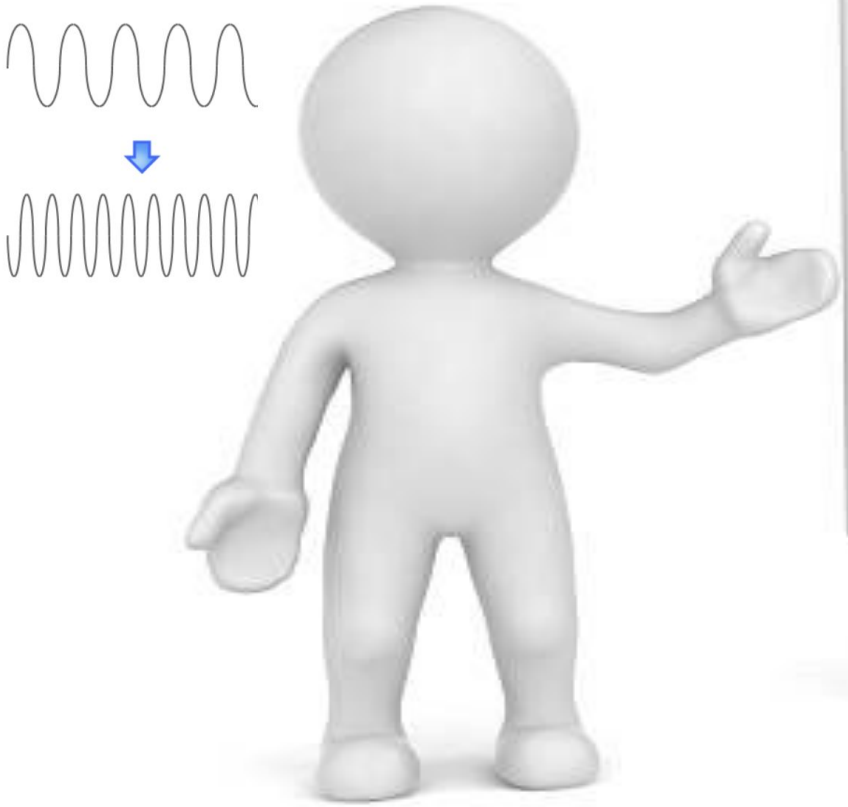
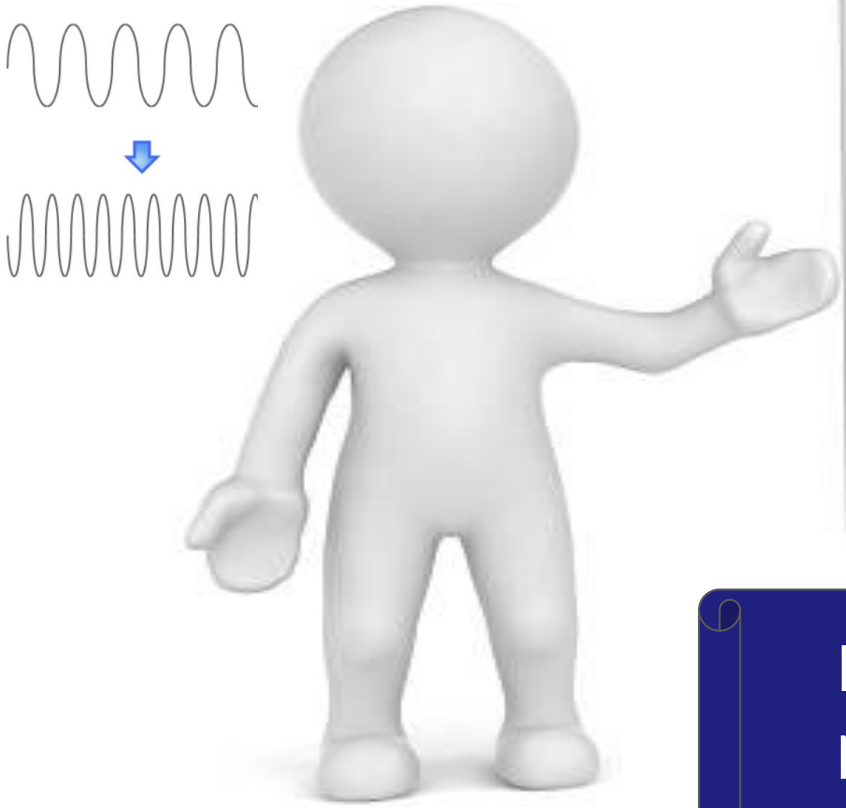
Evolving Constants by Rewriting Source Code

Dconf 2025 London — Guillaume Piolat


Nature
doesn’t have
low-latency pitch-shifting.



Changing the pitch
of a sound signal
is damn hard.



Changing the pitch of a sound signal is damn hard.

Especially real-time, with polyphonic input.

Zynaptiq Pitchshift Pro (2024)

Zplane Elastique Pro V3 Engine (2015)

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Zplane Elastique Pro V3 Engine (2015)

Inner Pitch v1 (2023)

- 17ms latency

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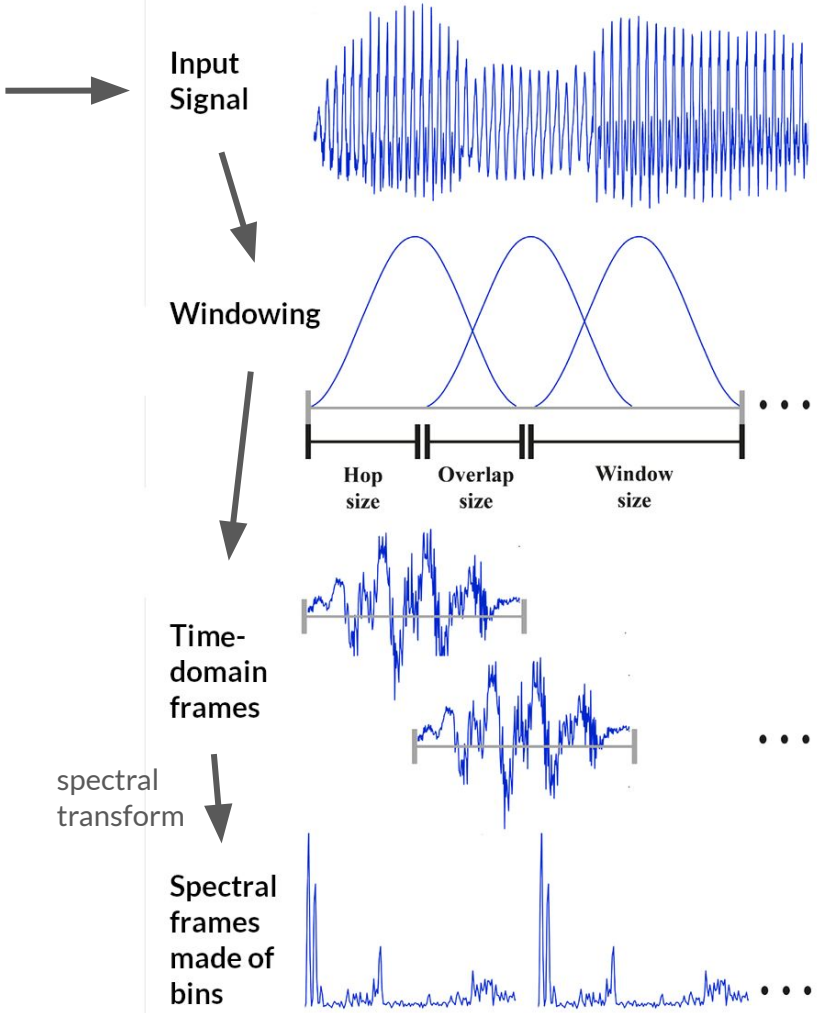
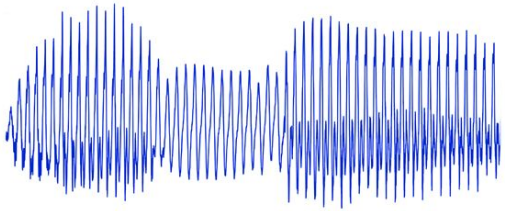
Inner Pitch v1 (2023)

- 17ms latency
- Competitive and written in D 
- All the State of the Art algorithms are spectral aka **Phase Vocoders**

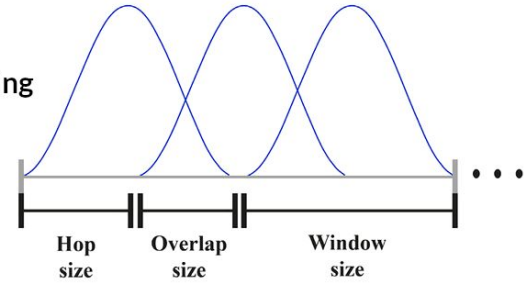

What does a
Phase Vocoder
do anyway?


Input Signal
Windowing
Hop size
Overlap size
Window size
Time-domain frames
spectral transform
Spectral frames made of bins

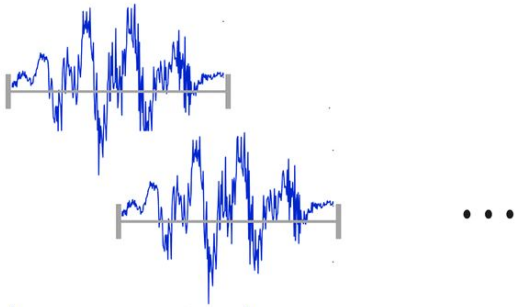
Input Signal



Windowing

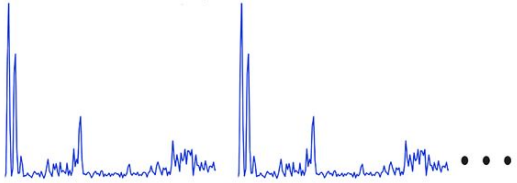


Time-domain frames

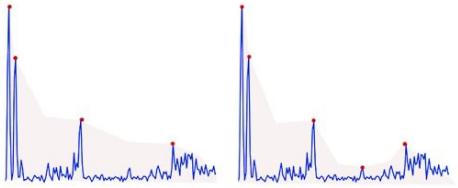


spectral transform

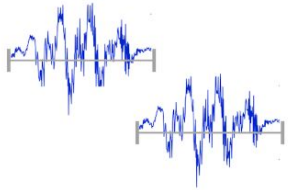
Spectral frames made of bins



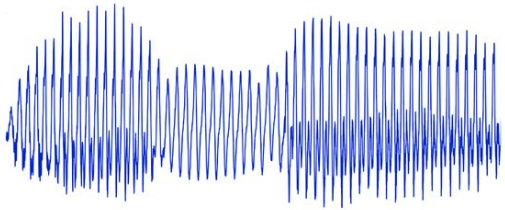
Phase vocoder algorithm

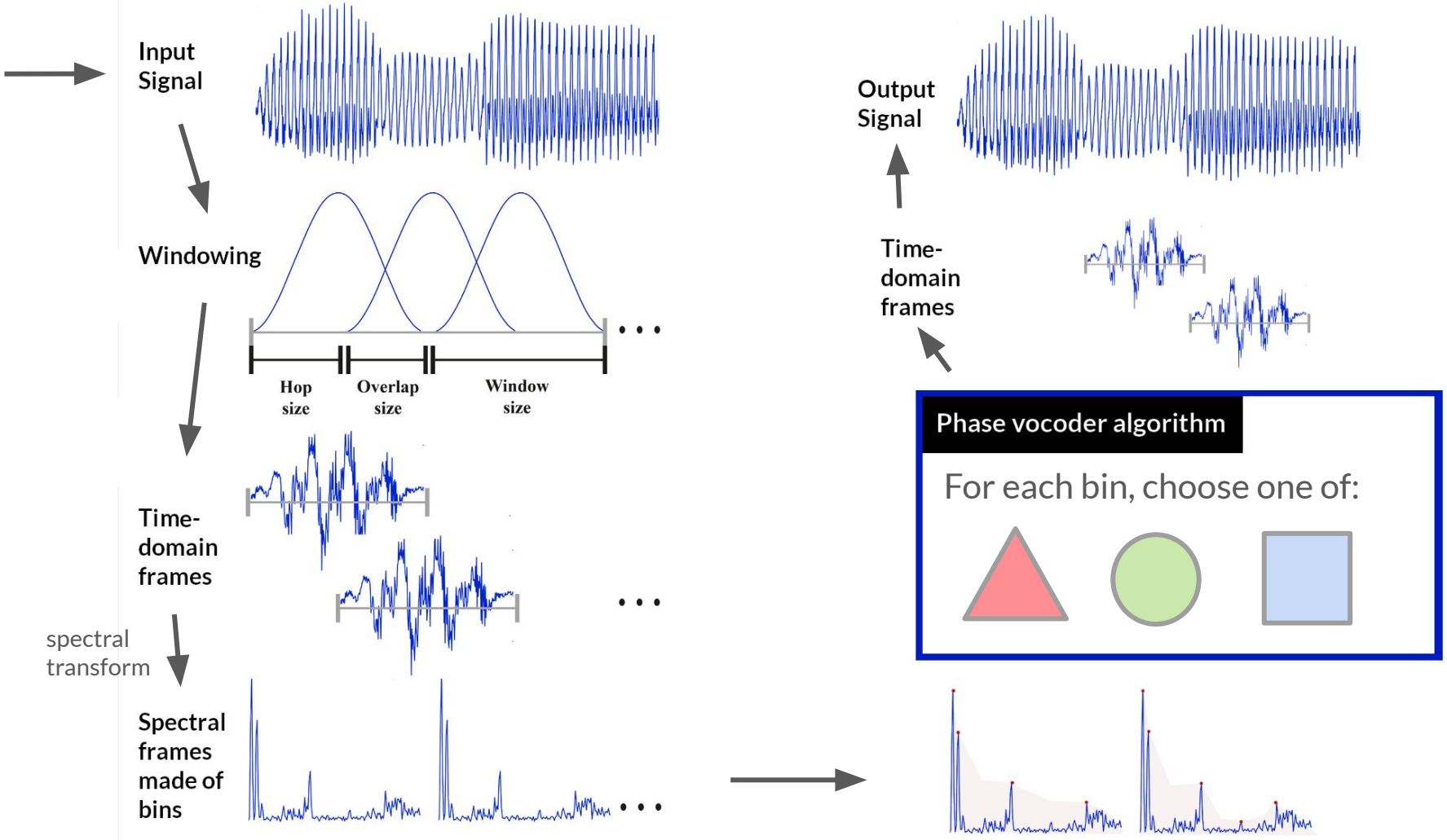


Time-domain frames



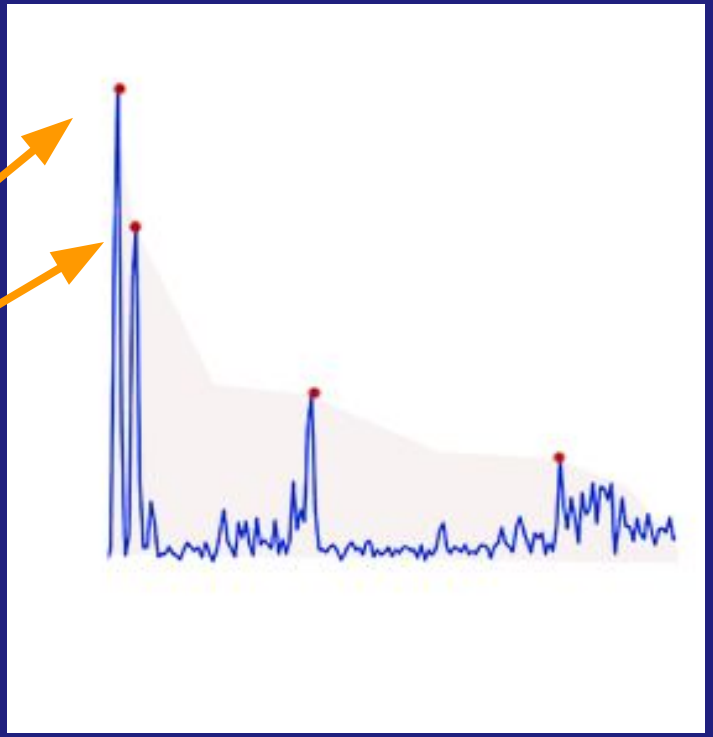
Output Signal




Input Signal
Windowing
Hop size
Overlap size
Window size
Time-domain frames
spectral transform
Spectral frames made of bins
Phase vocoder algorithm
For each bin, choose one of:
Time-domain frames
Output Signal

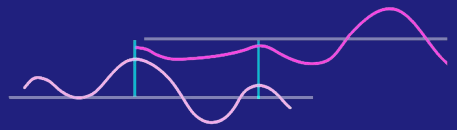
Phase propagation in the Phase Vocoder

Foreach bin in the spectral frame



Phase propagation in the Phase Vocoder

HORIZONTAL CHOICE
Favors continuity with previous frame.



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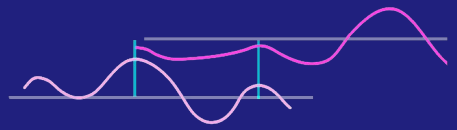
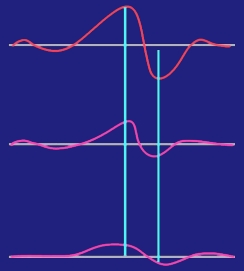


VERTICAL CHOICE
Do like strong neighbour bins do.



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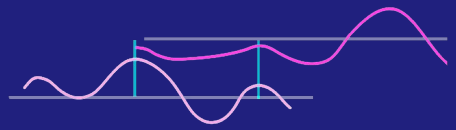

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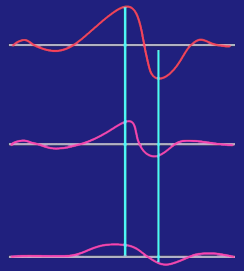
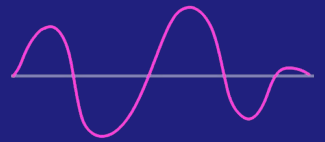


TRANSIENT CHOICE
Favors this bin phase information.
Ignore neighbours or previous frame.



End results in 2023

- Complex algorithm

End results in 2023

- Complex algorithm
- ~100 magic constants this time


```
/// Wins little bit of clarity.
@tuning enum float BLEND_NEAREST_SAMPLING_V1 = 0.05;
```

Such as this one.

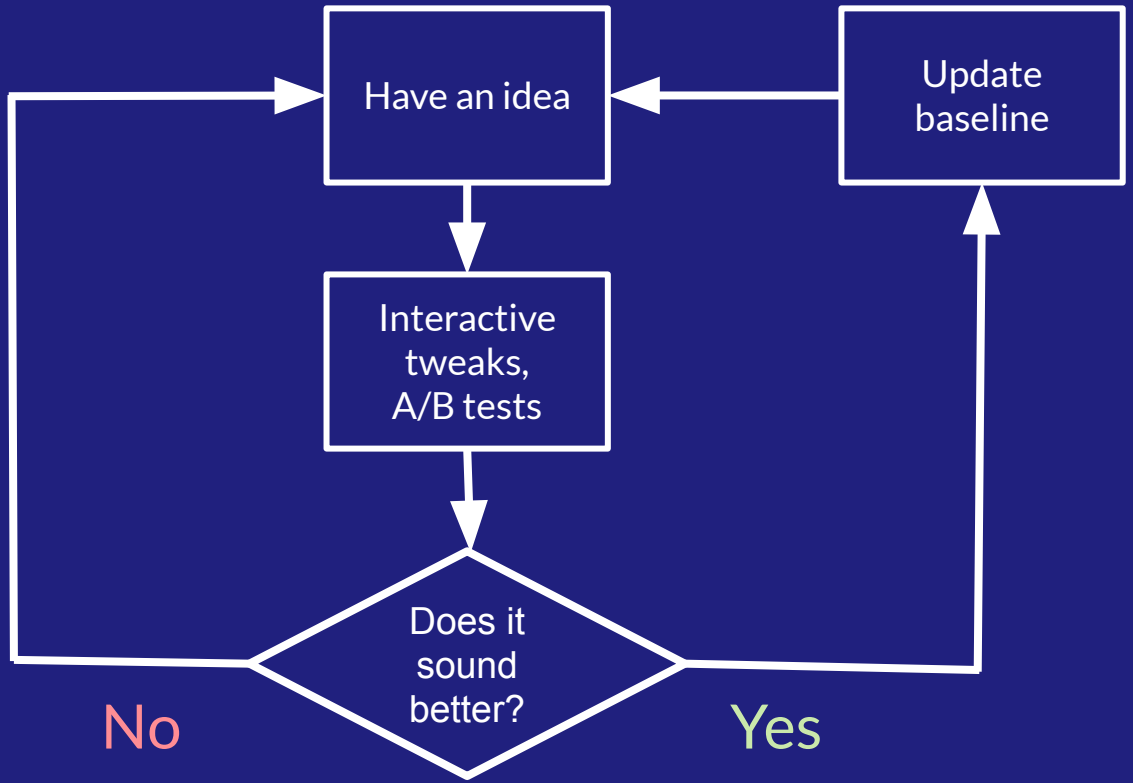
End results in 2023

- Complex algorithm
- ~100 magic constants this time
- Most of the time is spent **making these constants appear**, and **finding good values** for them.


```
/// Wins little bit of clarity.
@tuning enum float BLEND_NEAREST_SAMPLING_V1 = 0.05;
```

Such as this one.

Tuning process



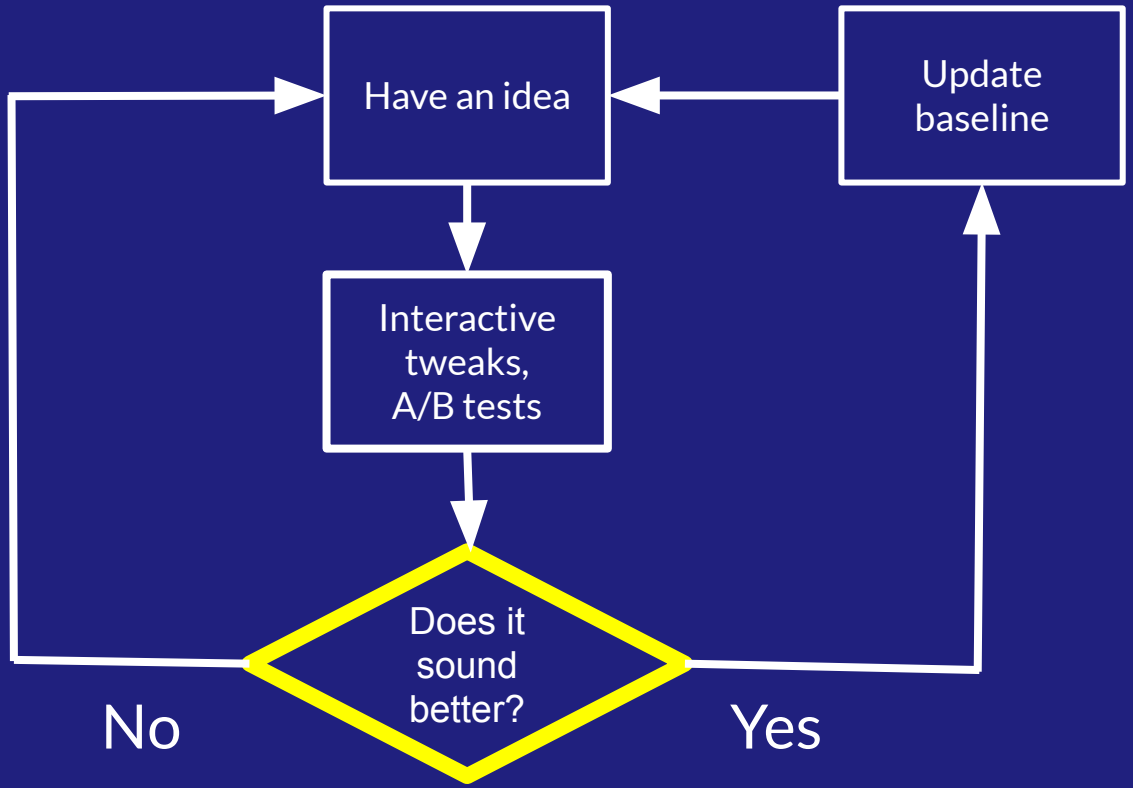
A sort of manual gradient descent

- Need to retune already tuned constants.
- Sometimes need to kill bad concepts and step back in sound quality.
- Some constants are “covering up” bad values of other constants

A sort of manual gradient descent

- Need to retune already tuned constants.
- Sometimes need to kill bad concepts and step back in sound quality.
- Some constants are “covering up” bad values of other constants
- Human audition degrades with age

Replace this step?




We are living in the future.

OPTICOM
THE PERCEPTUAL QUALITY EXPERTS

Efforts initiated in 1994 by the ITU-R to identify and recommend a method for the objective measurement of perceived audio quality culminated in 2001 with recommendation BS.1387 [1], most commonly known as the Perceptual Evaluation of Audio Quality (PEAQ) method. This method is based on generally accepted psychoacoustic principles and has successfully been adopted by the perceptual audio codec development and the broadcasting industries [2].

PEAQ (1998)

2020 Twelfth International Conference on Quality of Multimedia Experience (QoMEX)

ViSQOL v3: An Open Source Production Ready Objective Speech and Audio Metric

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dia Audio
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sco, USA
s@google.com

Nikita Gureev
Hangouts Meet
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feargusog@gmail.com, andrew.hines@ucd.ie

ptual quality in audio and speech thods. The combined v3 release of r speech and audio, respectively.) revious versions, in terms of both source C++ library or binary OL can now be deployed beyond action usage. The feedback from oogle has helped to improve this cases where it is most applicable, s. The new model is benchmarked luation purposes. The trends and ussed.
io quality assessment, mean opin- udio, ViSQOL, PESQ, POLQA,

the waveform by sampling from a distribution of learned parameters. One example is the WaveNet-based low bitrate coder [10], which is generative in nature. There are other DNN-based generative models, including SampleRNN [11] and WaveGlow [12], with promising results that suggest that this trend will continue. These generative models typically do not lend themselves to being analyzed well by existing full reference speech quality metrics. While the work described in this paper does not propose a solution to the generative problem, the limitations of the current model should be acknowledged to encourage development of solutions.

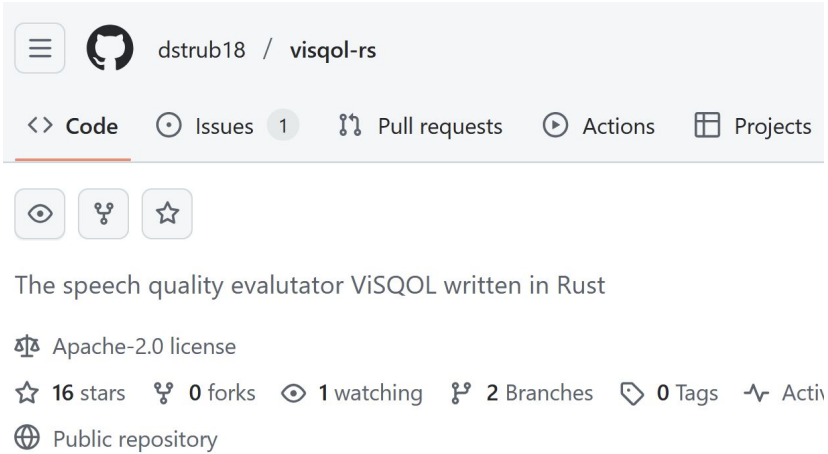
ViSQOL was originally designed with a polynomial map-

ViSQOL v3 (2020)

Perceptual objective measures are a thing

Settled on **visqol-rs**

- written in **Rust**
- based upon Google research and **AI** model



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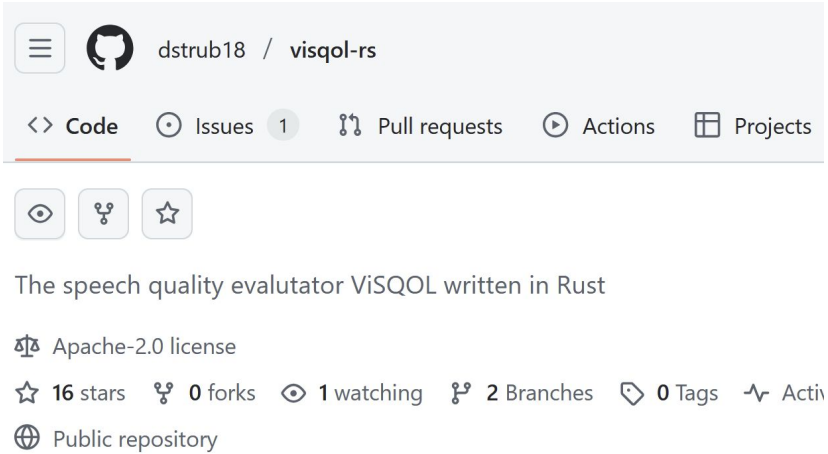
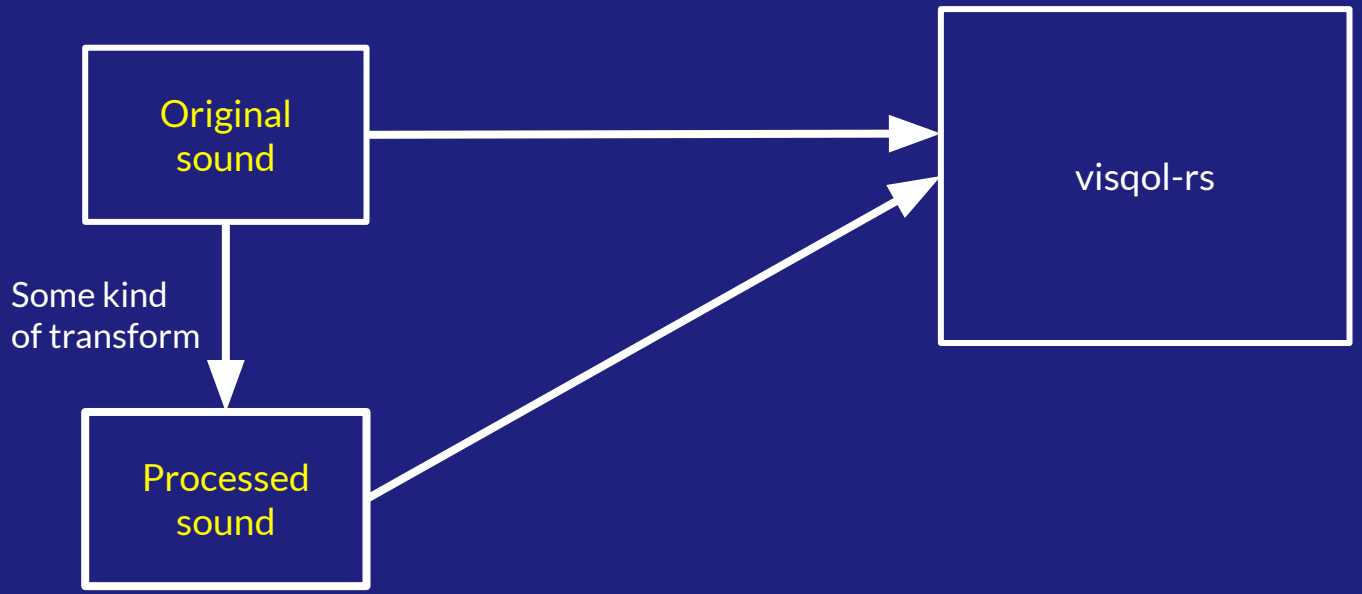


My stuff is finally running through a neural network!

How would a machine know what is “good sound”?

Audio objective measures



Audio objective measures

That, I evaluated.

Original sound

Some kind of transform

Processed sound

visqol-rs

"Processed sound is

94.65%

like the original,

according to AI."

Audio objective measures

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Original sound

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"Processed sound is

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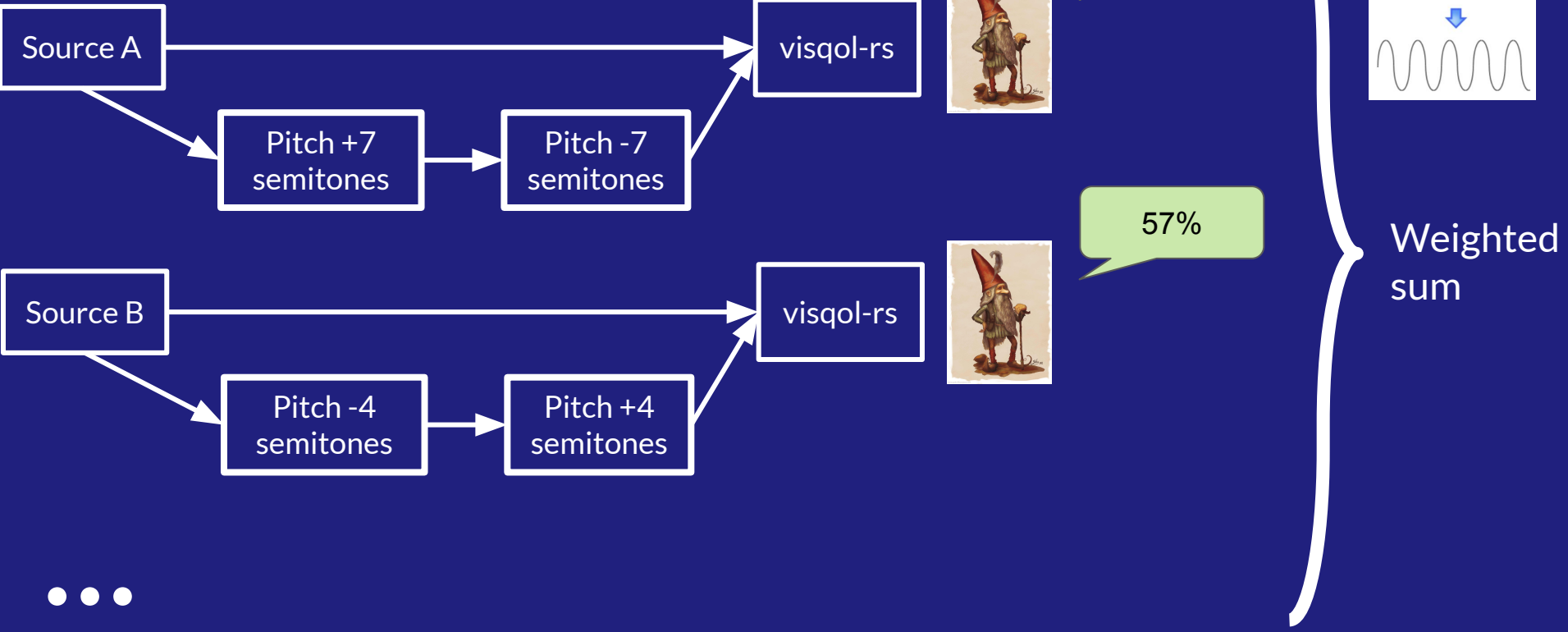
In the codec sense, quality is meant as reproduction.

Pitch-shifting objective measure



Yes, but…

You’re going to want to modify all my magic constants, right?

```
enum float NOISE_BINS_ANGLE_EXTENT = 0.9922;
```

You’re going to like… modify all my magic constants, right?

Well, yeah. That’s what optimization does.

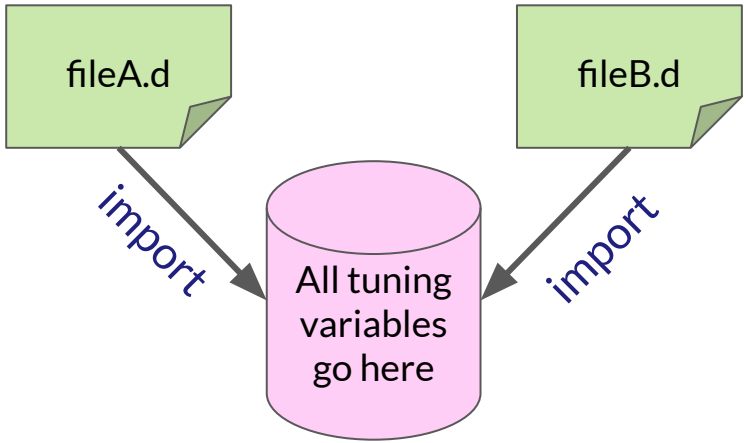
constants, right?

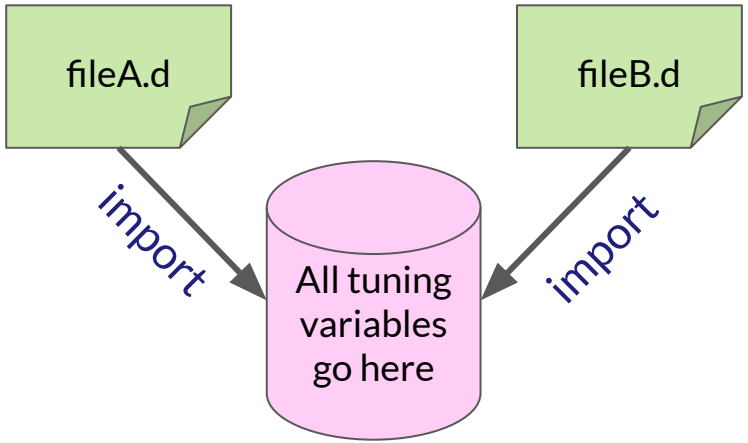
Well, yeah. That's what optimization does.

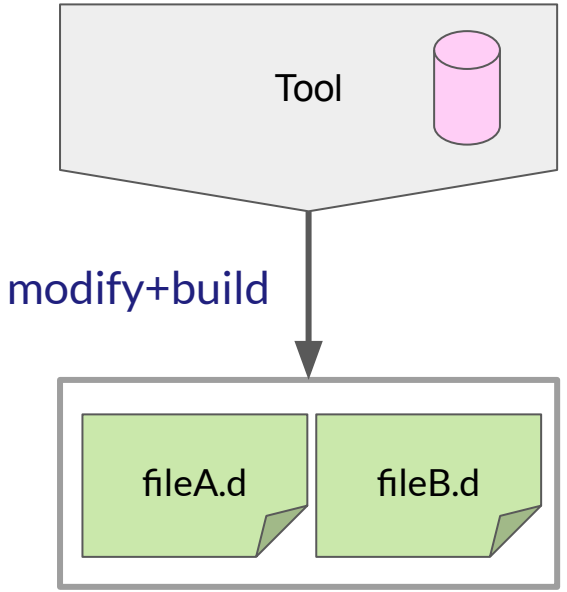
Take a look at the codebase. There is no way to put them all in one place.

Take a loot at the codebase. There is no way to put them all in one place.

I see. We'll find a way to just add some UDAs.


fileA.d
fileB.d
import
import
All tuning variables go here





Instead of gathering all parameters in one single place:

- 1. Parse source file
- 2. Regenerate source code
- 3. Rebuild and evaluate

```
 ___                 _
| __|__ __ ___ | |__ __  ___
| _| \ V // _ \| |\ V / / -_)
|___| \_/ \___/|_| \_/  \___|
```

Let's present evolve,
a solution for this problem.

➡️ WHAT'S THIS?

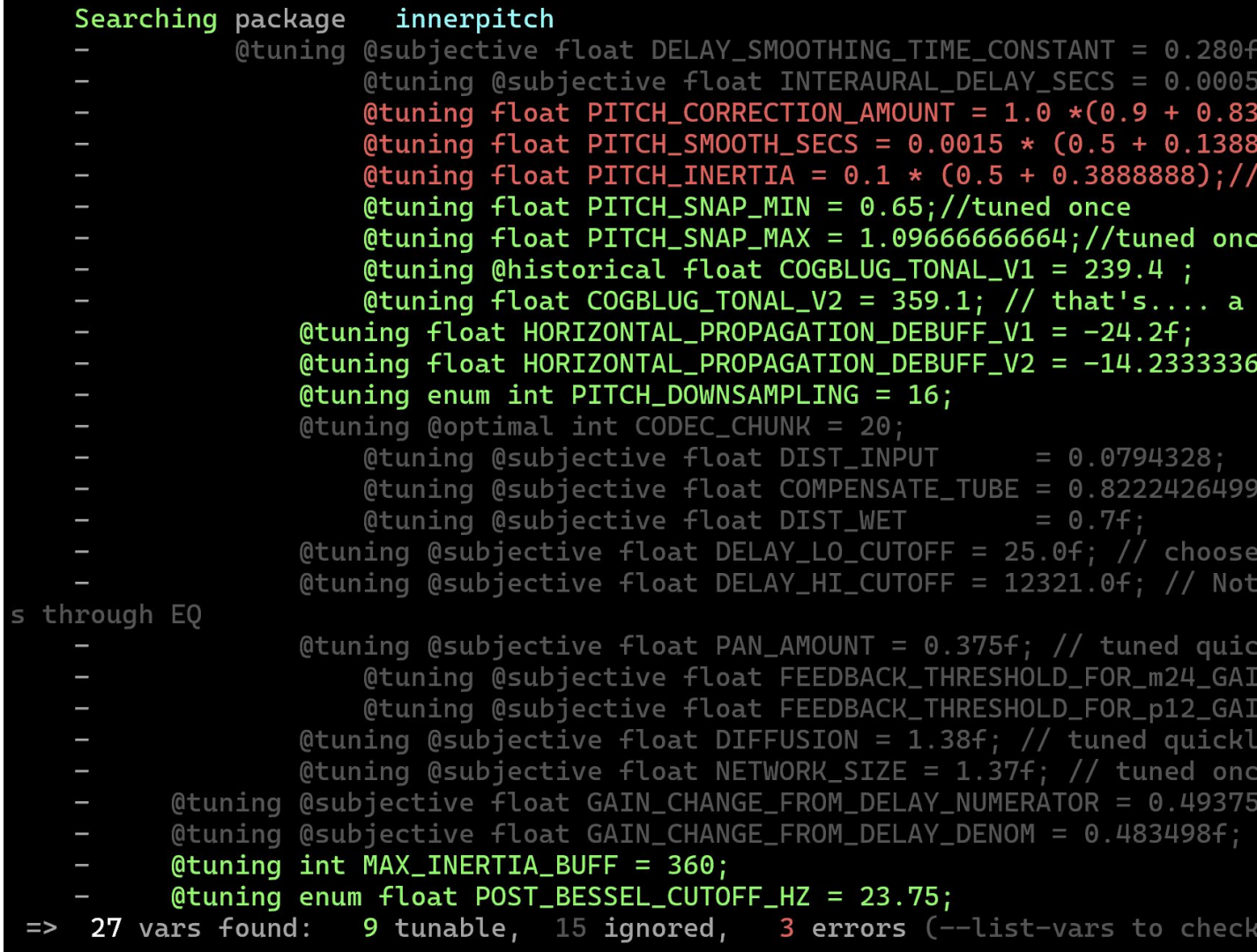
✨evolve✨ optimizes your magic constants with **gradient descent**.

➡️ HOW IT WORKS

✨evolve✨ builds a **D program** repeatedly while changing float/double non-array variables and constants, marked with the @tuning user-defined attribute (called *variables* below).

```
// ---------------- source.d -----------------
import dplug.dsp.udas;
@tuning float MY_MAGIC_CONSTANT0       = 0.10;
@tuning double MY_MAGIC_CONSTANT1      = 0.28;
@tuning enum float MY_MAGIC_CONSTANT2  = 0.30;
@tuning enum double MY_MAGIC_CONSTANT3 = 0.45;
```

evolve tool can list tuning variables and has semantic UDAs for ignoring some if needed.



Bring your own fitness measure.

```
For evolution it needs a ⚡fitness measure⚡to evaluate each build.
✨evolve✨ runs from within a DUB project directory, and uses the git
working copy as temporary state.
https://code.dlang.org/
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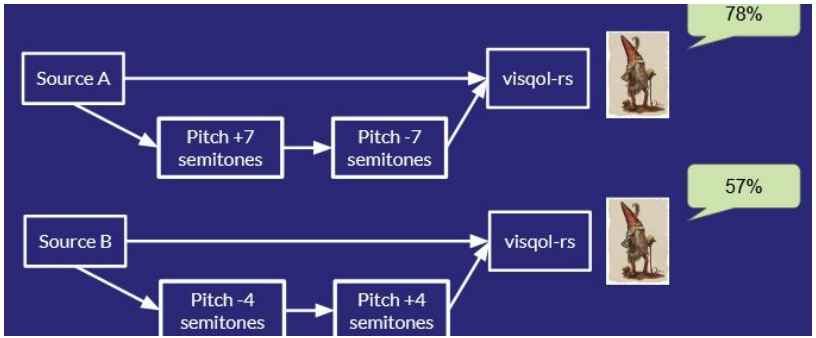


In pitch-shifting case

Bring your own fitness measure.

For evolution it needs a ⚡fitness measure⚡to evaluate each build.
✨evolve✨ runs from within a **DUB project** directory, and uses the **git** working copy as temporary state.
https://code.dlang.org/

Fitness program must return a **fitness.xml** file with one number.

```
// Write final XML

File results = File("fitness.xml", "w");
results.writeln(`<?xml version="1.0" encoding="UTF-8"?>`);
results.writeln(`<results>`);
results.writefln(`    <metric name="dummy" value="%.13f" />`, totalFitness);
results.writeln(`</results>`);

return 0;
```

```
// ------------------------ evolve.xml ------------------------

<?xml version="1.0" encoding="UTF-8"?>
<training>
  <!-- Both these variable will be evolved -->
  <var name="MY_VAR" />
  <var name="MY_OTHER_VAR" />

  <!-- How to build and ⚡evaluate⚡ the program -->
  <fitness-command>mytest -param</fitness-command>
  <build-command-windows>dub -b release</build-command-windows>
  <build-command-macos>dub</build-command-macos>

  <!-- Ignored package for parsing variables -->
  <exclude-package name="gamut" />
  <exclude-package name="dplug:dsp" />
</training>
```

Configuration file.

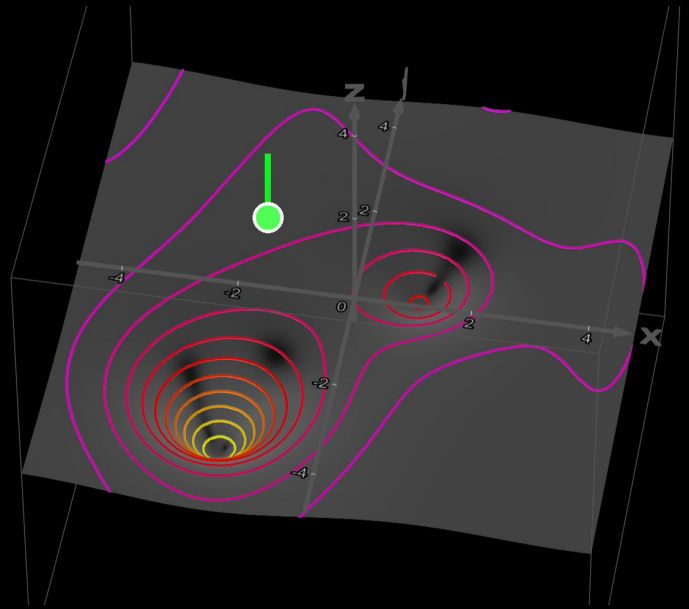
Tool here: https://github.com/AuburnSounds/Dplug/tree/master/tools/evolve

4 search algorithms in evolve tool

```
here        Compute fitness here, with current local changes.
            Do not change working copy.
```

```
evolve -a here
```

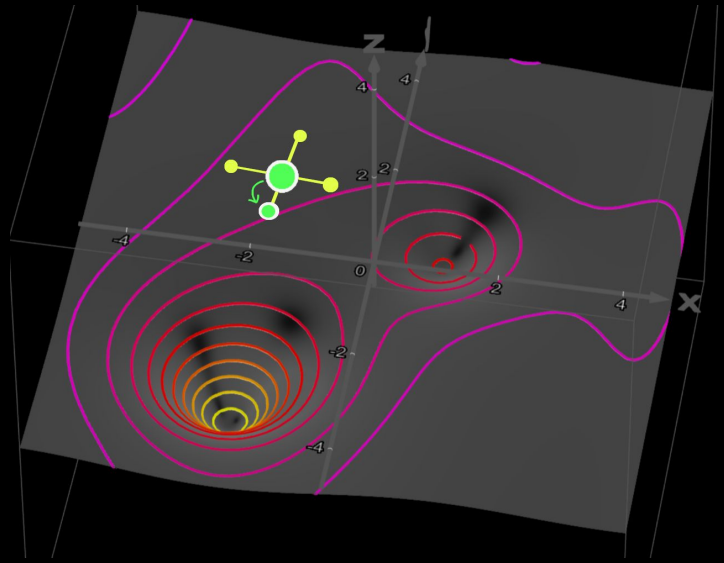
= Just displays current fitness.



4 search algorithms in evolve tool

```
gradient  Use --pattern search here, then change the best
          variable once all are evaluated.
```

Pattern search in each dimension. Pick single best improvement.

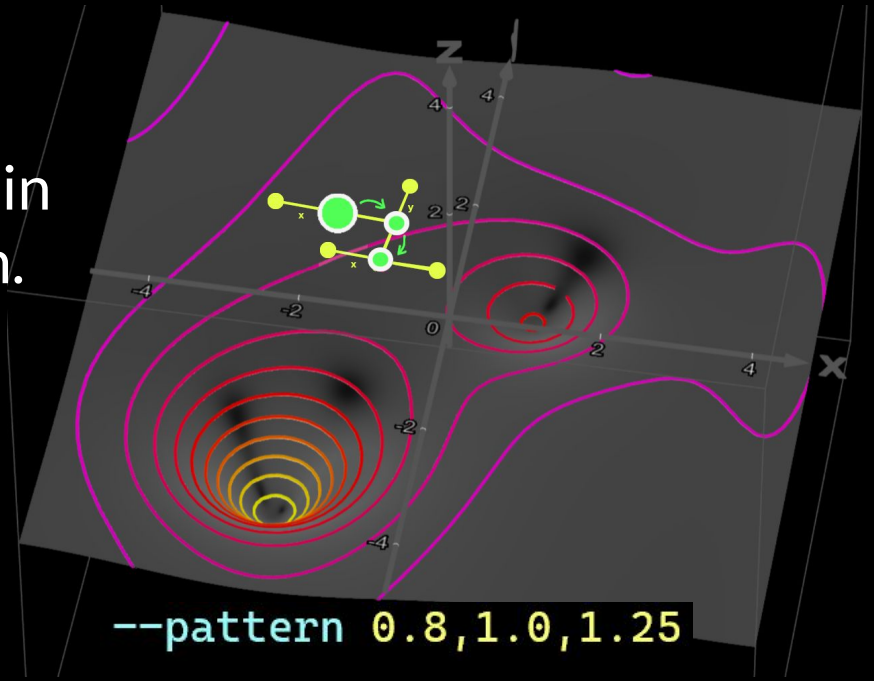



```
--pattern 0.8,1.0,1.25
```

4 search algorithms in evolve tool

```
whirlpool Use --pattern search here, then change each
          tested variable immediately after evaluation.
```

Pattern search in each dimension. Change immediately to better value.

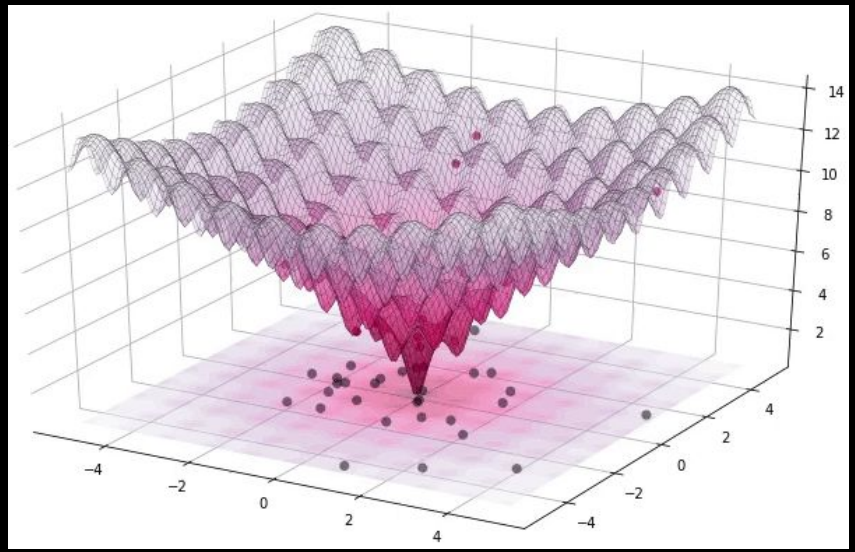


4 search algorithms in evolve tool

```
diffevol  Use "Differential Evolution" algorithm.
          https://en.wikipedia.org/wiki/Differential_evolution
```

A famously simple metaheuristic optimization algorithm.

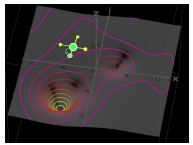
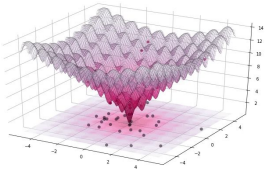
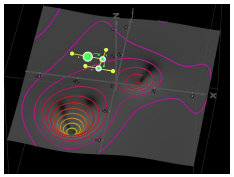
Typical population > 15



(source = https://pablormier.github.io/2017/09/05/a-tutorial-on-differential-evolution-with-python/)

QUIZZ QUESTION

What will be the most useful algorithm in practice?

- A. gradient 
- B. differential evolution 
- C. whirlpool 

QUIZZ ANSWER

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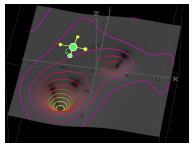
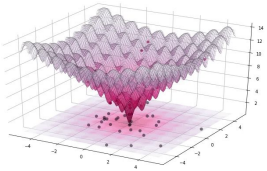
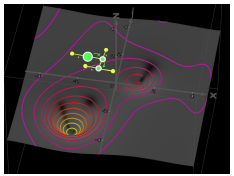
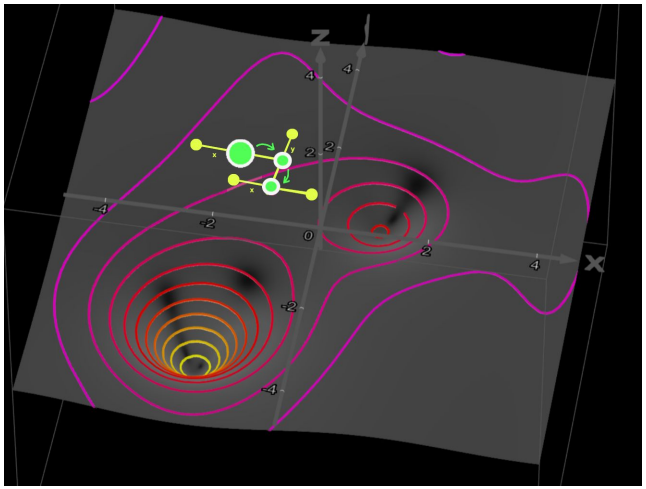
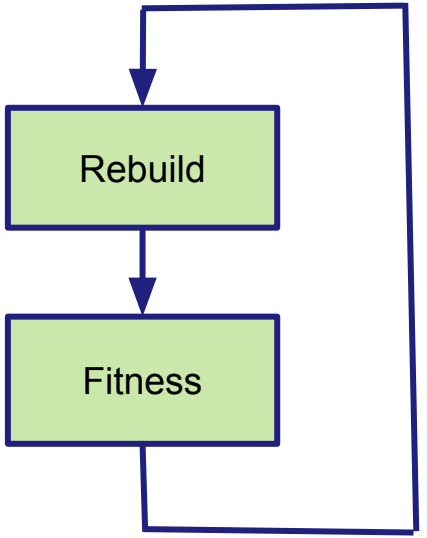
- A. gradient 
- B. differential evolution 
- **C. whirlpool** 

Build times are pretty slow

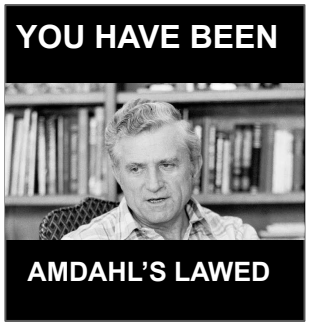
“whirlpool” method just makes less evaluations and **move on.**

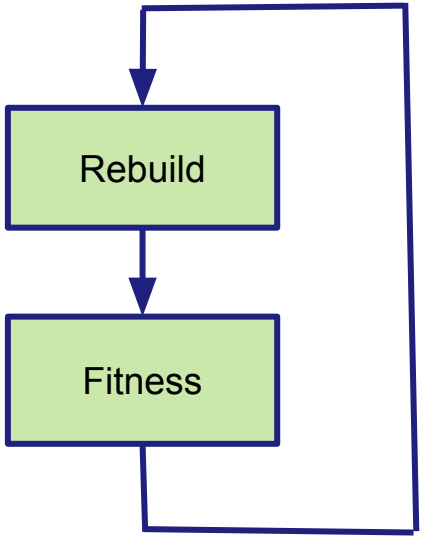




Build times are pretty slow

- Fitness evaluation may be fast, but it doesn't matter since rebuilding is rather slow.
- Might as well have a **slow fitness evalution**





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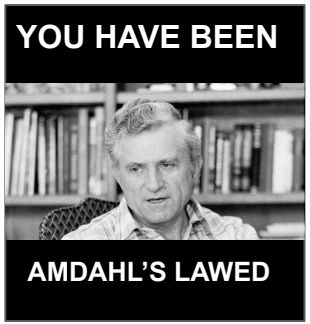
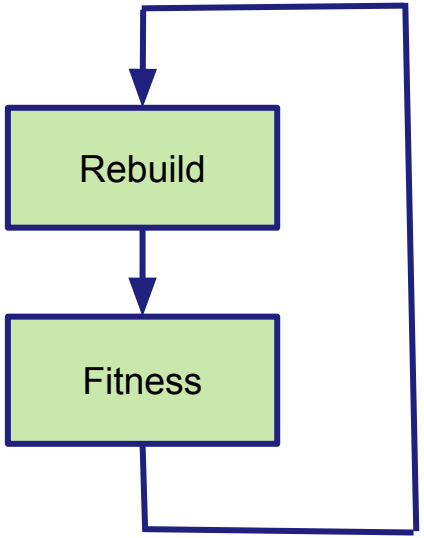


The evolve tool is applicable
where the fitness evaluation
is slow,
such as perceptual measures.

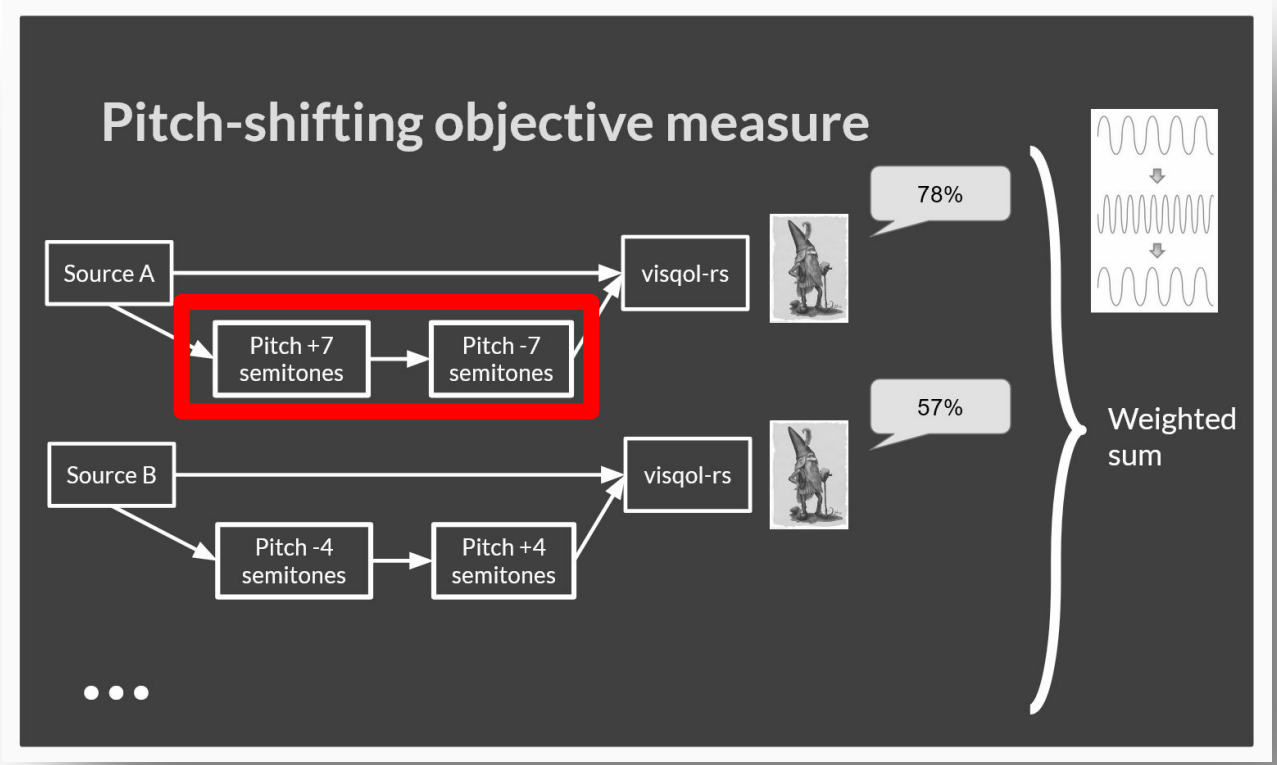


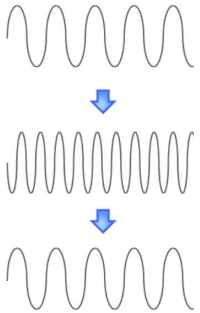
Just 26 man-days after starting the automatic optimization effort,

Just 26 man-days after starting the automatic optimization effort, sound quality had actually decreased.

Everything that went wrong

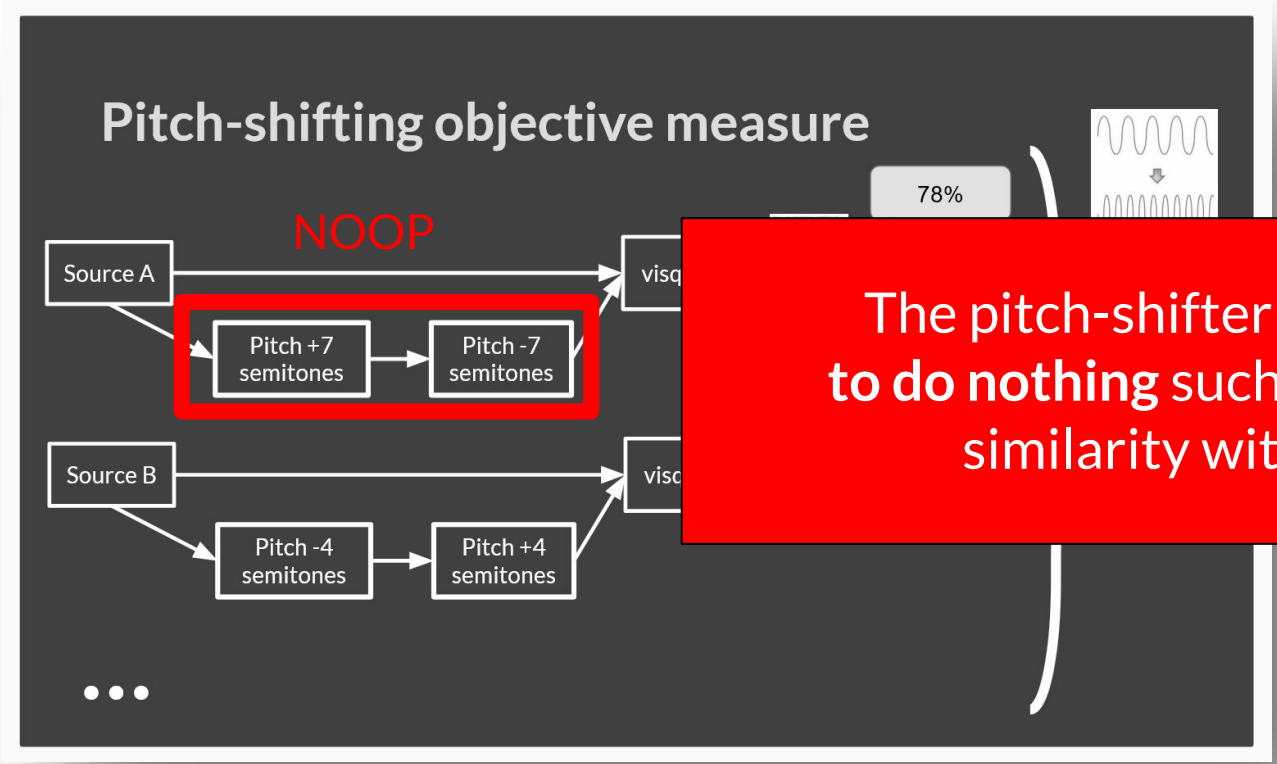
A. Remember slide 33?



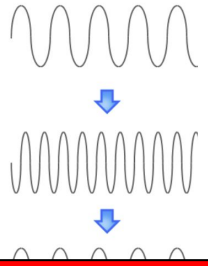


We proposed to shift +N then -N to return close to the original sound.

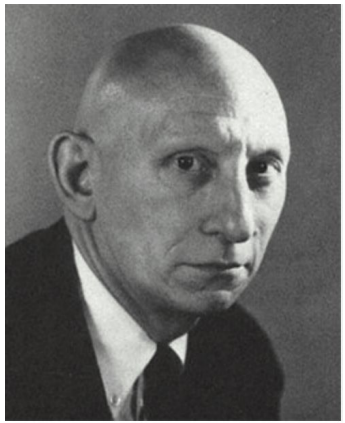
A. Remember slide 33?



The pitch-shifter quickly learnt **to do nothing** such as to maximize similarity with original.



then -N
to return close
to the original
sound.



When we write programs that "learn", it turns out that we do and they don't.

— Alan Perlis

B. Remember slide 19?

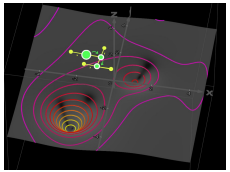
End results in 2023

- Complex algorithm
- ~100 magic constants this time


```
/// Wins little bit of clarity.
@tuning enum float BLEND_NEAREST_SAMPLING_V1 = 0.05;
```

Such as this one.

We said to have many parameters to evolve, and that whirlpool was used.



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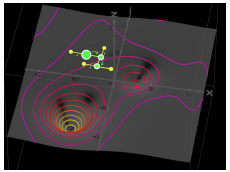
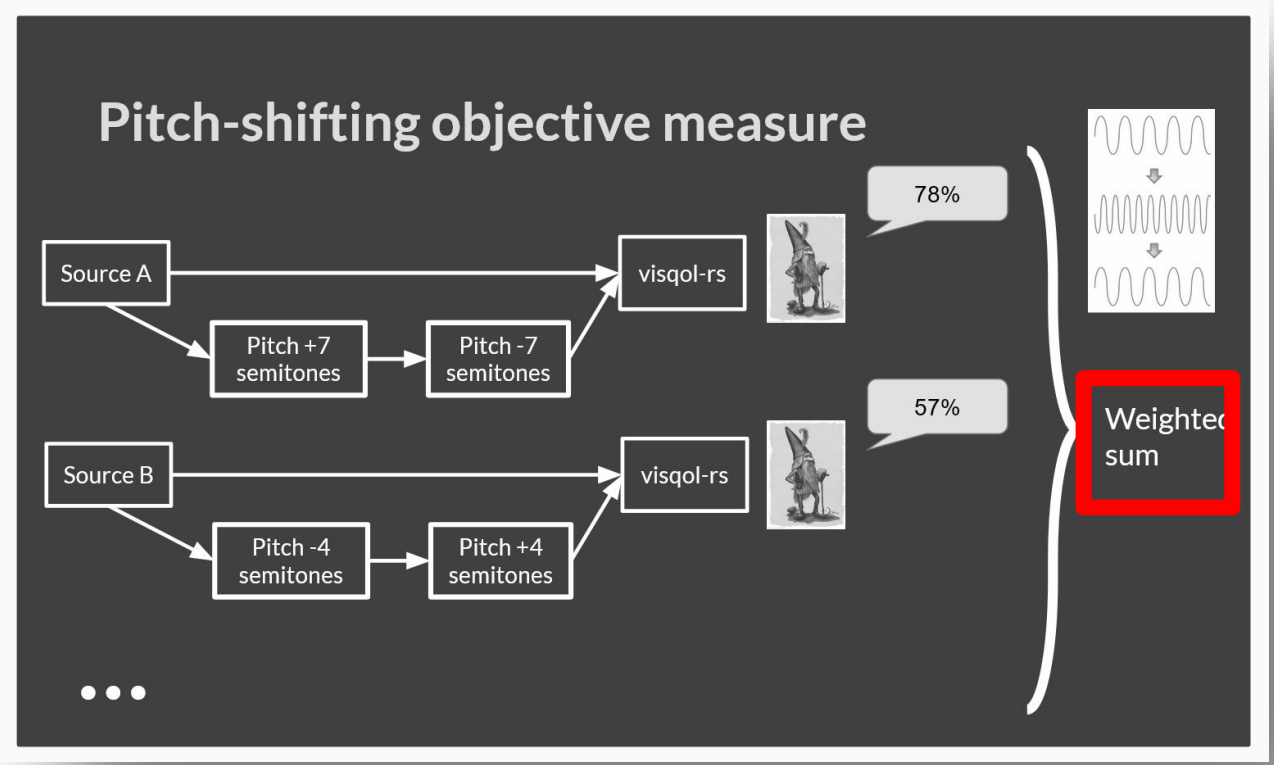


HIGH DIMENSIONS ARE NOT ADVISED

Easy to make minor “progress” indefinitely.

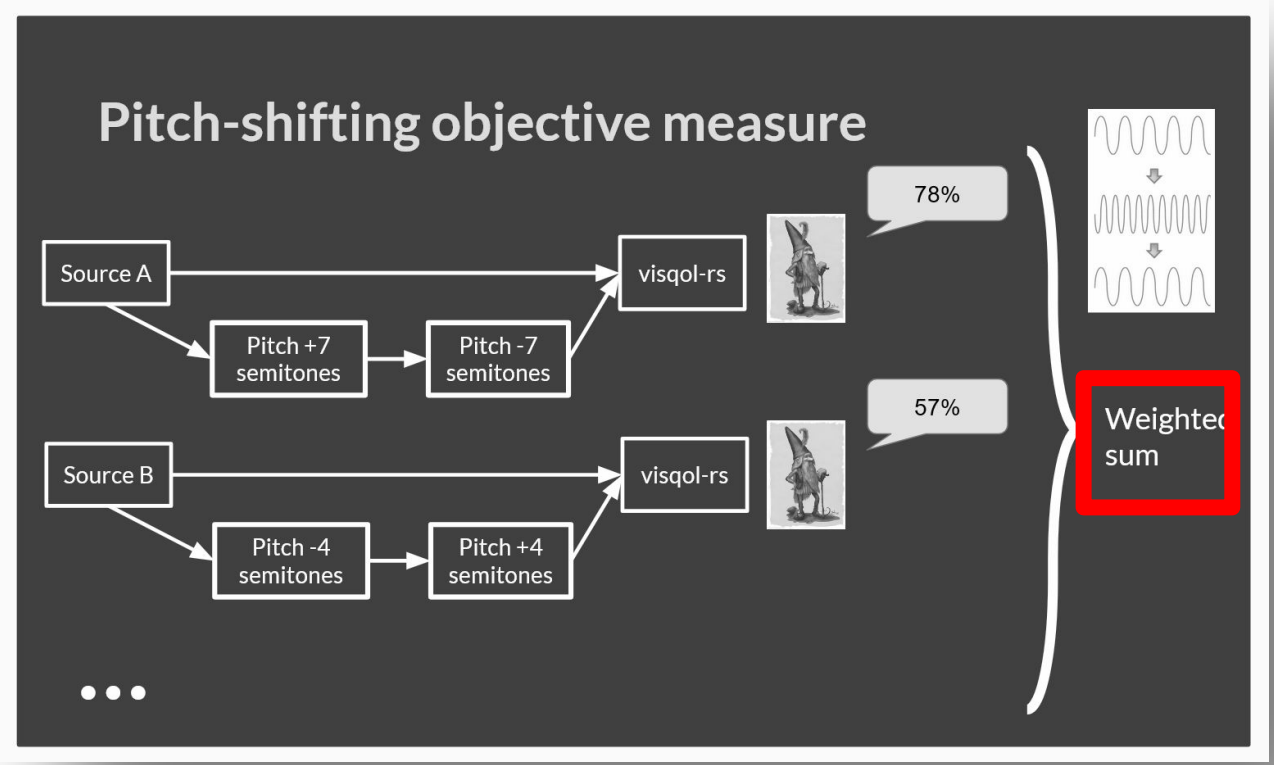
C. Again slide 33



Fitness evaluation used 11 meaningful and different sources to compute the ViSQOL v3 measure.

Each source is used for 4 different shifting.

C. Again slide 33



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Need more data else overfitting.

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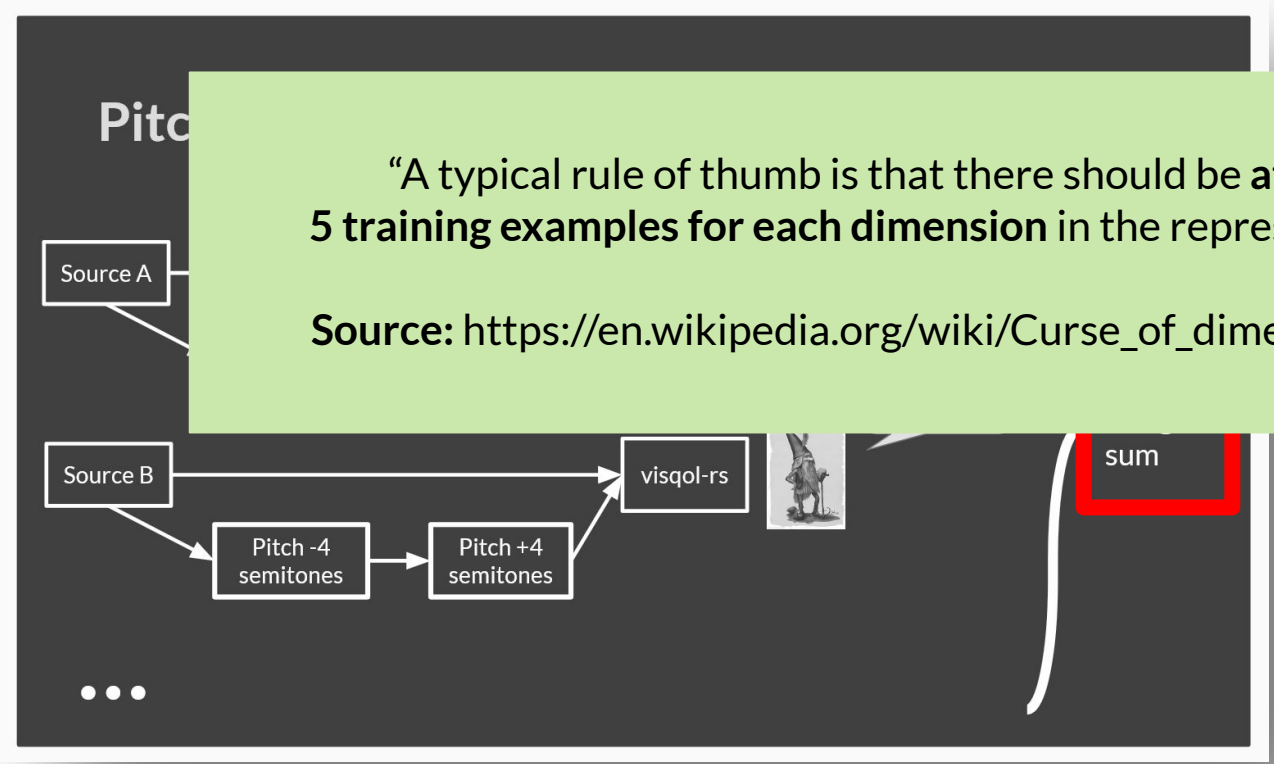

"A typical rule of thumb is that there should be **at least 5 training examples for each dimension** in the representation."

Source: https://en.wikipedia.org/wiki/Curse_of_dimensionality

Fitness evaluation
used 11 meaningful
nt
compute
L v3

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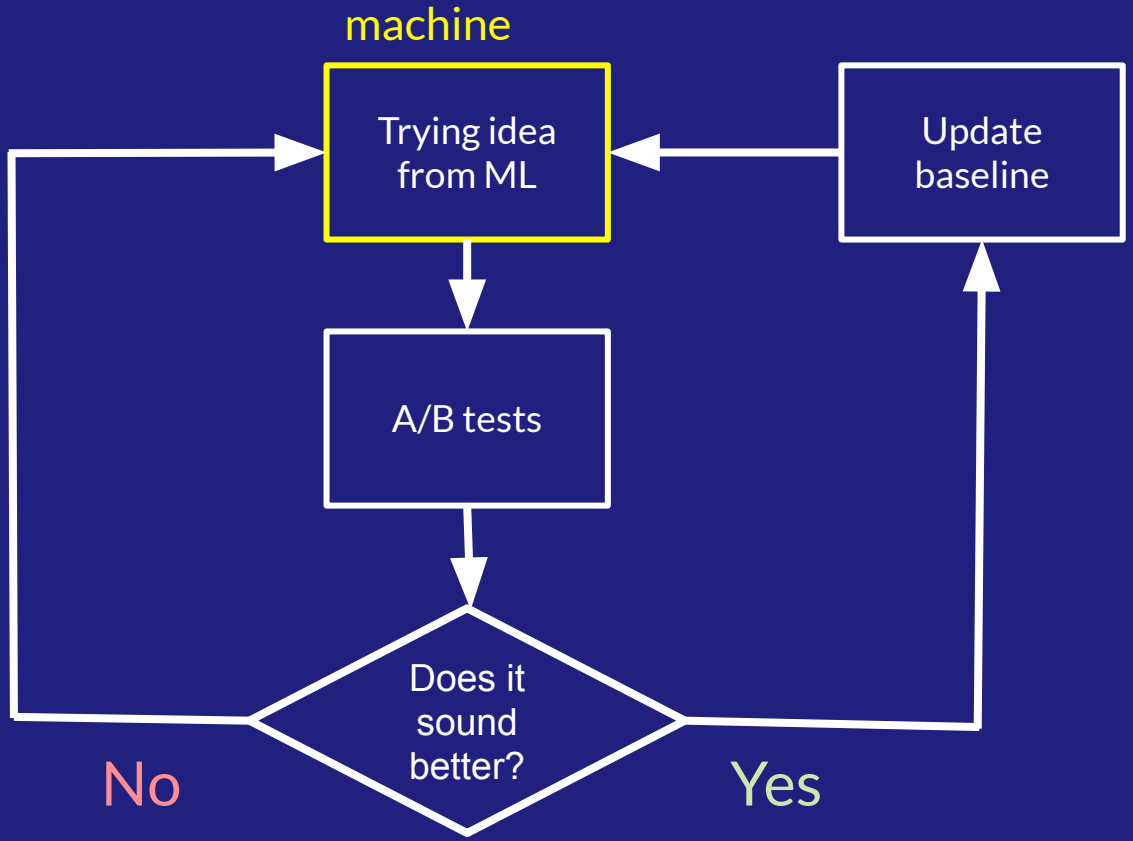
Epilogue

Had to assess each change manually



- Shipped in Inner Pitch v2 (Feb 2025)

Tool is on GitHub

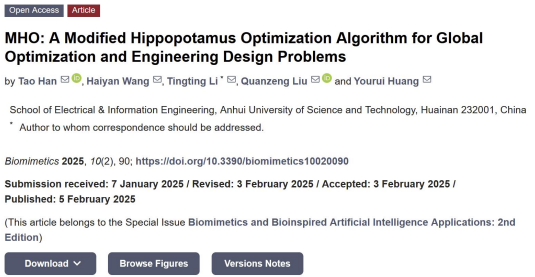
https://github.com/AuburnSounds/Dplug/tools

Questions?

Bonus slide

- Hippopotamus optimization
- Squid Game Optimizer
- Political Optimizer
- Emperor Penguins Colony
- Dujiangyan Irrigation System
- Cuckoo Optimization Algorithm
- Cuckoo Search

All are real meta-heuristic algorithms.



Open Access Article

MHO: A Modified Hippopotamus Optimization Algorithm for Global Optimization and Engineering Design Problems

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Hippotamus algorithm, one of the most downloaded papers of 2024.

Bonus slide

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