

# Bless You!

- a CBR based sneeze detector
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# Background

- Early indications of events in the society.
- One such event is the outbreak of an influenza.
- Sneeze detectors can early detect such an outbreak.

# Use Case

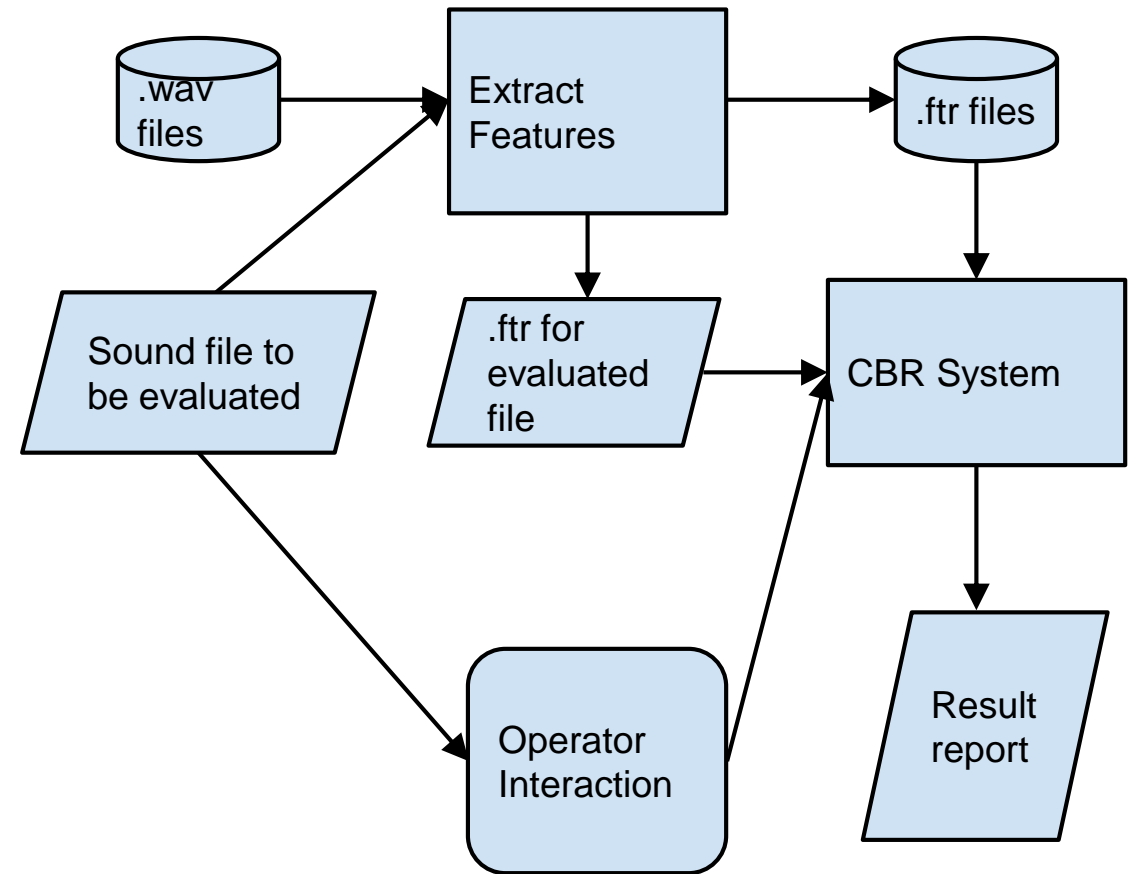
- A microphone listens to the sound in a public place.
- When it detects that someone sneezes a counter is incremented.
- A supervisory system reads the sneeze count at cyclic intervals.
- The read counter values can be used to detect if a flu is in progress.

# Our Program

- Limitations
  - No continuous evaluation
  - Uses sound samples
  - Console interface

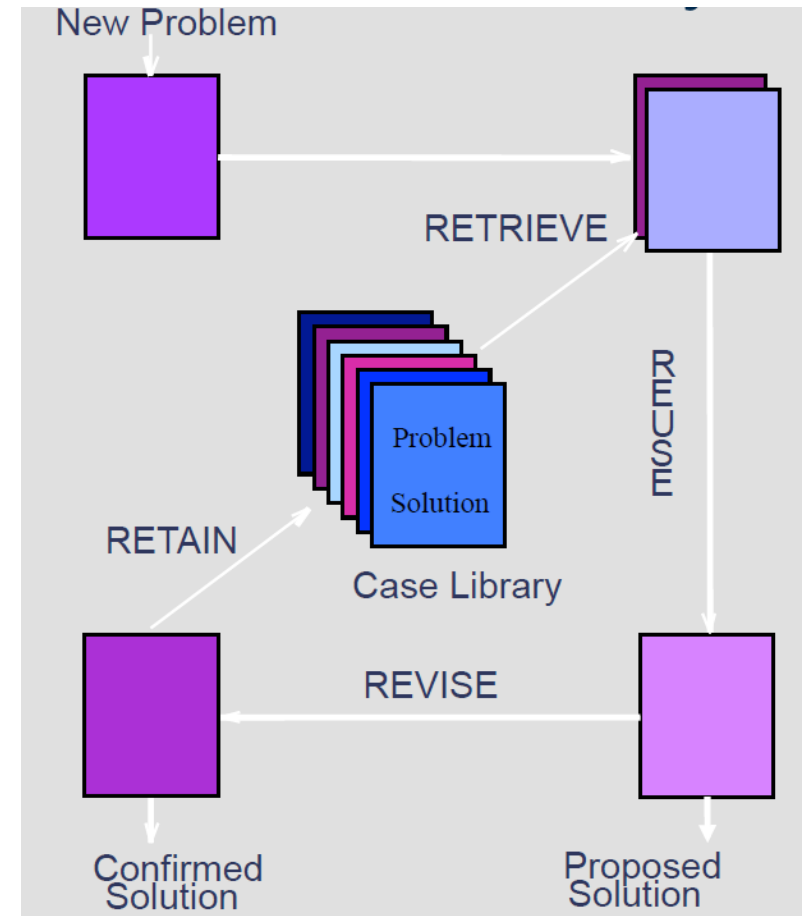
# System Overview

- Command Line Program
  - Batch mode
  - Interactive mode
- About 160 sound samples
  - 100 sneezes
  - 60 none-sneezes
- Text report:
  - Proposed sound type
  - Statistics



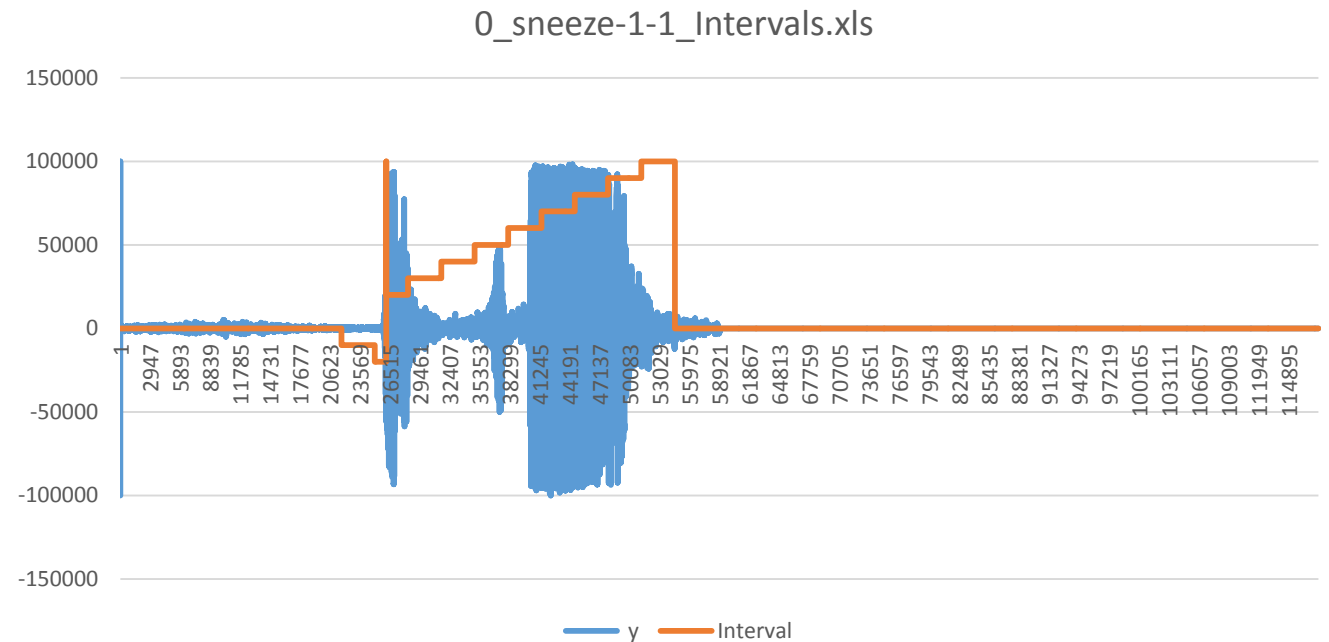
# Classic CBR Overview

- New Problem: sound sample
- Case Library:
  - Confirmed sneezes (50)
  - Confirmed none-sneezes (50)
- Proposed Solution:
  - Proposed Sneeze
  - Proposed None-Sneeze



# Extract Features from Sound Sample

- .wav-files PCM, 44.1 kHz, 16 bits, 1 or 2 channels (mono/stereo)
- Stereo -> Mono
- Normalize to 100000
- Analyse
  - Trigger On
  - Trigger Off
  - Split into intervals



# Feature types per Interval

- Peak values
- Average values
- RMS values (effect)
- Peak to peak values
- CF - Crest Factor values
- PZ - Passage through Zero values
- FFT values (Fast Fourier Transform)
  - Sample length  $2^{12}$
  - Sample length  $2^{14}$
  - Sample length  $2^{16}$

$$x_{\text{rms}} = \sqrt{\frac{1}{n} (x_1^2 + x_2^2 + \cdots + x_n^2)}.$$

$$C = \frac{|x|_{\text{peak}}}{x_{\text{rms}}}.$$

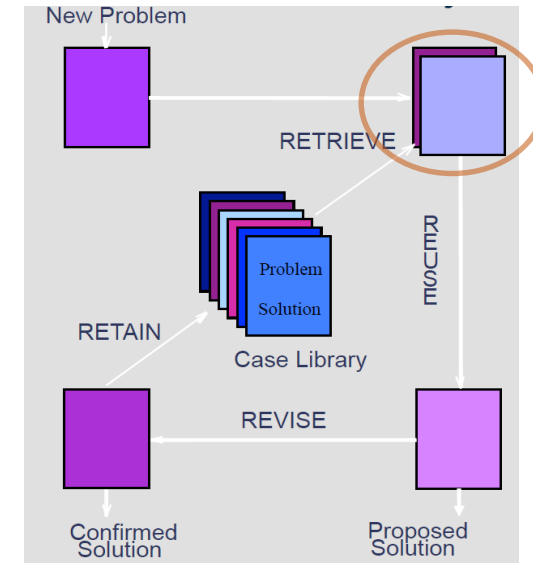
$$X_k = \sum_{n=0}^{N-1} x_n e^{-i2\pi k \frac{n}{N}} \quad k = 0, \dots, N-1.$$

Feature values normalized within intervals (0.0 ... 1.0)



# CBR System

- Retrieve (find best SF value cases)
  - Compare new problem ( $N$ ) to case library cases ( $R$ ).
  - Calculate Similarity Function ( $SF$ ) over all features and intervals.



- $d_k(N, R) = \sum_{i=0}^n |n_i - r_i|$   $i = \text{interval}$

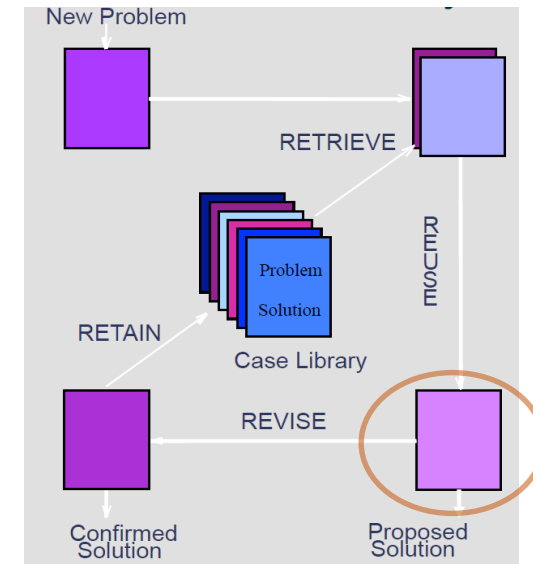
- $f_k(N, R) = \frac{1}{1 + d_k(N, R)}$   $(0.0 \dots 1.0)$

- $SF(N, R) = \sum_{k=0}^n w_k * f_k(N, R)$   $k = \text{feature type}, w_k = \text{feature type weight}$

- $\sum_{k=0}^n w_k = 1.0$

# CBR System

- Reuse (generate proposed solution)
  - Pick 5 best cases, highest Similarity Function
  - Use majority vote to decide if proposed sneeze/none-sneeze



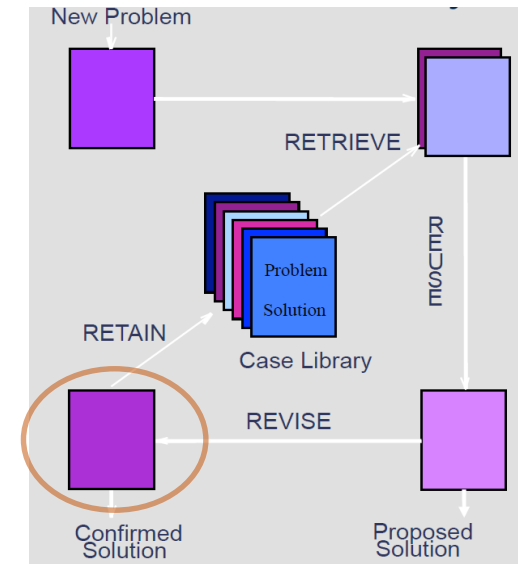
# CBR System - Maintenance

- Revise

- Add a new known case => 101 cases in library
- Iterate each of 101 cases in case library while calculating SF.
  - Obtain worst case
    - Prio 1: case that have participated in voting and voted wrong every time
    - Prio 2: case that never participated in voting and has lowest SF value
    - Prio 3: case that has the lowest SF value

- Retain

- Retain the best cases while preserving sneeze/none-sneeze ratio, i.e. remove the worst => back to 100 cases in library.



# Build Environment

- Visual Studio
- C#

# Demonstration



Ta-da!



# What we learned

- Github (Conflicts!)
- Sound file structure
- FFT (Fast Fourier transform)
- CBR Systems
  - Distance calculation
  - Similary functions

The End

Thanks for listening!