

Time-lagged partial correlations of financial time series with high dimensional conditions

Econophysics PY538

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Outline

Data

Partial correlation

Results - Synchronous correlation

Results - Time-lagged correlations

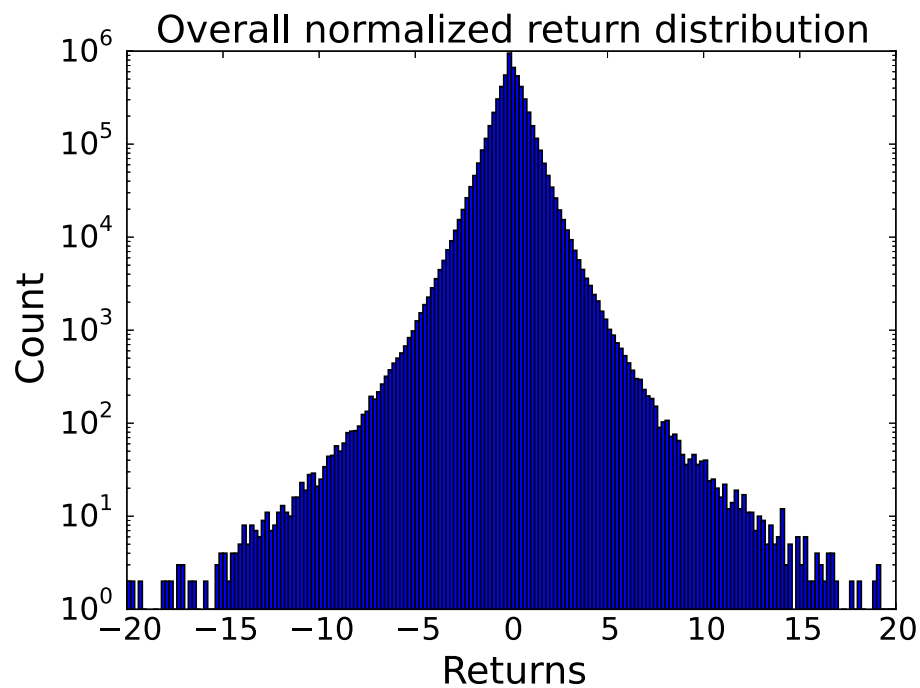
Conclusion

Data

- ▶ New York Stock Exchange 2001-2003
- ▶ Returns of the $N = 100$ largest capitalized stocks
- ▶ 748 trading days, 78 data points per day, 5 min interval
- ▶ Total: $T = 58344$ data points
- ▶ Data matrix X with dimension $N \times T$

Return distribution

- ▶ Rescaled data: zero mean, unit variance $x_i(t) = \frac{\tilde{x}_i(t) - \mu_{\tilde{x},i}}{\sigma_{\tilde{x},i}}$



Market mode

Covariance & Correlation matrix

$$\Sigma(X, X) = \rho(X, X) = \frac{1}{T} X X^T$$

with Eigenvalues $\lambda_1 \geq \lambda_2 \geq \dots$ and eigenvectors u_1, u_2, \dots

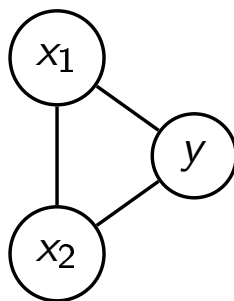
Market mode

$$x_m(t) = \sum_{j=1}^N u_{1j} x_j(t) \quad \Rightarrow \quad x_i(t) = \underbrace{\alpha_i}_{=0} + \beta_i x_m(t) + \epsilon_i(t)$$

→ Market mode removed data X_{res} with $\epsilon_i(t)$

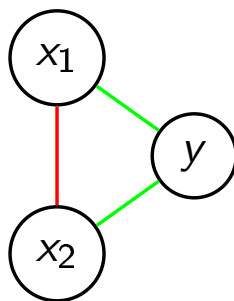
Partial Correlation

- ▶ **Question:** What is the correlation between two variables x_1, x_2 given y , a third one?



Partial Correlation

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- ▶ **Answer:** Partial correlation $\rho(x_1, x_2|y)$

Partial Correlation

Conditional mean

$$\hat{x}_i(y) = \underbrace{\mathbb{E}(x_i)}_{=0} + \frac{\sigma(x, y)}{\sigma(y, y)} \left(y - \underbrace{\mathbb{E}(y)}_{=0} \right)$$

Partial covariance

$$\sigma(x_1, x_2|y) = \text{Cov}(x_1 - \hat{x}_1(y), x_2 - \hat{x}_2(y))$$

Partial Correlation

Conditional mean for $X = \{x_1, x_2\}$ and $Y = \{y_1, y_2, \dots, y_m\}$

$$\hat{X}(Y) = \Sigma_{XY} \Sigma_{YY}^{-1} Y$$

Partial covariance

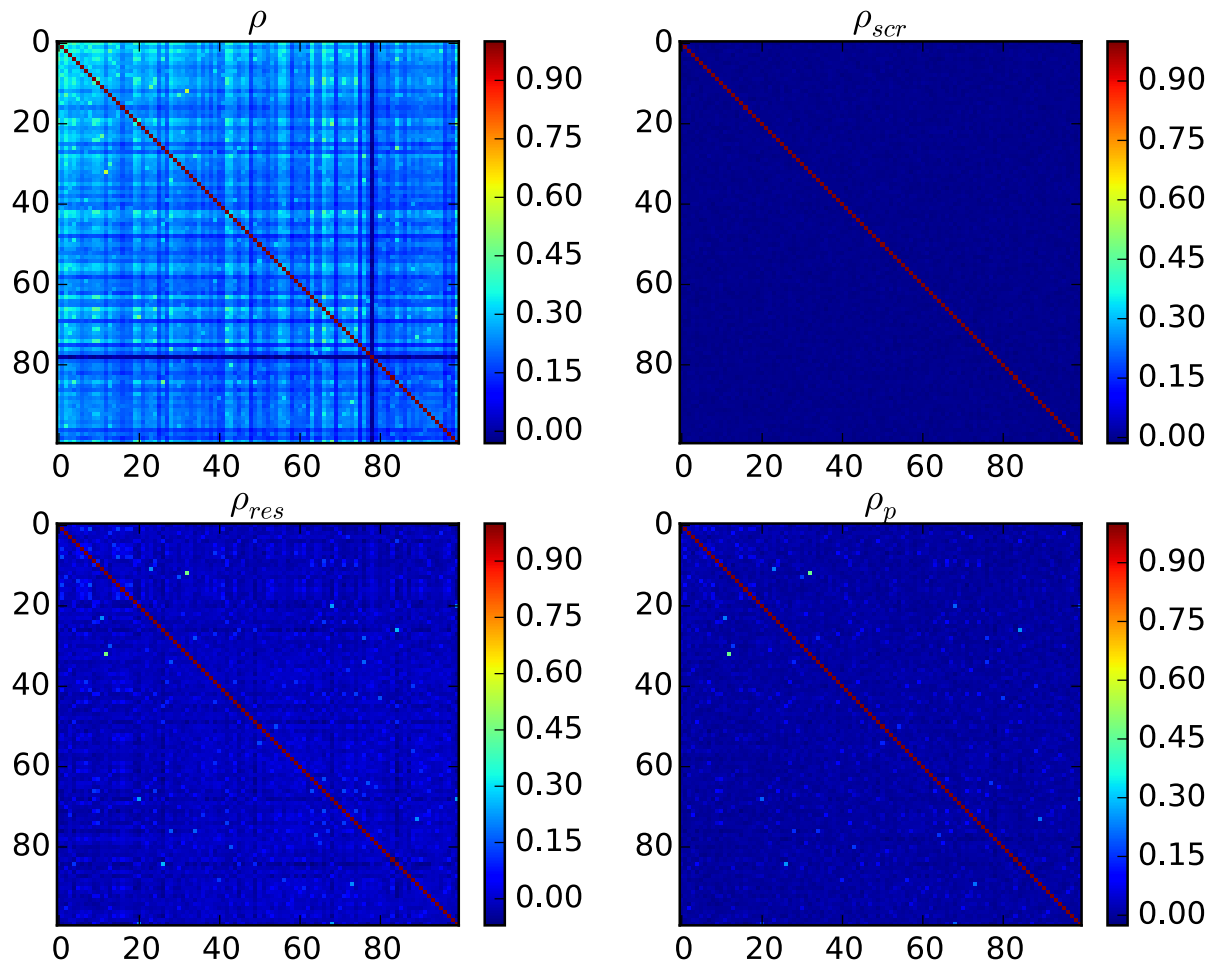
$$\begin{aligned} \Sigma_{XX|Y} &= \text{Cov} \left(X - \hat{X}(Y), X - \hat{X}(Y) \right) \\ &= \Sigma_{XX} - \Sigma_{XY} \Sigma_{YY}^{-1} \Sigma_{YX} = \begin{pmatrix} \sigma_{11|Y} & \sigma_{12|Y} \\ \sigma_{21|Y} & \sigma_{22|Y} \end{pmatrix} \end{aligned}$$

Partial correlation

$$\rho_{12|Y} = \frac{\sigma_{12|Y}}{\sqrt{\sigma_{11|Y} \sigma_{22|Y}}}$$

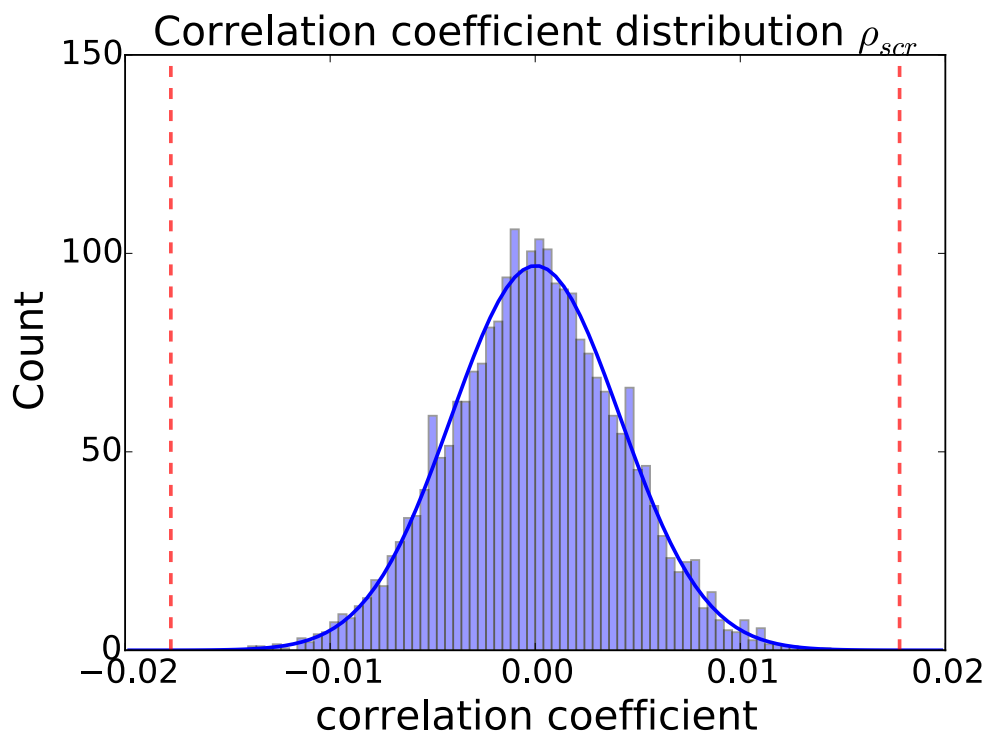
Synchronous Correlation

Correlation matrices



Synchronous Correlation

- ▶ Noise limit: $\rho_{\max} \sim \sqrt{2 \ln(N^2) / T} = 0.01777$



Time-lagged correlation

Market mode removed data X_{res}

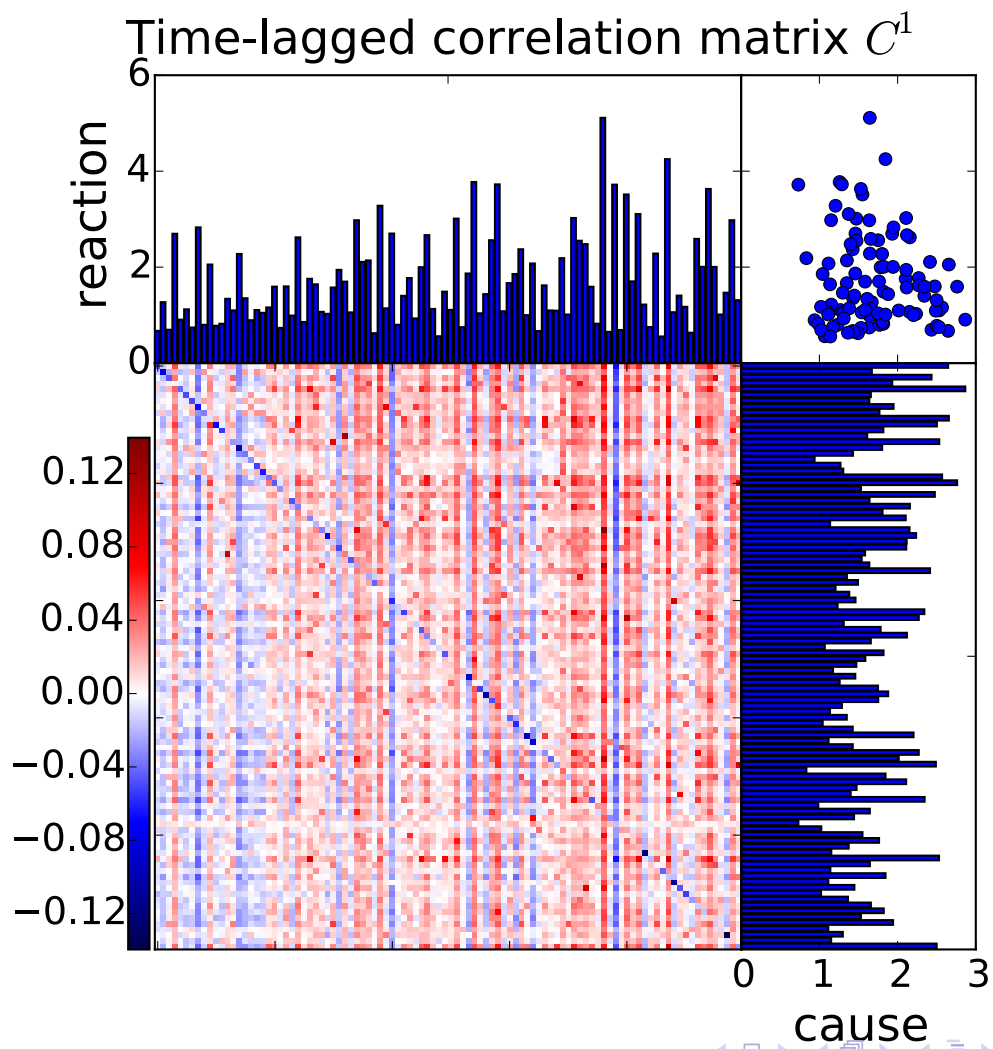
$$C_{res}^{\tau} = \frac{1}{T - \tau} \sum_{t=1}^{T-\tau} X_{res}(t) X_{res}^T(t + \tau)$$

Time-lagged partial correlation

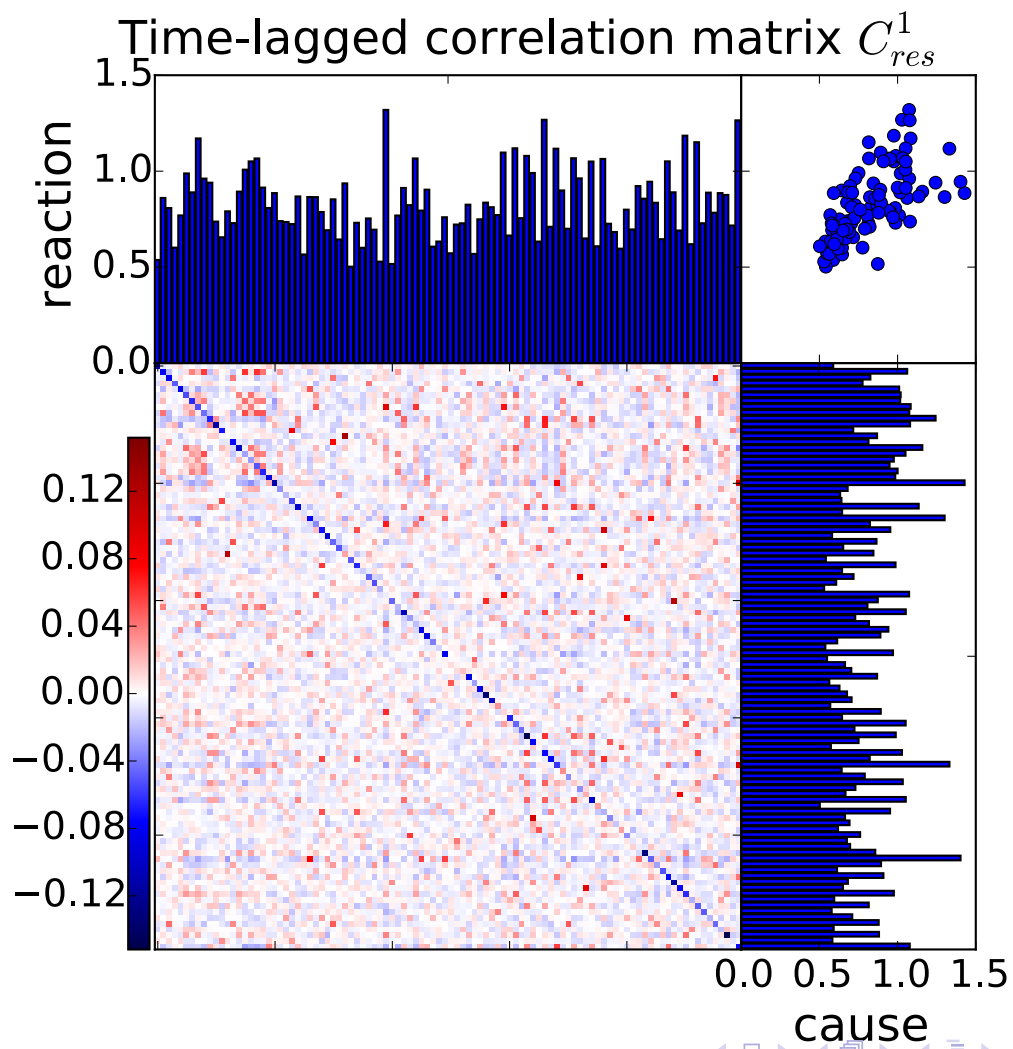
High dimensional condition vector, dim: $(\tau N - 2) \times (T - \tau)$

$$Y = \left\{ \begin{array}{l} x_1(t), \dots, x_{i-1}(t), x_{i+1}(t), \dots, x_N(t), \dots, \\ x_1(t + (\tau - k)), \dots, x_N(t + (\tau - k)), \dots, \\ x_1(t + \tau), \dots, x_{j-1}(t + \tau), x_{j+1}(t + \tau), \dots, x_N(t + \tau) \end{array} \right\}$$

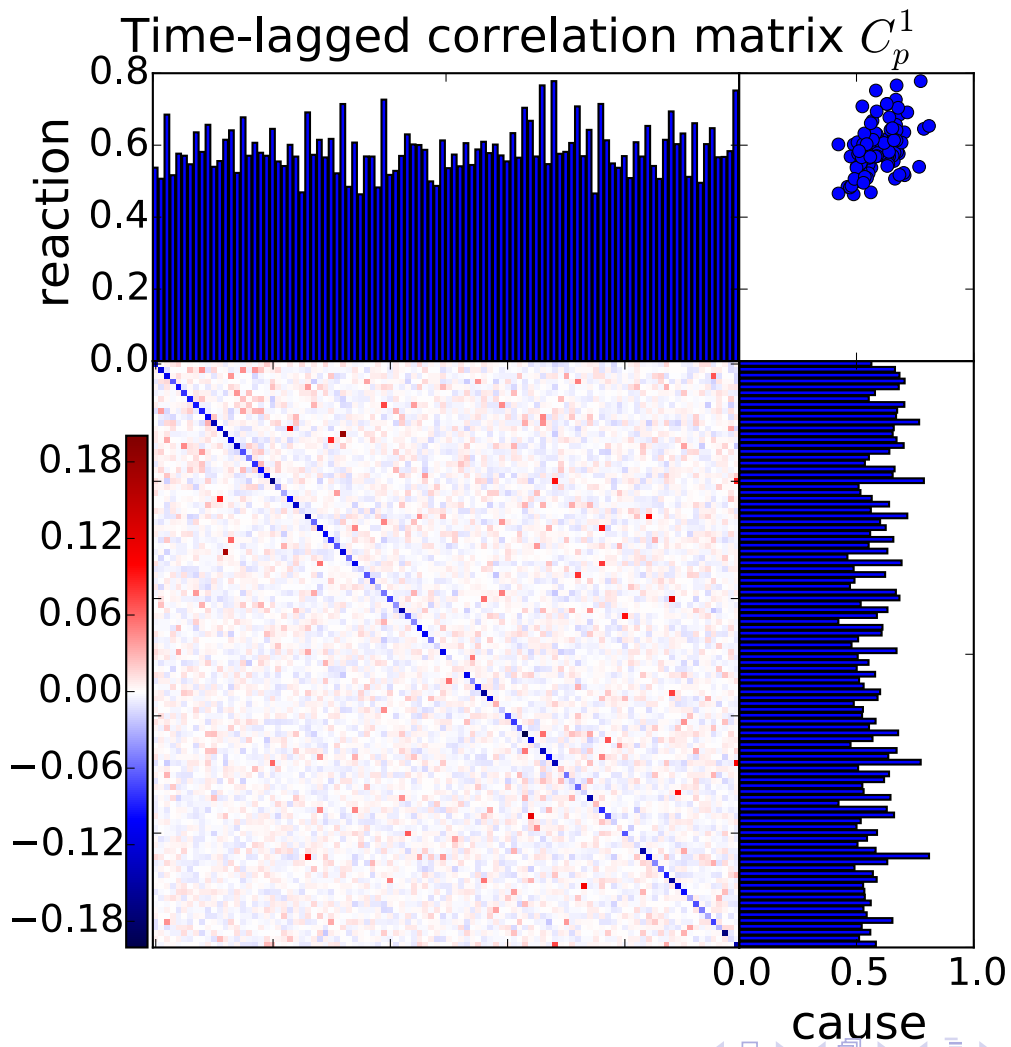
Time-lagged Correlation ($\tau = 1$)



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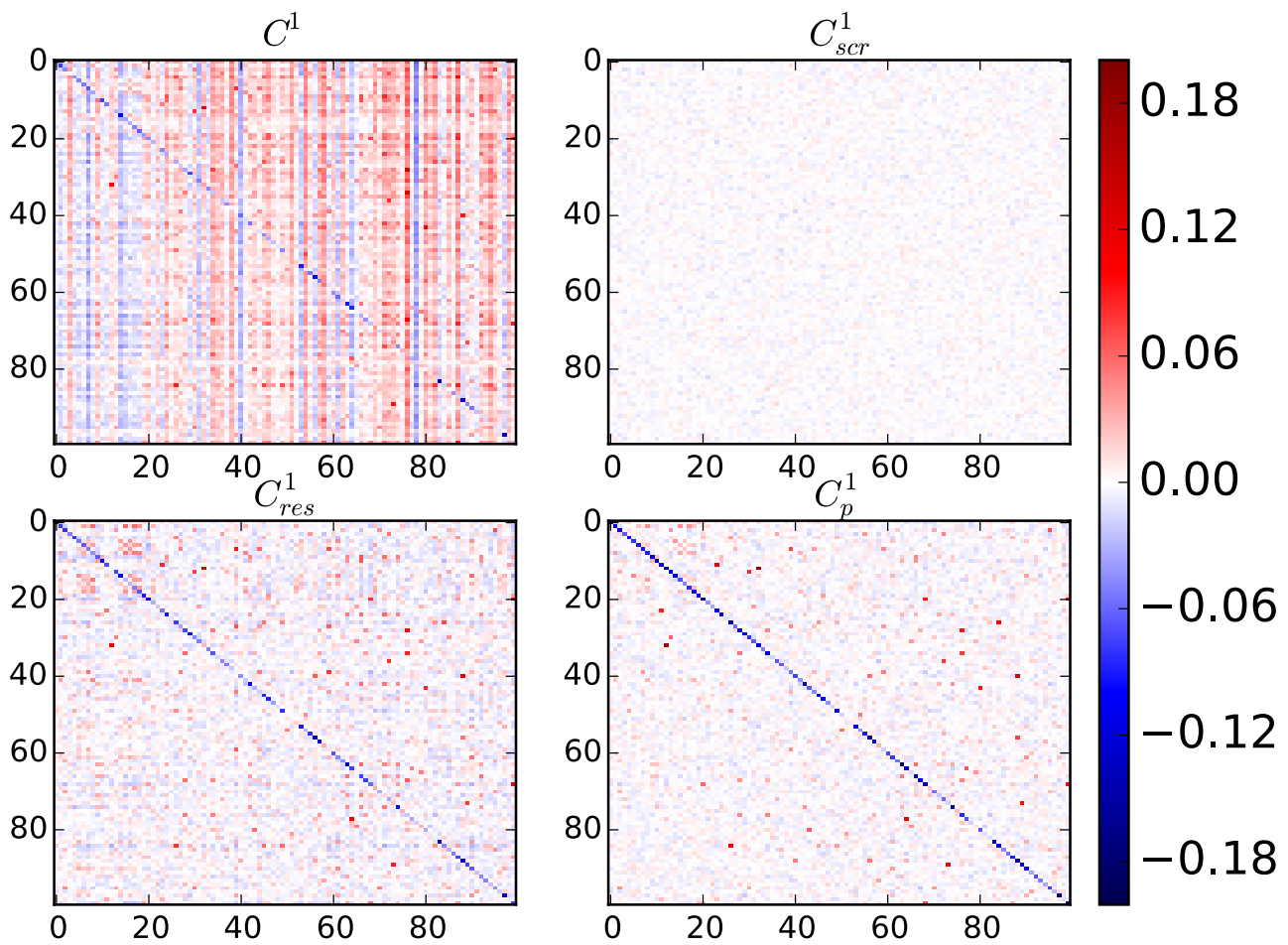


Time-lagged Correlation ($\tau = 1$)



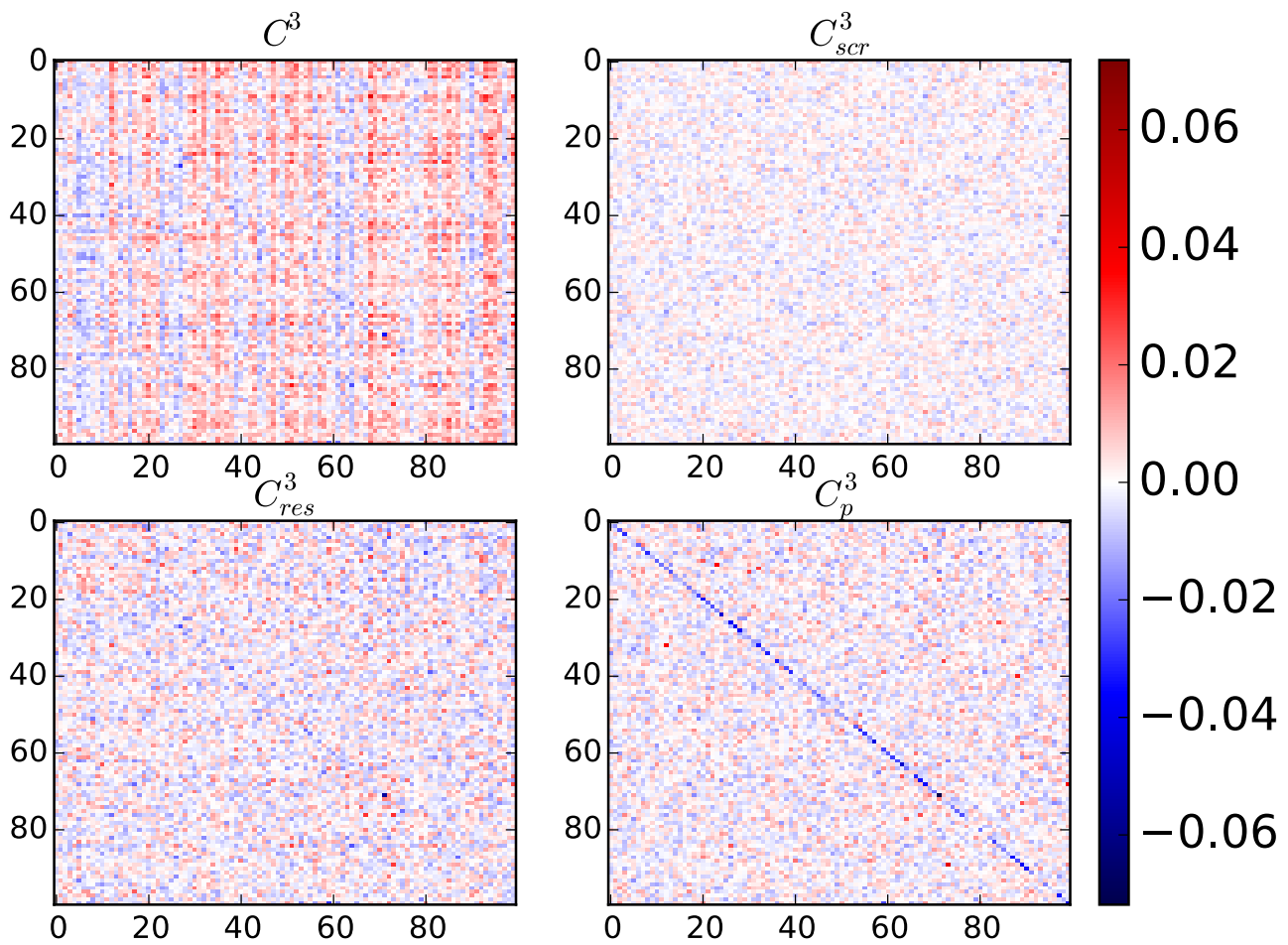
Time-lagged Correlation ($\tau = 1$)

Time-lagged Correlation matrices for lag 1



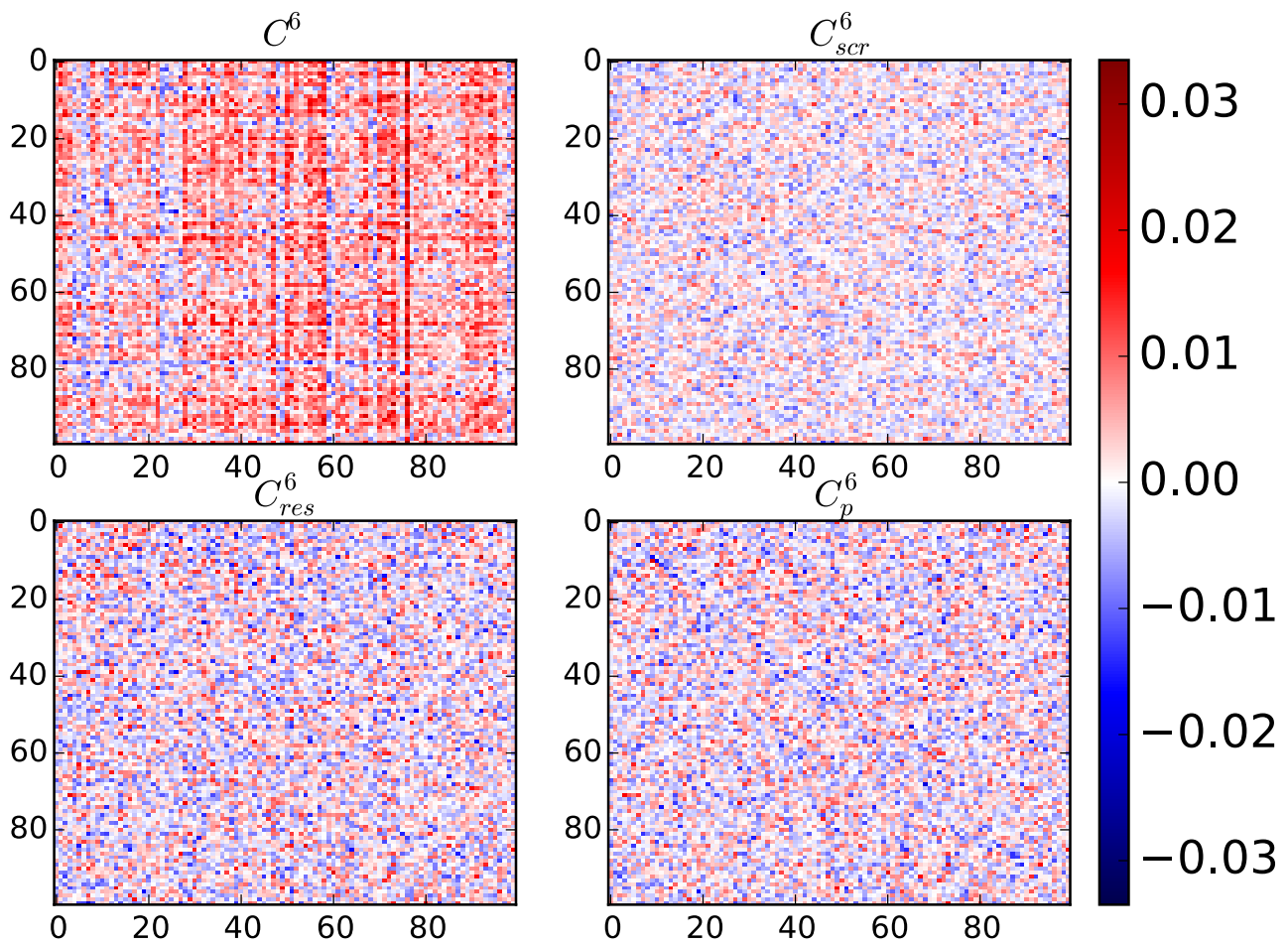
Time-lagged Correlation ($\tau = 3$)

Time-lagged Correlation matrices for lag 3



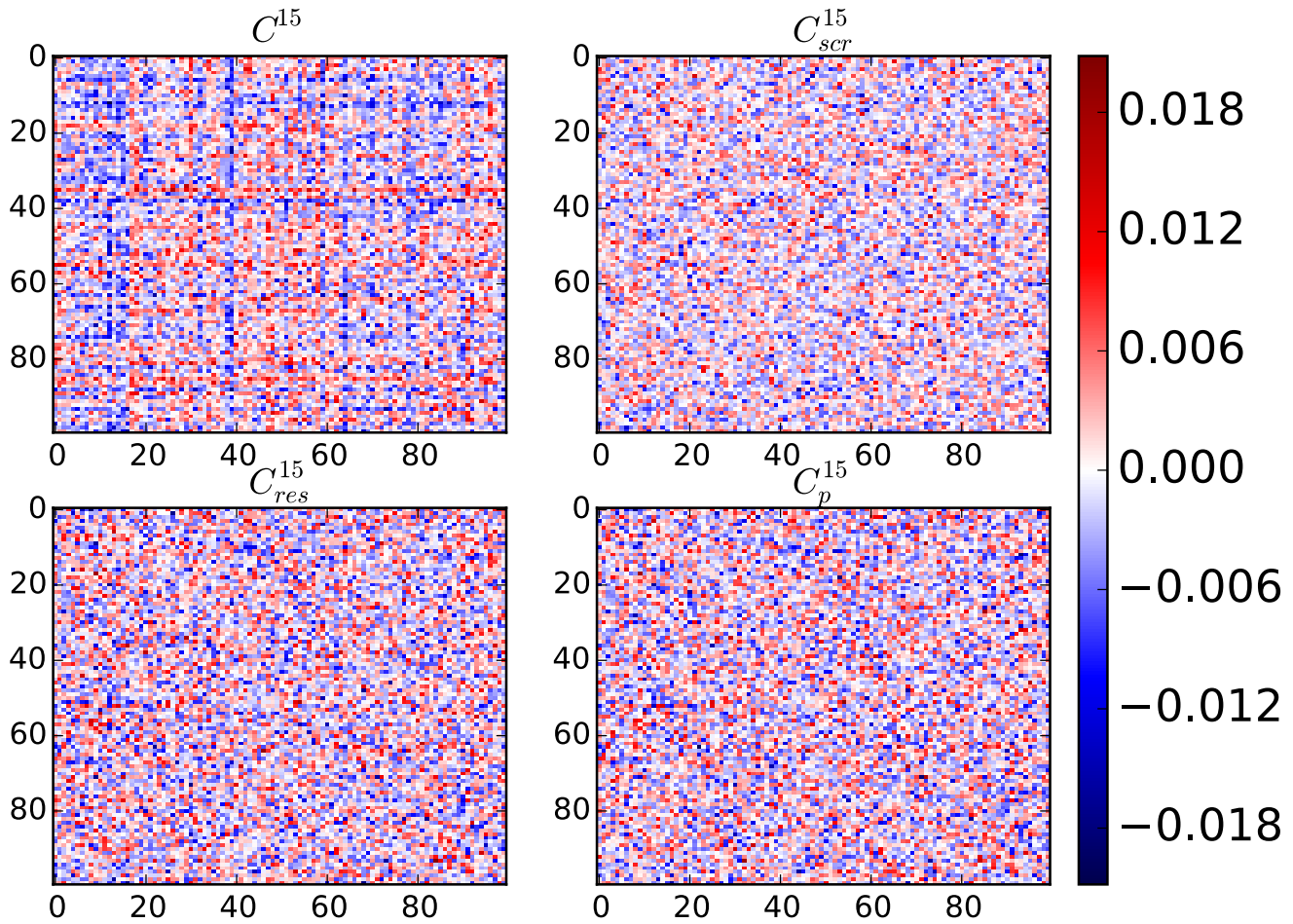
Time-lagged Correlation ($\tau = 6$)

Time-lagged Correlation matrices for lag 6

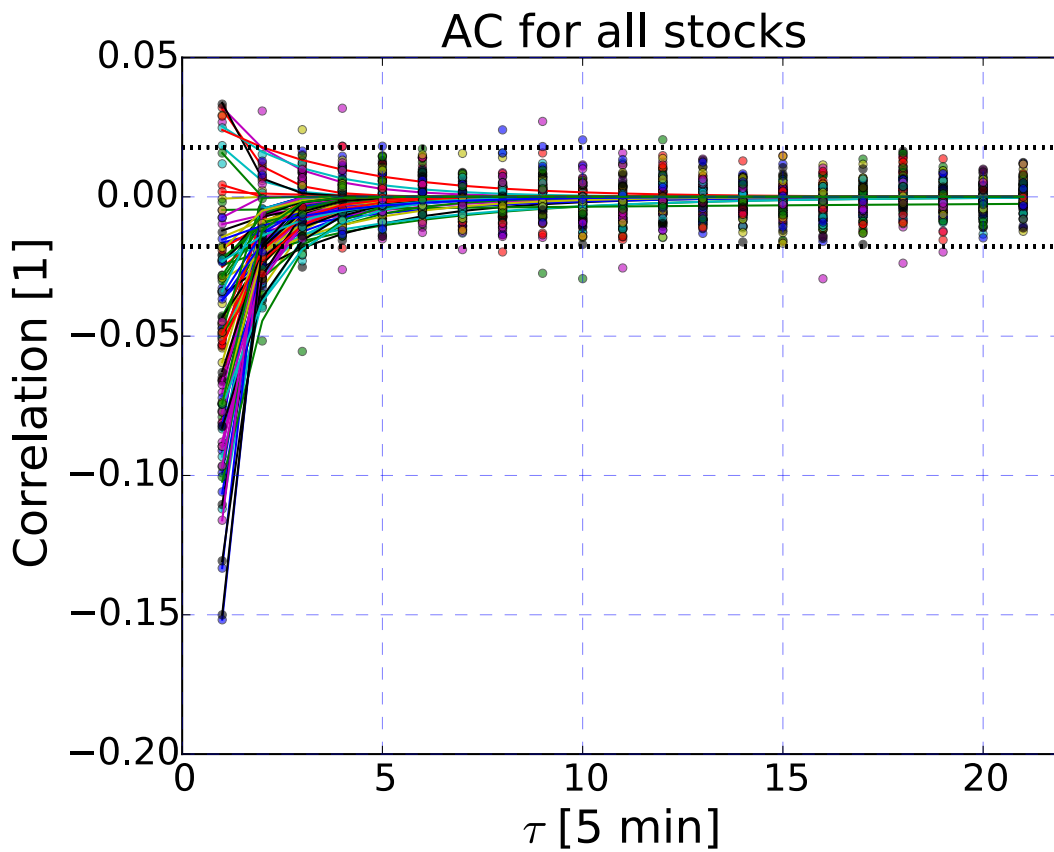


Time-lagged Correlation ($\tau = 15$)

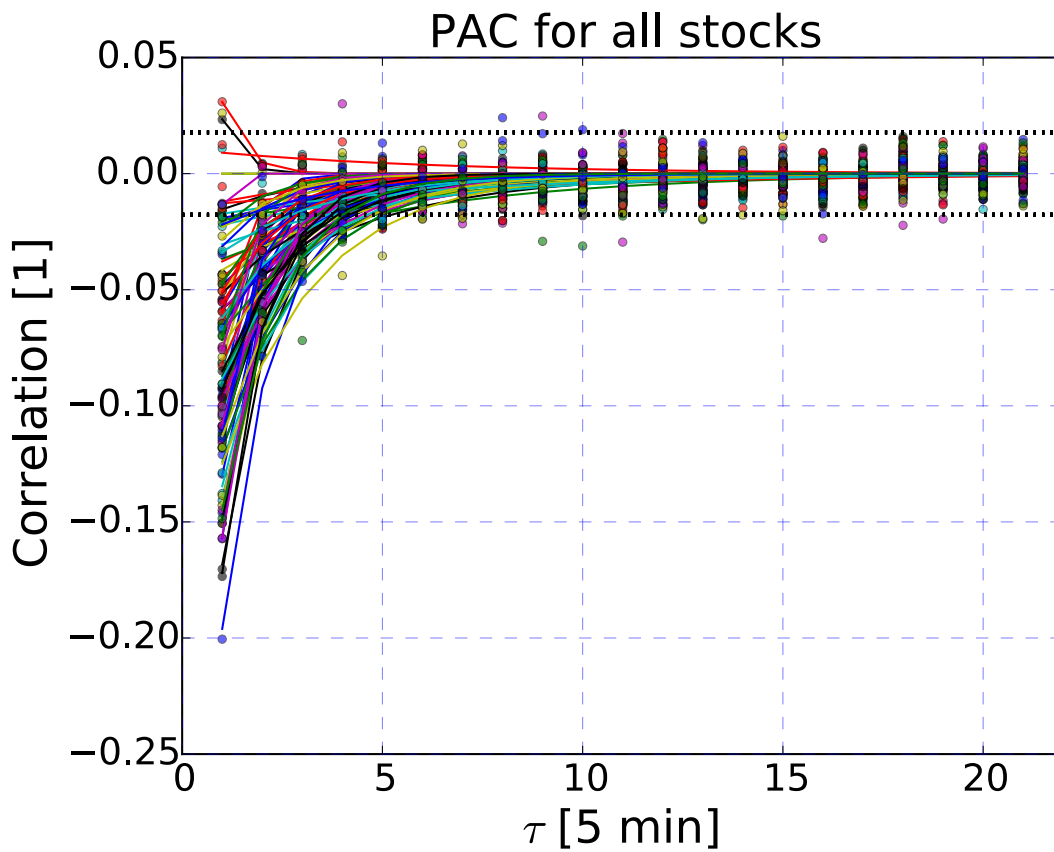
Time-lagged Correlation matrices for lag 15



Autocorrelations - market mode removed

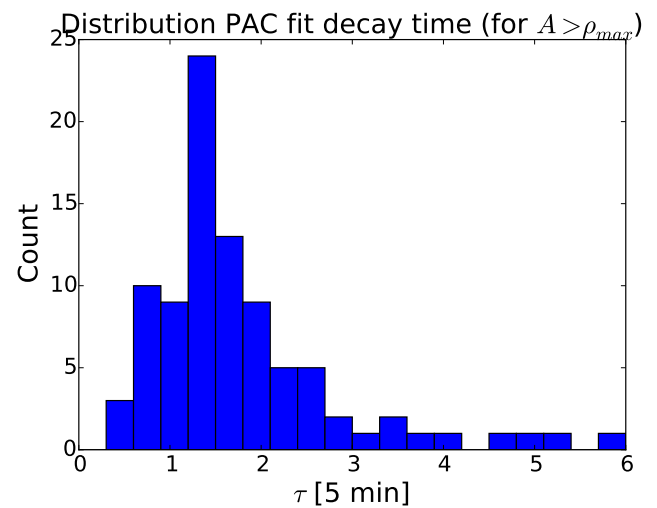
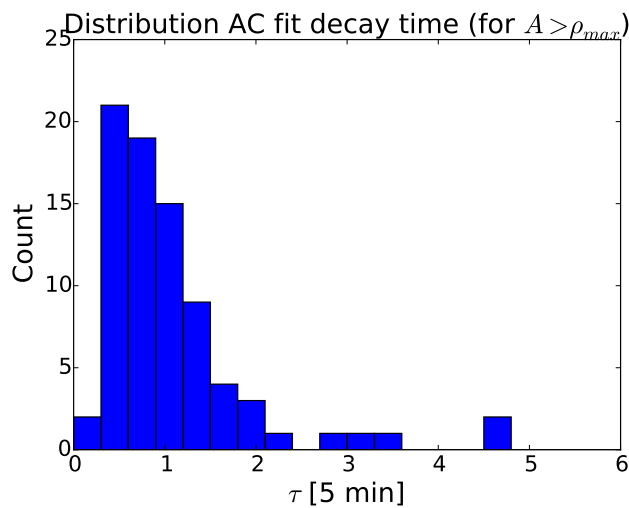


Partial autocorrelations



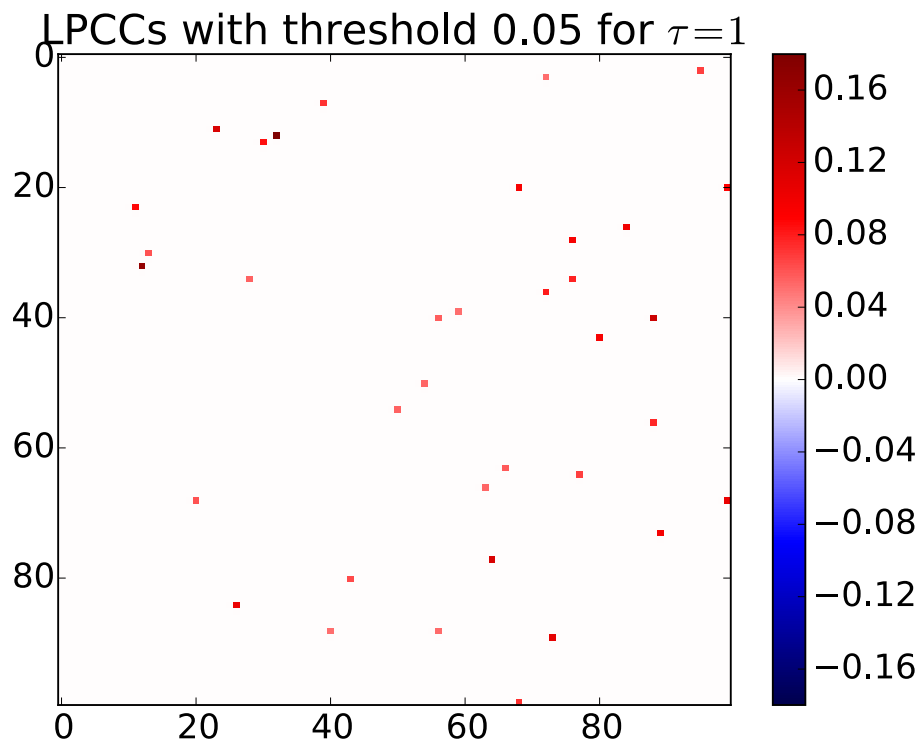
Fit parameter: exponential decay time

- ▶ Consider only if fit amplitude A is outside noise region
- ▶ AC: decay time $\sim 3 - 5$ min
- ▶ PAC: decay time ~ 7 min

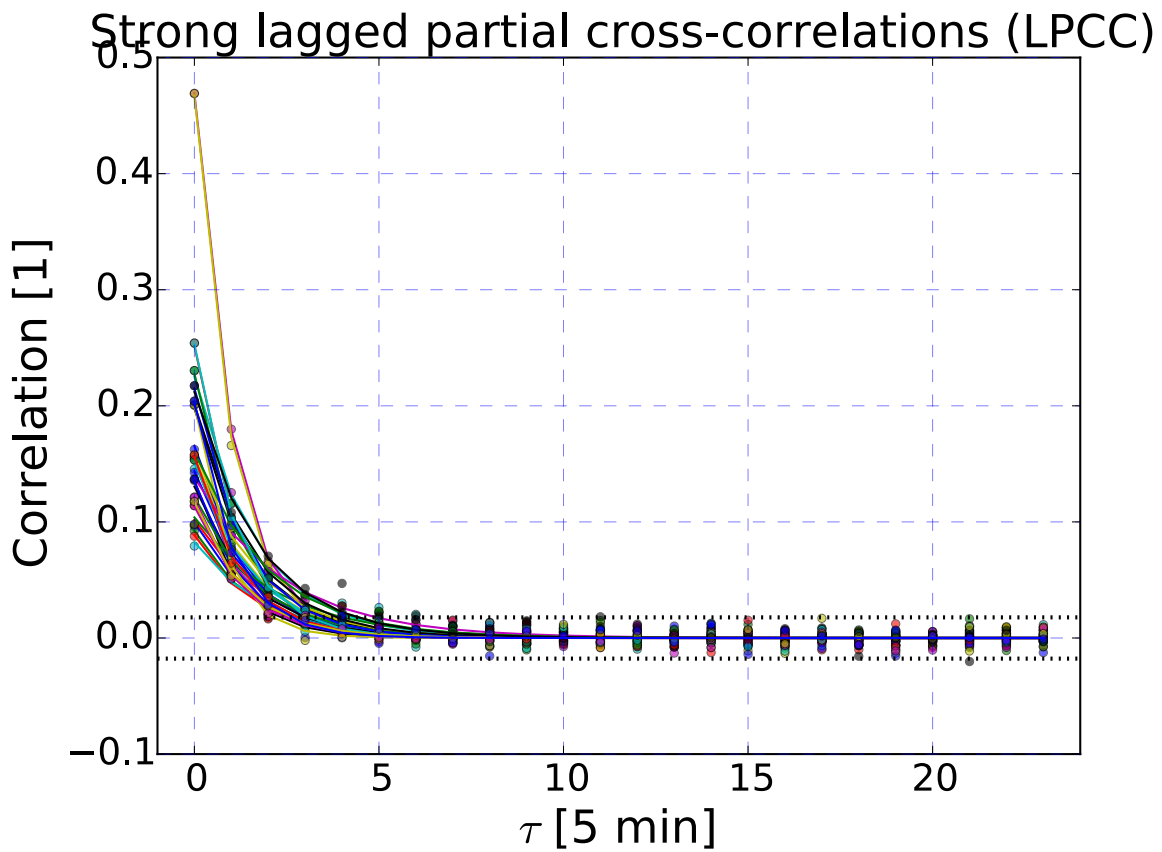


Strong partial cross-correlations

- ▶ Filter threshold for lag 1: $0.05 \approx 3\rho_{max}$

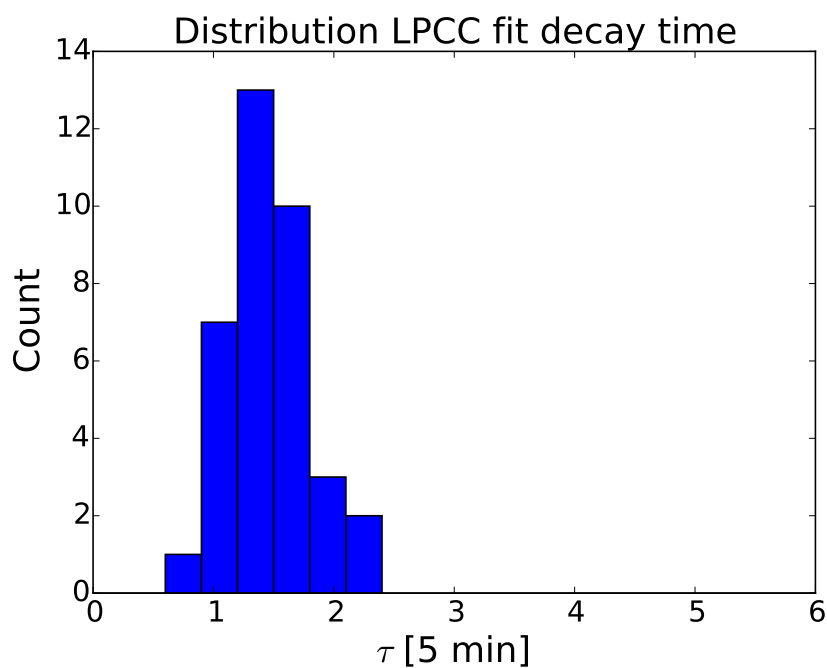


Partial cross-correlations

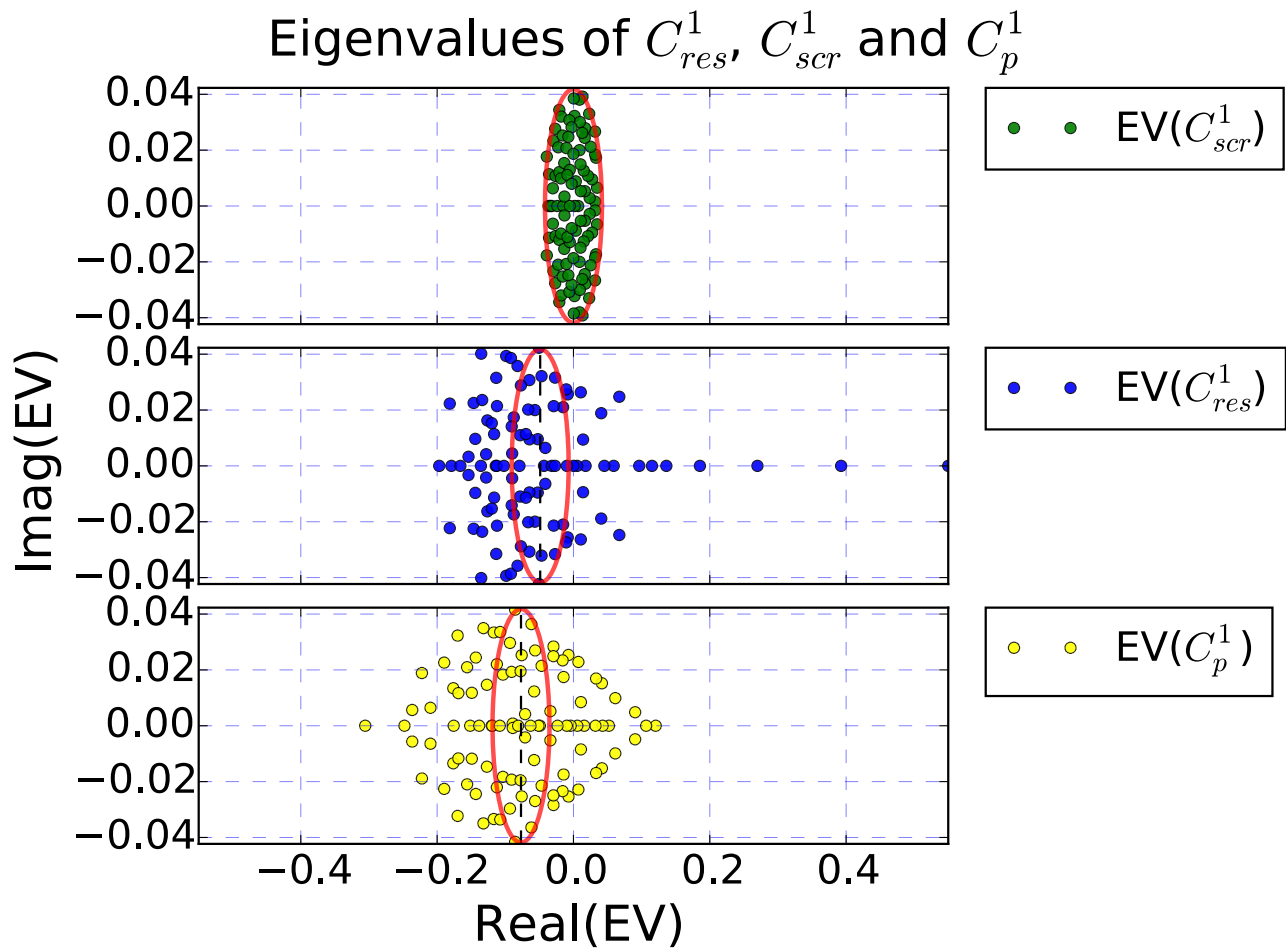


Partial cross-correlations

- ▶ Same decay time scale as partial autocorrelations, $\tau \approx 7$ min

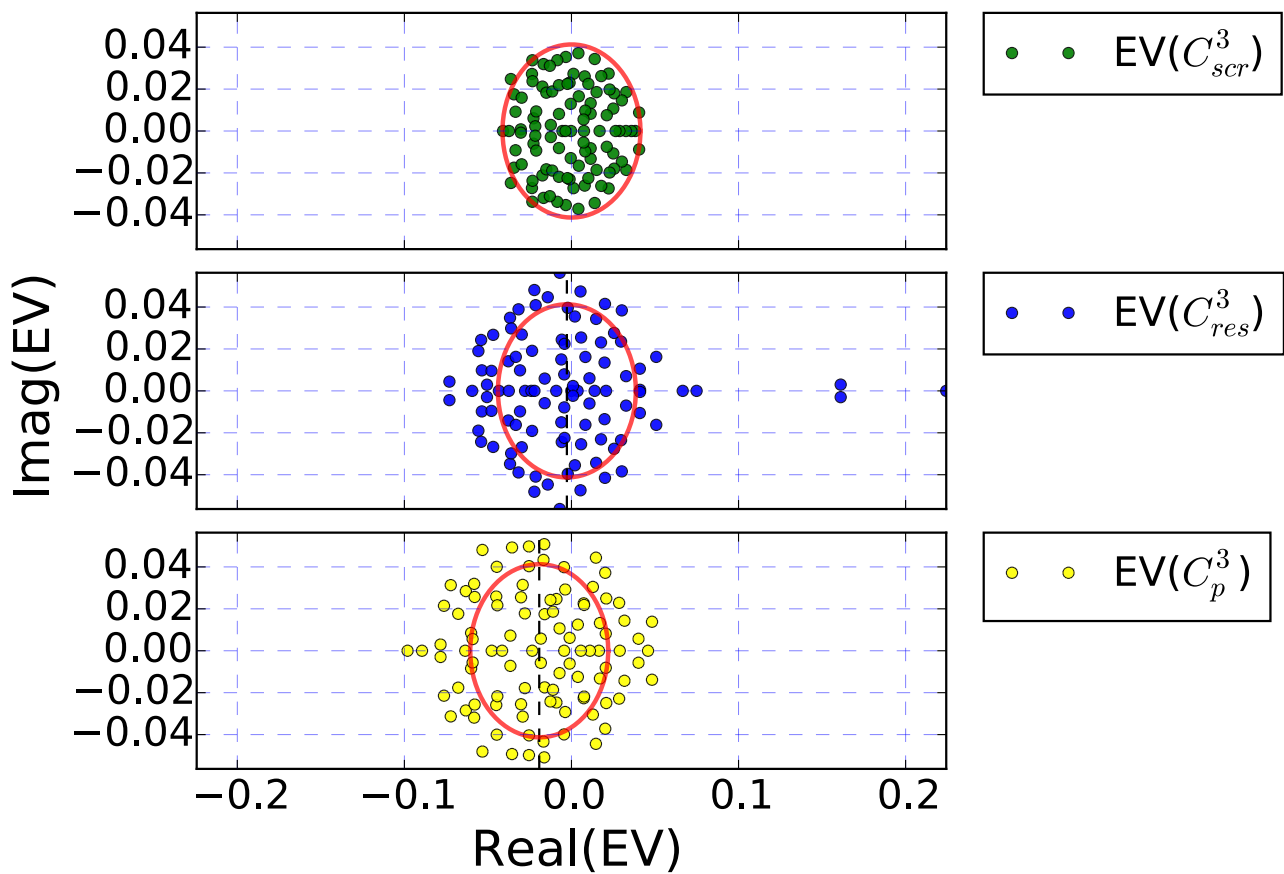


Eigenvalue distribution ($\tau = 1$)



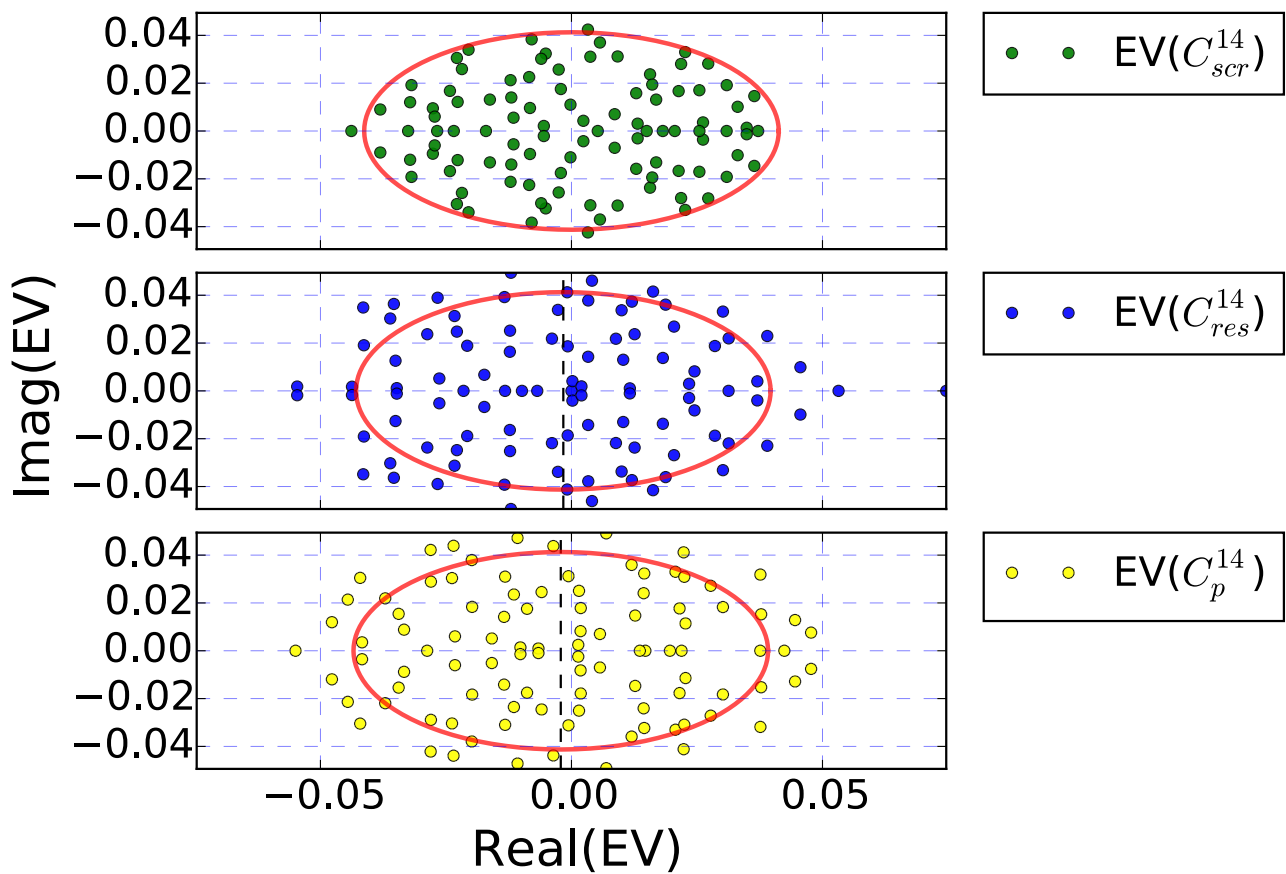
Eigenvalue distribution ($\tau = 3$)

Eigenvalues of C_{res}^3 , C_{scr}^3 and C_p^3



Eigenvalue distribution ($\tau = 14$)

Eigenvalues of C_{res}^{14} , C_{scr}^{14} and C_p^{14}



Conclusion

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Stock market - NYSE

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- ▶ raw correlation damped by mutual third party correlations

Conclusion

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- ▶ good tool to investigate underlying correlation network of a system
- ▶ conditions can be extended arbitrarily

Stock market - NYSE

- ▶ typical decay time for correlations: 7 min
- ▶ raw correlation damped by mutual third party correlations
- ▶ almost no negative time-lagged cross-correlations

Outlook

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Stock market - NYSE

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- ▶ include time-lagged partial correlations in cluster identification → new dimension

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Stock market - NYSE

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- ▶ include time-lagged partial correlations in cluster identification → new dimension
- ▶ Plot correlation network with time dimension

Outlook

General

- ▶ Parallel computing could speed up calculations

Stock market - NYSE

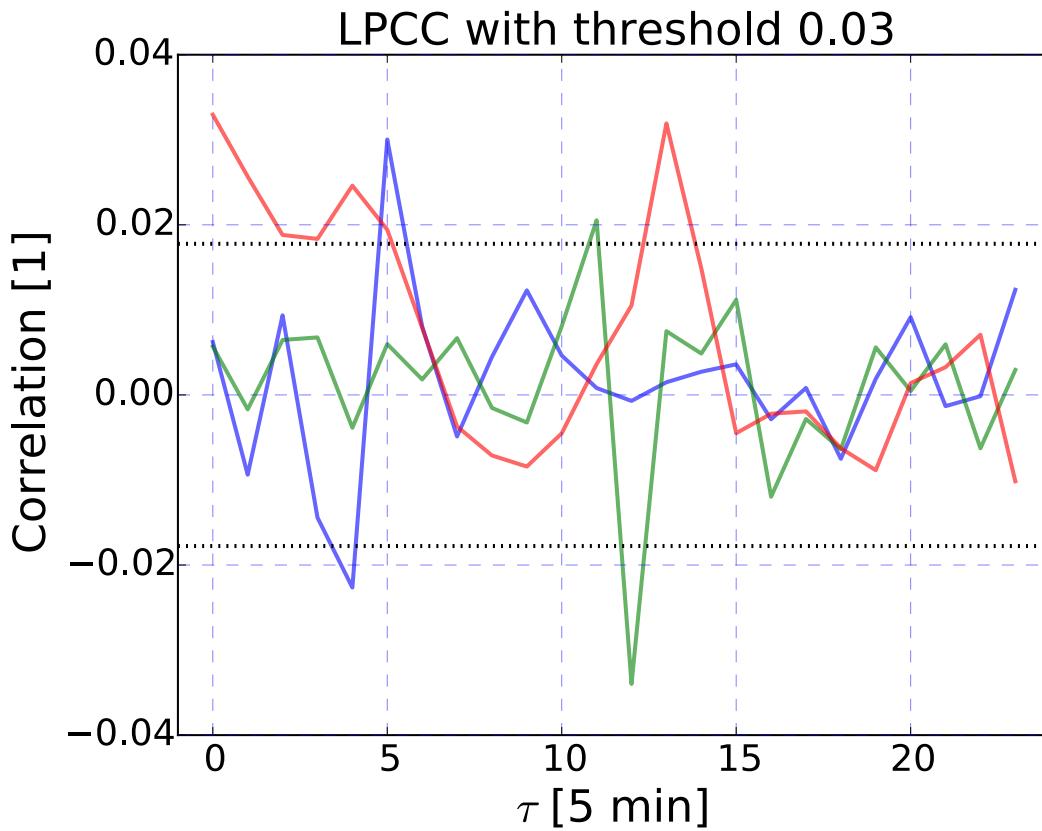
- ▶ identify sectors and subsectors with synchronous partial correlations and compare to older results
- ▶ include time-lagged partial correlations in cluster identification → new dimension
- ▶ Plot correlation network with time dimension
- ▶ Study SVD decompositions

End

Thank you for your attention!

And thanks to Chester!

Backup slides



Backup slides

