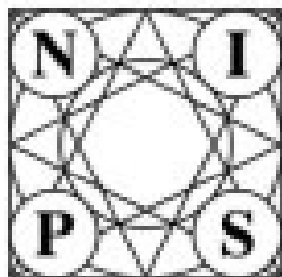


A Maximum Likelihood Approach to Multiple F0 Estimation From the Amplitude Spectrum Peaks

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Neural Information
Processing Systems
Conference



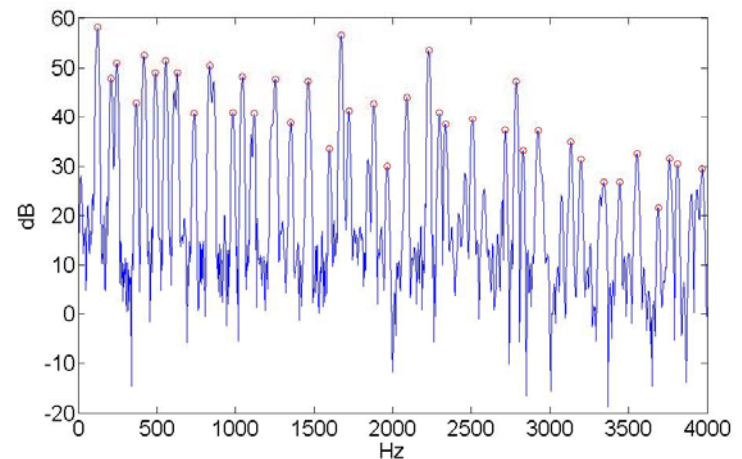
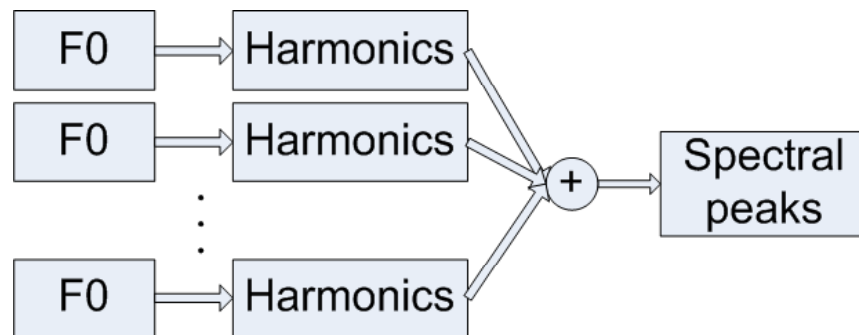


Multiple F0 Estimation

- A sound with mixed tones, tone 1 (F3), tone 2 (C4)
 - Estimate the **polyphony** (number of tones)
 - Estimate the **frequencies** of these tones
- How do musicians do this?
 - Analyze the frequency components by ears
 - Infer the frequencies by the brain
- Can computers also do this?
 - Analyze the frequency components by STFT
 - Infer the frequencies by a Maximum Likelihood method

Problem Formulation

- Parameters to be estimated
 - Number of F0s: N
 - F0s: f_0^1, \dots, f_0^N
- Observation
 - frequencies and amplitudes of the peaks in the amplitude spectrum



Likelihood Function

$$\begin{aligned}\mathcal{L}(\theta) &= p(f_1, A_1, \dots, f_K, A_K | f_0^1, \dots, f_0^N) \\ &= \sum_{I_1, \dots, I_K} p(f_1, A_1, I_1, \dots, f_K, A_K, I_K | f_0^1, \dots, f_0^N) \\ &\stackrel{\text{(assum.)}}{=} \sum_{I_1, \dots, I_K} \prod_{i=1}^K p(f_i, A_i, I_i | f_0^1, \dots, f_0^N) \\ &= \prod_{i=1}^K \sum_{I_i} p(f_i, A_i, I_i | f_0^1, \dots, f_0^N)\end{aligned}$$

- A peak

- “True”: $I_i = 1$: generated by a harmonic
- “False”: $I_i = 0$: caused by detection errors



Likelihood Function (a peak)

$$\sum_{I_i} p(f_i, A_i, I_i | f_0^1, \dots, f_0^N)$$
$$= \underbrace{\{p(f_i, A_i | I_i = 1; f_0^1, \dots, f_0^N)\}}_{\text{“true” peak part}} p(I_i = 1) + \underbrace{\{p(f_i, A_i | I_i = 0)\}}_{\text{“false” peak part}} p(I_i = 0)$$

- Learn the parameters from the training data
 - Training data: the monophonic note samples
 - Easy to know whether a peak is “true” or “false”
 - $p(I_i = 1) = 0.964$

True Peak Part

$$\underbrace{\{p(f_i, A_i | I_i = 1; f_0^1, \dots, f_0^N) p(I_i = 1) + (f_i, A_i | I_i = 0) p(I_i = 0)\}}$$

$$\begin{aligned} p(f_i, A_i | I_i = 1; f_0^1, \dots, f_0^N) &\stackrel{(\text{assum.})}{=} p(f_i, A_i | f_0^{l(i)}) \\ &= \underbrace{p(A_i | f_i, f_0^{l(i)})}_{\text{amplitude}} \underbrace{p(f_i | f_0^{l(i)})}_{\text{frequency}} \end{aligned}$$

- Assume that each “true” peak is generated by **only one** F0
 - 50dB + 30dB = 50.8dB



True Peak Part (amplitude)

$$p\left(A_i | f_i, f_0^{l(i)}\right) = p\left(A_i | f_i, h_i(f_0^{l(i)})\right)$$

- Replace F0 with hi: harmonic number of the peak i

	A	f	f_0	h	d
A	1.00	-0.72	-0.04	-0.61	0.00
f	-0.72	1.00	0.42	0.55	-0.00
f_0	-0.04	0.42	1.00	-0.40	0.01
h	-0.61	0.55	-0.40	1.00	-0.01
d	0.00	-0.00	0.01	-0.01	1.00

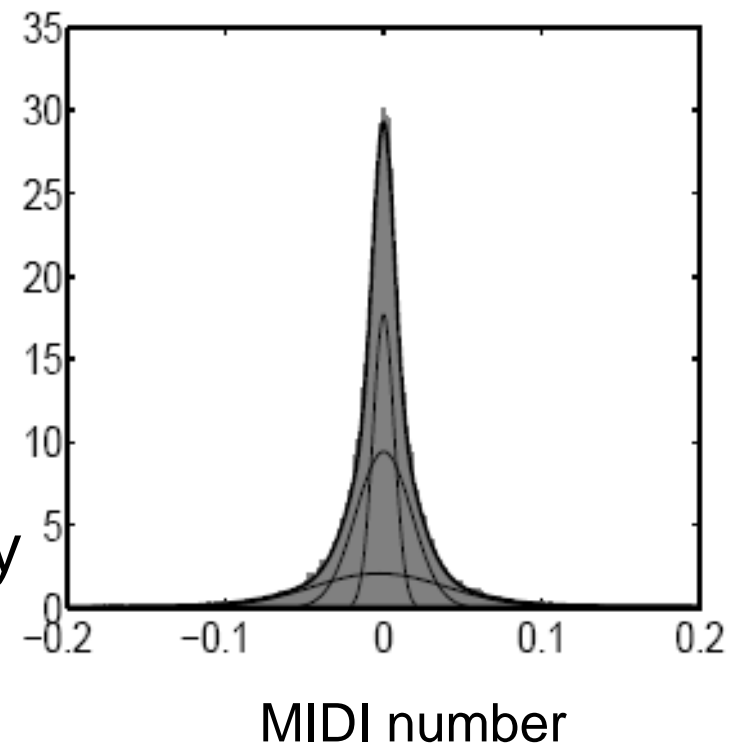
- Estimate $p\left(A_i, f_i, h_i(f_0^{l(i)})\right)$ from the training data
 - A Parzen window (11*11*5)

True Peak Part (frequency)

- Convert the peak frequency into the **frequency deviation** of the peak from the nearest harmonic position of F0

$$\begin{aligned} p\left(f_i | f_0^{l(i)}\right) &\stackrel{(assum.)}{=} p\left(d_i | f_0^{l(i)}\right) \\ &\stackrel{(assum.)}{=} p\left(d_i\right) \end{aligned}$$

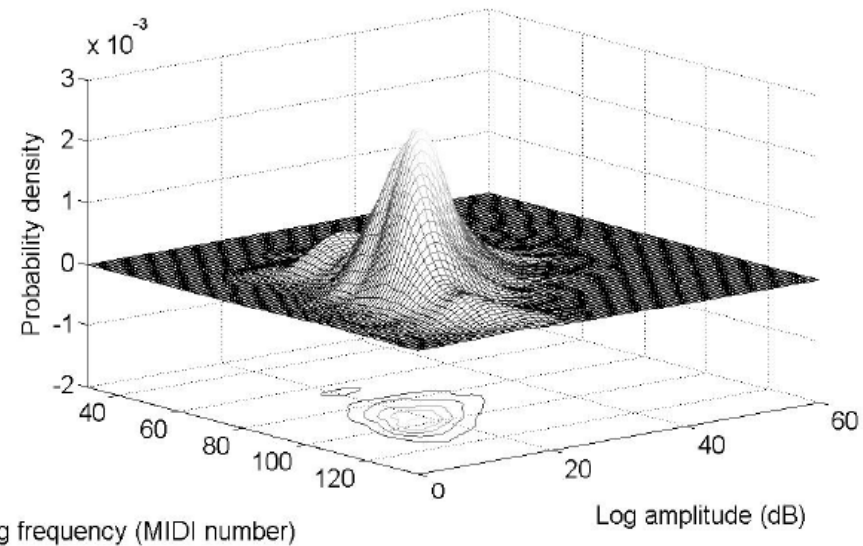
- Estimated from training data
- Symmetric, long tailed, not spiky
- A Gaussian Mixture Model (4 kernels)



False Peak Part

$$\{p(f_i, A_i | I_i = 1; f_0^1, \dots, f_0^N) p(I_i = 1) + \underbrace{(f_i, A_i | I_i = 0) p(I_i = 0)}\}$$

- Estimated from training data
- A Gaussian distribution
 - Mean $(92.7, 20.3)$
 - covariance $\begin{pmatrix} 208.5 & -43.0 \\ -43.0 & 41.0 \end{pmatrix}$



Estimating the Polyphony

- The likelihood will increase with the number of F0s (**overfitting**)
- A weighted Bayesian Information Criteria (BIC)
 - K: number of peaks; N: polyphony

$$BIC = \underbrace{\ln p(f_1, A_1, \dots, f_K, A_K | f_0^1, \dots, f_0^N)}_{\text{Log likelihood}} - \underbrace{2K^{0.45}}_{\text{weight}} \cdot \underbrace{\frac{1}{2}N \ln(2K)}_{\text{BIC penalty}}$$

- Search the F0s and the polyphony to maximize BIC
 - A combinational explosion problem
 - Greedy search: Start from N=1; add F0 one by one

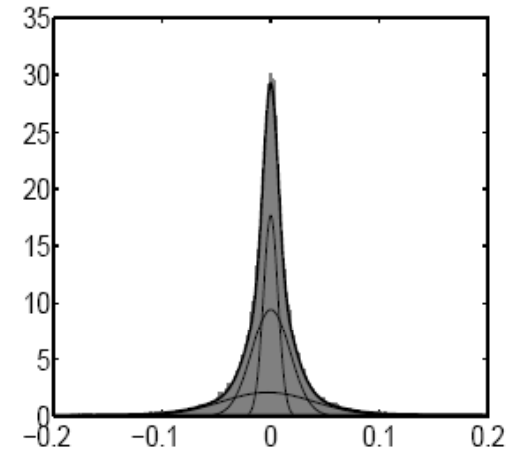
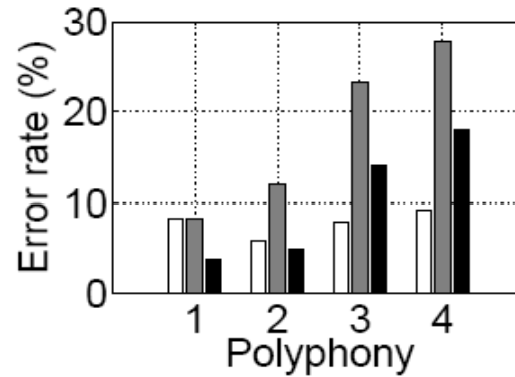
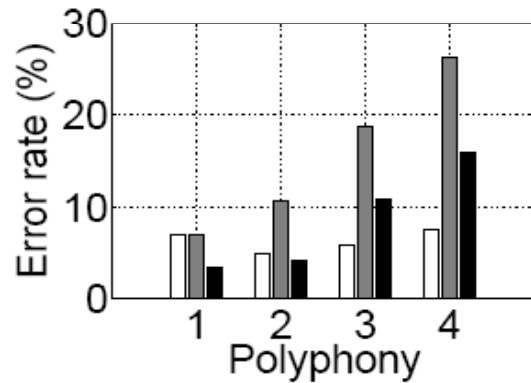


Experiments (1)

- Acoustic materials: 1500 note samples from Iowa music database
 - 18 wind and arco-string instruments
 - Pitch range: C2 (65Hz) – B6 (1976Hz)
 - Dynamic: mf, ff
- Training data: 500 notes
- Testing data: generated using the other 1000 notes
 - Mixed with equal mean square level and no duplication in pitch
 - 1000 mixtures each for polyphony 1, 2, 3 and 4.

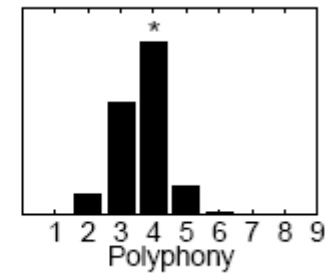
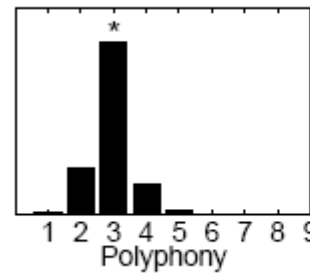
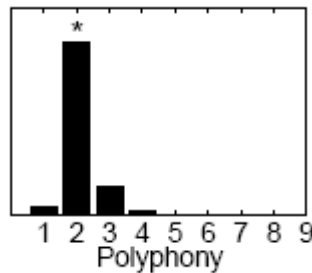
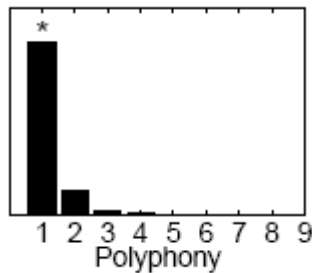
Experiments (2)


- Frequency estimation



(a) $p(d_i)$ in a learned GMM (b) $p(d_i)$ in a non-informative Gaussian

- Polyphony estimation





Thank you!
Welcome to my poster!