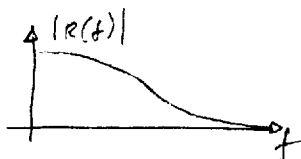


(c) Reflection is a low-pass filter:



- \Rightarrow higher frequency modes are attenuated relative to lowest mode at each reflection
- \Rightarrow for large T , waveform is dominated by fundamental mode \rightarrow sinusoidal standing waves

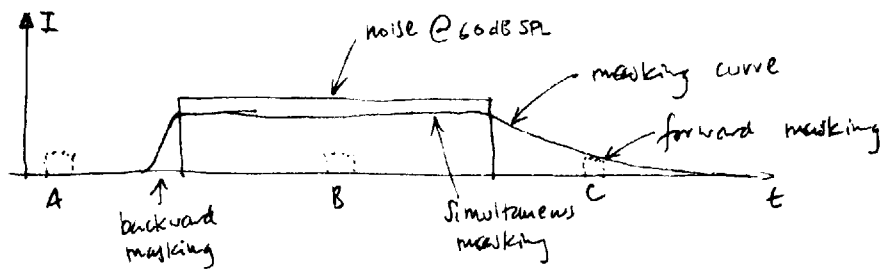


2. (a) A is a narrow-band spectrogram (analysis time window ~ 30 ms)
 B is a wide-band spectrogram (analysis time window ~ 5 ms)
 B also has a lower energy floor in its gray-scale mapping
 eg. A shows energy in range 30-60 dB, B shows 0-60 dB (or whatever)
 hence can see more low-level energy features in B.

- (b) From A, we have a pitched vowel between $t = 0.2 \dots 0.35$.
 Fundamental frequency peak ~ 220 Hz (by dividing freq. of 10th harmonic by 10). Pitch contour rises then falls.
 From B, have noise between $t = 0.1$ and 0.2 , the vowel (formants at 600 Hz, 1400 Hz, 2800 Hz); then a stop, then more fricative (sibilant). [Actually, the speech is "six"]

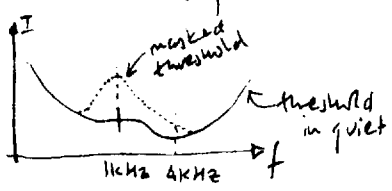
- (c) A shows harmonics and hence pitch, but pitch is ignored in English ASR.
 B shows the formants, which are the important features for ASR
 \rightarrow B is preferable for speech recognition.

3.



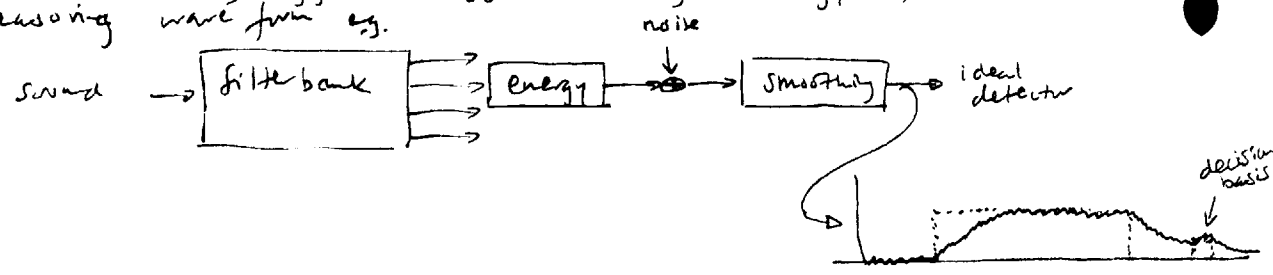
- (a) Pip at A is unmasked \rightarrow just audible at around 0 dB SPL (depending on listener, conditions)
- B is simultaneously masked \rightarrow inaudible until within ~ 10 dB of masker \rightarrow just noticeable ~ 50 dB SPL
- C is subject to forward masking - less than simultaneous \rightarrow somewhere in between - maybe 20-30 dB SPL?

(b) Masking effect is much less at 4 kHz for energy at 1 kHz \therefore in a separate critical band. Ear is more sensitive at 4 kHz



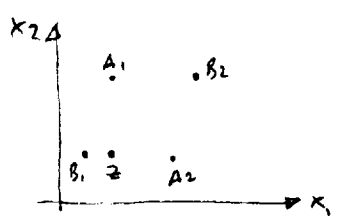
- \rightarrow A noticeable @ ~ -5 dB SPL
- B ~ 20 dB SPL or less
- C ~ 10 dB SPL or less.

- (c) Simultaneous
 - Masking suggests detection via energy envelope displacement.
 - Frequency selectivity \rightarrow band-pass filtering before energy
 - Forward masking suggests sluggish smoothing of energy before measuring wave form e.g.



4. (a) in (x_1, x_2) space, z is nearest $A_1 \rightarrow$ classify to ω_1

(b) Equalizing variance in each dimension "stretches" graph vertically so 4 points lie roughly in a square:



Because transformation matrix is diagonal there is no rotation or shear - just stretch in x_1 and x_2

Now z is closest $B_1 \rightarrow$ classify to ω_2

- (c) Alternative approaches: Decision boundaries from
 - linear - (no perfect fit)
 - Gauss models
 - Neural network etc.
- Could try k-nearest neighbors e.g. find 3 nearest neighbors and vote.

Basically, problem is nearly impossible to decide with so little training data and apparently highly overlapped classes.