

# Part 1: Serra's Analysis

In this notebook we demonstrate our implementation of Serra's Sine Plus Noise analysis.

**Set the paths below to point to the directories of the Philharmonia Cello and Guitar samples.**

```
In [2]: %matplotlib inline
%load_ext autoreload
%autoreload 2
from pathlib import Path
CELLO_PATH = Path("/home/lukas/BA/philharmonia-samples/cello")
GUITAR_PATH = Path("/home/lukas/BA/philharmonia-samples/guitar")

# Output directories for figures and wavfiles
GFX_PATH = Path("/home/lukas/BA/report/gfx/")
WAVS_PATH = Path("/home/lukas/BA/report/wavs/")

# Whether to rebuild the dataset (The dataset needs to be built
# at least once. Building from the commandline seems to be
# significantly faster than from Jupyter, though I don't know why.
# So I advise to set this to False and build from the commandline
# instead.)
REBUILD_DATASETS = False
```

The autoreload extension is already loaded. To reload it, use:  
%reload\_ext autoreload

```
In [3]: %%capture

import pya
aserver = pya.Aserver()
aserver.boot()
pya.Aserver.default = aserver
```

```
ALSA lib pcm_dmix.c:1090:(snd_pcm_dmix_open) unable to open slave
ALSA lib pcm.c:2660:(snd_pcm_open_noupdate) Unknown PCM cards.pcm.rear
ALSA lib pcm.c:2660:(snd_pcm_open_noupdate) Unknown PCM cards.pcm.center_lfe
ALSA lib pcm.c:2660:(snd_pcm_open_noupdate) Unknown PCM cards.pcm.side
ALSA lib pcm_route.c:877:(find_matching_chmap) Found no matching channel map
ALSA lib dlmisc.c:345:(snd_dlobj_cache_get0) Cannot open shared library libasound_module_rate_samplerate.so (libasound_module_rate_samplerate.so: libasound_module_rate_samplerate.so: cannot open shared object file: No such file or directory)
ALSA lib pcm_rate.c:1468:(snd_pcm_rate_open) Cannot find rate converter
connect(2) call to /dev/shm/jack-1000/default/jack_0 failed (err=No such file or directory)
attempt to connect to server failed
connect(2) call to /dev/shm/jack-1000/default/jack_0 failed (err=No such file or directory)
attempt to connect to server failed
ALSA lib pcm_oss.c:397:(_snd_pcm_oss_open) Cannot open device /dev/dsp
ALSA lib pcm_oss.c:397:(_snd_pcm_oss_open) Cannot open device /dev/dsp
ALSA lib dlmisc.c:345:(snd_dlobj_cache_get0) Cannot open shared library libasound_module_pcm_pipewire.so (libasound_module_pcm_pipewire.so: libasound_module_pcm_pipewire.so: cannot open shared object file: No such file or directory)
ALSA lib dlmisc.c:345:(snd_dlobj_cache_get0) Cannot open shared library libasound_module_pcm_pipewire.so (libasound_module_pcm_pipewire.so: libasound_module_pcm_pipewire.so: cannot open shared object file: No such file or directory)
ALSA lib pcm_a52.c:823:(_snd_pcm_a52_open) a52 is only for playback
ALSA lib dlmisc.c:345:(snd_dlobj_cache_get0) Cannot open shared library libasound_module_pcm_speex.so (libasound_module_pcm_speex.so: libasound_module_pcm_speex.so: cannot open shared object file: No such file or directory)
ALSA lib dlmisc.c:345:(snd_dlobj_cache_get0) Cannot open shared library libasound_module_pcm_speex.so (libasound_module_pcm_speex.so: libasound_module_pcm_speex.so: cannot open shared object file: No such file or directory)
ALSA lib pcm_usb_stream.c:486:(_snd_pcm_usb_stream_open) Invalid type for card
ALSA lib pcm_usb_stream.c:486:(_snd_pcm_usb_stream_open) Invalid type for card
ALSA lib pcm_dmix.c:1090:(snd_pcm_dmix_open) unable to open slave
```

```
In [4]: # Initialization
import numpy as np
import matplotlib
import matplotlib.pyplot as plt
import pandas as pd

import principal_harmonics as ph

for path in [GFX_PATH, WAVS_PATH]:
    if path.exists() and not path.is_dir():
        raise NotADirectoryError(path)
    if not path.exists():
        path.mkdir()
```

```
ALSA lib pcm.c:8545:(snd_pcm_recover) underrun occurred
ALSA lib pcm.c:8545:(snd_pcm_recover) underrun occurred
ALSA lib pcm.c:8545:(snd_pcm_recover) underrun occurred
```

## Pitch estimation

The YIN Pitch tracker provides reliable pitch estimates for loud parts of the signal. For quieter parts, the estimation tends to jump - this is especially problematic for the guitar, where the signal decays rapidly.

- Use `principal_harmonics.pvoc.get_pitch()` to get the YIN pitch estimate of a signal
- Use `principal_harmonics.pvoc.constant_pitch()` to get a weighted-median estimate of a constant pitch given a time-varying pitch estimate and an expected frequency.

```
In [5]: cello_path = CELLO_PATH / 'cello_A4_15_forte_arco-normal.mp3'
guitar_path = GUITAR_PATH / 'guitar_A4_very-long_forte_normal.mp3'
cello_asig = pya.Asig(str(cello_path))
guitar_asig = pya.Asig(str(guitar_path))
cello_pitch = ph.pvoc.get_pitch(cello_asig)
guitar_pitch = ph.pvoc.get_pitch(guitar_asig)
constant_guitar_pitch = ph.pvoc.constant_pitch(guitar_pitch, expected=440.0)

In [6]: fig, ((cello_rms_ax, guitar_rms_ax), (cello_pitch_ax, guitar_pitch_ax)) = \
        plt.subplots(nrows=2, ncols=2, sharex='col', sharey='row', figsize=(8.2, 3))

plt.sca(cello_rms_ax)
cello_asig.window_op(fn='rms', nperseg=512, stride=256).plot(x_as_time=False, label='RMS amplitude')
cello_rms_ax.set_ylabel("RMS Amplitude")
cello_rms_ax.set_title(cello_path.name)

cello_pitch_ax.plot(cello_pitch, label='YIN pitch estimation', c='orange')
cello_pitch_ax.set_ylabel("Pitch in Hz")
cello_pitch_ax.set_ylim(270, 600)
cello_pitch_ax.set_xlabel("Frame index")

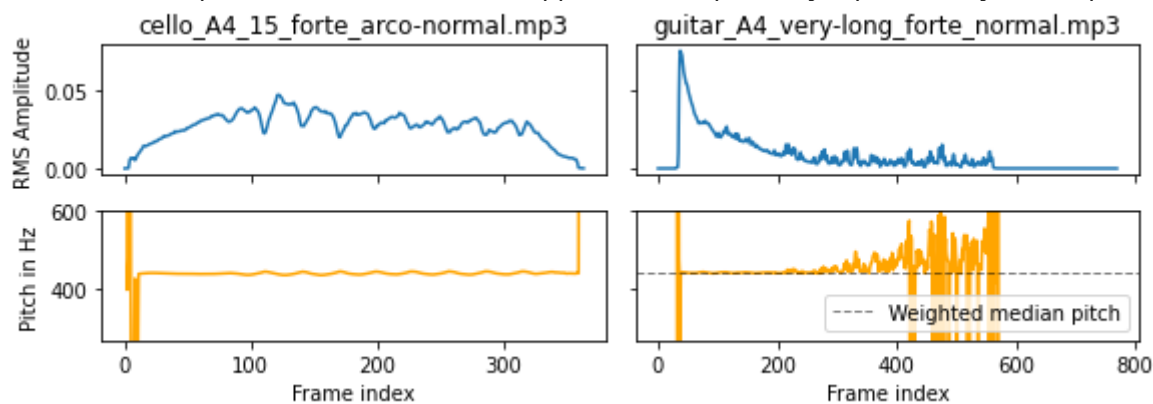
plt.sca(guitar_rms_ax)
guitar_asig.window_op(fn='rms', nperseg=512, stride=256).plot(x_as_time=False, label='RMS amplitude')
guitar_rms_ax.set_title(guitar_path.name)

guitar_pitch_ax.plot(guitar_pitch, c='orange')
guitar_pitch_ax.set_xlabel("Frame index")
guitar_pitch_ax.axhline(constant_guitar_pitch, linestyle='--', linewidth=1, c='black', label='Weighted median pitch')
guitar_pitch_ax.legend(loc=4)

fig.tight_layout()

plt.savefig(GFX_PATH / '1-pitch-estimation.eps')
```

The PostScript backend does not support transparency; partially transparent artists will be rendered opaque.



## Sinusoidal analysis

Let us analyze a single sample using the sinusoidal analysis. We visualize the timbre vectors in two different ways:

*Plotting the detected peak frequencies vs. time.* The color denotes the decibel-scaled amplitude of the peak. This can give a holistic view of the structure of the deterministic component we extracted, including vibrato, jumping partials, missing values, etc. We can use this to judge the overall performance of the analysis and to tune analysis parameters to the data at hand. However, we cannot identify finer structures in the trajectory of the peak amplitudes. *Plotting (a subset) of the detected peak amplitudes vs time*, we can better observe the behaviour of the overtone amplitudes.

```
In [7]: # using simple peak matching (default)
cello_pitch = ph.pvoc.get_pitch(cello_asig)
cello_freqs, cello_coefs = ph.pvoc.sinusoidal_analysis(cello_asig, cello_pitch)
start, stop = ph.pvoc.ClipStationary().clip(cello_coefs)
cello_freqs, cello_coefs = cello_freqs[start:stop], cello_coefs[start:stop]

cello_asig_clipped = cello_asig[start*256:stop*256] # we'll need this for later...

# ... and using tracking peak matching
cello_freqs_tracking, cello_coefs_tracking = ph.pvoc.sinusoidal_analysis(cello_asig, cello_pitch, peak_matching='tr
cello_freqs_tracking, cello_coefs_tracking = cello_freqs_tracking[start:stop], cello_coefs_tracking[start:stop]

# using simple peak matching (default)
guitar_pitch = ph.pvoc.constant_pitch(ph.pvoc.get_pitch(guitar_asig), expected=440.0)
guitar_freqs, guitar_coefs = ph.pvoc.sinusoidal_analysis(guitar_asig, guitar_pitch)
start, stop = ph.pvoc.ClipTransient().clip(guitar_coefs)
guitar_freqs, guitar_coefs = guitar_freqs[start:stop], guitar_coefs[start:stop]
```

```
In [8]: fig = plt.figure(figsize=(8.2, 4.5))

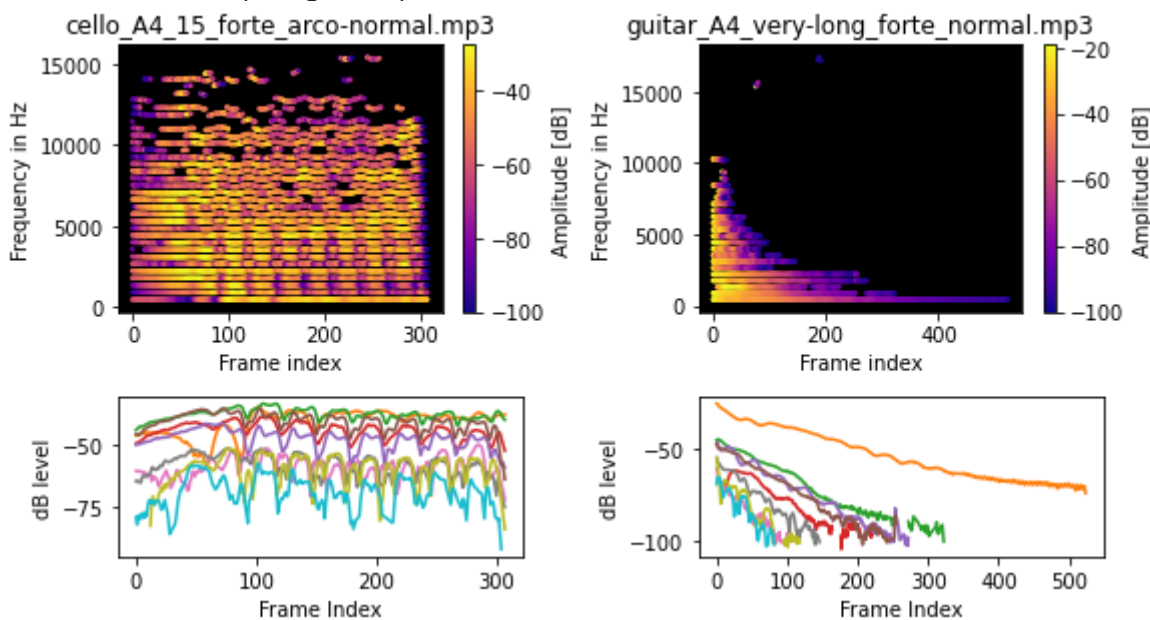
gs = matplotlib.gridspec.GridSpec(2, 2, height_ratios=[0.5, 0.3])
ax = fig.add_subplot(gs[0,0])
plt.sca(ax)
plt.title(cello_path.name)
ph.plots.plot_timbre_vectors(cello_freqs, cello_coefs)

ax = fig.add_subplot(gs[0,1])
plt.sca(ax)
plt.title(guitar_path.name)
ph.plots.plot_timbre_vectors(guitar_freqs, guitar_coefs)

ax = fig.add_subplot(gs[1,0])
plt.sca(ax)
ph.plots.plot_ampls(np.abs(cello_coefs[:, :10]), dbify=True, write_numbers=False)
plt.xlabel("Frame Index")
ax = fig.add_subplot(gs[1,1])
plt.sca(ax)
ph.plots.plot_ampls(np.abs(guitar_coefs[:, :10]), dbify=True, write_numbers=False)
plt.xlabel("Frame Index")

plt.tight_layout()
plt.savefig(GFX_PATH / '1-timbre-vectors.png', dpi=300)
```

/home/lukas/miniconda3/envs/ba-venv/lib/python3.9/site-packages/pya/helper/helpers.py:55: RuntimeWarning: divide by zero encountered in log10  
return 20 \* np.log10(amp)

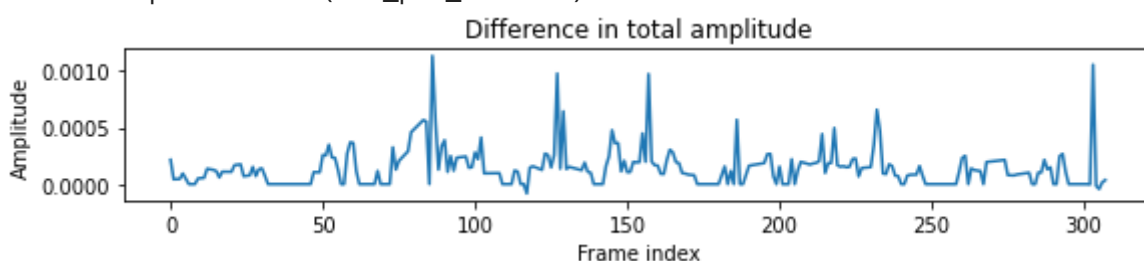


Comparing the performance of the simple and tracking peak matchers, we see little difference in extracted amplitude.

```
In [9]: plt.figure(figsize=(8.2, 2))
plt.title("Difference in total amplitude")
plt.xlabel("Frame index")
plt.ylabel("Amplitude")
cello_ampls_simple = np.nansum(np.abs(cello_coefs), axis=1)
cello_ampls_tracking = np.nansum(np.abs(cello_coefs_tracking), axis=1)
plt.plot(cello_ampls_simple - cello_ampls_tracking)

plt.tight_layout()
plt.savefig(GFX_PATH / "1-peak-matching-eval.eps")
```

ALSA lib pcm.c:8545:(snd\_pcm\_recover) underrun occurred



We can analyze the inner workings of the analysis by having it return the spectrum computation for each frame...

```
In [10]: freqs, coefs, steps = ph.pvoc.sinusoidal_analysis(cello_asig, cello_pitch, return_steps=True)
step = steps[42]
```

```
In [11]: %matplotlib inline
import scipy.signal
import matplotlib.gridspec

fig = plt.figure(figsize=(8.2, 8))
gs = matplotlib.gridspec.GridSpec(4, 1, height_ratios=[1,0.5,1,1])
plt.suptitle(cello_path.name)

ax = fig.add_subplot(gs[0, 0])
step.asig_slice.plot(x_as_time=False, ax=ax)
ax.set_title("Portion of a cello signal")
ax.set_xlabel("Sample index $t$")
ax.set_ylabel("$s[t]$")

ax = fig.add_subplot(gs[1, 0])
win = scipy.signal.get_window('blackmanharris', step.asig_slice.samples)
ax.plot(win)
ax.set_title("Blackman-Harris 92db analysis window")
ax.set_xlabel("Sample index $t$")
ax.set_ylabel("$w[t]$")

ax = fig.add_subplot(gs[2, 0])
(step.asig_slice * win).plot(x_as_time=False, ax=ax)
ax.set_title("Windowed signal")
ax.set_xlabel("Sample index $t$")
ax.set_ylabel("$s[t] * w[t]$")

ax = fig.add_subplot(gs[3, 0])
fft_freqs = step.fft_freqs
fft_dbs = pya.ampdb(np.abs(step.fft_coefs))
ax.plot(step.fft_freqs, fft_dbs)

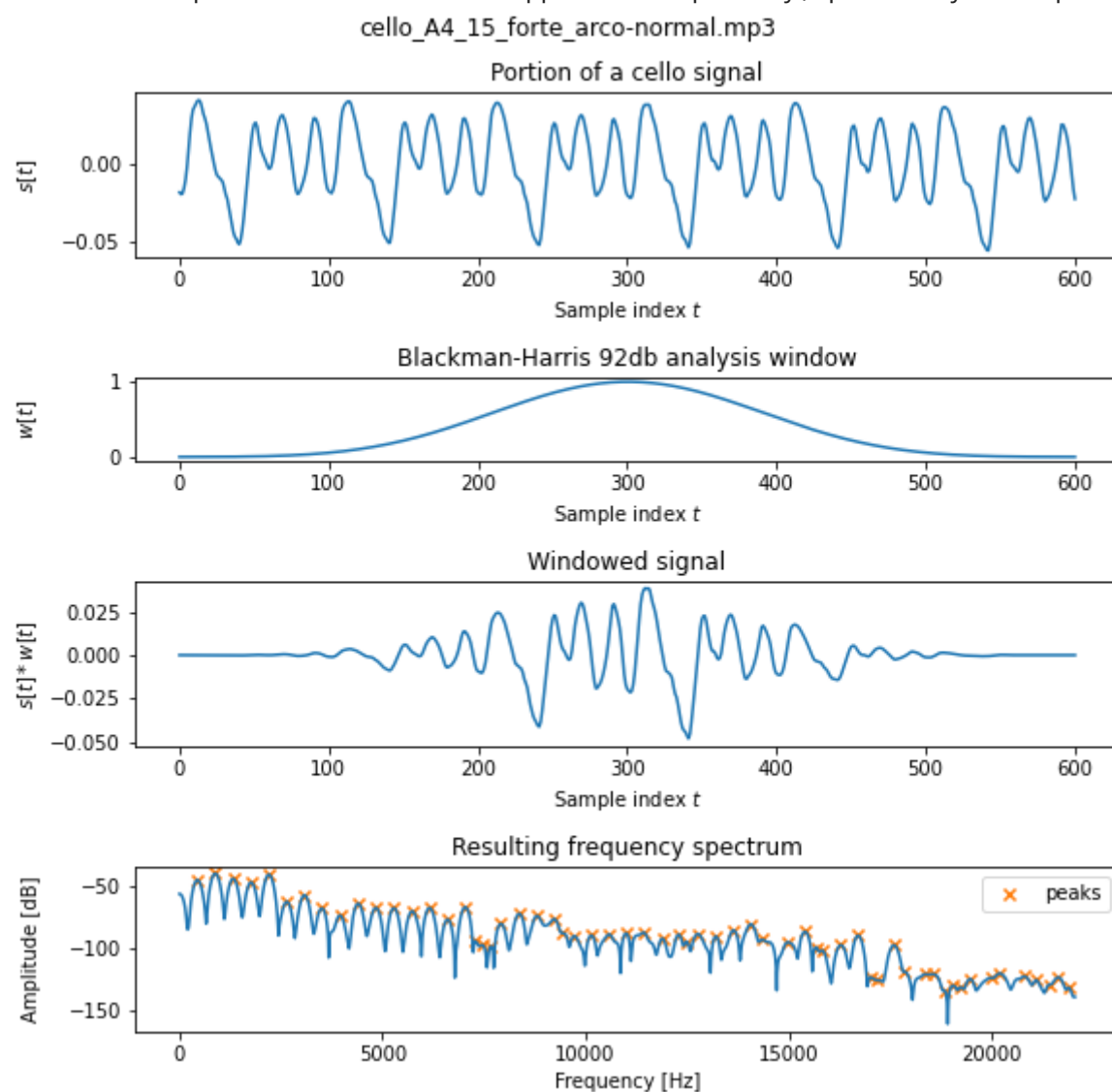
maxs = scipy.signal.argrelmax(fft_dbs)
ax.scatter(fft_freqs[maxs], fft_dbs[maxs], marker='x', label='peaks', color='tab:orange')

ax.set_ylabel("Amplitude [dB]")
ax.set_xlabel("Frequency [Hz]")
ax.legend()
ax.set_title("Resulting frequency spectrum")

fig.align_ylabels()
plt.tight_layout()

plt.savefig(GFX_PATH / "1-spectrum.eps")
```

The PostScript backend does not support transparency; partially transparent artists will be rendered opaque.



This figure illustrates the simple peak matcher. (Not used in the report)



```
In [12]: import matplotlib.gridspec as gridspec
import matplotlib.image as img
rng = np.random.default_rng(seed=42)

plt.figure(figsize=(8.2, 3))
fundamentals = np.array([220, 230, 200, 210, 230])
n = fundamentals.shape[0]
ks = np.arange(1, 5).reshape((1, -1))
m = ks.shape[1]
expected_freqs = fundamentals.reshape((-1, 1)) * ks
devs = np.random.uniform(-30, 30, size=expected_freqs.shape)
devs[1, 2] = np.nan
devs[4, 1] = -80
actual_freqs = expected_freqs + devs

xs = np.broadcast_to(np.arange(n).reshape((-1, 1)), actual_freqs.shape)

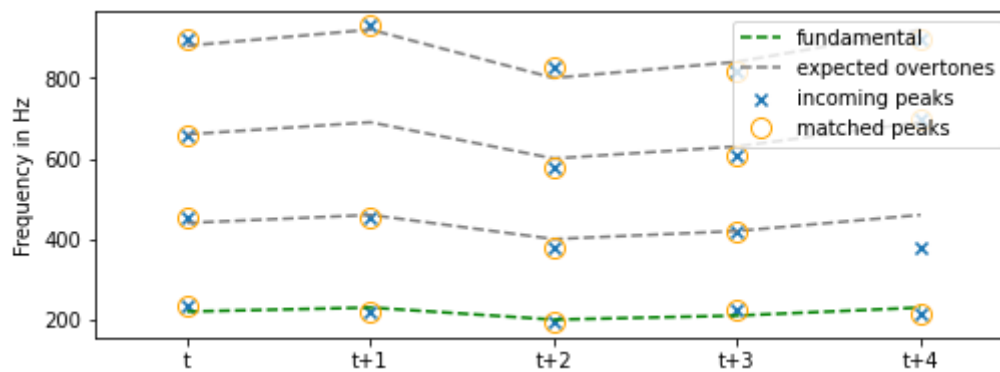
plt.ylabel("Frequency in Hz")
plt.xlim(-0.5, n - 0.5)
plt.xticks(np.arange(n), labels=['t', 't+1', 't+2', 't+3', 't+4'])

plt.plot(np.arange(n), fundamentals, label='fundamental', c='green', linestyle='--')
for i in range(1, m):
    line = plt.plot(xs[:, i], expected_freqs[:, i], c='gray', linestyle='--')[0]
    if i == 1:
        line.set_label('expected overtones')

plt.scatter(xs.ravel(), actual_freqs.ravel(), label='incoming peaks', marker='x')
matches = np.where(np.abs(expected_freqs - actual_freqs) < 30)
plt.plot(xs[matches].ravel(), actual_freqs[matches].ravel(), c='orange', marker='o', markersize=10, label='matched peaks')

plt.legend(loc=1)
plt.savefig(GFX_PATH / '1-expected-overtone-peak-matcher.eps')
```

The PostScript backend does not support transparency; partially transparent artists will be rendered opaque.



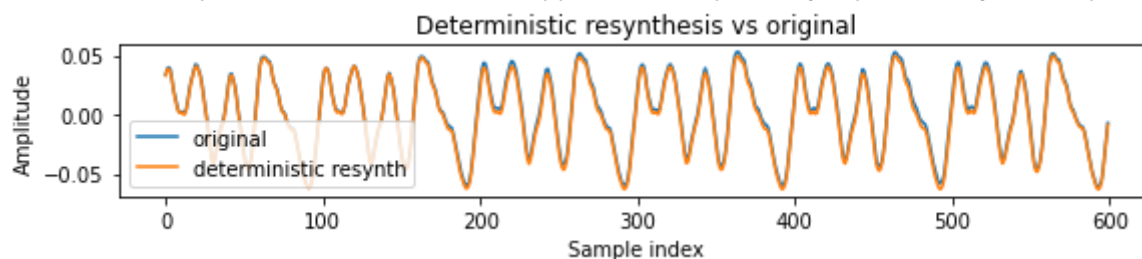
Having gathered timbre vectors, let's check that the resynthesis worked:

```
In [13]: resynth = ph.pvoc.phase_correct_resynth(cello_freqs, cello_coefs)

In [14]: plt.figure(figsize=(8.2, 2))
plt.title("Deterministic resynthesis vs original")
plt.ylabel("Amplitude")
t = 42*256
n = int((6*step.n_samples)//2)
cello_asig_clipped[t-n : t+n].plot(label='original', x_as_time=False)
resynth[t-n:t+n].plot(label='deterministic resynth', x_as_time=False)
plt.xlabel("Sample index")
plt.legend(loc=0)

plt.tight_layout()
plt.savefig(GFX_PATH / '1-phase-correct-resynth.eps')
```

The PostScript backend does not support transparency; partially transparent artists will be rendered opaque.



We can subtract the synthesized signal from the original to get the stochastic residue.

```
In [15]: guitar_freqs_tracking, guitar_coefs_tracking = ph.pvoc.sinusoidal_analysis(guitar_asig, guitar_pitch, peak_matching)
guitar_resynth = ph.pvoc.phase_correct_resynth(guitar_freqs_tracking, guitar_coefs_tracking)
(guitar_resynth.x - guitar_asig.x).play()
```

```
Out[15]: Asig('_subtracted'): 1 x 197120 @ 44100Hz = 4.470s cn=['0']
```

Using plain additive synthesis without considering the phases, we get a signal that is perceptually the same.

```
In [16]: guitar_resynth_additive = ph.pvoc.additive_resynth(guitar_freqs_tracking, np.abs(guitar_coefs_tracking))
guitar_resynth_additive.play()
```

```
(1, 1) (770, 40) (770, 40)
```

```
ALSA lib pcm.c:8545:(snd_pcm_recover) underrun occurred
```

```
Out[16]: Asig(''): 1 x 196864 @ 44100Hz = 4.464s cn=['0']
```

## Building the dataset

Having looked at the analysis of single samples, let us now build the whole dataset. If `REBUILD_DATASET` was set to `True`, this might take a while.

```
In [17]: if REBUILD_DATASETS:
          for path in [CELLO_PATH, GUITAR_PATH]:
              # delete labels file if it exists
              if (path / 'labels.csv').exists():
                  (path / 'labels.csv').unlink()
          ph.dataset.build_labels(base_dir=CELLO_PATH, glob='cello*_15*_arco-normal.mp3', parser='philharmonia')
          ph.dataset.build_labels(base_dir=GUITAR_PATH, glob='guitar*_very-long*_normal.mp3', parser='philharmonia')

          ph.dataset.build_dataset(GUITAR_PATH, pitch_mode='constant', clip_strategy='transient')
          ph.dataset.build_dataset(CELLO_PATH, pitch_mode='variable', clip_strategy='stationary')
```

We can load the whole dataset from disk using `ph.dataset.open_dataset`. The dataframe that is returned contains all analysis results and the labels. Frequencies and Coefficients are stored as `np.ndarray objects`, i.e. they need to be unwrapped before they can be used for analysis in a pandas Series. This is because the analysis results might not have the same dimensionality (the stochastic analysis might have a different hop size than the deterministic analysis))

```
In [18]: cello_df = ph.dataset.open_dataset(CELLO_PATH)
          guitar_df = ph.dataset.open_dataset(GUITAR_PATH)
          pd.options.display.min_rows=30
          pd.options.display.max_rows=30
          cello_df
```

Out[18]:

	Unnamed: 0	midi	exclude	other	instrument	duration	dynamic	note	clipped_wav	resynth_wav	r
filename											
cello_A2_15_forte_arco-normal.mp3	0	45.0	False	arco-normal	cello	15	forte	A2	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_A2_15_pianissimo_arco-normal.mp3	1	45.0	False	arco-normal	cello	15	pianissimo	A2	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_A2_15_piano_arco-normal.mp3	2	45.0	False	arco-normal	cello	15	piano	A2	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_A3_15_forte_arco-normal.mp3	3	57.0	False	arco-normal	cello	15	forte	A3	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_A3_15_mezzo-piano_arco-normal.mp3	4	57.0	False	arco-normal	cello	15	mezzo-piano	A3	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_A4_15_forte_arco-normal.mp3	5	69.0	False	arco-normal	cello	15	forte	A4	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_A4_15_mezzo-piano_arco-normal.mp3	6	69.0	False	arco-normal	cello	15	mezzo-piano	A4	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_A5_15_forte_arco-normal.mp3	7	81.0	False	arco-normal	cello	15	forte	A5	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_A5_15_mezzo-piano_arco-normal.mp3	8	81.0	False	arco-normal	cello	15	mezzo-piano	A5	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_A5_15_pianissimo_arco-normal.mp3	9	81.0	False	arco-normal	cello	15	pianissimo	A5	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_As2_15_forte_arco-normal.mp3	10	46.0	False	arco-normal	cello	15	forte	A#2	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_As2_15_fortissimo_arco-normal.mp3	11	46.0	False	arco-normal	cello	15	fortissimo	A#2	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_As2_15_pianissimo_arco-normal.mp3	12	46.0	False	arco-normal	cello	15	pianissimo	A#2	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_As2_15_piano_arco-normal.mp3	13	46.0	False	arco-normal	cello	15	piano	A#2	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_As3_15_forte_arco-normal.mp3	14	58.0	False	arco-normal	cello	15	forte	A#3	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
...	...	...	...	...	...	...	...	...	...	...	
cello_Gs2_15_fortissimo_arco-normal.mp3	166	44.0	False	arco-normal	cello	15	fortissimo	G#2	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_Gs2_15_pianissimo_arco-normal.mp3	167	44.0	False	arco-normal	cello	15	pianissimo	G#2	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_Gs2_15_piano_arco-normal.mp3	168	44.0	False	arco-normal	cello	15	piano	G#2	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar
cello_Gs3_15_forte_arco-normal.mp3	169	56.0	False	arco-normal	cello	15	forte	G#3	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/rBA/phi sar

	Unnamed: 0	midi	exclude	other	instrument	duration	dynamic	note	clipped_wav	resynth_wav	r
filename											
cello_Gs3_15_fortissimo_arco-normal.mp3	170	56.0	False	arco-normal	cello	15	fortissimo	G#3	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/h
cello_Gs3_15_pianissimo_arco-normal.mp3	171	56.0	False	arco-normal	cello	15	pianissimo	G#3	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/h
cello_Gs3_15_piano_arco-normal.mp3	172	56.0	False	arco-normal	cello	15	piano	G#3	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/h
cello_Gs4_15_forte_arco-normal.mp3	173	68.0	False	arco-normal	cello	15	forte	G#4	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/h
cello_Gs4_15_fortissimo_arco-normal.mp3	174	68.0	False	arco-normal	cello	15	fortissimo	G#4	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/h
cello_Gs4_15_mezzo-piano_arco-normal.mp3	175	68.0	False	arco-normal	cello	15	mezzo-piano	G#4	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/h
cello_Gs4_15_pianissimo_arco-normal.mp3	176	68.0	False	arco-normal	cello	15	pianissimo	G#4	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/h
cello_Gs5_15_forte_arco-normal.mp3	177	80.0	False	arco-normal	cello	15	forte	G#5	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/h
cello_Gs5_15_fortissimo_arco-normal.mp3	178	80.0	False	arco-normal	cello	15	fortissimo	G#5	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/h
cello_Gs5_15_mezzo-piano_arco-normal.mp3	179	80.0	False	arco-normal	cello	15	mezzo-piano	G#5	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/h
cello_Gs5_15_pianissimo_arco-normal.mp3	180	80.0	False	arco-normal	cello	15	pianissimo	G#5	/home/lukas/BA/philharmonia-samples/cello/cell...	/home/lukas/BA/philharmonia-samples/cello/cell...	/h

181 rows × 16 columns

We can unwrap ndarrays by using `ph.dataset.expand_ndarrays()` , yielding a `MultiIndex` ed dataframe:

```
In [19]: ph.dataset.expand_ndarrays(cello_df, ['freqs', 'coefs'])
```



Out[19]:

		freqs-0	freqs-1	freqs-2	freqs-3	freqs-4	freqs-5	freqs-6	freqs-7	freqs-
filename		n								
cello_A2_15_forte_arco-normal.mp3	0	NaN	104.137558	220.598289	318.077413	NaN	528.929370	635.073708	741.260676	848.03838
	1	NaN	104.147027	216.604819	319.800956	NaN	528.279552	638.403465	743.001433	849.11652
	2	NaN	104.319988	215.776632	321.678334	NaN	529.864419	640.562735	745.262760	850.35237
	3	NaN	104.724961	215.664823	323.746908	442.430948	533.007135	641.596162	749.592093	852.29140
	4	NaN	105.260677	216.328448	325.791948	440.828898	536.100527	642.077293	756.082938	854.95418
	5	NaN	105.572623	217.762039	327.478049	439.372924	538.714698	642.731984	762.984872	857.05224
	6	NaN	105.823635	219.582341	328.353235	439.457084	541.261270	643.852288	765.782298	858.11330
	7	NaN	106.518020	220.518023	328.042049	441.372627	543.214222	645.470405	761.970830	859.67437
	8	NaN	107.561063	220.101268	326.858993	442.859676	544.279501	647.537026	755.484977	862.08608
	9	NaN	108.587171	219.288079	325.775657	441.748662	544.841642	649.172645	756.678083	863.19765
	10	NaN	109.621103	218.927402	325.263526	439.400788	545.034300	649.124158	764.487761	862.68353
	11	NaN	110.476806	218.887618	325.374477	438.485908	544.778753	648.301910	770.963124	861.93363
	12	NaN	110.710476	218.861005	325.714377	439.259736	544.406793	648.548954	758.970842	861.88556
	13	NaN	110.205722	218.821524	325.838015	440.156431	544.108501	650.280453	744.371597	862.52448
	14	NaN	109.388313	218.787805	325.903406	440.237244	543.966887	651.907194	NaN	863.18691
...	...	...	...	...	...	...	...	...	...	...
cello_Gs5_15_pianissimo_arco-normal.mp3	224	NaN	826.871125	1674.464250	NaN	3329.788233	4161.082620	4967.450000	5763.077669	6670.02495
	225	NaN	827.197457	1660.688193	NaN	3313.823610	4134.538414	4907.475654	5762.337811	6628.72063
	226	NaN	828.207768	1654.150380	NaN	3313.022546	4142.826831	5005.622510	5761.597952	6586.99151
	227	NaN	827.432263	1655.191645	NaN	3318.297161	4164.283833	NaN	5760.858094	6651.17151

		freqs-0	freqs-1	freqs-2	freqs-3	freqs-4	freqs-5	freqs-6	freqs-7	freqs-8
filename	n									
	228	NaN	829.678140	1650.765588	NaN	3320.502648	4135.942630	NaN	5760.118236	6636.88464
	229	NaN	829.861801	1658.678979	NaN	3317.341215	4147.344627	NaN	5822.171627	6617.81116
	230	NaN	832.514302	1668.432379	NaN	3333.890426	4163.951625	NaN	5802.470982	6638.67744
	231	NaN	830.266029	1660.239430	NaN	3320.030134	4159.836821	NaN	5870.150936	6675.11936
	232	NaN	832.361679	1670.256452	NaN	3337.697085	4165.031531	NaN	5876.821312	6666.67858
	233	NaN	832.177705	1664.896523	NaN	3332.317098	4170.635537	NaN	5819.169078	6636.19170
	234	NaN	832.599091	1672.435802	NaN	3337.399565	4174.660442	NaN	NaN	6681.90450
	235	NaN	830.661842	1666.078569	NaN	3331.438144	4178.729425	NaN	NaN	6713.34786
	236	NaN	832.651638	1661.470263	NaN	3338.471721	4171.561577	NaN	NaN	6670.74008
	237	NaN	828.948936	1660.014666	NaN	3334.191989	4177.842297	NaN	NaN	6673.44022
	238	NaN	828.786480	1655.368995	NaN	3309.162089	4136.953845	NaN	NaN	6585.83736

55842 rows × 94 columns

Let's look at the results for the cello. Notice that the r2 scores are degraded only for C2, C#2, D2 notes.

In [20]:

cello\_df.sort\_values('harmonic\_r2')[['note', 'harmonic\_r2']]

Out[20]:

filename	note	harmonic_r2
cello_Cs2_15_fortissimo_arco-normal.mp3	C#2	0.000184
cello_Cs2_15_piano_arco-normal.mp3	C#2	0.000417
cello_Cs2_15_forte_arco-normal.mp3	C#2	0.000643
cello_C2_15_piano_arco-normal.mp3	C2	0.021169
cello_D2_15_pianissimo_arco-normal.mp3	D2	0.022226
cello_C2_15_forte_arco-normal.mp3	C2	0.109291
cello_D2_15_mezzo-piano_arco-normal.mp3	D2	0.116769
cello_D2_15_forte_arco-normal.mp3	D2	0.151544
cello_Cs2_15_pianissimo_arco-normal.mp3	C#2	0.167291
cello_C2_15_fortissimo_arco-normal.mp3	C2	0.368731
cello_C2_15_pianissimo_arco-normal.mp3	C2	0.779312
cello_Ds2_15_fortissimo_arco-normal.mp3	D#2	0.794196
cello_Ds2_15_piano_arco-normal.mp3	D#2	0.834249
cello_Ds2_15_forte_arco-normal.mp3	D#2	0.852125
cello_Ds2_15_pianissimo_arco-normal.mp3	D#2	0.885786
...	...	...
cello_Fs3_15_fortissimo_arco-normal.mp3	F#3	0.998054
cello_Gs3_15_forte_arco-normal.mp3	G#3	0.998054
cello_Gs2_15_forte_arco-normal.mp3	G#2	0.998257
cello_Ds4_15_mezzo-piano_arco-normal.mp3	D#4	0.998317
cello_F3_15_piano_arco-normal.mp3	F3	0.998342
cello_E3_15_forte_arco-normal.mp3	E3	0.998353
cello_Ds3_15_forte_arco-normal.mp3	D#3	0.998408
cello_Ds3_15_fortissimo_arco-normal.mp3	D#3	0.998559
cello_E3_15_fortissimo_arco-normal.mp3	E3	0.998592
cello_Ds4_15_forte_arco-normal.mp3	D#4	0.998685
cello_F2_15_piano_arco-normal.mp3	F2	0.998764
cello_Fs3_15_forte_arco-normal.mp3	F#3	0.998830
cello_F2_15_forte_arco-normal.mp3	F2	0.999023
cello_F3_15_forte_arco-normal.mp3	F3	0.999042
cello_F3_15_fortissimo_arco-normal.mp3	F3	0.999061

181 rows × 2 columns

The guitar shows worse R2 results. However, notice that the samples with the worst R2 scores are almost all `piano` samples, and the samples with the best scores are almost all `forte` samples.

```
In [21]: guitar_df.sort_values('harmonic_r2')[['note', 'harmonic_r2']]
```

Out[21]:

	note	harmonic_r2
filename		
guitar_E5_very-long_piano_normal.mp3	E5	0.198769
guitar_E2_very-long_piano_normal.mp3	E2	0.239426
guitar_Ds5_very-long_piano_normal.mp3	D#5	0.249378
guitar_Gs5_very-long_forte_normal.mp3	G#5	0.268161
guitar_C3_very-long_piano_normal.mp3	C3	0.312250
guitar_F3_very-long_piano_normal.mp3	F3	0.322729
guitar_E3_very-long_piano_normal.mp3	E3	0.336654
guitar_G4_very-long_piano_normal.mp3	G4	0.395874
guitar-Cs3_very-long_piano_normal.mp3	C#3	0.405350
guitar_Ds5_very-long_forte_normal.mp3	D#5	0.425177
guitar_B2_very-long_piano_normal.mp3	B2	0.429724
guitar_G5_very-long_forte_normal.mp3	G5	0.455253
guitar_Fs3_very-long_piano_normal.mp3	F#3	0.495844
guitar_D3_very-long_piano_normal.mp3	D3	0.504071
guitar_Ds3_very-long_piano_normal.mp3	D#3	0.519614
...	...	...
guitar_Fs3_very-long_forte_normal.mp3	F#3	0.879171
guitar_A4_very-long_forte_normal.mp3	A4	0.881476
guitar_D5_very-long_forte_normal.mp3	D5	0.891243
guitar_As3_very-long_piano_normal.mp3	A#3	0.895590
guitar_As2_very-long_forte_normal.mp3	A#2	0.900786
guitar_F3_very-long_forte_normal.mp3	F3	0.903941
guitar_Gs3_very-long_forte_normal.mp3	G#3	0.906259
guitar_Ds3_very-long_forte_normal.mp3	D#3	0.910498
guitar_B2_very-long_forte_normal.mp3	B2	0.910852
guitar_D3_very-long_forte_normal.mp3	D3	0.917608
guitar_C6_very-long_piano_normal.mp3	C6	0.928321
guitar_A3_very-long_forte_normal.mp3	A3	0.929812
guitar_B3_very-long_forte_normal.mp3	B3	0.939934
guitar_G2_very-long_forte_normal.mp3	G2	0.957464
guitar_As3_very-long_forte_normal.mp3	A#3	0.958670

71 rows × 2 columns

If we interpret the dynamic as `piano == 0` and `forte == 1`, we get a 0.52 corellation between the dynamic and the r2 score.

In [22]:

guitar\_df['dynamic\_int'] = guitar\_df.dynamic.apply(lambda d: 1.0 if d == 'forte' else 0)  
guitar\_df.convert\_dtypes()[['dynamic\_int', 'harmonic\_r2']].corr()

Out[22]:

	dynamic_int	harmonic_r2
dynamic_int	1.000000	0.518304
harmonic_r2	0.518304	1.000000

Let's build all stats for the report:

```
In [23]: import librosa

quantiles = [0, 0.25, 0.5, 0.75, 1]
cello_r2s = cello_df.harmonic_r2.quantile(quantiles)
cello_over_d2_r2s = cello_df[cello_df.midi > librosa.note_to_midi('D2')].harmonic_r2.quantile(quantiles)

guitar_r2s = guitar_df.harmonic_r2.quantile(quantiles)
guitar_piano_r2s = guitar_df[guitar_df.dynamic == 'piano'].harmonic_r2.quantile(quantiles)
guitar_forte_r2s = guitar_df[guitar_df.dynamic == 'forte'].harmonic_r2.quantile(quantiles)

r2_stats = pd.DataFrame(
    [
        cello_r2s,
        cello_over_d2_r2s,
        guitar_r2s,
        guitar_piano_r2s,
        guitar_forte_r2s
    ],
    index=[
        'Cello',
        'Cello > D2',
        'Guitar',
        r'Guitar \textbf{\textit{p}}',
        r'Guitar \textbf{\textit{f}}'
    ]
)
r2_stats.columns = [
    r'\(q=0\) (min)',
    r'\(q=0.25\) ',
    r'\(q=0.5\) (median)',
    r'\(q=0.75\) ',
    r'\(q=1\) (max)'
]
r2_stats
```

Out[23]:

	$q = 0$ (min)	$q = 0.25$	$q = 0.5$ (median)	$q = 0.75$	$q = 1$ (max)
Cello	0.000184		0.978829	0.991228	0.996171
Cello > D2	0.794196		0.983347	0.992441	0.996351
Guitar	0.198769		0.549556	0.761197	0.865428
Guitar $\textbf{\textit{p}}$	0.198769		0.446254	0.606756	0.751304
Guitar $\textbf{\textit{f}}$	0.268161		0.772095	0.856637	0.900786

```
In [24]: r2_stats.style.to_latex(GFX_PATH / '1-r2-stats.tex', hrules=True,
                                label='tab:r2-stats',
                                caption=r'\(R^2\) scores of the deterministic synthesis, in quantiles.',
                                position_float='centering')
```

ALSA lib pcm.c:8545:(snd\_pcm\_recover) underrun occurred

Copying the analysis results that are referenced in the report...

```
In [25]: def write_results(df, fname):
    asig = pya.Asig(str(df.loc[fname + '.mp3'].clipped_wav))
    asig.save_wavfile(str(WAVS_PATH / ('ch3_' + fname + '_clipped.wav')))
    asig = pya.Asig(str(df.loc[fname + '.mp3'].resynth_wav))
    asig.save_wavfile(str(WAVS_PATH / ('ch3_' + fname + '_harmonic-analysis.wav')))
    asig = pya.Asig(str(df.loc[fname + '.mp3'].noise_wav))
    asig.save_wavfile(str(WAVS_PATH / ('ch3_' + fname + '_noise-residue.wav')))

    for fname in [
        "cello_A4_15_forte_arco-normal",
        "cello-Cs3_15_fortissimo_arco-normal",
        "cello-Cs3_15_pianissimo_arco-normal",
        "cello_E4_15_mezzo-piano_arco-normal",
    ]:
        write_results(cello_df, fname)

    for fname in [
        "guitar_As4_very-long_forte_normal",
        "guitar_Ds3_very-long_piano_normal"
    ]:
        write_results(guitar_df, fname)
```

ALSA lib pcm.c:8545:(snd\_pcm\_recover) underrun occurred