

Reducing the "Horseness" of Music Information Retrieval methods

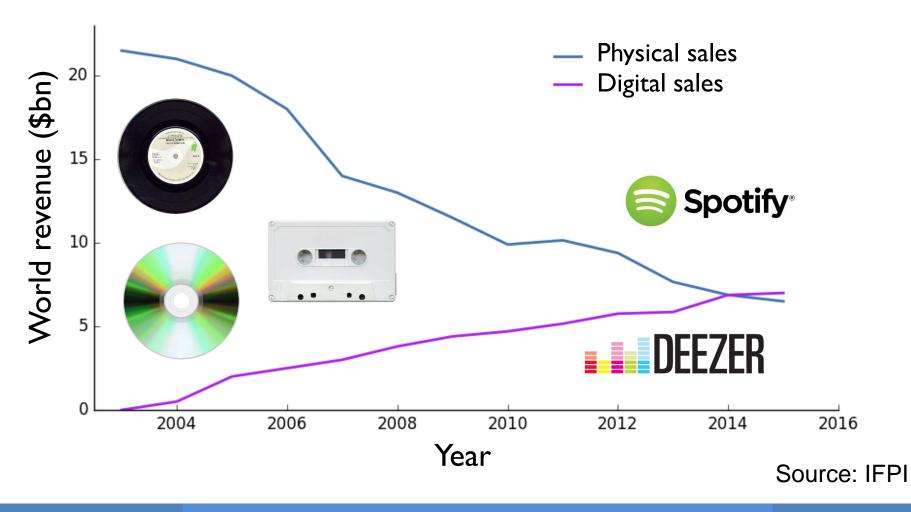
PhD Thesis in I.T. applied to music



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Musical industry



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Preventing « Horses » in MIR tasks

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Streaming

Recommendation **Playlist ∫ Genre** (Rock, Blues, …) ♪ Mood (Joy, Nostalgia, …) **∫ Activity** (Sport, Work, ...) J Top 100 ♪ Celebrities (« Obama », …) **Tag tracks**

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Methods	Advantages	Drawbacks	Examples
Manual (editor)	Precise	Little	PANDORA
Manual (community)	Plenty	Incorrect Ambiguous Abuse	SOUNDCLOUD lost.fm
Automatic (data usage)	Precise	Coverage	
Automatic (autotagging)	Coverage	Precise	DEEZER

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Goal

Enhance autotagging for music recommendations

Focus on Instrumentals and Songs

Tools for development

- ♪ Database Management
 - ♪ Signal processing
 - ♪ Machine learning
 - ♪ Statistical analysis

Test with industrial partners





How to guarantee « Horsefree » methods?

"a horse is just a system that is not actually addressing the problem it appears to be solving." (Sturm 2014)



Song/Instrumental classification

Precision on Instrumental detection

	Dataset	Algorithm	Precision (%)
	I,677 tracks (MSD)	SVMBFF	82.0
*25 41,491 tracks (SATIN)	(Gouyon et al., 2014)	12.5	
	Random prediction	11.0	
	Bayle et al., (2017)	82.5	

- ♪ SVMBFF: 68 features per track
- Proposed algorithm: 39 features per frame

Is bigger better?

Dataset

- ♪ Diversified
 - ♪ Sources (Cross-dataset comparison)
 - ♪ Samples (Representative)
- ♪ Deep learning approaches require a lot of data

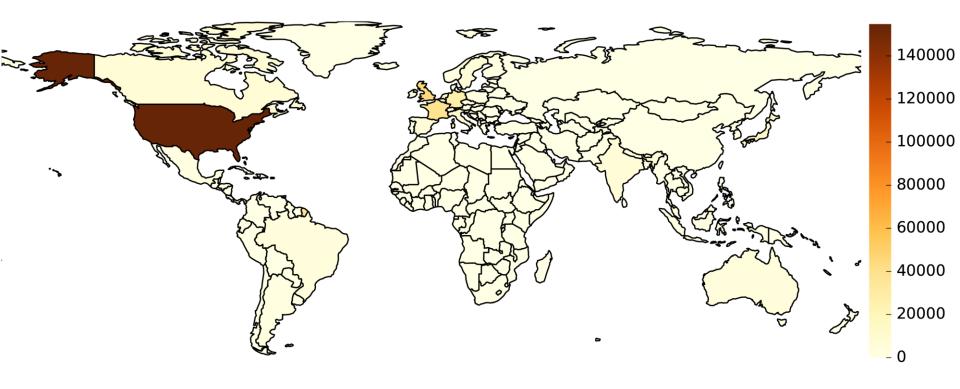
Image research field

- ♪ ~2bn images
- ♪ Duplicate Discovery on 2 Billion Internet Images (Wang et al., 2013)

Music research field

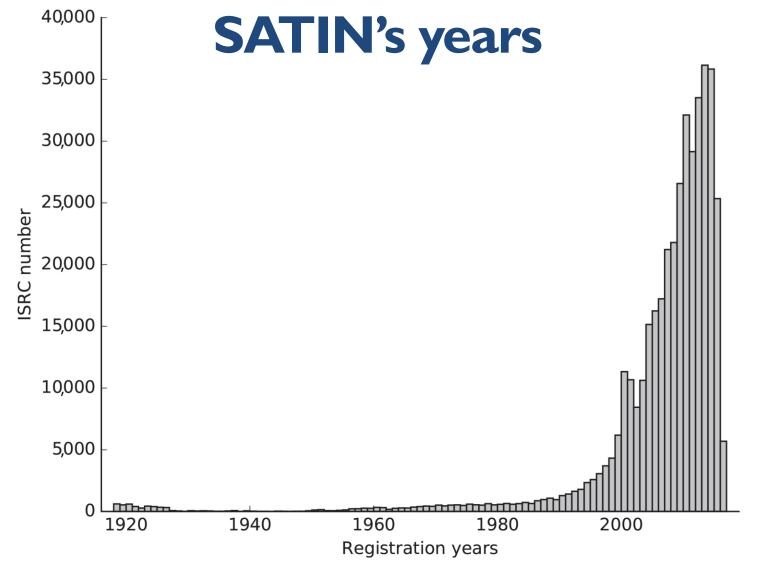
- ♪ Deezer: 40M tracks under copyright
- AcousticBrainz: features for 2.7M tracks
- ♪ FMA: 106k tracks available for the research community

SATIN's world repartition



♪ Bias toward western music





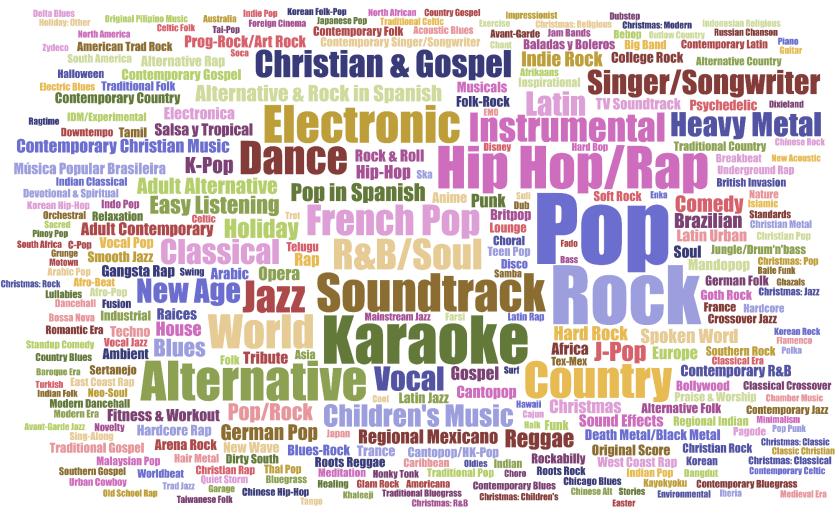
♪ Bias toward 21st century music

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SATIN's wordcloud



♪ Reduce genre bias



Bigger but not too big!

Artist and album filtering

- ♪ A closer look on artist filters for musical genre classification (Flexer 2007)
- Detect studio recording and mastering signature
- ♪ Up to which point to filter?
- I Human can distinguish song from same artist with 20 albums?
- ♪ Filtering reduce the dataset

Data augmentation

- ♪ Copyright restriction and filtering reduce the dataset size
- ♪ Artificially increase the dataset (pitch, speed, add noise, filter,...)
- ♪ A software framework for musical data augmentation (McFee et al., 2015)
- Vork in progress: Adding phase-based data augmentation for NN with raw signal as input

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Human annotations

Quality

- ♪ Track-level (track from 30s to 12m)
- Frame-level (sample precise to seconds)
- Evaluating Hierarchical Structure in Music Annotations (McFee et al., 2017)
- From ground truths to L-measure: multi-annotators and multi-level aggregation.

Objective and subjective

- ♪ Subjective: Genre, Mood, Activity...
- Solution Objective: Instrumental/Song
- I "The tags Vocals and Non-Vocals are well-defined and relatively objective, mutually exclusive, and always relevant." (Gouyon et al., 2014)

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Definitions

Oxford dictionary

- **Song**: A short poem or other set of words set to music or meant to be sung
- Instrumental: music performed on instruments, with no vocals

Notes

- The voice is an instrument
- What about humming?
- Scat: Improvised jazz singing in which the voice is used in imitation of an instrument
- A Song is a musical piece containing human voice, whereas an Instrumental does not.

Examples

- Joe Satriani Crow chant (cf music excerpt)
- Michael Gregorio (cf video)
- **Objective** definition but **subjective** perception?

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Can we measure "Horseness"?

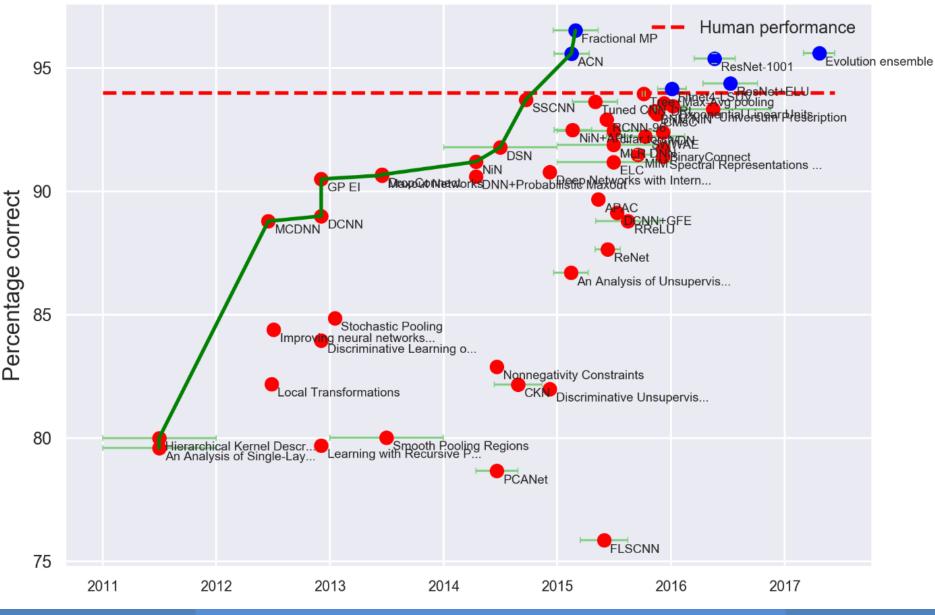
Comparison to baseline

- ♪ Human detection performances
- **Random** classification (on the **dataset**)
- **Random** input (in the **system**)

Project « AI Metrics »

- Human detection threshold comparison
- ♪ State-of-the-art per task in multiple fields
 - ♪ video games, image, video, music,...
- <u>https://github.com/ai-metrics/ai-metrics</u>

CIFAR-10 Image Recognition

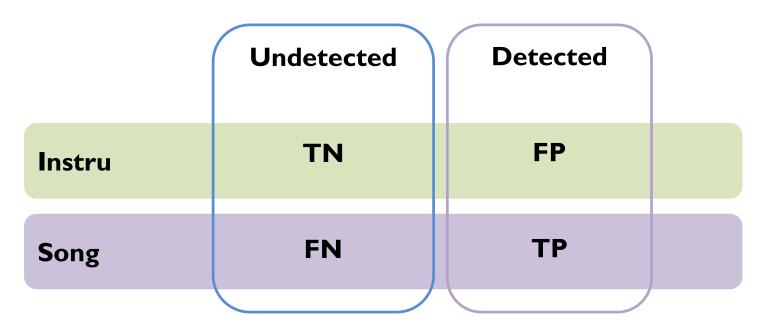


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Horse and metrics



- ightharpoons **Recall** = TP / (TP + FN)
- ♪ Accuracy, F-Measure,... but:
 - ♪ Medecine: 0 false negative required
 - ♪ Music recommendation: minimum of false positive needed

Horse and metrics

Checklist to diminish horseness of a method

- ♪ Metric with statistic and math
- ♪ User listening experience
 - ♪ Subjective
 - ♪ Different expectation
 - ♪ Time-consuming
 - ♪ Too few number of participants

Scientist validation

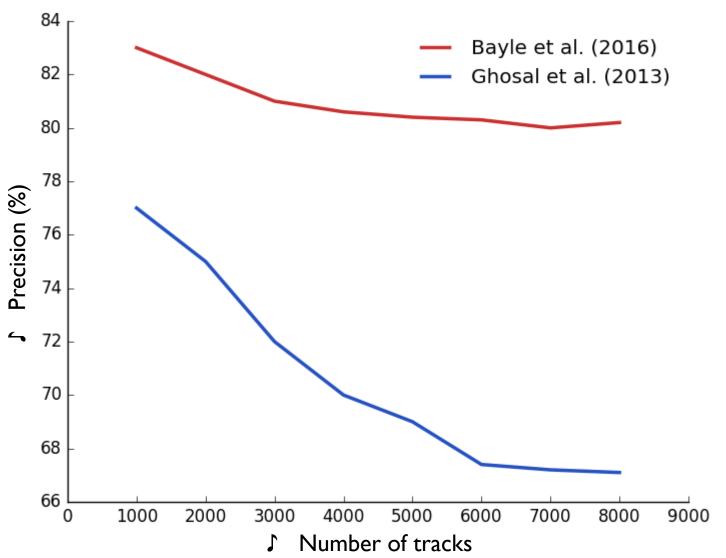
- ♪ Check the results or what the ML is learning?
- ♪ Auralisation of deep convolutional neural networks: listening to learned features (Choi et al., 2015)

Reproducibility and replicability

Examples in Song/Instrumentals classification

- ♪ A hierarchical approach for speech-instrumental-song classification (Ghosal et al., 2013)
 - ♪ Precision @ 95%
 - ♪ 540 excerpts of 30s: « inhouse dataset »
- - ♪ Source code in matlab
 - ♪ Crash for more than 1k tracks
 - ♪ Cannot run on industrial server with 40k tracks

Reproducibility and replicability



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Reproducibility and replicability

Materials

- Replicability is not reproducibility: nor is it good science (Drummond 2009)
- <u>https://github.com/audiolabs/APSRR-2016</u>
- <u>https://infoscience.epfl.ch/record/136640</u>
- <u>https://github.com/faroit/reproducible-audio-research</u>
- <u>https://rescience.github.io/</u>
- <u>https://github.com/Cloud-CV/EvalAl</u>

Conclusion and solutions Ideas

Checklist to diminish « horseness » of a method

- Definition of the problem/task/goal
- ♪ Objective/subjective tag ⇔ objective/subjective solution?
- ♪ Dataset
 - ♪ Bigger
 - Diversified
 - ♪ Sources (Cross-dataset comparison)
 - ♪ Samples (representative)
- Data augmentation
- ♪ Cross-validation
- Preprocessing

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- Normalise signal/spectrograms
- ♪ Comparison to baseline
 - ♪ Human performances
 - Random classification (on the dataset)
 - Random input (in the system)
- Auralisation of deep convolutional neural networks: listening to learned features (Choi 2015)
- ♪ Reproducible research and replicable code
- User listening experiment for validation?
- ♪ Ground truth and L-measure

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