Wind speed turbulence system identification and signal generator

Stéphane VICTOR



M. Haijem, S. Victor, P. Melchior, P. Lanusse, and L. Thomas (2023), "Wind turbulence modeling for real-time simulation," Fractional Calculus & Applied Analysis vol. 26, pp.1632-1662 doi: 10.1007/s13540-023-00165-0.





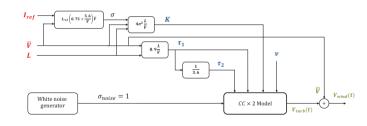






Summary

- Introduction
- 2 State of the art
 - Wind speed components
 - von Kármán model
 - Model identification
- Improved fractional models
 - Davidson-Cole model
 - Cole-Cole model
 - Improved Cole-Cole model
- Wind turbulence model design
 - Model parameter tuning
 - Time simulation
 - Wind speed generator
 - Application on a 3-axis antenna





Introduction

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 - von Kármán model
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Introduction

Wind turbulence = main disturbance for many mechanical systems operating outdoors



wind turbines, wind response of solar panels, kites for traction or electric power generation, sailing, ...

Introduction

• Wind turbulence = main disturbance for many

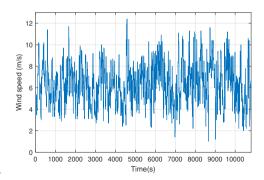
- mechanical systems operating outdoors wind turbines, wind response of solar panels, kites for traction or electric power generation, sailing, ...
- A need to study wind effects in order to design a suitable control law capable of rejecting them
- Wind turbulence = non-stationary and stochastic process ⇒ Models based on spectral characteristics
- Design of models capable of time simulating wind turbulence effects on systems

Objectives

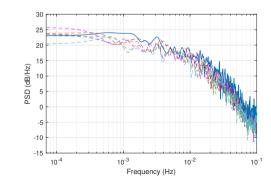
Design an accurate model to well simulate realistic wind speed and provide a wind turbulence generator.

Experimental protocol

- 28 datasets recorded near Montpelier, France
- Ultrasonic anemometer integrated into a Vaisala WXT520 weather transmitter range: [0, 60]m/s; accurace: ±3% in [0, 36]m/s and ±5% in [36, 60]m/s
- Wind speed studied over a period of 3h
 ⇒ modeling only short-term fluctuations



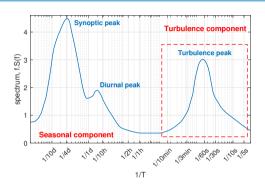
 Study focused on spectral characteristics (PSD) at frequencies below 0.1 Hz with a quasi-constant average speed





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- Spectral analysis of wind power over a long period of time \Rightarrow Several scales of fluctuations [van der Hoven (1957)]
- Spectrum divided into 3 parts
 - Seasonal: long periods (T > 3h)
 - Energy gap (15min < T < 2h)
 - Turbulence: short periods (T < 10min)
- Turbulence peak has a period of 76s
- Frequency range $\left[\frac{1}{10 \text{min}}, \frac{1}{5 \text{g}}\right] = \left[1.6 \, 10^{-3} \text{Hz}, 0.2 \text{Hz}\right]$

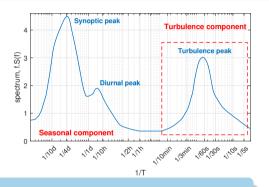


Wind speed components

- Spectral analysis of wind power over a long period of time \Rightarrow Several scales of fluctuations [van der Hoven (1957)]
- Spectrum divided into 3 parts
 - Seasonal: long periods (T > 3h)

State of the art

- Energy gap (15min < T < 2h)
- Turbulence: short periods (T < 10min)
- Turbulence peak has a period of 76s
- Frequency range $\left[\frac{1}{10 \text{ min}}, \frac{1}{50}\right] = \left[1.6 \cdot 10^{-3} \text{Hz}, 0.2 \text{Hz}\right]$



Wind speed components

Wind speed can be expressed as:

$$V(t) = \overline{V}(t) + V_{turb}(t) \tag{1}$$

where $\overline{V}(t)$ is the long-term component and $V_{turb}(t)$ the turbulence component.



I. van der Hoven (1957), "Power Spectrum of Horizontal Wind Speed in the Frequency Range from 0.0007 to 900 Cycles Per Hour." Journal of Atmospheric Sciences 14(2) 160–164 v9(4) 341–351

Wind turbulence PSD approximation

• The von Kármán model approximates the wind speed turbulence PSD [Kármán (1948)]:

$$S_{VK}(f) = \frac{4\sigma^2 \frac{L}{\overline{V}}}{\left(1 + 70.8 \left(\frac{L}{\overline{V}}f\right)^2\right)^{5/6}}$$
(2)

where σ is the turbulence standard deviation, L is the length scale and \overline{V} is the mean speed

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- The model is tuned by 2 parameters:
 - length scale L = average size of the fluctuations in terms of distance from Engineering Sciences Data Unit (ESDU) report:

$$L_{ESDU} = 25z^{0.35}z_0^{-0.063} (3$$

where z is the altitude and z_0 is a surface roughness parameter.



T. von Kármán (1948). "Progress in the Statistical Theory of Turbulence." Proceedings of the National Academy of Sciences vol. 34, no.

11, pp.530 LP - 539.

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• turbulence intensity $I = \frac{\sigma}{V}$ from International Electrotechnical Commission (IEC) report is:

$$I_{IEC} = I_{ref} \left(0.75 + \frac{5.6}{\overline{V}} \right)$$
 (4)

T von Kármán (1948). "Progress in the Statistical Theory of Turbulence." Proceedings of the National Academy of Sciences vol. 34, no.

Shaping filter

Wind PSD approximation ⇒ Time simulation ©

⇒ Need to use **shaping filters**

Random white noise
$$(S_{WN}(f) = 1)$$
 Wind speed $(S_{VK}(f))$ $(S_{VK}(f))$

 $H_{VK}(jf)$ = the filter transfer function stemming from the von Kármán model

$$S_{VK}(f) = \frac{K}{\left(1 + (\tau f)^2\right)^{5/6}} \tag{5}$$

where K = $4\sigma^2\frac{L}{\overline{V}}$ and τ = $\sqrt{70.8}\frac{L}{\overline{V}}$

$$S_{VK}(f) = S_{WN}(f)|H_{VK}(f)|^2 = |H_{VK}(f)|^2 = H_{VK}(jf)H_{VK}^*(jf)$$
(6)

The resulting transfer function:

$$H_{VK}(jf) = \frac{\sqrt{K}}{\left(1 + j\tau f\right)^{5/6}} \tag{7}$$

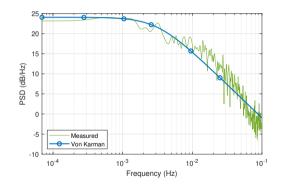
Identification results

• Minimization of the cost function:

$$J = \frac{1}{N} \sum_{i=1}^{N} \left| S_{dB}(f_i) - \widehat{S}_{dB}(f_i, K, \tau) \right|^2$$
 (8)

by using the nonlinear simplex algorithm

- The von Kármán model provides a good fit in low frequency
- \bullet Need to improve the fit in middle and high frequency High frequency slope fixed to 5/6
 - \Rightarrow Constraint that can be released according to real data



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A direct extension is to use a Davidson-Cole form with a variable non-integer order ν :

$$H_{DC}(s) = \frac{\sqrt{K}}{\left(1 + \frac{\tau}{2\pi}s\right)^{\nu}}.$$
 (9)

The resulting PSD expression:

$$S_{DC}(f) = H_{DC}(jf)H_{DC}(jf)^*$$

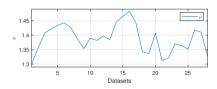
with

$$H_{DC}(jf) = \frac{\sqrt{K}}{(1+j\tau f)^{\nu}}$$



Davidson, D. W. and Cole, R. H. (1951), "Dielectric relaxation in glycerol, propylene glycol, and n-propanol." *The Journal of Chemical Physics*, 19(12):1484–1490, (1951).

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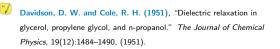
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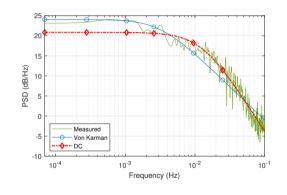
$$S_{DC}(f) = H_{DC}(jf)H_{DC}(jf)^*$$

with

$$H_{DC}(jf) = \frac{\sqrt{K}}{(1+j\tau f)^{\nu}}$$



Identification results:



• The DC model fits well the measurements in high frequency (f > 0.01Hz) thanks to the estimated $\nu \in [1.30, 1.48] \neq 5/6 \ (\approx 0.833)$

Cole-Cole model

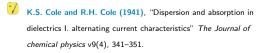
An improved model transfer function based on the Cole-Cole form (CC model):

$$H_{CC}(s) = \frac{\sqrt{K}}{1 + \left(\frac{\tau}{2\pi}s\right)^{\nu}}.$$
 (10)

The resulting PSD expression:

$$S_{CC}(f) = H_{CC}(jf)H_{CC}(jf)^*$$

$$= \frac{K}{1 + 2\cos\left(\nu\frac{\pi}{2}\right)(\tau f)^{\nu} + (\tau f)^{2\nu}}$$



Cole-Cole model

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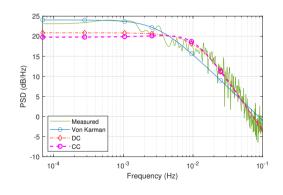
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K.S. Cole and R.H. Cole (1941), "Dispersion and absorption in dielectrics I. alternating current characteristics" *The Journal of chemical physics* v9(4), 341–351.

Identification results:



• The CC model fits well the measurements in high frequency (f > 0.01Hz) thanks to the estimated $\nu \in [1.16, 1.25] \neq 5/6 \ (\approx 0.833)$

Improved Cole-Cole model: CC×2 model

Addition of a second fractional order cell to the Cole-Cole model (CCx2 model):

$$H_{CC\times 2}(s) = \frac{\sqrt{K}}{(1 + (\frac{\tau_1}{2\pi}s)^{\nu})(1 + (\frac{\tau_2}{2\pi}s)^{2\nu})}$$
(11)

⇒ A second slope for a better fit in middle frequency

The resulting PSD expression:

$$S_{CC \times 2}(f) = \frac{K}{D_1(f)D_2(f)}$$
 (12)

with

$$D_1(f) = 1 + 2\cos\left(\nu \frac{\pi}{2}\right) (\tau_1 f)^{\nu} + (\tau_1 f)^{2\nu}$$

$$D_2(f) = 1 + 2\cos(\nu\pi)(\tau_2 f)^{2\nu} + (\tau_2 f)^{4\nu}$$

Identification results:

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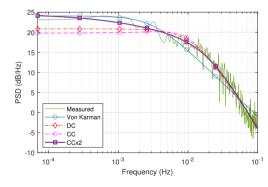
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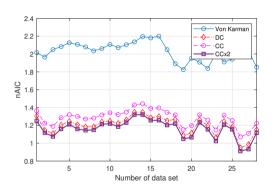
Identification results:



- Good fit in high frequency
- Static gain equal to the von Kármán one
- Improved fit in middle frequency

Table: Comparison between the mean nAIC of the different models

Model	Number of parameters	Mean nAIC
von Kármán	2	2.027
Davidson-Cole	3	1.200
Cole-Cole	3	1.283
$Cole\text{-}Cole{\times}2$	4	1.162



⇒ The CC×2 model reproduces accurately the wind turbulence spectral characteristics

$$nAIC = N\log(\det(\frac{1}{N}E^{t}E)) + \frac{2n_{p}}{N}$$
(13)

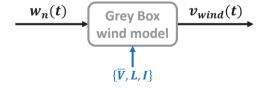
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Wind grey box model



Wind grey box model:

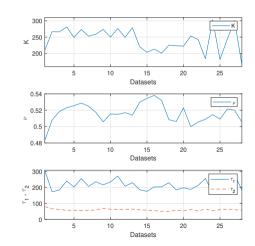
- same transfer function form as the black box model
- parameters obtained from the wind known physical characteristics
- ⇒ Generating random realistic wind speed turbulence without reference data



System Identification for 28 datasets

Selected model: Wind CC×2 model:

$$H_{CC\times 2}(s) = \frac{\sqrt{K}}{(1 + (\frac{\tau_1}{2\pi}s)^{\nu})(1 + (\frac{\tau_2}{2\pi}s)^{2\nu})}$$



System Identification for 28 datasets

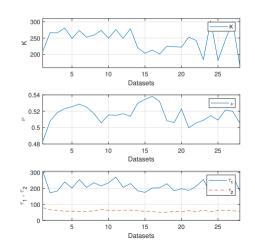
Selected model: Wind CC×2 model:

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 Same static gain expression as the Von Kármán model:

$$K = 4\sigma^2 \frac{L}{\overline{V}}$$

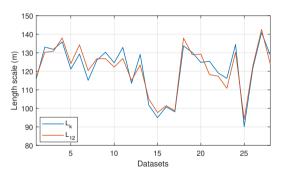
- The mean value of $\nu = 0.516$ ($\nu \in [0.49, 0.53]$)
- The ratio between τ_1 and τ_2 is in [2.6, 4.6]: average value **3.6**



Model parameter tuning

(Von Kármán model:
$$K$$
 = $4\sigma^2\frac{L}{\overline{V}}$ and τ = $8.41\frac{L}{\overline{V}}$ $\Longrightarrow L_K = \frac{K\overline{V}}{4\sigma^2}$.)

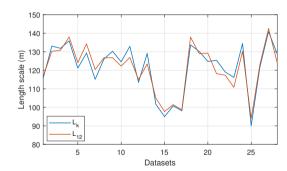
• Validation by comparing the length scales L_k and $L_{12} = \frac{\overline{V}\sqrt{\tau_1\tau_2}}{4.7}$ obtained from τ_1 and τ_2 for 28 different datasets



Model parameter tuning

(Von Kármán model:
$$K=4\sigma^2\frac{L}{\overline{V}}$$
 and $\tau=8.41\frac{L}{\overline{V}} \Longrightarrow L_K=\frac{K\overline{V}}{4\sigma^2}.)$

 Validation by comparing the length scales L_k and $L_{12} = \frac{\overline{V}\sqrt{\tau_1\tau_2}}{4.7}$ obtained from τ_1 and τ_2 for 28 different datasets



Grey-box wind turbulence model

Wind CC×2 model defined by:

$$K = 4\sigma^2 \frac{L}{V}, \qquad \tau_1 = 8.9 \frac{L}{V}, \qquad \tau_2 = \frac{\tau_1}{3.6}, \qquad \nu = 0.516$$
 (14)

Recursive distribution of zeros and poles

The frequency band $[\omega_A, \omega_B] = [10^{-1}, 10^1]$. Therefore, the rational approximation will be carried out on $[10^{-2}, 10^2]$. s^{ν} is approximated by:

$$s^{\nu} \to s^{\nu}_{[\omega_A, \omega_B]} = C_0 \left(\frac{1 + \frac{s}{\omega_b}}{1 + \frac{s}{\omega_b}} \right)^{\nu} \approx C_0 \prod_{k=1}^{N} \frac{1 + \frac{s}{\omega_k'}}{1 + \frac{s}{\omega_b}}$$
 (15)

where:

$$\omega_1' = \eta^{1/2} \omega_b, \ \omega_1 = \alpha \omega_1',$$

$$\omega_{k+1}' = \alpha \eta \omega_k', \ \omega_{k+1} = \alpha \eta \omega_k,$$

$$\nu = \frac{\log \alpha}{\log (\alpha \eta)}, \ \alpha = \left(\frac{\omega_h}{\omega_b}\right)^{\frac{\nu}{N}}, \ \eta = \left(\frac{\omega_h}{\omega_b}\right)^{\frac{1-\nu}{N}}$$

and

$$C_0 = \left| \frac{1 + \mathrm{j} \frac{1}{\omega_h}}{1 + \mathrm{j} \frac{1}{\omega}} \right|^{\nu} = \left(\frac{\omega_h}{\omega_h} \right)^{\nu} \left(\frac{1 + \omega_b^2}{1 + \omega_b^2} \right)^{\nu/2}.$$



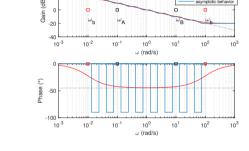


Figure: Fractional integrator $s^{-0.5}$ and its approximation $s_{[0,1,10]}^{-0.5}$ based on Oustaloup's recursive poles and zeros approximation

A. Oustaloup, F. Levron, B. Mathieu, and F. M. Nanot (2000), "Frequency-band complex noninteger differentiator: characterization and synthesis" IEEE Transactions on Circuits and Systems I: Fundamental Theory and Applications, 47(1):25-39, DOI:10.1109/81.817385.

Time simulation

Measured wind characteristics:

$$\overline{V} = 6.6 \text{m/s}, \ \sigma = 1.92, \ L = 120 m$$

Identified CC×2 model:

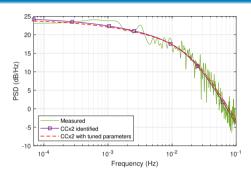
$$H_{BB}(s) = \frac{17.35}{48.72s^{1.554} + 8.59s^{1.036} + 5.67s^{0.518} + 1}$$

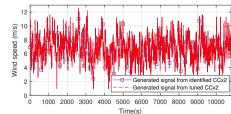
• Tuned grey box model:

$$K = 269.94$$
, $\tau_1 = 161.74s$, $\tau_2 = 44.93s$

$$H_{GB}(s) = \frac{16.43}{40.7s^{1.548} + 7.62s^{1.032} + 5.34s^{0.516} + 1}$$

⇒ Both models provide wind speed with a PSD similar to the measured wind PSD





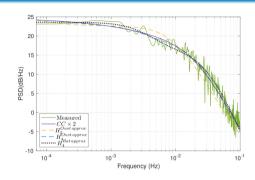
Comparison with rational models

Rational approximation from Oustaloup's approximation:

$$H_{N_{cells}}^{Oust\,approx} = H_0 \prod_{k=1}^{N} \frac{1 + \frac{s'}{\omega_k'}}{1 + \frac{s}{\omega_k}}, \quad (16)$$

Rational approximation:

$$H_4^{Rat\,approx} = H_0 \frac{1 + \frac{s}{\omega_2}}{\left(1 + \frac{s}{\omega_1}\right)\left(1 + \frac{s}{\omega_3}\right)}, \qquad (17)$$



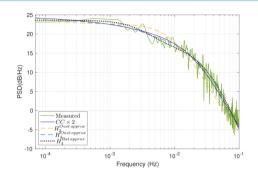
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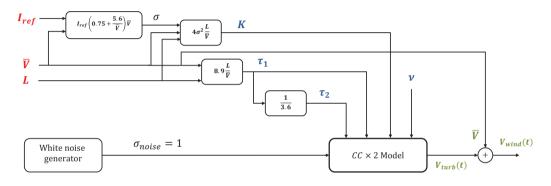
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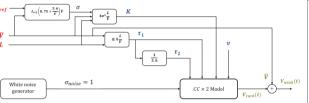


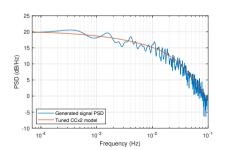
Model	Number of parameters	Fit (%)	nAIC
Cole-Cole×2	4	96.473	1.193
Oust. Approx $H_2^{Oustapprox}$ Oust Approx $H_4^{Oustapprox}$	14	91.514	2.083
Oust Approx $H_4^{Oustapprox}$	26	96.502	1.214
Rat. Approx $H_4^{\widetilde{R}atapprox}$	4	73.166	3.222

Wind speed generator



Wind speed generator



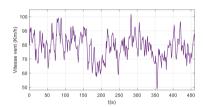


```
%% Model inputs
  L = 110: %Length scale (m)
  Vm = 80/3.6; %Mean speed (m/s)
  I_ref = 0.12: %Turbulence intensity reference (low: 0.12, mean: 0.14,
                 % high: 0.16)
6 Ts = 1;
7 %% Tuning model parameters
  I = I_ref.*(0.75+5.6/Vm);
  sigma = T+Vm:
  K = 4.*(sigma)^2.*L./Vm;
  Tau1 = 8.*L./Vm;
  Tau2 = Tau1./3:
  nn = 0.51:
14
  %% Fractional transfer function
  num=sgrt(K);
  denl=frac_poly_exp([(Tau1/2/pi)^nu 1],[nu 0]);
  den2=frac_polv_exp([(Tau2/2/pi)^(2*nu) 1],[2*nu 0]);
19
  sys = frac_tf(num,den1)*frac_tf(1,den2);
  %% Approximated rational transfer function
  set (sys.'N'.10): %Cell number
  set(sys, 'band', [1e-4,10]); %Frequency range + 2 decades
  set(sys, 'sim', 'Oustaloup');
27 H_CCx2 = frac2int(sys):
```

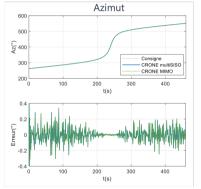


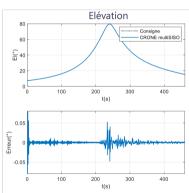
http://cronetoolbox.ims-bordeaux.fr

Application on a 3-axis antenna



- Wind turbulence average: $\bar{V} = 80km/h$
- Wind turbulence speed: [50, 100]km/h
- MIMO CRONE control for turbulence rejection





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 - Model parameter tuning
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 - Wind speed generator
 - Application on a 3-axis antenna
- Conclusion

Conclusion

- Wind models based on the spectral characteristics and not on time series data, then shaping filters are synthesized for time simulation
- Study and identification of the Von Kármán approximation and its limits
- Improved fractional models with: Davidson-Cole, Cole-Cole and Cole-Cole x2 models
- Tuning methodology to transform the CC×2 black box model into a grey box model ⇒ not limited by access to reference data
- A long period wind mean speed generator can be developed for more realistic scenarios
- M. Haiiem. S. Victor. P. Melchior, P. Lanusse, and L. Thomas (2023), "Wind turbulence modeling for real-time simulation." Fractional Calculus & Applied Analysis vol. 26, pp.1632–1662 doi: 10.1007/s13540-023-00165-0.