

EBU6018 Advanced Transform Methods

Week 4.3 – Wigner-Ville Distribution (WVD)

Dr Yixuan Zou

Lecture Outline

- Wigner-Ville Distribution
 - □ Definition
 - ☐ Pros and cons
 - □ Connection to other transform methods



Wigner-Ville Distribution - Background

- So far we have looked at transforms that compute the correlation between a signal and basis functions that are functions of time and frequency (or of scale and translation). The time-frequency resolution is determined by the basis functions
 - E.g., STFT, Wavelet Transform
- An alternative approach is to compute directly:
 - Time-frequency energy density- signal's energy density in both time and frequency
 - In contrast to power spectrum: energy in frequency only
 - An example of this is the Wigner-Ville distribution



Wigner-Ville Distribution - Background

- Comparison of STFT and CWT
 - □ Similarities:
 - They are both Windowed Transforms
 - signal is multiplied by a function, and the transform is computed separately for different segments of signals.
 - can be written in inner product form

$$STFT(b,\omega) = \left\langle s(t), \gamma(t-b)e^{j\omega t} \right\rangle$$

$$CWT(b,a) = \left\langle s(t), \frac{1}{\sqrt{a}}\psi\left(\frac{t-b}{a}\right) \right\rangle$$

- □ Difference:
 - Fixed time duration and frequency bandwidths of $\gamma(t)$
 - Variable time duration and bandwidth of $\psi(t)$

Wigner-Ville Distribution – Energy Distribution

- The purpose of energy distributions is to distribute the energy of the signal over time and frequency.
- The energy of a signal s(t) can be found from the squared modulus of either the signal or its Fourier Transform:
- $E_S = \int |s(t)|^2 dt = \int |s(\omega)|^2 d\omega$
- $|s(t)|^2$ and $|s(\omega)|^2$ can be interpreted as energy densities in time and frequency respectively.



Wigner-Ville Distribution – Wiener-Khinchin Theorem

 The energy density and autocorrelation function of a signal are related by the Wiener-Khinchin Theorem.

 According to the Wiener-Khinchin theorem, the power spectrum is the Fourier Transform of the Autocorrelation Function.

$$P(\omega) = |s(\omega)|^2 = \int_{-\infty}^{\infty} R(\tau) e^{-j\omega t} d\tau$$

(For an example of the derivation look at:

https://mathworld.wolfram.com/Wiener-Khinchin Theorem.html)



Wigner-Ville Distribution – Autocorrelation

$$R(\tau) = \int s(t)s(t+\tau)dt$$

 τ is the shift of the signal with respect to itself.

- In the standard autocorrelation function, time is integrated out of the result, and $R(\tau)$ is a function of only the time lag τ .
- There is a class of distribution called Cohen Class (of which the Wigner-Ville Distribution is a member) that uses a variation of the autocorrelation function where time remains in the result, this is the Instantaneous Autocorrelation Function:

$$R(t, \tau) = s(t + \tau/2)s^*(t - \tau/2)$$
 (If the signal is real, then $s^* = s$)

• where τ is the time lag and * represents the complex conjugate of signal s.



Wigner-Ville Distribution – Instantaneous power spectrum

Recall the power spectrum:

$$P(\omega) = |S(\omega)|^2 = \int_{-\infty}^{\infty} R(\tau)e^{-j\omega\tau}d\tau$$

Power spectrum of a signal is the Fourier Transform of its autocorrelation function

where $R(\tau)$ is the autocorrelation function (acf)

$$R(\tau) = \int_{-\infty}^{\infty} s(t)s^*(t-\tau)dt = \int_{-\infty}^{\infty} s(t+\tau/2)s^*(t-\tau/2)dt$$

What happens if we use *instantaneous* autocorrelation:

$$R(t,\tau) = s(t + \tau/2)s*(t - \tau/2)$$

instead of
$$R(\tau) = \int_{-\infty}^{\infty} R(t, \tau) dt$$
 ? We get:

$$WVD_s(t,\omega) = \int_{-\infty}^{\infty} s(t+\tau/2)s^*(t-\tau/2)e^{-j\omega\tau}d\tau$$

which is the Wigner - Ville Distribution (WVD).

The WVD is the Fourier Transform of the instantaneous autocorrelation function



 The Wigner-Ville Distribution is <u>not</u> a Windowed Transform in a similar way to the STFT

 However, the shifted version of the same signal in the autocorrelation function could be considered as a window.

- The WVD compares the information in the signal with its own information at other times and frequencies.
- The WVD has some interesting and useful properties.



Cross-WVD vs auto-WVD

Since we can define a cross-correlation, we can also define a cross-Wigner-Ville distribution:

$$WVD_{s,g}(t,\omega) = \int_{-\infty}^{\infty} s(t+\tau/2)g * (t-\tau/2)e^{-j\omega\tau}d\tau$$

Taking complex conjugates we find

$$WVD_{s,g}(t,\omega) = WVD *_{g,s} (t,\omega)$$

So for the usual WVD (auto-WVD) we have

$$WVD_{s}(t,\omega) = WVD_{s,s}(t,\omega) = WVD *_{s}(t,\omega)$$

so the auto-WVD is always real.



Example: Gaussian function

Signal
$$s(t) = \sqrt[4]{\frac{\alpha}{\pi}} e^{-\alpha t^2/2}$$
 (normalized to unit energy)
For WVD, we get

$$WVDs(t,\omega) = \sqrt{\frac{\alpha}{\pi}} \int_{-\infty}^{\infty} \exp\left\{-\frac{\alpha}{2} \left[\left(t + \frac{\tau}{2}\right)^2 + \left(t - \frac{\tau}{2}\right)^2 \right] \right\} e^{-j\omega\tau} d\tau$$

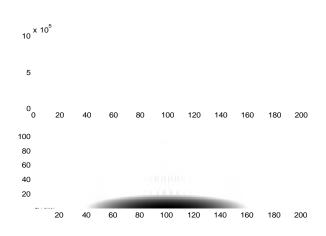
$$= e^{-\alpha t^2} \sqrt{\frac{\alpha}{\pi}} \int_{-\infty}^{\infty} \exp\left\{-\frac{\alpha}{4}\tau^2\right\} e^{-j\omega\tau} d\tau \quad \text{Gaussian in time,}$$

$$= 2 \exp\left\{-\left[\alpha t^2 + \frac{1}{\alpha}\omega^2\right] \right\}$$

i.e. concentrated around (0,0). α controls spread:

"time-width":
$$|t| < \sqrt{\frac{1}{\alpha}}$$

"freq-width": $|\omega| < \sqrt{\alpha}$ Similar to $\Delta t.\Delta \omega = k$ from the UP



Example: Gaussian chirplet

Signal:
$$s(t) = \sqrt[4]{\frac{\alpha}{\pi}} \exp\left\{-\frac{\alpha}{2}t^2 + j\frac{\beta}{2}t^2\right\}$$
Power spectrum: $|S(\omega)|^2 = \sqrt{\frac{4\pi(\alpha^2 + \beta^2)}{\alpha}} \exp\left\{-\frac{\alpha}{\alpha^2 + \beta^2}\omega^2\right\}$

tells us which freqs s(t) contains, not when. Compare:

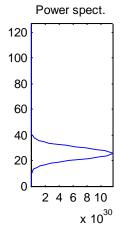
$$WVDs(t,\omega) = \sqrt{\frac{\alpha}{\pi}} \int_{-\infty}^{\infty} e^{-\frac{\alpha}{2} \left[\left(t + \frac{\tau}{2} \right)^2 + \left(t - \frac{\tau}{2} \right)^2 \right] + \frac{j\beta}{2} \left[\left(t + \frac{\tau}{2} \right)^2 - \left(t - \frac{\tau}{2} \right)^2 \right]} e^{-j\omega\tau} d\tau$$

$$= e^{-\alpha t^2} \sqrt{\frac{\alpha}{\pi}} \int_{-\infty}^{\infty} e^{-\frac{\alpha}{4}\tau^2} e^{-j(\omega - \beta t)\tau} d\tau$$

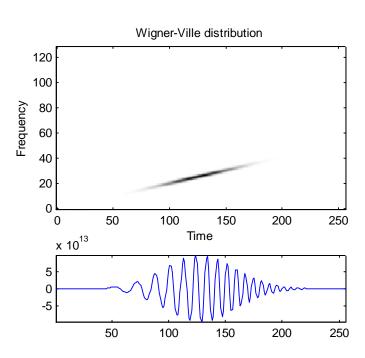
$$= 2e^{-\left[\alpha t^2 + \frac{1}{\alpha}(\omega - \beta t)^2 \right]}$$

i.e. energy concentrated at $\omega = \beta t$, changing with time.

Illustration: chirplet



$$s(t) = \sqrt[4]{\frac{\alpha}{\pi}} \exp\left\{-\frac{\alpha}{2}(t - t_0)^2 + j\frac{\beta}{2}t^2\right\}$$





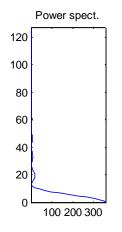
- Time-limited or band-limited signals
 - If s(t) is time—limited, i.e. s(t) = 0 outside some interval $[t_0, t_1]$, then the WVD is also time—limited, i.e.

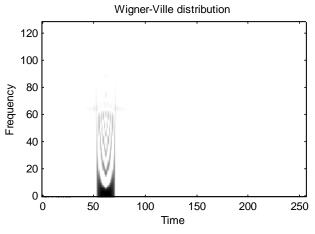
$$WVD_s(t,\omega)=0$$
 for $t\notin [t_0,t_1]$ since no value for τ can make both $s(t+\tau/2)$ and $s(t-\tau/2)$ non-zero if t is outside this range. (Either $s(t+\tau/2)$ or $s(t-\tau/2)$ will be zero)

A similar result is true for frequency band—limited signals.

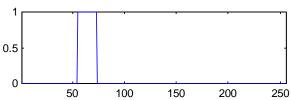


Example: WVD of a Time-limited signal





The isolated pulse is time-bound but not frequency-bound.
The FT is a sinc function.





WVD Properties: Time Marginal Condition

$$\frac{1}{2\pi} \int_{-\infty}^{\infty} WVD_{s}(t,\omega)d\omega = \int_{-\infty}^{\infty} s\left(t + \frac{\tau}{2}\right) s * \left(t - \frac{\tau}{2}\right) \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-j\omega\tau}d\omega d\tau$$
$$= \int_{-\infty}^{\infty} s\left(t + \frac{\tau}{2}\right) s * \left(t - \frac{\tau}{2}\right) \delta(\tau)d\tau$$
$$= |s(t)|^{2}$$

So integral over frequency of WVD is the signal power density at time *t*

Compare similar result for probability densities:

$$p_X(x) = \int_{-\infty}^{\infty} p_{X,Y}(x,y) dy$$

WVD Properties: Frequency Marginal Condition

$$\int_{-\infty}^{\infty} WVD_{s}(t,\omega)dt = \int_{-\infty}^{\infty} s\left(t + \frac{\tau}{2}\right)s * \left(t - \frac{\tau}{2}\right)\int_{-\infty}^{\infty} e^{-j\omega\tau}dt d\tau$$

$$= \int_{-\infty}^{\infty} e^{-j\omega\tau} \int_{-\infty}^{\infty} s(t)s * (t - \tau)dt d\tau$$

$$= \int_{-\infty}^{\infty} e^{-j\omega\tau}R(\tau)d\tau$$

$$= |S(\omega)|^{2}$$

So integral over time of WVD is the power spectral density We also have:

$$\frac{1}{2\pi} \int_{-\infty}^{-\infty} \int_{-\infty}^{\infty} WVD_s(t,\omega)dt \, d\omega = \frac{1}{2\pi} \int_{-\infty}^{-\infty} |S(\omega)|^2 d\omega = \int_{-\infty}^{\infty} |s(t)|^2 dt$$

i.e. the WVD is unitary: the energy in $WVD_s(t, \omega)$ is equal to energy in original signal s(t).



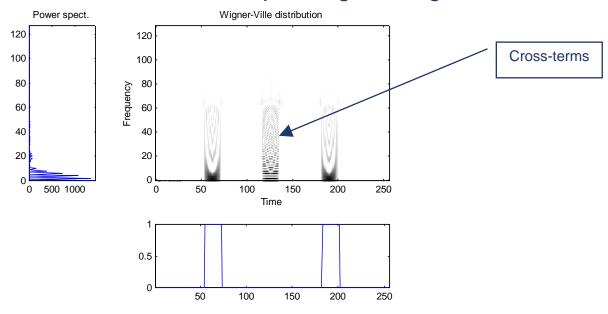
- Time-shift & Freq-moduln. invariant
 - If the WVD of s(t) is $WVD_s(t,\omega)$, then the WVD of time—shifted signal $s_0(t) = s(t-t_0)$ is a time—shifted WVD: $WVD_{s_0}(t,\omega) = WVD_s(t-t_0,\omega)$
 - Further, the WVD of frequency—modulated signal $s_1(t) = s(t)e^{j\omega_1 t}$ is a frequency—shifted WVD: $WVD_{s_1}(t,\omega) = WVD_s(t,\omega-\omega_1)$

(Both follow immediately from the formulas for WVD)

- > WVD of multiple signals: Cross-terms
 - Wigner-Ville Distribution has many useful properties, and better resolution than STFT spectrogram. <u>BUT</u>
 - Applications are limited due to cross-term interference.
 - Consider composite signal $s(t) = s_1(t) + s_2(t)$. Then $WVD_s(t,\omega) = WVD_{s_1}(t,\omega) + WVD_{s_2}(t,\omega) + 2 \operatorname{Re}\{WVD_{s_1,s_2}(t,\omega)\}$ i.e. not only the sum of WVDs, but also the cross—term $WVD_{s_1,s_2}(t,\omega)$
 - Also, cross—term is included at double the magnitude of the auto—terms, so often obscures useful patterns.

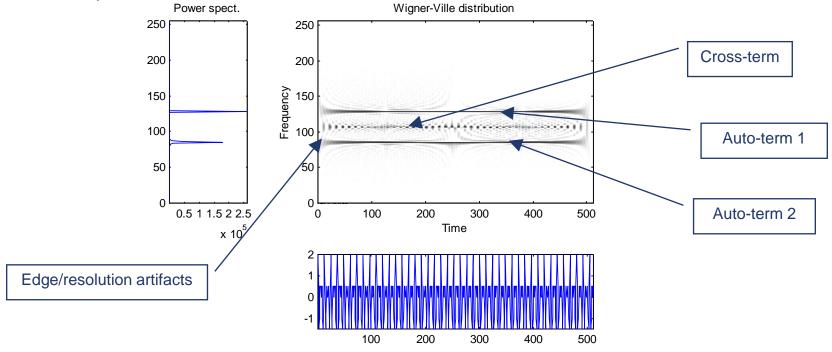


- Example of cross-terms
 - WVD gives cross-terms for all but simplest signals, e.g.:





> Example: sum of two sinusoids





- Example: sum of two sinusoids
 - If $s(t)=\exp(j\omega_0t)$ then $WVD_s(t,\omega)=\int_{-\infty}^{\infty}\exp\left\{j\omega_0\left(t+\frac{t}{2}-t+\frac{t}{2}\right)\right\}e^{-j\omega\tau}d\tau=2\pi\delta(\omega-\omega_0)$ i.e. the WVD is a "ridge" along frequency ω_0 .

• Now let $s(t) = \exp(j\omega_1 t) + \exp(j\omega_2 t)$. The power spectrum is $|S(\omega)|^2 = 2\pi\delta(\omega_1) + 2\pi\delta(\omega_2)$ while the WVD is $WVD_S(t,\omega) = 2\pi\delta(\omega-\omega_1) + 2\pi\delta(\omega-\omega_2) + 4\pi\delta(\omega-\omega_\mu)\cos(\omega_d t)$ where $\omega_\mu = \frac{\omega_1 + \omega_2}{2}$ and $\omega_d = \omega_1 - \omega_2$.

Example: cross-term (cont)

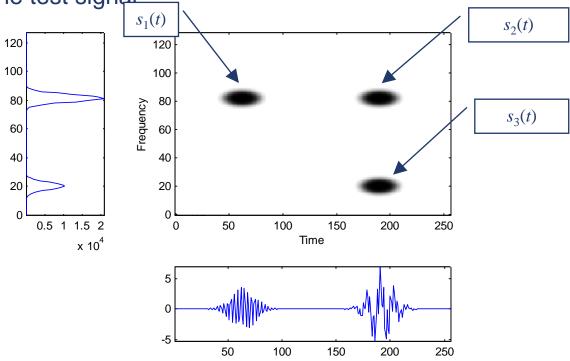
Get a large cross—term $4\pi\delta(\omega-\omega_{\mu})\cos(\omega_{d}t)$ which varies as $\cos((\omega_{1}-\omega_{2})t)$ at ω_{d} , mid—way between the auto—terms. While the auto—terms are + ve, this oscillates + ve & \rightleftarrows —ve Average of the cross—term is zero:

$$\int_{-\infty}^{\infty} 4\pi \delta(\omega - \omega_{\mu}) \cos(\omega_{d} t) dt = 0 \qquad \omega_{d} \neq 0$$

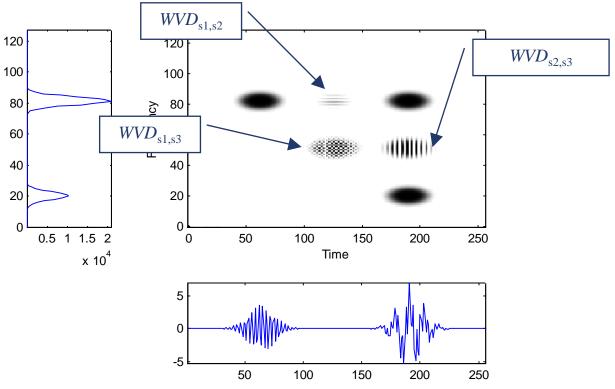
This suggests we may be able to remove these by smoothing (see later.)



> Example: 3-tone test signal

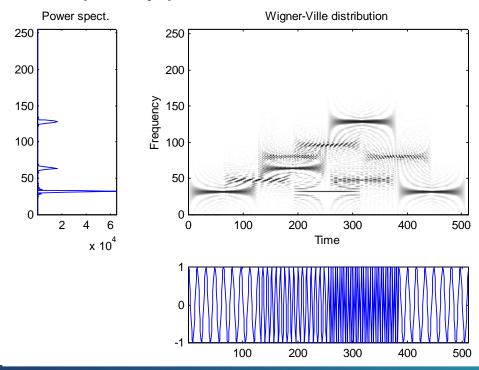






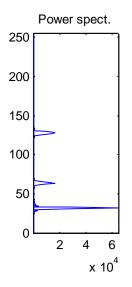


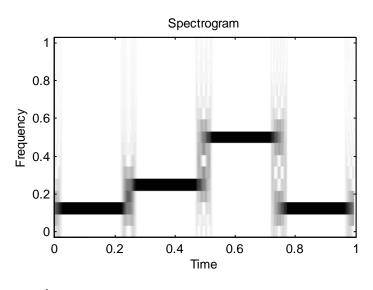
Another example: Frequency pulses



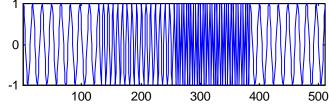


STFT spectrogram





Less resolution, but no cross-terms





Wigner-Ville Distribution - Smoothing

- So, there is a trade-off between the beneficial properties of the WVD and the interference caused by the cross terms.
- To try to reduce the interference problem, we could use a windowed version of the WVD, called the Pseudo-WVD (PWVD):

$$PWVD_{S}(t, \omega) = \frac{1}{2\pi} \int s\left(t + \frac{\tau}{2}\right) s^{*}\left(t - \frac{\tau}{2}\right) h(\tau) e^{-j\omega\tau} d\tau$$

 $h(\tau)$ is the window function.

This is equivalent to frequency smoothing (low-pass filtering) the WVD



 Since the cross—terms WVD are usually strongly oscillating, try removing them by using 2D low—pass filtering, to give a "smoothed Wigner—Ville distribution" (SWVD):

$$SWVD_S(t, \omega) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \varphi(x, y)WVD_S(t - x, \omega - y)dx dy$$

where $\varphi(x, y)$ is a 2D low–pass filter.

Example: 2D Gaussian

$$\varphi(x,y) = e^{-\alpha t^2 - \beta \omega^2} \qquad \alpha, \beta > 0$$

We have a trade—off:

more smoothing → less cross—terms, BUT more smoothing → reduced resolution i.e. not as good



- STFT Spectrogram from WVD
 - The STFT spectrogram is a smoothed WVD, with the WVD of the analysis function $\gamma(t)$ doing the smoothing:

$$|STFT_s(t,\omega)|^2 = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} WVD_{\gamma}(x,y)WVD_s(t-x,\omega-y)dx\,dy$$

Convolution of the WVD of s(t) and the WVD of the STFT window function



- Wavelet Scalogram from WVD
 - ➤ The scalogram (square of the wavelet transform) can be written in terms of the WVD:

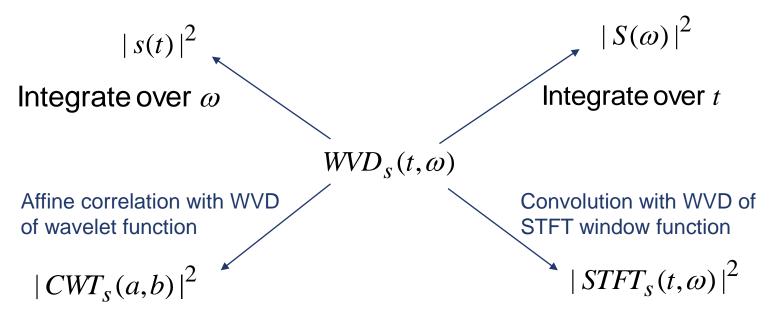
$$SCAL_{S}(a,b) = |CWT_{S}(a,b)|^{2}$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} WVD_{S}(x,y)WVD_{\psi}\left(\frac{x-b}{a},ay\right)dx\,dy$$
where $WVD_{S}(x,y)$ is the WVD of the signal $s(t)$ and $WVD_{\psi}\left(\frac{x-b}{a},ay\right)$ is the WVD of the mother wavelet $\psi(t)$.

- This operation is known as affine correlation.
 - Affine means "related to" and in mathematics can refer to a transformation that maps from one space to another while preserving shape.



Wigner-Ville Distribution - From WVD to ...?



Both the STFT spectrogram and the WT scalogram are smoothed versions of the WVD, explaining why the WVD has the best time-frequency resolution. It is still constrained by the Uncertainty Principle.



Wigner-Ville Distribution - Summary

Kind of Decomposition

Time-Frequency

Analyzing Function

Uses the signal itself. Motivated by time-frequency energy density (c.f. a probability density).

Variable

Time and Frequency. Has high resolution in both time and frequency.

Suited for

Simple (i.e. not composite) signals (non-stationary), e.g. linear chirp, gaussian pulse

Notes

More "complex" (composite) signals lead to undesired "cross-terms". Can be suppressed with smoothing, but lose high resolution in the process.



