${MODEL\_DIR\_REGEX}/EVAL\_SOFTMAX\_mistunedharm\_v01\_bestckpt.json" \ -p \${PRIOR\_RANGE} \ -j \${job\_idx} #### Experiment E (Oxenham et al., 2004) singularity exec \ -B /om2/user/\$USER/pitchnet/ \ -B /om2/user/\$USER/pitchnet/packages:/packages \ -B /om2/user/\$USER/pitchnet/models:/models \ tensorflow-1.13.1-pitchnet.simg \ python assets\_psychophysics/f0dl\_transposed\_tones.py \ -r "\${MODEL\_DIR\_REGEX}/EVAL\_SOFTMAX\_transposedtones\_v01\_bestckpt.json" \

-p \${PRIOR\_RANGE} \

-j \${job\_idx}

In [ ]:

**Pitchnet DEMO Notebook** 

\$ singularity exec --nv \

-B \$(pwd)/packages:/packages \

-B \$(pwd) \

To run this notebook within the provided Singularity environment:

tensorflow-1.13.1-pitchnet.simg jupyter notebook

1. Demonstrates how to evauate the pre-trained pitch network models on new stimuli

3. Provides a sample SLURM sbatch script for evaluating networks on psychophysical experiment stimuli 4. Provides a sample SLURM sbatch script for running psychophysical experiments and plotting results

2. Provides a sample SLURM sbatch script for training new pich network models

This notebook: