

Analyse av koblinger mellom lyd og bevegelse

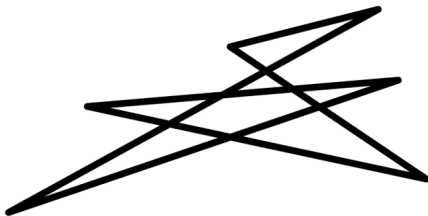
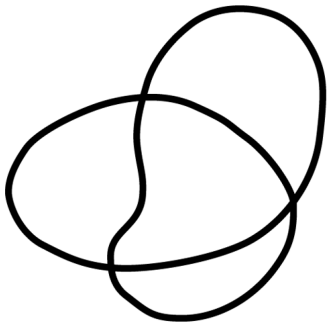
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12 februar 2013

Cross-modality



Takete and Maluma



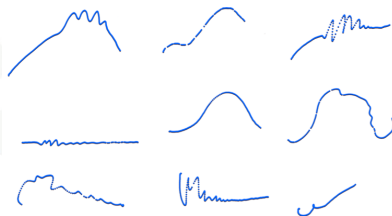
Sonic objects and action objects

If musical sound is made up from sonic objects, maybe we could learn more about how people perceive musical sound by observing their responses to short sonic objects?

⇒ “Sound-tracing”

Sound-tracing on a digital tablet

Godøy, Haga, Jensenius (2006): "Exploring music-related gestures by sound-tracing. A preliminary study", in 2nd ConGAS International Symposium on Gesture Interfaces for Multimedia Systems, Leeds, UK.



Experiment design

Some things to think about when designing a sound-tracing experiment...

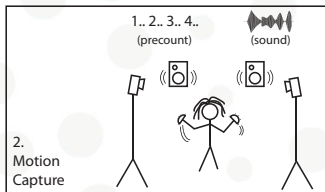
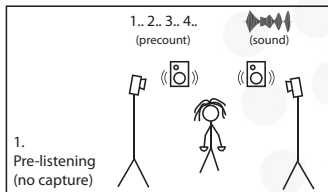
- Sounds
 - how many?
 - what types?
 - controlled (e.g. sine tones) or natural?
- Motion
 - How to record?
- Participants
- Questionnaire?

+ more....

All depends on what you want to study...

E.g. If the goal is to study differences between experienced musicians and non-musicians, your subjects should be chosen accordingly.

3D sound-tracing



3D sound-tracing

Experiment 1 (2009):

- 15 participants (4 female)
- 10 sounds
- 3 recordings per sound
- 450 recordings in total
- Rod used for recording

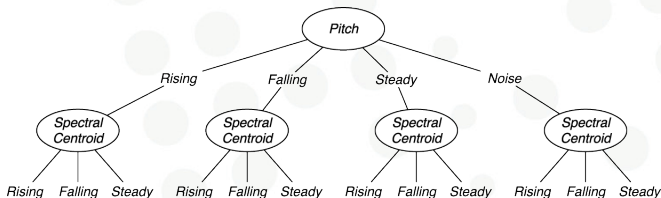


Experiment 2 (2010):

- 38 participants (9 female)
- 18 sounds
- 1 recordings per sound
- 684 recordings in total
- Two handles used for recording



Sounds



Sound	Pitch	Sp.Centroid	Dyn.Env.	Sound	Pitch	Sp.Centroid	Dyn.Env.
1	Rising	Falling	Non-impulsive ¹	10	Noise	Falling	Non-impulsive ¹
2	Falling	Rising	Non-impulsive ¹	11	Noise	Rising	Non-impulsive ²
3	Falling	Falling	Non-impulsive ¹	12	Noise	Steady	Non-impulsive ²
4	Rising	Rising	Non-impulsive ¹	13	Steady	Rising slightly	Non-impulsive ²
5	Rising	Steady	Non-impulsive ²	14	Steady	Falling slightly	Non-impulsive ²
6	Falling	Steady	Non-impulsive ²	15	Rising	Falling	Impulsive
7	Steady	Falling	Non-impulsive ¹	16	Steady	Steady	Impulsive
8	Steady	Rising	Non-impulsive ¹	17	Noise	Steady	Impulsive
9	Steady	Steady	Non-impulsive ²	18	Noise	Falling	Impulsive

Example 1: Impulsive sound



Example 2: Steady sound



Example 3: Opposing ideas

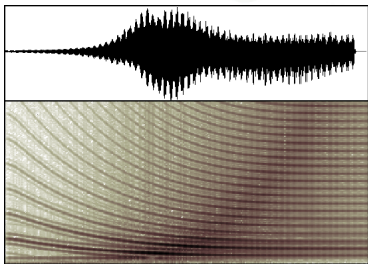


Example 4: Engine metaphor



Visualisations:

Sounds represented by waveform and spectrogram

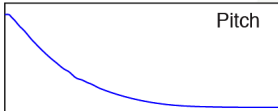
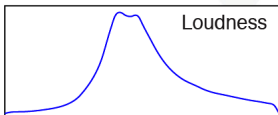
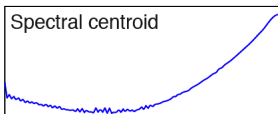


Movement represented by motion images and motiongram

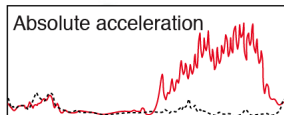
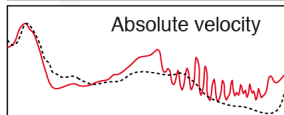
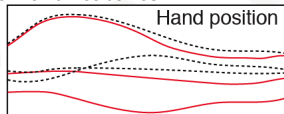


Feature Extraction: Time-varying features (examples)

Sound features



Movement features



Feature Extraction: Global features (examples)

Sound features

- **Categorical:**
 - “Ballistic”, “Sustained”, or “Iterative”
 - “Rising” pitch
- **Calculations:**
 - Duration
 - Overall dynamic energy
 - Onset rise time

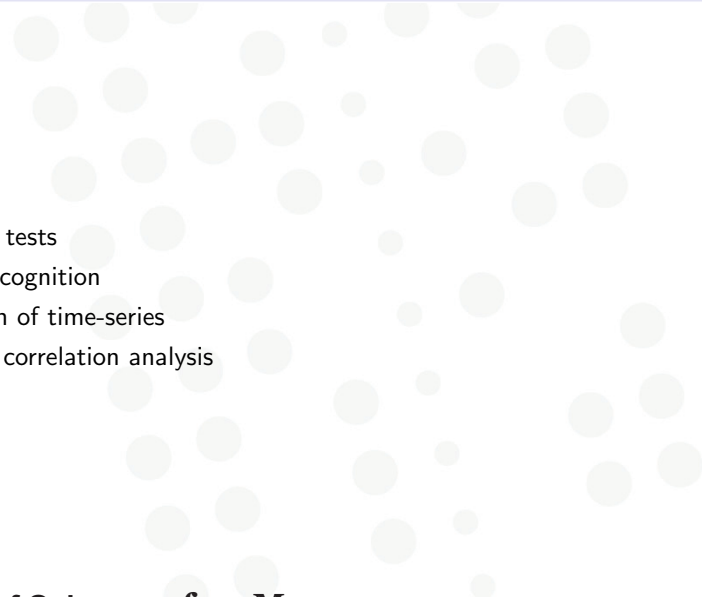
Movement features

- **Categorical:**
 - “Ballistic”, “Sustained”, or “Iterative”
 - “Two-handed”
- **Calculations:**
 - Mean vertical velocity
 - Mean absolute acceleration
 - Mean change in distance between hands

Analysis

Methods:

- Statistical tests
- Pattern recognition
- Correlation of time-series
- Canonical correlation analysis



Statistical tests

Example experiment:

Ask people to rate the expressivity of some musical performance between 1 and 10. Compare the results of “musical experts” to “non-experts”

non-experts	experts
5	6
7	8
8	5
5	5
5	5
2	6
6	5
6	7

mean non-experts: 5.5

mean experts: 5.9

Statistical tests

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- T-test

Statistical tests

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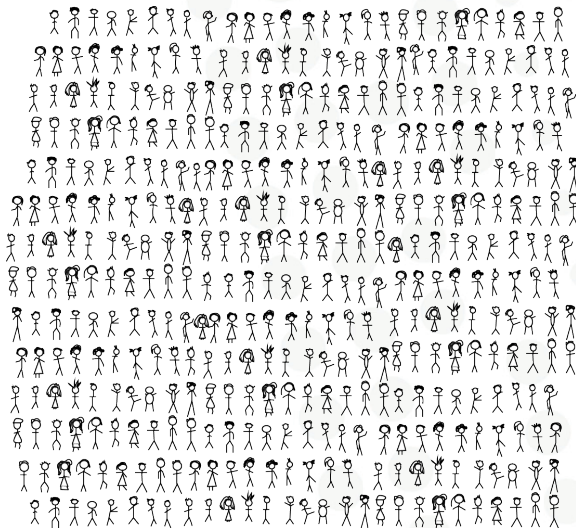
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5	6
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- T-test
- Anova

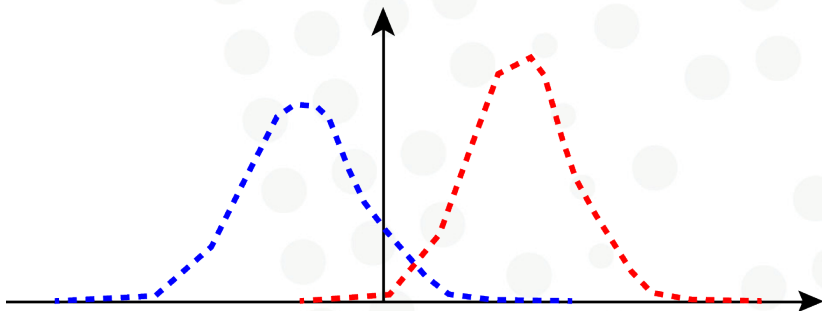
Statistical tests



Statistical tests

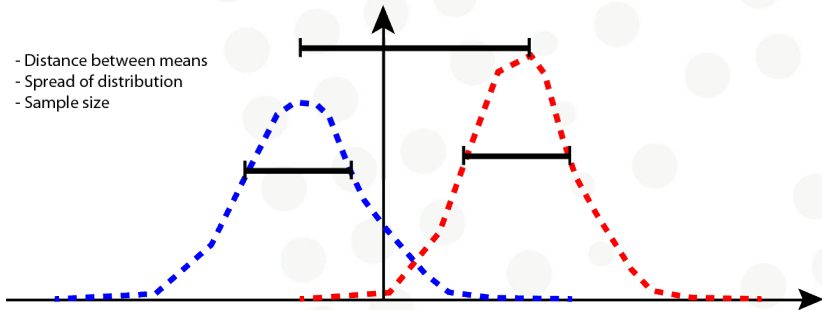


Statistical tests

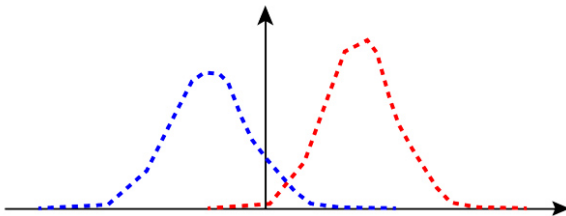


Statistical tests

- Distance between means
- Spread of distribution
- Sample size

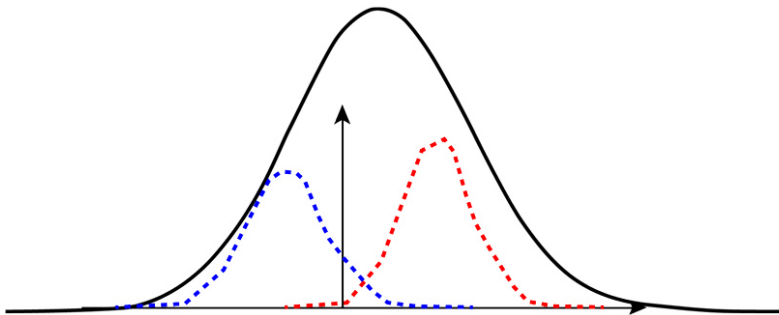


The meaning of the p-value



p : the probability that the two selections stem from the same population
Typical thresholds for p to claim statistical significance: 0.05, 0.01, 0.001

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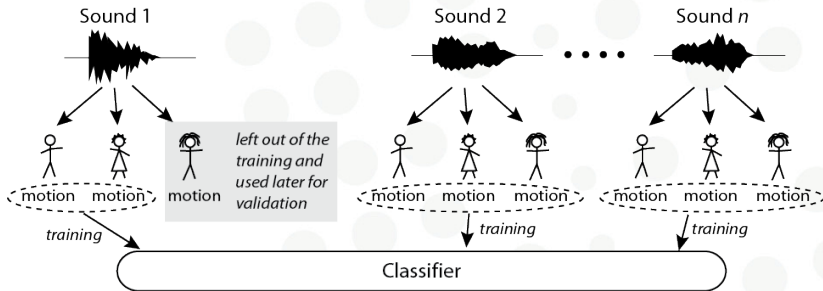


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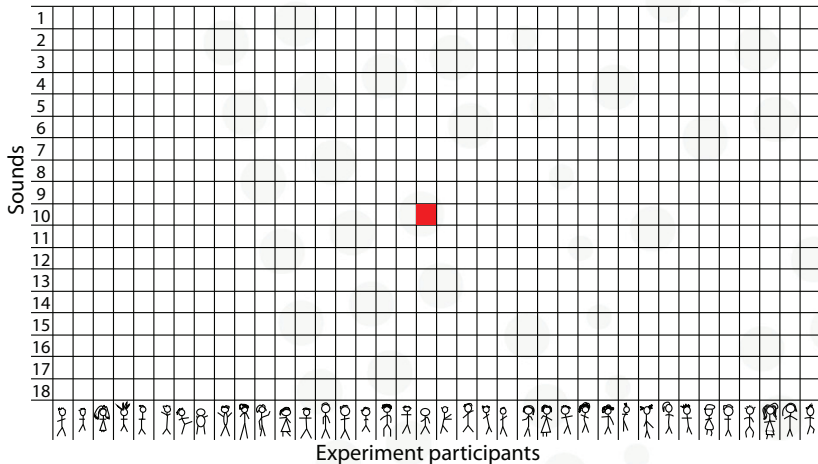
My results

Motion feature	Comparison	<i>df</i>	<i>t</i>	<i>p</i>
OnsetAcceleration	Impulsive vs non-impulsive sounds	526	13.65	< 0.01
VerticalVelocityMean	Rising vs falling sounds	284	18.89	< 0.01
AbsAccelerationMean	Pitched vs noise-based sounds	179	5.53	< 0.01

Pattern recognition classifier



Pattern recognition classifier



Classification results: based on all features

	True 1	True 2	True 3	True 4	True 5	True 6	True 7	True 8	True 9	True 10	Class Precision
Pred. 1	34	6	1	1	1	0	0	0	0	4	72.3 %
Pred. 2	9	36	0	0	0	1	0	2	0	0	75.0 %
Pred. 3	0	0	36	2	0	2	0	0	0	0	90.0 %
Pred. 4	0	0	2	32	1	0	1	3	0	0	82.1 %
Pred. 5	0	0	1	2	31	6	1	2	1	0	70.5 %
Pred. 6	1	0	3	0	6	32	0	1	2	0	71.1 %
Pred. 7	0	0	0	0	1	0	40	3	0	0	90.9 %
Pred. 8	1	0	0	6	3	1	0	34	0	0	75.6 %
Pred. 9	0	1	0	0	2	2	0	0	36	6	76.6 %
Pred. 10	0	0	0	0	0	0	0	0	6	34	85.0 %
Class Recall	75.6 %	83.7 %	83.7 %	74.4 %	68.9 %	72.7 %	95.2 %	75.6 %	80.0 %	77.3 %	

Overall classification accuracy: 78.6 %

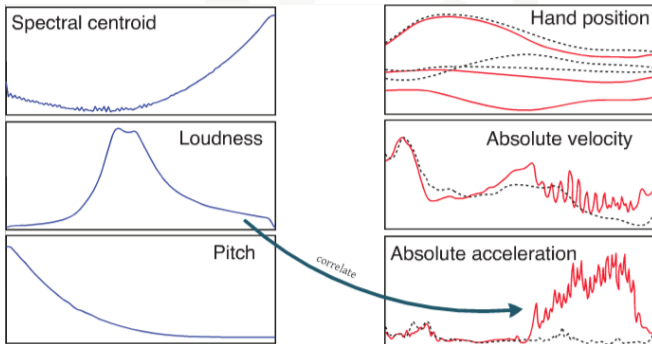
Classification results: based on only vertical features

	True 1	True 2	True 3	True 4	True 5	True 6	True 7	True 8	True 9	True 10	Class Precision
Pred. 1	7	11	0	0	0	0	1	0	0	4	30.4 %
Pred. 2	7	5	0	1	0	0	0	0	0	0	38.5 %
Pred. 3	0	0	34	0	2	10	0	1	1	0	70.8 %
Pred. 4	0	2	1	22	6	1	1	3	0	0	61.1 %
Pred. 5	0	1	0	3	5	4	0	9	0	0	22.7 %
Pred. 6	0	0	3	2	3	2	2	0	0	0	16.7 %
Pred. 7	9	5	2	6	2	6	16	3	6	6	26.2 %
Pred. 8	0	0	0	4	11	0	0	21	0	1	56.8 %
Pred. 9	14	15	3	1	12	14	7	6	31	18	25.6 %
Pred. 10	8	4	0	4	4	7	15	2	7	15	22.7 %
Class Recall	15.6 %	11.6 %	79.1 %	51.2 %	11.1 %	4.6 %	38.1 %	46.7 %	68.9 %	34.1 %	

Overall classification accuracy: 36 %

For only the sounds with changing pitch: 61 %

Correlation of time-series



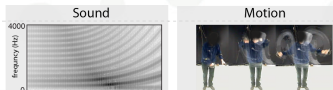
How related are the two variables?

Canonical Correlation

Look at correlation between two sets of variables

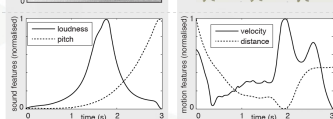
1

Starting with a sound file and a corresponding motion recording.



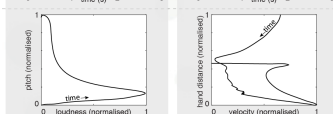
2

The sound and the motion are represented by features (only two sound features and two motion features are used in this example for the sake of clarity).



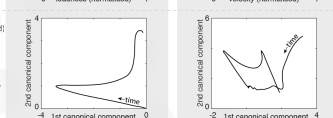
3

To see more clearly what happens, we represent the two features in a two-dimensional feature space.



4

The features are projected (i.e. scaled, stretched, and rotated) into a new space spanned by the canonical components. This space is found by seeking the maximum correlation between the first canonical components. The process is handled by the *canoncorr* function in Matlab.



5

By plotting the canonical components on a timeline, we see that the 1st canonical components bear similarities to each other. In this case, the correlation between the 1st canonical components is 0.75, and between the 2nd canonical components correlation is 0.25.

