Text S1

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Maximization of the absolute value of R

Here we derive eqs. (5) of the main manuscript that maximize the Pearson coefficient R. Without loss of generality let us assume that the variance of f(t) equals unity. Then, R reads

$$R = \frac{\operatorname{cov}(f, p_a)}{\sigma_a} = \frac{\sum_{i=1}^d \alpha_i \operatorname{cov}(f, p_i)}{\left[\sum_{i,j=1}^d \alpha_i \alpha_j \operatorname{cov}(p_i, p_j)\right]^{1/2}}$$
(1)

Note that R is invariant to a scaling of all coefficients α_i by a constant and hence, $\sum_i \alpha_i^2$ must not be treated as a constraint for the optimization. The maximum of the absolute value of R can be found by maximizing R^2 . For compact notation, the abbreviations $c_i := \operatorname{cov}(f, p_i)$ and $\gamma_{ij} := \operatorname{cov}(p_i, p_j)$ are introduced. Then, the square of the correlation coefficient reads $R^2 = \left[\sum_{i=1}^d \alpha_i c_i\right]^2 / \sum_{i,j=1}^d \alpha_i \alpha_j \gamma_{ij}$, and its derivate with respect to α_ℓ is

$$\frac{\partial[R^2]}{\partial\alpha_\ell} = \frac{2\sum_i \alpha_i c_i}{\left(\sum_{ij} \alpha_i \alpha_j \gamma_{ij}\right)^2} \left[c_\ell \left(\sum_{ij} \alpha_i \alpha_j \gamma_{ij}\right) - \left(\sum_i \alpha_i c_i\right) \left(\sum_i \alpha_i \gamma_{i\ell}\right) \right]. \tag{2}$$

 R^2 is maximized by setting all d derivatives of R^2 to equal zero. $\partial[R^2]/\partial \alpha_\ell = 0$ is satisfied for all ℓ if the α_i obey

$$\kappa \sum_{i} \alpha_{i} \gamma_{i\ell} = c_{\ell}, \ \ell = 1, \dots, d.$$
(3)

Here, κ is an arbitrary constant which can be used to normalize the collective vector **a** (such that $\sum_i \alpha_i^2 = 1$), but which does not affect R. For $\kappa = 1$, eqs. (3) are equivalent to equivalent to eqs. (5) of the main manuscript. To verify that the set of α_i computed via eqs. (3) defines a local maximum in R^2 , the eigenvalues of the Hessian $H_{k\ell}$ of R^2 must be non-positive at the maximum. A straightforward calculation yields

$$H_{k\ell}\Big|_{\max} = \left. \frac{\partial^2 [R^2]}{\partial \alpha_k \partial \alpha_\ell} \right|_{\max} = -2\kappa^2 \left[\gamma_{k\ell} - \frac{c_k c_\ell}{\sum_i \kappa \alpha_i c_i} \right],\tag{4}$$

where $|_{\max}$ indicates that the α_i were computed according to eqs. (3). The eigenvalues of $H_{k\ell}|_{\max}$ can be computed numerically which has been implemented into the FMA tool. For the examples presented in this study all eigenvalues of $H_{k\ell}|_{\max}$ were negative, always with the exception of one zero eigenvalue corresponding the the scaling of all α_i by a scalar factor. This finding is expected because R is invariant to a scaling of the coefficients α_i .

Construction of best linear model $m_f(t)$

Let $D = \langle [f - m_f]^2 \rangle$ denote the MSD between the functional quantity f(t) and its model $m_f(t) = \langle f \rangle + \sum_{i=1}^d \beta_i (p_i(t) - \langle p_i \rangle)$. D is minimized with respect to the d parameters β_i by setting the d partial derivatives $\partial D/\partial \beta_\ell$ to zero, immediately yielding eqs. (7) of the main manuscript. It is straightforward to show that the second derivatives of D are twice the elements of the covariance matrix of the principal

components $p_i(t)$, i.e., $\partial^2 D / \partial \beta_\ell \partial \beta_k = 2 \operatorname{cov}(p_\ell, p_k)$. Because covariance matrices are positive semidefinite [1], the set of β_i defined by eqs. (7) of the main manuscript correspond to a local minimum of D.

References

1. Mardia KV, Kent JT, Bibby JM (1979) Multivariate Analysis. San Diego: Academic Press.