

An Alternative Approach to Covariance Propagation (and a Generalized Gaspari-Cohn Correlation Function)

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An Alternative to Standard Covariance Propagation

Covariance propagation, e.g.,

$$\mathbf{P}_{k+1} = \mathbf{M}_{k+1,k}(\mathbf{M}_{k+1,k}\mathbf{P}_k)^T + \mathbf{Q}_k$$

Issues with Covariance Propagation:

- Inaccurate variance propagation
- Computational expense

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“Local covariance evolution” (Cohn, 1993) \Rightarrow Parametric Kalman Filter (e.g., Pannekoucke et al. 2016)

Local Covariance Evolution, 1D

Consider covariances $P = P(x_1, x_2, t)$ associated with states $q = q(x, t)$ on the unit circle (S_1^1),

$$q_t + vq_x + bq = 0,$$

$$q(x, t_0) = q_0(x)$$

$$P_t + v_1 P_{x_1} + v_2 P_{x_2} + (b_1 + b_2)P = 0,$$

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Variance Equation:

$$\sigma_t^2 + v\sigma_x^2 + 2b\sigma^2 = 0,$$

$$\sigma^2(x, t_0) = \sigma_0^2(x)$$

Correlation Length Equation:

$$L_t + vL_x - v_x L = 0,$$

$$L(x, t_0) = L_0(x)$$

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1. Evolve σ^2 and L from initial condition $P_0(x_1, x_2) = \sigma_0(x_1)C_0(x_1, x_2)\sigma_0(x_2)$.
2. Approximate $P(x_1, x_2, t) = \sigma(x_1, t)C(x_1, x_2, t)\sigma(x_2, t)$ with evolved σ^2 and L using a *parametric correlation function*.

The Gaspari and Cohn (1999) Correlation Function

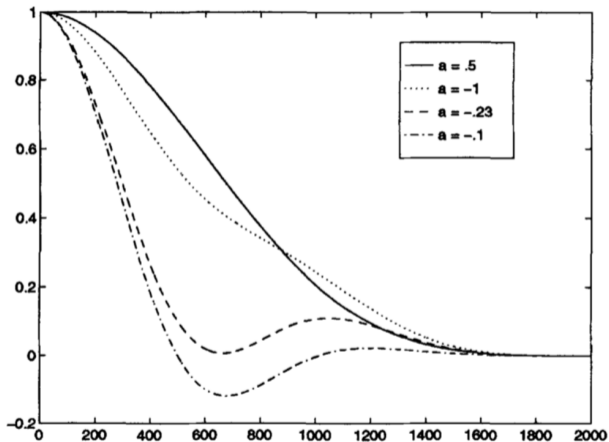


Figure 7. The function $C_0(z, a, c)$ of the example in section 4(c) for $c = 1000$ km and various values of a . See text for explanation.

Figure 1: Figure 7 from Gaspari and Cohn (1999). The function $C_0(z, a, c)$ is the general form of the compactly-supported, fifth-order, piecewise rational correlation function derived in their Sec. 4(c). Typically, $a = 1/2$ (solid black).

The Generalized Gaspari-Cohn (GenGC) Correlation Function

Correlation length for the compactly-supported, piecewise rational:

$$L = c \left(\frac{3(22a^2 + 3a + 1)}{40(8a^2 - 2a + 1)} \right)^{1/2}, \quad a = 1/2 \Rightarrow L = \sqrt{0.3}c$$

Need to generalize this correlation function to allow for *variable* L .

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Generalized Gaspari-Cohn

Allow $a = a_k$ and $c = c_k$ to vary over the spatial index k .
Now $L = L_k$ can vary!

Construction of the GenGC Correlation Function

For fixed $c_k > 0$ and $a_k \in \mathbb{R}$, define the following compactly-supported, radially symmetric functions \mathbb{R}^3 :

$$h_k(\mathbf{r}; a_k, c_k) = \begin{cases} (2(a_k - 1)\|\mathbf{r}\|/c_k + 1)n_k, & 0 \leq \|\mathbf{r}\| \leq c_k/2, \\ 2a_k n_k(1 - \|\mathbf{r}\|/c_k), & c_k/2 \leq \|\mathbf{r}\| \leq c_k, \\ 0, & c_k \leq \|\mathbf{r}\|, \end{cases}$$

with $n_k = (44a_k^2 + 6a_k + 2)^{-1/2}$.

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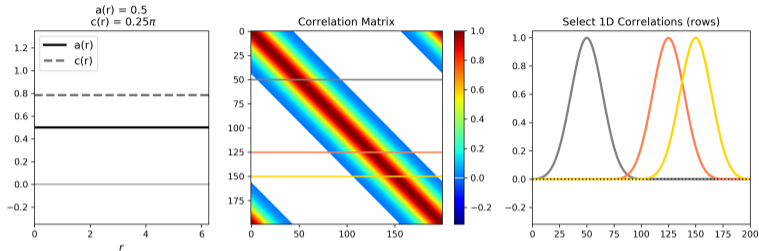
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For each fixed $k, \ell = 1, 2, \dots, m$, GenGC is defined by the convolution,

$$B_{k\ell}(\mathbf{r}, \mathbf{s}) = (h_k * h_\ell)(\mathbf{r} - \mathbf{s}) = \int_{\mathbb{R}^3} h_k(\mathbf{v}) h_\ell(\mathbf{r} - \mathbf{s} - \mathbf{v}) d\mathbf{v}.$$

Generalized Gaspari-Cohn (GenGC), 1D Example

Gaspari and Cohn (1999) on S_1^1 for Constant $a(r)$, $c(r)$ (200 grid points, Chordal Distance Norm)



Generalized Gaspari Cohn on S_1^1 for Continuous $a(r)$, $c(r)$ (200 grid points, Chordal Distance Norm)

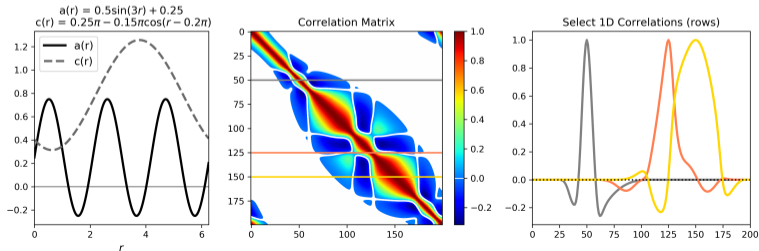


Figure 2: Correlations constructed on the unit circle (S_1^1) for constant a and c (top row) and spatially-varying, continuous a and c (bottom row). White regions in the correlation matrix (middle column) correspond to correlations between -0.003 and 0.003 .

Demonstration

Correlations from Direct, Full Rank Covariance Propagation, $t_f = T$

Correlations Extracted from Full Rank Covariance Propagation

$$c_0 = 0.25, a_0 = 0.5, \sigma_0^2(x) = 1$$

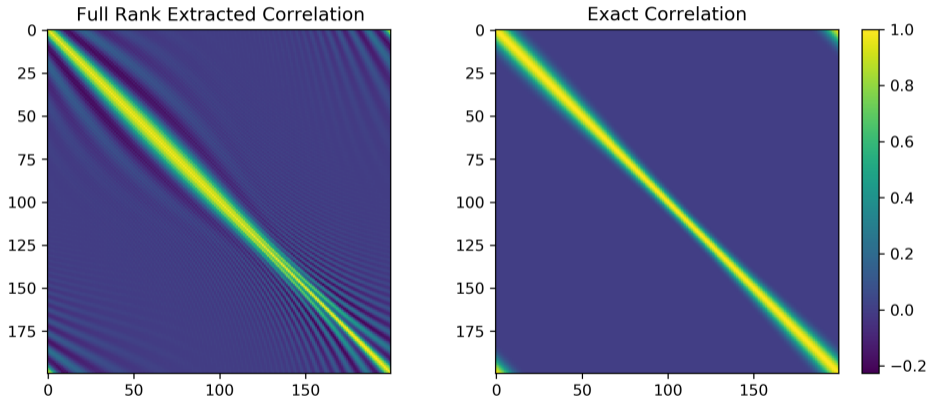


Figure 3: Adapted from Gilpin et al. (2022): Correlation matrices extracted from covariances evolved from the Gaspari-Cohn correlation function for $a_0 = 1/2$, $c_0 = 0.25$, and $\sigma_0^2 = 1$, evolved in time up to slightly after a full time period. Errors in full rank propagation are between -0.713 and 0.450

Current Work, Correlation Reconstruction with GenGC, $t_f = T$

LCE Correlation Test: GenGC with $a = 0.5$ (constant),
evolved L , $L_0 = 0.137$, $c_0 = 0.25$

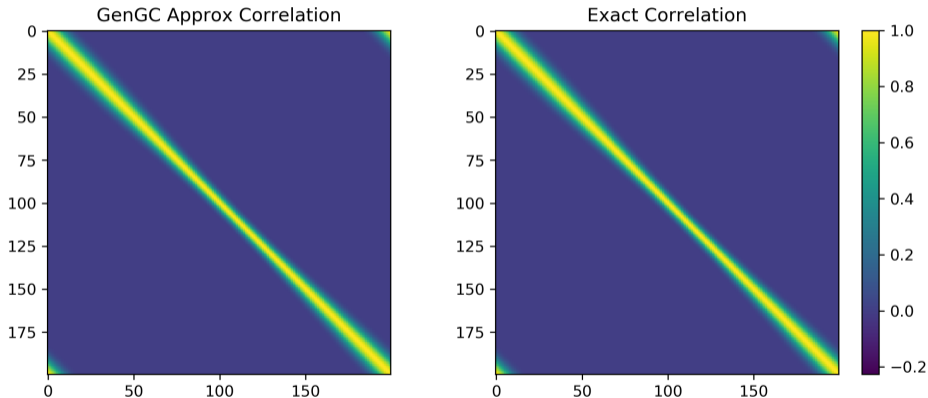


Figure 4: Left: correlation matrix approximated with GenGC using evolved correlation lengths L , $a = 0.5$ constant, $c_0 = 0.25$. Right: the exact correlation matrix. Errors in the GenGC approximation are between -0.0009 and 0.0008 .

Correlations from Direct, Full Rank Covariance Propagation, $t_f = T/2$

Correlations Extracted from Full Rank Covariance Propagation

$$c_0 = 0.25, a_0 = 0.5, \sigma_0^2(x) = 1$$

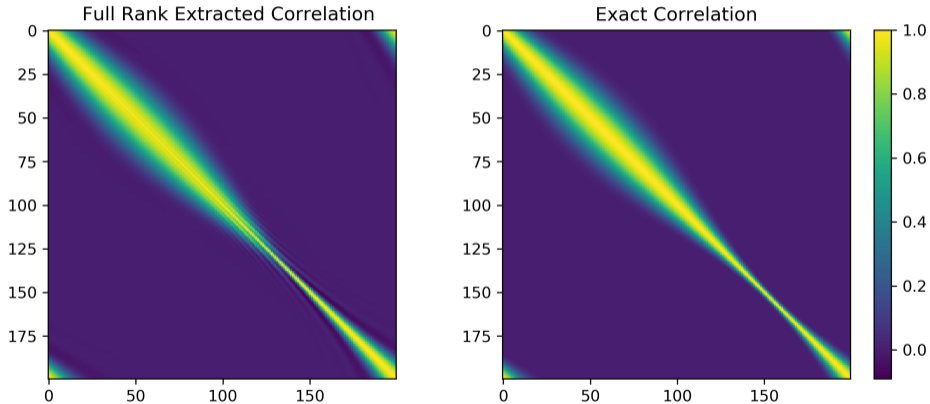


Figure 5: Adapted from Gilpin et al. (2022): Correlation matrices extracted from covariances evolved from the Gaspari-Cohn correlation function for $a_0 = 1/2$, $c_0 = 0.25$, and $\sigma_0^2 = 1$, evolved in time up to slightly after half a time period. Errors in full rank propagation are between -0.557 and 0.339

Current Work, Correlation Reconstruction with GenGC, $t_f = T/2$

LCE Correlation Test: GenGC with $a = 0.5$ (constant),
evolved L , $L_0 = 0.137$, $c_0 = 0.25$

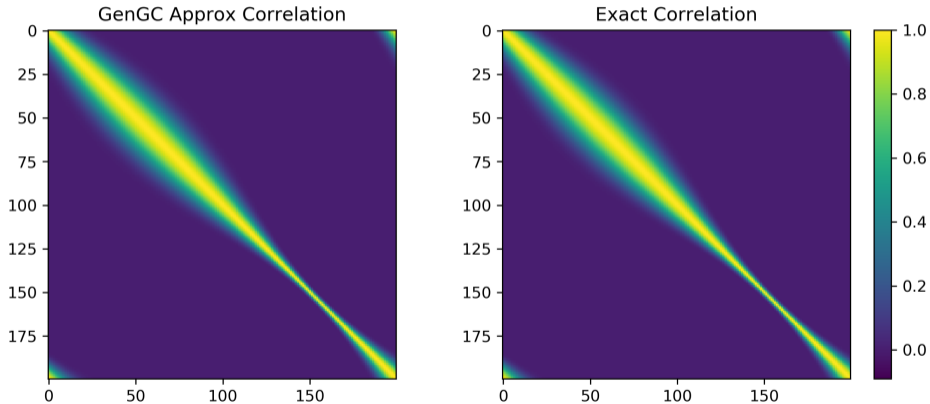


Figure 6: Left: correlation matrix approximated with GenGC using evolved correlation lengths L , $a = 0.5$ constant, $c_0 = 0.25$. Right: the exact correlation matrix. Errors in the GenGC approximation are between -0.0124 and 0.0107 .

Concluding Remarks and Further Investigation

- Local covariance evolution is an alternative means of mitigating problems associated with covariance propagation.
- The Generalized Gaspari-Cohn correlation function has several additional applications (e.g., covariance modeling, localization, coupled data assimilation).

Acknowledgements and Questions

For more information or further discussion, contact Shay at **shay.gilpin@colorado.edu**

Relevant work:

Gilpin, Matsuo, and Cohn, (2023): *A generalized, compactly-supported correlation function for data assimilation applications*, submitted to QJRMS.

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Extra Slides

Generalized Gaspari-Cohn (GenGC), 2D Example

