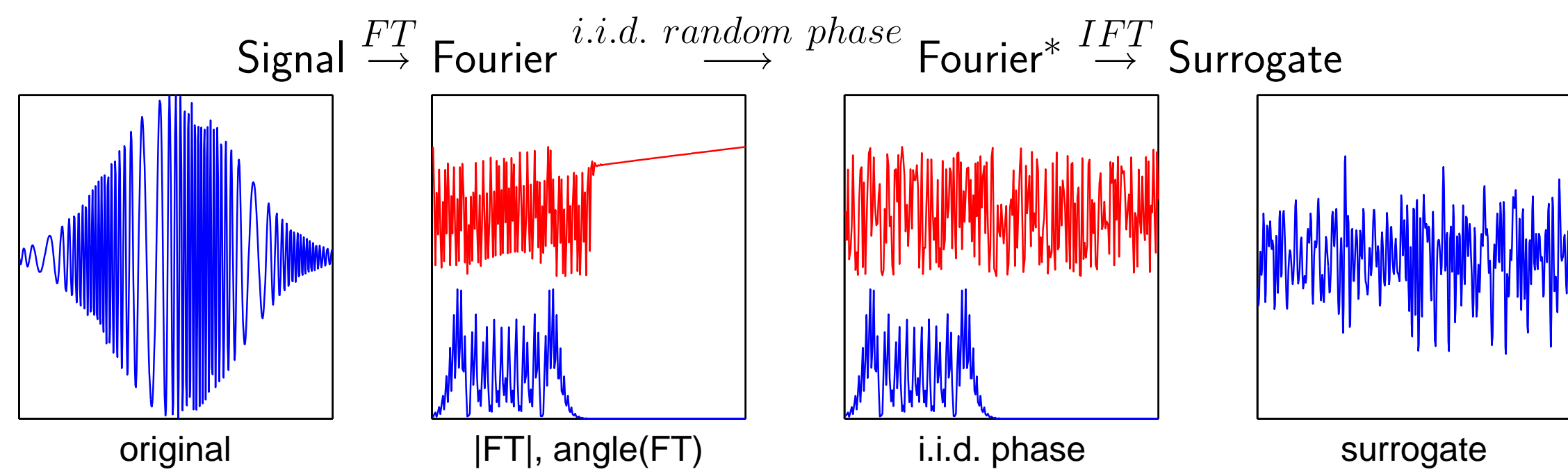


1. Revisiting Surrogates

Original context — Surrogate data were introduced in [Theiler et al., 1992] for **testing nonlinearity**.

- This is a method of Monte-Carlo sampling aiming at creating new samples from the data;
- The new samples are constrained to satisfy a given null hypothesis, here = linearity of the system.

Generate Surrogate Data — Simplest form: Fourier-based Surrogates



Use of Surrogates for Non-linearity tests. — The Non-linearity test is applied to a set of surrogates.

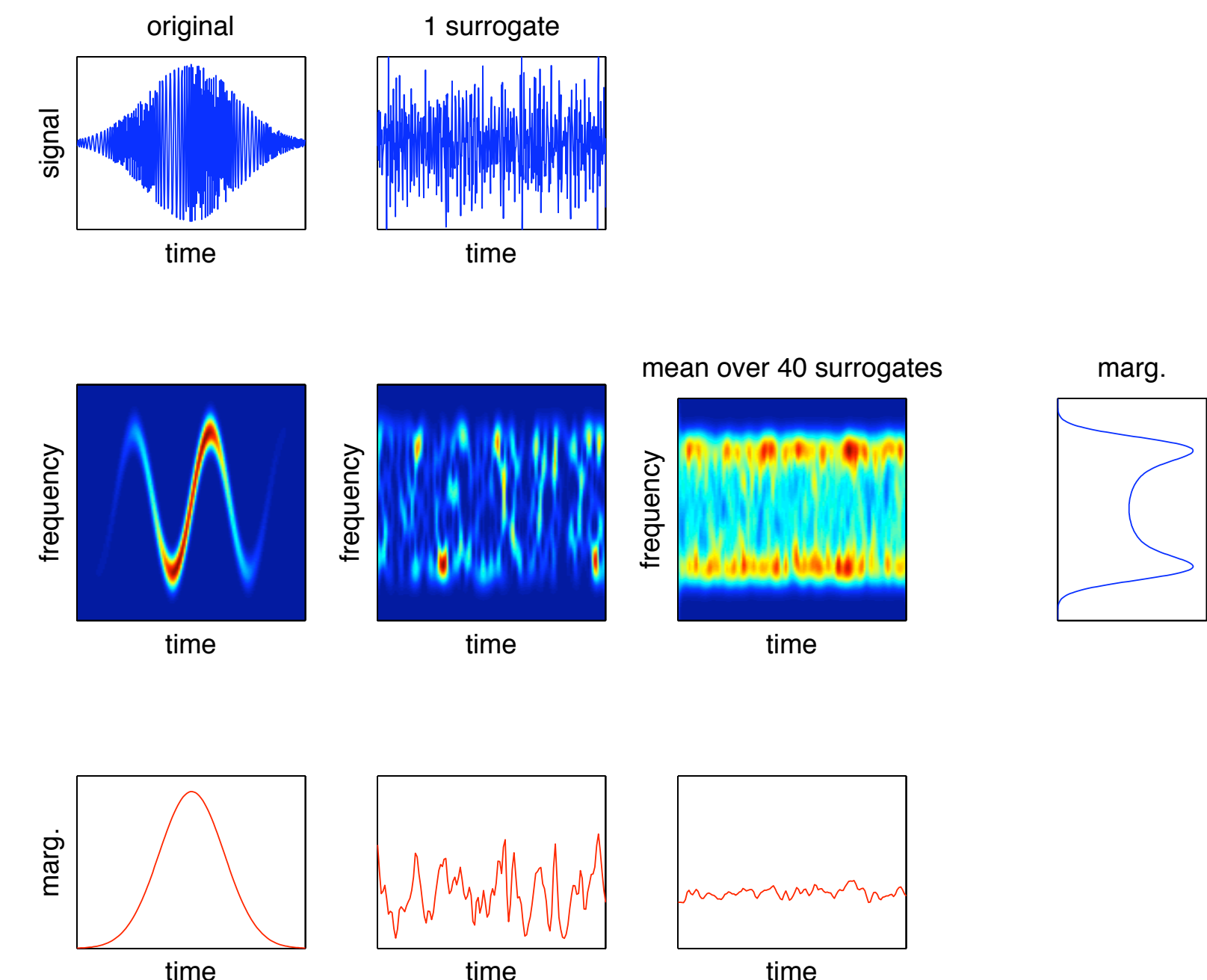
- Surrogates → set of data under the null hypothesis for the test → experimental threshold.

Variations. — Other surrogate methods with added constraints from the original data:

- Improved Surrogates having same probability distribution, see review in [Schreiber & Schmitz, 1996],
- Preservation of the mean and variance structure → preserves nonstationary evolution [Keylock, 2006].

2. Stationarization via Surrogates

New idea — Surrogates act as null hypothesis for **stationary signals**.



Rationale — For a same spectrum density, “nonstationary” signals differ from “stationary” ones by temporal *structures* encoded in the spectrum phase.

3. Stationarity Testing with Surrogates

Objective — Deal with “stationarity” in an operational sense, including the possibility of its test relatively to a given observation scale and that would both encompass stochastic and deterministic variants.

Time-Frequency framework — Natural language for nonstationary signals and processes: Time-frequency spectra (and their estimates) should undergo no evolution in “stationary” situations.

- **Multitaper spectrograms** are defined as

$$S_{x,K}(t, f) = \frac{1}{K} \sum_{k=1}^K \left| \int_{-\infty}^{+\infty} x(s) h_k(s-t) e^{-i2\pi f s} ds \right|^2,$$

with K short-time windows $h_k(t)$ chosen as the K first Hermite functions [Bayram & Baraniuk, 2000].

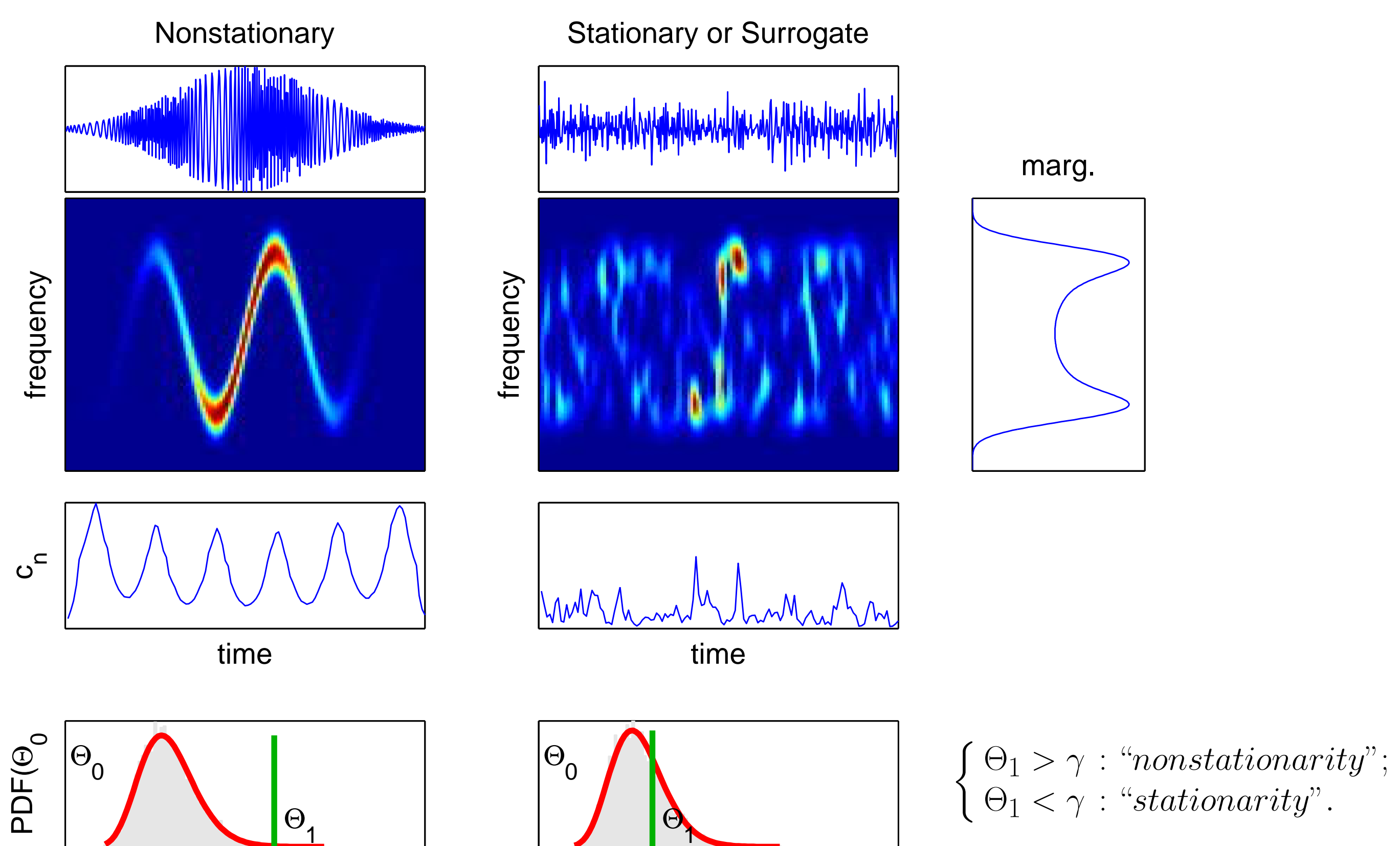
- **Advantage:** *estimates* of the Wigner-Ville Spectrum for stochastic processes and *reduced interference distributions* for deterministic signals, with reduced estimation variance without some extra time-averaging (unappropriate in a nonstationary context).
- **In practice:** spectrograms evaluated only at N time positions $\{t_n, n = 1, \dots, N\}$, with a spacing $t_{n+1} - t_n$ which is a fraction of the width of the K windows $h_k(t)$, K chosen in between 5 and 10.

Principle of the test — Compare *local features* vs. *global ones* obtained by *marginalization over time*, relatively to a chosen observation scale [Xiao et al. 2007 A].

- Contrast Global/Local with some TFD distance κ : $c_n^{(x)} := \kappa(S_{x,K}(t_n, \cdot), \langle S_{x,K}(t_n, \cdot) \rangle_{n=1, \dots, N})$,
- Fluctuations in time of these divergences $c_n^{(x)}$: $L(g_n, h_n) := \frac{1}{N} \sum_{n=1}^N (g_n - h_n)^2$,
- Test statistics = dispersion of the fluctuations: $\Theta_1 = L(c_n^{(x)}, \langle c_n^{(x)} \rangle_{n=1, \dots, N})$.

Null hypothesis — provided by the collection of Test Statistics of the surrogates s_j :

$$\left\{ \Theta_0(j) = L(c_n^{(s_j)}, \langle c_n^{(s_j)} \rangle_{n=1, \dots, N}) \right\}, j = 1, \dots, J.$$



Stationarity Test — Form the one-sided test, with threshold γ obtained from the null hypothesis.

Variant — Use One-Class SVM to learn the distribution of stationary signals [Xiao et al. 2007 B].

4. Variations in Time-Frequency Contexts

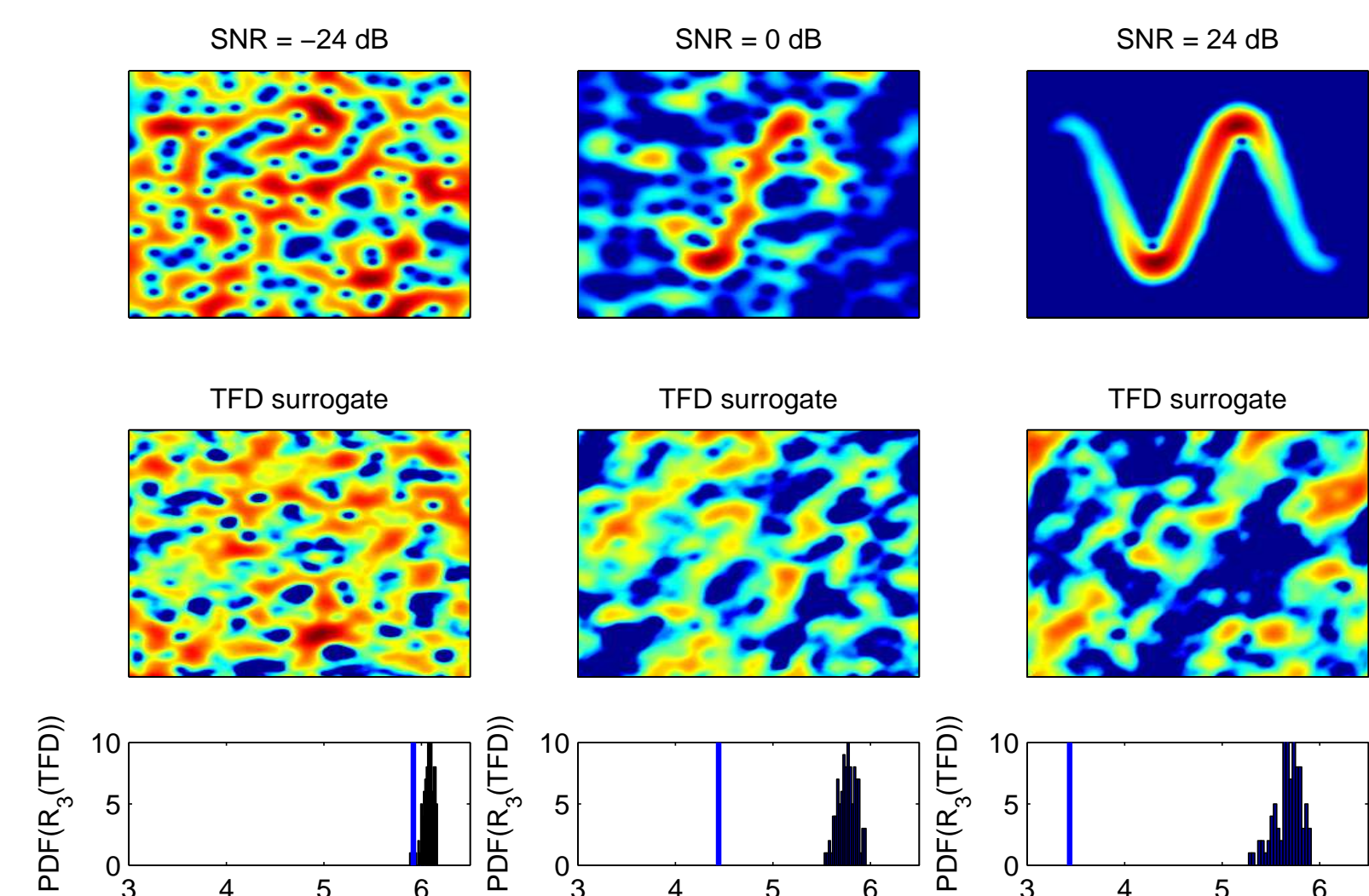
4.a For Transient Detection

Objective — Assess the level of significance of a “time-frequency” patch in a fluctuating background.

Principle — Construct a **Time-Frequency Distribution Surrogate**, directly via phase randomization of the 2D Fourier transform.

- Constraints: preserve the amplitude of the Ambiguity function; positivity of the TFD Surrogate.
- Procedure: iterative correction of the phase and thresholding negative values of the TFD.

Statistical Test — with Renyi entropy: $R_\alpha(TFD) = \frac{1}{1-\alpha} \iint \left(\frac{TFD(t, f)}{\iint TFD(t, f) dt df} \right)^\alpha dt df$.



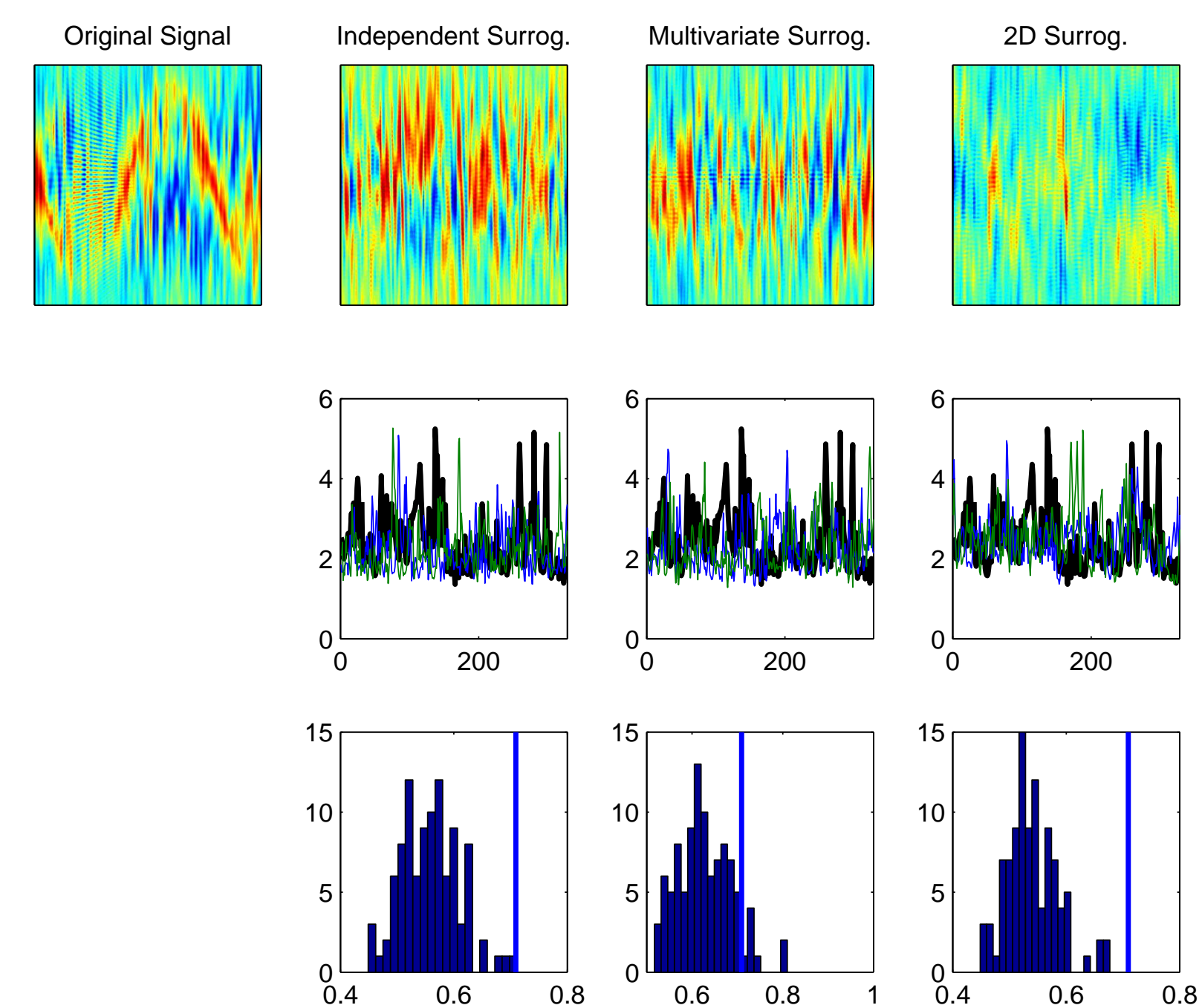
4.b For Nonstationary Correlations

Objective — Assess the existence and evolution of cross-correlations in multivariate data.

Principle — Compare the cross-correlations of original data with the one of surrogates.

- Multivariate formulation of original surrogates keeps the marginal cross-spectrum [Prichard & Theiler 1994]. But it does not keep the ‘geometrical’ structure of the quadratic distribution in the plane.
- **Proposition:** directly design time-lag surrogates for the cross-correlations via a Fourier phase randomization of the 2D cross-correlations.

Statistical Test — with Kurtosis of the divergence between local and averaged cross-correlations.



Open question: — Comparison to other resampling plans: Bootstrap; Jackknife; Cross-Validation [Efron, 1982].

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[Xiao et al., 2007 A] J. Xiao, P. Borgnat and P. Flandrin, “Testing stationarity with time-frequency surrogates,” *Proc. EUSIPCO-07*, Poznan (PL).

[Xiao et al., 2007 B] J. Xiao, P. Borgnat, P. Flandrin and C. Richard, “Testing stationarity with surrogates – A one-class SVM approach,” *Proc. IEEE Stat. Sig. Proc. Workshop SSP-07*, Madison (WI).