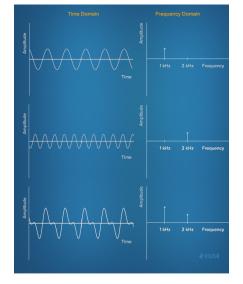
Music



http://www.toptal.com/algorithms/
shazam-it-music-processing-fingerprinting-and-recognition
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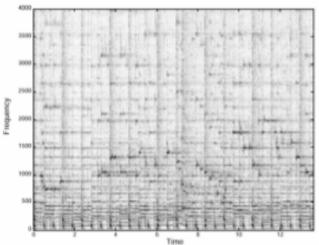
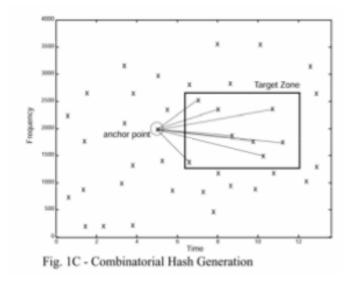


Fig. 1A - Spectrogram



https://www.ee.columbia.edu/~dpwe/papers/Wang03-shazam.pdf

Time Series

This is hard, but it depends on your goals. And on context.

Definition (discrete time series):

$$\{s_t \mid t \in \mathbb{R}^+ \land s \in \mathbb{R}\}$$

(though *s* in any vector space is fine)

Examples domains:

- Weather
- Economics
- Industry (e.g., factories)
- Medicine
- Web
- Biological processes

Why?

- Predict
- Control
- Understand
- Describe

Some strategies:

• Differencing:

$$y_t' = y_t - y_{t-1}$$

• Second-order differencing:

$$y_t'' = (y_t - y_{t-1}) - (y_{t-1} - y_{t-2}) = y_t - 2y_{t-1} + y_{t-2}$$

Some strategies:

- Clustering
- Hidden Markov Models (HMM)
- Recurrent neural networks (RNN)
- Autoregressive integrated moving average (ARIMA)
 - · Generalisation of autoregressive moving average (ARMA) model
 - Regress on series' own lag

One model:

$$s_t = g(t) + \phi_t$$

where

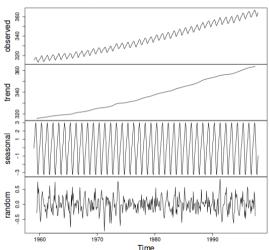
g(t) is deterministic: signal (or trend)

 ϕ_t is stochastic noise

Variation types:

- Trend (*g*)
- Seasonal effect (g)
- Irregular fluctuation (residuals: ϕ)

Decomposition of additive time series



http://www.ulb.ac.be/di/map/gbonte/ftp/time_ser.pdf

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Some easy things to try

- Introduce features to break out seasonality
- Introduce lags as features
- Some domain-specific transformation

Neural Networks