

Noncentral Skewed HEAVY Model for High-Dimensional Realized Returns and Covariances Matrices

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Abstract

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JEL subject classifications:

Keywords:

1 Introduction

2 Noncentral HEAVY GAS skew tF Model

Assumptions:

- (A1) The vector of asset returns $\mathbf{y}_t \in \mathbb{R}^p$ conditionally on the observation set \mathcal{F}_{t-1} is assumed to follow a normalized multivariate skew Student's t distribution with density given by

$$f_{\mathbf{y}}(\mathbf{y}_t | \Sigma_t, \mathcal{F}_{t-1}; \nu_0) = 2f_{T_{\nu_0}}(\mathbf{y}_t | \Sigma_t, \mathcal{F}_{t-1}) F_{T_{\nu_0+p}} \left(\frac{\boldsymbol{\alpha}^\top \mathbf{y}_t}{(\nu_0 - 2 + \mathbf{y}_t^\top \Sigma_t^{-1} \mathbf{y}_t)^{1/2}} \sqrt{\nu_0 + p} \right), \quad (2.1)$$

where the function $F_{T_{\nu_0}}(\cdot)$ denotes the c.d.f. of the standard univariate t -distribution with ν_0 degrees of freedom and $f_{T_{\nu_0}}(\cdot)$ is density of the (standardized) multivariate central t distribution with ν_0 degrees of freedom given by

$$f_{T_{\nu_0}}(\mathbf{y}_t | \Sigma_t, \mathcal{F}_{t-1}) = \frac{\Gamma[(\nu_0 + p)/2]}{\Gamma[\nu_0/2][\pi(\nu_0 - 2)]^{p/2}} |\Sigma_t|^{-1/2} \left(1 + \frac{\mathbf{y}_t^\top \Sigma_t^{-1} \mathbf{y}_t}{\nu_0 - 2} \right)^{-(\nu_0 + p)/2}. \quad (2.2)$$

The vector $\boldsymbol{\alpha}$ stays for the skewness parameters.

Basic properties of multivariate skew t :

- (i) $\boldsymbol{\alpha} = \mathbf{0}$ implies the central multivariate t -distribution with ν_0 degrees of freedom.
- (ii) $\lim_{\nu_0 \rightarrow \infty} f_{T_{\nu_0}}(\mathbf{y}_t) = 2\phi_{\mathbf{0}, \Sigma_t}(\mathbf{y}_t | \Sigma_t, \mathcal{F}_{t-1}) \Phi(\boldsymbol{\alpha}^\top \mathbf{y}_t)$, i.e., the multivariate skew t tends to the multivariate skew normal as $\nu_0 \rightarrow \infty$.
- (iii) Let $\sigma_{t,jj}$ is the j -th diagonal element of Σ_t . Then $\sigma_{t,jj}^{-1} y_{t,j}^2 \sim F(1, \nu_0)$ and it is independent of $\boldsymbol{\alpha}$ and of \mathcal{F}_{t-1} .
- (iv) the quadratic form $p^{-1} \mathbf{y}_t^\top \Sigma_t^{-1} \mathbf{y}_t \sim F(p, \nu_0)$ and it is independent of $\boldsymbol{\alpha}$ and of \mathcal{F}_{t-1} .
- (v) At last, the multivariate skew t -distribution with ν_0 degrees of freedom obeys the following stochastic representation

$$\mathbf{y}_t \stackrel{d}{\sim} \frac{\mathbf{x}_t}{\sqrt{W_t/\nu_0}}$$

with $\mathbf{x}_t | \Sigma_t, \mathcal{F}_{t-1} \sim \mathcal{SN}(\mathbf{0}, \Sigma_t, \boldsymbol{\alpha})$ (multivariate skew normal) and $W_t \sim \chi_{\nu_0}^2$ independent of \mathbf{x}_t . Or, using the stochastic representation of \mathbf{x}_t , we get

$$\mathbf{y}_t \stackrel{d}{\sim} \frac{1}{\sqrt{W_t/\nu_0}} \left((\Sigma_t^{-1} + \boldsymbol{\alpha} \boldsymbol{\alpha}^\top)^{-1/2} \mathbf{z}_t + \frac{\Sigma_t \boldsymbol{\alpha}}{(1 + \boldsymbol{\alpha}^\top \Sigma_t \boldsymbol{\alpha})^{1/2}} |u_t| \right), \quad (2.3)$$

where $\mathbf{z}_t \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$, $u_t \sim \mathcal{N}(0, 1)$ and W_t are independent.

- (vi) The first two moments are given by

$$\mathbb{E}(\mathbf{y}_t | \Sigma_t, \mathcal{F}_{t-1}) = \sqrt{\frac{\nu_0}{\pi}} \frac{\Gamma(\frac{\nu_0-1}{2})}{\Gamma(\frac{\nu_0}{2})} \frac{\Sigma_t \boldsymbol{\alpha}}{(1 + \boldsymbol{\alpha}^\top \Sigma_t \boldsymbol{\alpha})^{1/2}}, \quad \nu_0 > 1 \quad (2.4)$$

$$\mathbb{Cov}(\mathbf{y}_t | \Sigma_t, \mathcal{F}_{t-1}) = \frac{\nu_0}{\nu_0 - 2} \Sigma_t - \frac{\nu_0}{\pi} \left(\frac{\Gamma(\frac{\nu_0-1}{2})}{\Gamma(\frac{\nu_0}{2})} \right)^2 \frac{\Sigma_t \boldsymbol{\alpha} \boldsymbol{\alpha}^\top \Sigma_t}{1 + \boldsymbol{\alpha}^\top \Sigma_t \boldsymbol{\alpha}}, \quad \nu_0 > 2. \quad (2.5)$$

- (vii) Thus, the multivariate skew t can be seen as a generalization of a non-central multivariate t -distribution.

(A2) The realized covariance matrix RC_t conditionally on \mathcal{F}_{t-1} is assumed to follow a rescaled matrix variate non-central F distribution with degrees of freedom ν_1, ν_2 and non-centrality matrix $\Sigma_t^{-1}\Omega_t$ with density given by

$$f_{RC}(RC_t|\Sigma_t, \Omega_t, \mathcal{F}_{t-1}; \nu_1, \nu_2) = K(\nu_1, \nu_2) \left| \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t^{-1} \right|^{\nu_1/2} \frac{|RC_t|^{(\nu_1 - p - 1)/2}}{\left| \mathbf{I} + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t^{-1} RC_t \right|^{(\nu_1 + \nu_2)/2}} \quad (2.6)$$

$$\times e^{\text{tr}(-\frac{1}{2}\Sigma_t^{-1}\Omega_t)} {}_1F_1\left(\frac{\nu_1 + \nu_2}{2}; \frac{\nu_1}{2}; \frac{1}{2} \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t^{-1} \Omega_t \Sigma_t^{-1} RC_t [\mathbf{I} + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t^{-1} RC_t]^{-1}\right) \\ = K(\nu_1, \nu_2) \left(\frac{\nu_1}{\nu_2 - p - 1}\right)^{p\nu_1/2} |RC_t|^{-(\nu_2 + p + 1)/2} |\Sigma_t|^{(\nu_1 + 2\nu_2)/2} \left| \Sigma_t RC_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t \right|^{-(\nu_1 + \nu_2)/2} \quad (2.7)$$

$$\times e^{\text{tr}(-\frac{1}{2}\Sigma_t^{-1}\Omega_t)} {}_1F_1\left(\frac{\nu_1 + \nu_2}{2}; \frac{\nu_1}{2}; \frac{1}{2} \frac{\nu_1}{\nu_2 - p - 1} \Omega_t [\Sigma_t RC_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t]^{-1}\right) \\ = \frac{K(\nu_1, \nu_2)}{\Gamma_p[(\nu_1 + \nu_2)/2]} \left(\frac{\nu_1}{\nu_2 - p - 1}\right)^{p\nu_1/2} |RC_t|^{-(\nu_2 + p + 1)/2} |\Sigma_t|^{(\nu_1 + 2\nu_2)/2} e^{\text{tr}(-\frac{1}{2}\Sigma_t^{-1}\Omega_t)} \quad (2.8)$$

$$\times \int_{\mathbf{V} > \mathbf{O}} \exp\left(-\frac{1}{2}\text{tr}((\Sigma_t RC_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t) \mathbf{V})\right) |\mathbf{V}|^{\frac{\nu_1 + \nu_2 - p - 1}{2}} {}_0F_1\left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} \Omega_t \mathbf{V}\right) d\mathbf{V} \\ = \frac{K(\nu_1, \nu_2)}{\Gamma_p[(\nu_1 + \nu_2)/2]} \left(\frac{\nu_1}{\nu_2 - p - 1}\right)^{p\nu_1/2} |RC_t|^{-(\nu_2 + p + 1)/2} |\Sigma_t|^{(\nu_1 + 2\nu_2)/2} \left| \Sigma_t RC_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t \right|^{-(\nu_1 + \nu_2)/2} \\ \times 2^{-(\nu_1 + \nu_2)/2} |\Omega_t|^{-(\nu_1 + \nu_2)/2} e^{\text{tr}(-\frac{1}{2}\Sigma_t^{-1}\Omega_t)} \int_{\mathbf{V} > \mathbf{O}} \exp\left(-\frac{1}{2}\text{tr}(\Omega_t^{-1} \mathbf{V})\right) |\mathbf{V}|^{\frac{\nu_1 + \nu_2 - p - 1}{2}} \\ \times {}_0F_1\left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} [\Sigma_t RC_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t]^{-1} \mathbf{V}\right) d\mathbf{V}, \quad (2.9)$$

where $K(\nu_1, \nu_2) = \frac{\Gamma_p[(\nu_1 + \nu_2)/2]}{\Gamma_p[\nu_1/2]\Gamma_p[\nu_2/2]}$ with $\Gamma_p(x) = \pi^{p(p-1)/4} \prod_{i=1}^p \Gamma(x - (i-1)/2)$ for $x > \frac{p-1}{2}$ is the multivariate Gamma function. The function ${}_1F_1(\cdot)$ is the matrix variate hypergeometric function

$${}_qF_k(a_1, \dots, a_k; b_1, \dots, b_k; \mathbf{A}) = \sum_{i=0}^{\infty} \sum_l \frac{(a_1)_l \dots (a_q)_l}{(b_1)_l \dots (b_k)_l} \frac{C_l(\mathbf{A})}{i!}, \quad (2.10)$$

where \mathbf{A} is a $p \times p$ symmetric matrix, \sum_l denotes the summation over all partitions l , $(c)_l = \frac{\Gamma_p(c, l)}{\Gamma_p(c)}$

with $\Gamma_p(c, l) = \pi^{p(p-1)/4} \prod_{j=1}^p \Gamma(c + i_j - (j-1)/2)$ for $c > (p-1)/2 - i_m$ and $l = (i-1, \dots, i_m)$

with $i_1 \geq i_2 \geq \dots i_m \geq 0$ and $\sum_j^p i_j = i$; $C_l(\mathbf{A})$ stands for the zonal polynomial.

We used Gupta & Nagar (2000), Theorem 1.6.2 to get from (2.7) to (2.8),

$${}_1F_1\left(\frac{\nu_1 + \nu_2}{2}; \frac{\nu_1}{2}; \frac{1}{2} \frac{\nu_1}{\nu_2 - p - 1} \Omega_t [\Sigma_t RC_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t]^{-1}\right) \\ = \frac{1}{\Gamma_p[(\nu_1 + \nu_2)/2]} \left| \Sigma_t RC_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t \right|^{\frac{\nu_1 + \nu_2}{2}} \\ \times \int_{\mathbf{V} > \mathbf{O}} \exp\left(-\frac{1}{2}\text{tr}((\Sigma_t RC_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t) \mathbf{V})\right) |\mathbf{V}|^{\frac{\nu_1 + \nu_2 - p - 1}{2}} {}_0F_1\left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} \Omega_t \mathbf{V}\right) d\mathbf{V},$$

and similarly to get from (2.7) to (2.9). (2.8) is useful to derive the score w.r.t. Σ_t , (2.9) is useful to derive the score w.r.t. Ω_t .

Basic properties of non-central matrix variate F distribution:

- (i) Obviously, if $\mathbf{\Omega}_t \equiv \mathbf{O}$ the distribution becomes the central one and the density function simplifies to

$$f_{RC}(RC_t|\mathbf{\Sigma}_t, \mathcal{F}_{t-1}; \nu_1, \nu_2) = K(\nu_1, \nu_2) \frac{\left| \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Sigma}_t^{-1} \right|^{\nu_1/2} |RC_t|^{(\nu_1 - p - 1)/2}}{\left| \mathbf{I} + \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Sigma}_t^{-1} RC_t \right|^{(\nu_1 + \nu_2)/2}}. \quad (2.11)$$

- (ii) For $\nu_2 \rightarrow \infty$ it collapses to a non-central Wishart distribution with ν_1 degrees of freedom.
 (iii) Let $\mathbf{\Omega}_t = \mathbf{\Lambda}_t \mathbf{\Lambda}_t^\top$ where $\mathbf{\Lambda}_t$ is a $p \times \nu_1$ matrix. The matrix F distribution obeys the following stochastic representation

$$RC_t \stackrel{d}{\sim} \frac{\nu_2 - p - 1}{\nu_1} \mathbf{\Sigma}_t^{1/2} \mathbf{V}_t^{1/2} \mathbf{T}_t^{-1} \mathbf{V}_t^{1/2} \mathbf{\Sigma}_t^{1/2}, \quad (2.12)$$

where $\mathbf{V}_t = (\mathbf{X}_t + \mathbf{\Sigma}_t^{-1/2} \mathbf{\Lambda}_t)(\mathbf{X}_t + \mathbf{\Sigma}_t^{-1/2} \mathbf{\Lambda}_t)^\top$ has a standard non-central Wishart distribution with ν_1 degrees of freedom and non-centrality matrix $\mathbf{\Sigma}_t^{-1/2} \mathbf{\Lambda}_t \mathbf{\Lambda}_t^\top \mathbf{\Sigma}_t^{-1/2}$, $\mathbf{X}_t \sim \mathcal{N}_{p, \nu_1}(\mathbf{O}, \mathbf{I}_p \otimes \mathbf{I}_{\nu_1})$ is matrix variate standard normal, $\mathbf{V}_t^{1/2}$ denotes a symmetric square root of \mathbf{V}_t and \mathbf{T}_t is a standard Wishart matrix with ν_2 degrees of freedom.

- (iv) The mean of $RC_t|\mathbf{\Sigma}_t, \mathcal{F}_{t-1}$ is equal to $\mathbf{\Sigma}_t + \frac{1}{\nu_1} \mathbf{\Omega}_t$.

(A3) Dynamics for $\mathbf{\Sigma}_t$ and $\mathbf{\Omega}_t$:

$$\mathbf{\Sigma}_t = \mathbf{\Xi} + \alpha \mathbf{S}_{t-1} + \beta \mathbf{\Sigma}_{t-1} \quad \text{GAS recursion} \quad (2.13)$$

$$\mathbf{\Omega}_t = \mathbf{M} RC_{t-1} \mathbf{M}^\top \quad \text{WAR specification} \quad (2.14)$$

The matrix \mathbf{M} could be assumed of some simple structure (diagonal, sparse etc.); $\mathbf{\Xi}$ is the unconditional covariance matrix and \mathbf{S}_t is the scaled score matrix.

(A4) Independence of error processes:

We assume that all five error processes $\{\mathbf{z}_t\}$, $\{u_t\}$, $\{W_t\}$, $\{\mathbf{V}_t\}$ (or $\{\mathbf{X}_t\}$), and $\{\mathbf{T}_t\}$ in (2.3) and (2.12) are mutually independent. Assumption (A4) in particular implies that the return vector \mathbf{y}_t and the realized covariance matrix RC_t conditionally on \mathcal{F}_{t-1} are independent.

2.1 t-noncentral-spiked-F scalar (p, q, r) model

The model of the t-noncentral-spiked-F (p, q, r) model dynamics (Eq. 2.9 and 2.10) are given by

$$\begin{aligned} \mathbf{\Sigma}_t &= \mathbf{\Xi} + \sum_{i=1}^p \alpha_i \mathbf{S}_{\mathbf{\Sigma}_{t-i}} + \sum_{j=1}^q \beta_j \mathbf{\Sigma}_{t-j} \\ \mathbf{\Omega}_t &= \sum_{i=1}^r \mathbf{M}_i \mathbf{S}_{\mathbf{\Omega}_{t-i}} \mathbf{M}_i', \end{aligned}$$

where $\mathbf{\Xi}$ is a positive definite p by p matrix, α_i and β_i are scalars and $\mathbf{M}_i = \mathbf{m}_i \mathbf{m}_i'$ and \mathbf{m}_i is a p by 1 vector, such that $\mathbf{\Omega}_t$ is rank 1 and the noncentral-F becomes the spiked noncentral-F.

Since (assuming stationarity),

$$\begin{aligned}
\mathbf{E}[RC_t] &= \mathbf{E}[\Sigma_t] + \frac{1}{\nu_1} \mathbf{E}[\Omega_t] \\
&= \frac{\Xi}{(1 - \sum_{j=1}^q \beta_j)} + \frac{1}{\nu_1} \mathbf{E} \left[\sum_{i=1}^r \mathbf{M}_i RC_{t-i} \mathbf{M}_i' \right] \\
&= \frac{\Xi}{(1 - \sum_{j=1}^q \beta_j)} + \frac{1}{\nu_1} \sum_{i=1}^r \mathbf{M}_i \mathbf{E}[RC_t] \mathbf{M}_i' \\
&\Leftrightarrow \\
\Xi &= \left(1 - \sum_{i=1}^q \beta_i \right) \left(\mathbf{E}[RC_t] - \frac{1}{\nu_1} \sum_{i=1}^r \mathbf{M}_i \mathbf{E}[RC_t] \mathbf{M}_i' \right),
\end{aligned}$$

we can use

$$\hat{\Xi} = \left(1 - \sum_{i=1}^q \beta_i \right) \left(\frac{1}{T} \sum_{t=1}^T RC_t - \frac{\sum_{j=1}^r \mathbf{M}_j \frac{1}{T} \sum_{t=1}^T RC_t \mathbf{M}_j'}{\nu_1} \right).$$

The the only parameters to estimate via ML are α_i , β_i and \mathbf{m}_i , such that the estimation is fast and the model is applicable in higher dimensions.

2.2 t-noncentral-F full (p, q, r) model

The model of the t-noncentral-spiked-F (p, q, r) full model dynamics (Eq. 2.9 and 2.10) are given by

$$\begin{aligned}
\Sigma_t &= \Xi + \sum_{i=1}^p \mathbf{A}_i RC_{t-i} \mathbf{A}_i' + \sum_{j=1}^q \mathbf{B}_j \Sigma_{t-i} \mathbf{B}_j' \\
\Omega_t &= \sum_{i=1}^r \mathbf{M}_i RC_{t-i} \mathbf{M}_i,
\end{aligned}$$

where \mathbf{A}_i , \mathbf{B}_i and \mathbf{M}_i are p by p matrices and Ξ is a positive definite p by p matrix. All parameters are estimated via one-step ML.

3 Likelihood function and score matrix

Let $\text{vech}(\mathbf{A}) = (a_{11}, \dots, a_{k1}, \dots, a_{ii}, \dots, a_{ki}, \dots, a_{kk})^\top$ for an arbitrary symmetric matrix $\mathbf{A} = (a_{ij})$. In the following we make use of the operator \mathbf{vec} defined by $\text{vec}(\mathbf{A}) = (a_{11}, \dots, a_{k1}, \dots, a_{1i}, \dots, a_{ki}, \dots, a_{kk})^\top$. For the properties of the operators \mathbf{vech} and \mathbf{vec} we refer to Harville (1997). Moreover, let \mathbf{D}_k be a $k^2 \times k(k+1)/2$ duplication matrix such that $\mathbf{D}_k \text{vech}(A) = \text{vec}(A)$ and $\mathbf{D}_k^+ = (\mathbf{D}_k^\top \mathbf{D}_k)^{-1} \mathbf{D}_k^\top$ with the property $\mathbf{D}_k^+ \text{vec}(A) = \text{vech}(A)$. Finally, the symbol \otimes stand for the Kronecker product.

In the derivation of the likelihood we use following two results (see, Harville (1997, p. 368))

$$\frac{\partial (\text{vec}(\mathbf{A}^{-1}))^\top}{\partial \text{vech}(\mathbf{A})} = -\mathbf{D}_k^\top (\mathbf{A}^{-1} \otimes \mathbf{A}^{-1}) \mathbf{D}_k^+{}^\top \mathbf{D}_k^\top. \quad (2.15)$$

and

$$\frac{\partial \log |\mathbf{A}|}{\partial \text{vech}(\mathbf{A})} = \mathbf{D}_k^\top \text{vec}(\mathbf{A}^{-1}), \quad (2.16)$$

where (2.16) follows from Magnus and Neudecker (1999, p. 171) since

$$d \log |\mathbf{A}| = \text{tr}(\mathbf{A}^{-1} d\mathbf{A}) = \text{vec}(\mathbf{A}^{-1})^\top d\text{vec}(\mathbf{A}) = \text{vec}(\mathbf{A}^{-1})^\top \mathbf{D}_k d\text{vech}(\mathbf{A})$$

The conditional log-likelihood of the considered model is given by

$$\mathcal{L}_t = \mathcal{L}_{t;1} + \mathcal{L}_{t;2} \quad (2.17)$$

with

$$\mathcal{L}_{t;1} = \log f_y(\mathbf{y}_t | \boldsymbol{\Sigma}_t, \mathcal{F}_{t-1}; \nu_0)$$

and

$$\mathcal{L}_{t;2} = \log f_{RC}(RC_t | \boldsymbol{\Sigma}_t, \boldsymbol{\Omega}_t, \mathcal{F}_{t-1}; \nu_1, \nu_2)$$

3.1 Skew-t score

3.1.1 Score w.r.t. $\boldsymbol{\Sigma}_t$

Then, we compute

$$\begin{aligned} \frac{\partial \mathcal{L}_{t;1}}{\partial \text{vech}(\boldsymbol{\Sigma}_t)} &= -\frac{1}{2} \frac{\partial \log |\boldsymbol{\Sigma}_t|}{\partial \text{vech}(\boldsymbol{\Sigma}_t)} - \frac{\nu_0 + p}{2} \frac{\partial \log \left(1 + \frac{\mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t}{\nu_0 - 2} \right)}{\partial \text{vech}(\boldsymbol{\Sigma}_t)} \\ &\quad + \frac{\partial \log \left(F_{T_{\nu_0+p}} \left(\frac{\boldsymbol{\alpha}^\top \mathbf{y}_t}{(\nu_0 - 2 + \mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t)^{1/2}} \sqrt{\nu_0 + p} \right) \right)}{\partial \text{vech}(\boldsymbol{\Sigma}_t)} \\ &= -\frac{1}{2} \mathbf{D}_p^\top \text{vec}(\boldsymbol{\Sigma}_t^{-1}) - \frac{\nu_0 + p}{2} \left(1 + \frac{\mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t}{\nu_0 - 2} \right)^{-1} \frac{1}{\nu_0 - 2} \frac{\partial (\text{vec}(\boldsymbol{\Sigma}_t^{-1}))^\top}{\partial \text{vech}(\boldsymbol{\Sigma}_t)} (\mathbf{y}_t \otimes \mathbf{y}_t) \\ &\quad - \frac{1}{2} \frac{f_{T_{\nu_0+p}} \left(\frac{\boldsymbol{\alpha}^\top \mathbf{y}_t}{(\nu_0 - 2 + \mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t)^{1/2}} \sqrt{\nu_0 + p} \right)}{F_{T_{\nu_0+p}} \left(\frac{\boldsymbol{\alpha}^\top \mathbf{y}_t}{(\nu_0 - 2 + \mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t)^{1/2}} \sqrt{\nu_0 + p} \right)} \frac{\sqrt{\nu_0 + p} \boldsymbol{\alpha}^\top \mathbf{y}_t}{(\nu_0 - 2 + \mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t)^{3/2}} \frac{\partial (\text{vec}(\boldsymbol{\Sigma}_t^{-1}))^\top}{\partial \text{vech}(\boldsymbol{\Sigma}_t)} (\mathbf{y}_t \otimes \mathbf{y}_t) \\ &= -\frac{1}{2} \mathbf{D}_p^\top \text{vec}(\boldsymbol{\Sigma}_t^{-1}) + \frac{1}{2} \frac{\sqrt{\nu_0 + p}}{\sqrt{\nu_0 - 2}} \left(1 + \frac{\mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t}{\nu_0 - 2} \right)^{-1} \\ &\quad \times \left(\frac{\sqrt{\nu_0 + p}}{\sqrt{\nu_0 - 2}} + \frac{f_{T_{\nu_0+p}} \left(\frac{\boldsymbol{\alpha}^\top \mathbf{y}_t}{(\nu_0 - 2 + \mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t)^{1/2}} \sqrt{\nu_0 + p} \right)}{F_{T_{\nu_0+p}} \left(\frac{\boldsymbol{\alpha}^\top \mathbf{y}_t}{(\nu_0 - 2 + \mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t)^{1/2}} \sqrt{\nu_0 + p} \right)} \frac{\boldsymbol{\alpha}^\top \mathbf{y}_t / (\nu_0 - 2)}{\sqrt{1 + \frac{\mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t}{\nu_0 - 2}}} \right) \\ &\quad \times \mathbf{D}_p^\top (\boldsymbol{\Sigma}_t^{-1} \otimes \boldsymbol{\Sigma}_t^{-1}) \mathbf{D}_p^{+\top} \mathbf{D}_p^\top (\mathbf{y}_t \otimes \mathbf{y}_t) \\ &= -\frac{1}{2} \mathbf{D}_p^\top \text{vec}(\boldsymbol{\Sigma}_t^{-1}) + \frac{1}{2} \frac{\sqrt{\nu_0 + p}}{\sqrt{\nu_0 - 2}} \left(1 + \frac{\mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t}{\nu_0 - 2} \right)^{-1} \\ &\quad \times \left(\frac{\sqrt{\nu_0 + p}}{\sqrt{\nu_0 - 2}} + \frac{f_{T_{\nu_0+p}} \left(\frac{\boldsymbol{\alpha}^\top \mathbf{y}_t}{(\nu_0 - 2 + \mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t)^{1/2}} \sqrt{\nu_0 + p} \right)}{F_{T_{\nu_0+p}} \left(\frac{\boldsymbol{\alpha}^\top \mathbf{y}_t}{(\nu_0 - 2 + \mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t)^{1/2}} \sqrt{\nu_0 + p} \right)} \frac{\boldsymbol{\alpha}^\top \mathbf{y}_t / (\nu_0 - 2)}{\sqrt{1 + \frac{\mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t}{\nu_0 - 2}}} \right) \mathbf{D}_p^\top (\boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t \otimes \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t), \end{aligned}$$

where in the last equality we used the following properties:

$$(i) \quad \mathbf{D}_k \mathbf{D}_k^+ = \frac{1}{2} (\mathbf{I}_{k^2} + \mathbf{K}_k),$$

- (ii) $\mathbf{K}_k = \mathbf{K}_k^\top = \mathbf{K}_k^{-1}$,
- (iii) $\mathbf{K}_k(\mathbf{A} \otimes \mathbf{a}) = (\mathbf{a} \otimes \mathbf{A})$ for an arbitrary $k \times n$ matrix \mathbf{A} and an arbitrary $k \times 1$ vector \mathbf{a} .

Now removing vech operators and denoting

$$\mathbf{Y}_t = \frac{1}{2} \boldsymbol{\Sigma}_t^{-1} \left(\mathbf{y}_t \mathbf{y}_t^\top \omega_t - \boldsymbol{\Sigma}_t \right) \boldsymbol{\Sigma}_t^{-1} \quad (2.18)$$

with

$$\omega_t = \frac{\sqrt{\nu_0 + p}}{\sqrt{\nu_0 - 2}} \left(1 + \frac{\mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t}{\nu_0 - 2} \right)^{-1} \left(\frac{\sqrt{\nu_0 + p}}{\sqrt{\nu_0 - 2}} + \frac{f_{T_{\nu_0+p}} \left(\frac{\boldsymbol{\alpha}_t^\top \mathbf{y}_t}{(\nu_0 - 2 + \mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t)^{1/2}} \sqrt{\nu_0 + p} \right)}{F_{T_{\nu_0+p}} \left(\frac{\boldsymbol{\alpha}_t^\top \mathbf{y}_t}{(\nu_0 - 2 + \mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t)^{1/2}} \sqrt{\nu_0 + p} \right)} \frac{\boldsymbol{\alpha}_t^\top \mathbf{y}_t / (\nu_0 - 2)}{\sqrt{1 + \frac{\mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t}{\nu_0 - 2}}} \right)$$

we obtain

$$\frac{\partial \mathcal{L}_{t;1}}{\partial \boldsymbol{\Sigma}_t} = 2\mathbf{Y}_t - \mathbf{Y}_t \circ \mathbf{I}.$$

3.1.2 Score w.r.t. α_t (if needed and if we will decide to replace α by α_t)

After replacing α by α_t we get

$$\begin{aligned} \frac{\partial \mathcal{L}_{t;1}}{\partial \alpha_t} &= -\frac{1}{2} \frac{\partial \log |\boldsymbol{\Sigma}_t|}{\partial \alpha_t} - \frac{\nu_0 + p}{2} \frac{\partial \log \left(1 + \frac{\mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t}{\nu_0 - 2} \right)}{\partial \alpha_t} \\ &\quad + \frac{\partial \log \left(F_{T_{\nu_0+p}} \left(\frac{\boldsymbol{\alpha}_t^\top \mathbf{y}_t}{(\nu_0 - 2 + \mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t)^{1/2}} \sqrt{\nu_0 + p} \right) \right)}{\partial \alpha_t} \\ &= \frac{f_{T_{\nu_0+p}} \left(\frac{\boldsymbol{\alpha}_t^\top \mathbf{y}_t}{(\nu_0 - 2 + \mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t)^{1/2}} \sqrt{\nu_0 + p} \right)}{F_{T_{\nu_0+p}} \left(\frac{\boldsymbol{\alpha}_t^\top \mathbf{y}_t}{(\nu_0 - 2 + \mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t)^{1/2}} \sqrt{\nu_0 + p} \right)} \frac{\sqrt{\nu_0 + p}}{\sqrt{\nu_0 - 2 + \mathbf{y}_t^\top \boldsymbol{\Sigma}_t^{-1} \mathbf{y}_t}} \mathbf{y}_t \end{aligned}$$

3.2 Noncentral-F Score

3.2.1 Score w.r.t. $\boldsymbol{\Sigma}_t$

For derivation of the noncentral matrix F score w.r.t. $\boldsymbol{\Sigma}_t$, p.d.f. representation (2.8) is used. The Log-Likelihood function w.r.t $\boldsymbol{\Sigma}$ is

$$\begin{aligned} \mathcal{L}_{t;2} &= \frac{\nu_1 + 2\nu_2}{2} \log (|\boldsymbol{\Sigma}_t|) - \frac{1}{2} \text{tr}(\boldsymbol{\Sigma}_t^{-1} \boldsymbol{\Omega}_t) \\ &\quad + \log \left(\int_{\mathbf{V} > \mathbf{O}} \exp \left(-\frac{1}{2} \text{tr}((\boldsymbol{\Sigma}_t R C_t^{-1} \boldsymbol{\Sigma}_t + \frac{\nu_1}{\nu_2 - p - 1} \boldsymbol{\Sigma}_t) \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 + \nu_2 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} \boldsymbol{\Omega}_t \mathbf{V} \right) d\mathbf{V} \right). \end{aligned} \quad (2.19)$$

Then the score obtains as

$$\begin{aligned} & \frac{\partial \mathcal{L}_{t;2}}{\partial \text{vech}(\mathbf{\Sigma}_t)} \\ &= \left(\frac{\nu_1}{2} + \nu_2 \right) \frac{\partial \log |\mathbf{\Sigma}_t|}{\partial \text{vech}(\mathbf{\Sigma}_t)} - \frac{1}{2} \frac{\partial \text{tr}(\mathbf{\Sigma}_t^{-1} \mathbf{\Omega}_t)}{\partial \text{vech}(\mathbf{\Sigma}_t)} \\ &+ \frac{\partial \log \left(\int_{\mathbf{V} > \mathbf{O}} \exp \left(-\frac{1}{2} \text{tr}((\mathbf{\Sigma}_t R C_t^{-1} \mathbf{\Sigma}_t + \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Sigma}_t) \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 + \nu_2 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Omega}_t \mathbf{V} \right) d\mathbf{V} \right)}{\partial \text{vech}(\mathbf{\Sigma}_t)} \end{aligned} \quad (2.20)$$

$$\begin{aligned} &= \left(\frac{\nu_1}{2} + \nu_2 \right) \mathbf{D}_p^\top \text{vec}(\mathbf{\Sigma}_t^{-1}) + \frac{1}{2} \mathbf{D}_p^\top (\mathbf{\Sigma}_t^{-1} \otimes \mathbf{\Sigma}_t^{-1}) \text{vec}(\mathbf{\Omega}_t) \\ &+ \frac{1}{\int_{\mathbf{V} > \mathbf{O}} \exp \left(-\frac{1}{2} \text{tr}((\mathbf{\Sigma}_t R C_t^{-1} \mathbf{\Sigma}_t + \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Sigma}_t) \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 + \nu_2 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Omega}_t \mathbf{V} \right) d\mathbf{V}} \\ &\times \frac{\partial \left(\int_{\mathbf{V} > \mathbf{O}} \exp \left(-\frac{1}{2} \text{tr}((\mathbf{\Sigma}_t R C_t^{-1} \mathbf{\Sigma}_t + \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Sigma}_t) \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 + \nu_2 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Omega}_t \mathbf{V} \right) d\mathbf{V} \right)}{\partial \text{vech}(\mathbf{\Sigma}_t)} \end{aligned} \quad (2.21)$$

$$\begin{aligned} &= \left(\frac{\nu_1}{2} + \nu_2 \right) \mathbf{D}_p^\top \text{vec}(\mathbf{\Sigma}_t^{-1}) + \frac{1}{2} \mathbf{D}_p^\top (\mathbf{\Sigma}_t^{-1} \otimes \mathbf{\Sigma}_t^{-1}) \text{vec}(\mathbf{\Omega}_t) \\ &+ \frac{1}{\int_{\mathbf{V} > \mathbf{O}} \exp \left(-\frac{1}{2} \text{tr}((\mathbf{\Sigma}_t R C_t^{-1} \mathbf{\Sigma}_t + \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Sigma}_t) \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 + \nu_2 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Omega}_t \mathbf{V} \right) d\mathbf{V}} \\ &\times \int_{\mathbf{V} > \mathbf{O}} \frac{\partial \exp \left(-\frac{1}{2} \text{tr}((\mathbf{\Sigma}_t R C_t^{-1} \mathbf{\Sigma}_t + \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Sigma}_t) \mathbf{V}) \right)}{\partial \text{vech}(\mathbf{\Sigma}_t)} |\mathbf{V}|^{\frac{\nu_1 + \nu_2 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Omega}_t \mathbf{V} \right) d\mathbf{V} \end{aligned} \quad (2.22)$$

$$\begin{aligned} &= \left(\frac{\nu_1}{2} + \nu_2 \right) \mathbf{D}_p^\top \text{vec}(\mathbf{\Sigma}_t^{-1}) + \frac{1}{2} \mathbf{D}_p^\top (\mathbf{\Sigma}_t^{-1} \otimes \mathbf{\Sigma}_t^{-1}) \text{vec}(\mathbf{\Omega}_t) \\ &- \frac{1}{2} \mathbf{D}_p^\top \frac{\int_{\mathbf{V} > \mathbf{O}} \mathbf{Z}(\mathbf{V}) |\mathbf{V}|^{\frac{\nu_2}{2}} \exp \left(-\frac{1}{2} \text{tr}((\mathbf{\Sigma}_t R C_t^{-1} \mathbf{\Sigma}_t + \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Sigma}_t) \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Omega}_t \mathbf{V} \right) d\mathbf{V}}{\int_{\mathbf{V} > \mathbf{O}} |\mathbf{V}|^{\frac{\nu_2}{2}} \exp \left(-\frac{1}{2} \text{tr}((\mathbf{\Sigma}_t R C_t^{-1} \mathbf{\Sigma}_t + \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Sigma}_t) \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Omega}_t \mathbf{V} \right) d\mathbf{V}} \end{aligned} \quad (2.23)$$

$$= \left(\frac{\nu_1}{2} + \nu_2 \right) \mathbf{D}_p^\top \text{vec}(\mathbf{\Sigma}_t^{-1}) + \frac{1}{2} \mathbf{D}_p^\top (\mathbf{\Sigma}_t^{-1} \otimes \mathbf{\Sigma}_t^{-1}) \text{vec}(\mathbf{\Omega}_t) - \frac{1}{2} \mathbf{D}_p^\top \mathbf{Z} \left(\frac{\mathbb{E} \left[|\mathbf{V}|^{\frac{\nu_2}{2}} \right]}{\mathbb{E} \left[|\mathbf{V}|^{\frac{\nu_2}{2}} \right]} \right) \quad (2.24)$$

$$\approx \left(\frac{\nu_1}{2} + \nu_2 \right) \mathbf{D}_p^\top \text{vec}(\mathbf{\Sigma}_t^{-1}) + \frac{1}{2} \mathbf{D}_p^\top (\mathbf{\Sigma}_t^{-1} \otimes \mathbf{\Sigma}_t^{-1}) \text{vec}(\mathbf{\Omega}_t) - \frac{1}{2} \mathbf{D}_p^\top \mathbf{Z} \left(\frac{(\bar{\mathbf{\Sigma}} + \frac{\bar{\mathbf{\Omega}}}{\nu_1}) \mathbf{I} + \frac{\bar{\mathbf{\Sigma}}^{-1} \bar{\mathbf{\Omega}}}{\nu_1} \left| \frac{\nu_2}{2} \right| (\nu_1 + \nu_2)}{e^{\text{tr}(-\frac{1}{2} \bar{\mathbf{\Sigma}}^{-1} \bar{\mathbf{\Omega}})} {}_1F_1 \left((\nu_1 + \nu_2)/2, \frac{1}{2} \nu_1, \frac{1}{2} \bar{\mathbf{\Sigma}}^{-1} \bar{\mathbf{\Omega}} \right)} \right),$$

see Appendix for last step

where in (2.23)

$$\frac{\partial \left(-\frac{1}{2} \text{tr}((\mathbf{\Sigma}_t R C_t^{-1} \mathbf{\Sigma}_t + \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Sigma}_t) \mathbf{V}) \right)}{\partial \text{vech}(\mathbf{\Sigma}_t)}$$

$$= -\frac{1}{2} \mathbf{D}_p^\top \left((\mathbf{V} \otimes R C_t^{-1}) \text{vec}(\mathbf{\Sigma}_t) + (R C_t^{-1} \otimes \mathbf{V}) \text{vec}(\mathbf{\Sigma}_t) + \frac{\nu_1}{\nu_2 - p - 1} \text{vec}(\mathbf{V}) \right) \quad (2.25)$$

$$\equiv -\frac{1}{2} \mathbf{D}_p^\top \mathbf{Z}(\mathbf{V}) \quad (2.26)$$

and the expectation in (2.24) are w.r.t. $\mathbf{V} \sim$ noncentral Wishart, since

$$\exp \left(-\frac{1}{2} \text{tr}((\mathbf{\Sigma}_t R C_t^{-1} \mathbf{\Sigma}_t + \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Sigma}_t) \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Omega}_t \mathbf{V} \right)$$

is the kernel of the p.d.f. of the noncentral Wishart distribution with covariance matrix

$$\bar{\mathbf{\Sigma}} = \left(\mathbf{\Sigma}_t R C_t^{-1} \mathbf{\Sigma}_t + \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Sigma}_t \right)^{-1},$$

non-centrality matrix

$$\bar{\mathbf{\Omega}} = \left(\mathbf{\Sigma}_t R C_t^{-1} \mathbf{\Sigma}_t + \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Sigma}_t \right)^{-1} \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Omega}_t \left(\mathbf{\Sigma}_t R C_t^{-1} \mathbf{\Sigma}_t + \frac{\nu_1}{\nu_2 - p - 1} \mathbf{\Sigma}_t \right)^{-1}$$

and

$$\mathbf{E}[\mathbf{V}] = \nu_1 \bar{\Sigma} + \bar{\Omega} \equiv \mathbf{Q}.$$

3.2.2 Score w.r.t. Ω_t

For derivation of the noncentral matrix F score w.r.t. Ω_t , p.d.f. representation (2.9) is used. The Log-Likelihood function w.r.t Ω_t is

$$\begin{aligned} \mathcal{L}_{t;2} = & -\frac{\nu_1 + \nu_2}{2} \log |\Omega_t| - \frac{1}{2} \text{tr}(\Sigma_t^{-1} \Omega_t) \\ & + \log \left(\int_{\mathbf{V} > \mathbf{O}} \exp \left(-\frac{1}{2} \text{tr}(\Omega_t^{-1} \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 + \nu_2 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} [\Sigma_t R C_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t]^{-1} \mathbf{V} \right) d\mathbf{V} \right). \end{aligned} \quad (2.27)$$

Then the score obtains as

$$\begin{aligned} \frac{\partial \mathcal{L}_{t;2}}{\partial \text{vech}(\Omega_t)} &= -\frac{\nu_1 + \nu_2}{2} \frac{\partial \log |\Omega_t|}{\partial \text{vech}(\Omega_t)} - \frac{1}{2} \frac{\partial \text{tr}(\Sigma_t^{-1} \Omega_t)}{\partial \text{vech}(\Omega_t)} \\ &+ \frac{\partial \log \left(\int_{\mathbf{V} > \mathbf{O}} \exp \left(-\frac{1}{2} \text{tr}(\Omega_t^{-1} \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 + \nu_2 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} [\Sigma_t R C_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t]^{-1} \mathbf{V} \right) d\mathbf{V} \right)}{\partial \text{vech}(\Omega_t)} \end{aligned} \quad (2.28)$$

$$\begin{aligned} &= -\frac{\nu_1 + \nu_2}{2} \mathbf{D}_p^\top \text{vec}(\Omega_t^{-1}) - \frac{1}{2} \mathbf{D}_p^\top \text{vec}(\Sigma_t^{-1}) \\ &+ \frac{1}{\int_{\mathbf{V} > \mathbf{O}} \exp \left(-\frac{1}{2} \text{tr}(\Omega_t^{-1} \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 + \nu_2 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} [\Sigma_t R C_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t]^{-1} \mathbf{V} \right) d\mathbf{V}} \\ &\times \frac{\partial \int_{\mathbf{V} > \mathbf{O}} \exp \left(-\frac{1}{2} \text{tr}(\Omega_t^{-1} \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 + \nu_2 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} [\Sigma_t R C_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t]^{-1} \mathbf{V} \right) d\mathbf{V}}{\partial \text{vech}(\Omega_t)} \end{aligned} \quad (2.29)$$

$$\begin{aligned} &= -\frac{\nu_1 + \nu_2}{2} \mathbf{D}_p^\top \text{vec}(\Omega_t^{-1}) - \frac{1}{2} \mathbf{D}_p^\top \text{vec}(\Sigma_t^{-1}) \\ &+ \frac{\int_{\mathbf{V} > \mathbf{O}} \frac{\partial \exp \left(-\frac{1}{2} \text{tr}(\Omega_t^{-1} \mathbf{V}) \right)}{\partial \text{vech}(\Omega_t)} |\mathbf{V}|^{\frac{\nu_1 + \nu_2 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} [\Sigma_t R C_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t]^{-1} \mathbf{V} \right) d\mathbf{V}}{\int_{\mathbf{V} > \mathbf{O}} \exp \left(-\frac{1}{2} \text{tr}(\Omega_t^{-1} \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 + \nu_2 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} [\Sigma_t R C_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t]^{-1} \mathbf{V} \right) d\mathbf{V}} \end{aligned} \quad (2.30)$$

$$\begin{aligned} &= -\frac{\nu_1 + \nu_2}{2} \mathbf{D}_p^\top \text{vec}(\Omega_t^{-1}) - \frac{1}{2} \mathbf{D}_p^\top \text{vec}(\Sigma_t^{-1}) \\ &+ \frac{1}{2} \mathbf{D}_p^\top (\Omega_t^{-1} \otimes \Omega_t^{-1}) \\ &\times \frac{\int_{\mathbf{V} > \mathbf{O}} \text{vec}(\mathbf{V}) |\mathbf{V}|^{\frac{\nu_2}{2}} \exp \left(-\frac{1}{2} \text{tr}(\Omega_t^{-1} \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} [\Sigma_t R C_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t]^{-1} \mathbf{V} \right) d\mathbf{V}}{\int_{\mathbf{V} > \mathbf{O}} |\mathbf{V}|^{\frac{\nu_2}{2}} \exp \left(-\frac{1}{2} \text{tr}(\Omega_t^{-1} \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} [\Sigma_t R C_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t]^{-1} \mathbf{V} \right) d\mathbf{V}} \end{aligned} \quad (2.31)$$

$$= -\frac{\nu_1 + \nu_2}{2} \mathbf{D}_p^\top \text{vec}(\Omega_t^{-1}) - \frac{1}{2} \mathbf{D}_p^\top \text{vec}(\Sigma_t^{-1}) + \frac{1}{2} \mathbf{D}_p^\top (\Omega_t^{-1} \otimes \Omega_t^{-1}) \text{vec} \left(\frac{\mathbb{E} \left[|\mathbf{V}|^{\frac{\nu_2}{2}} \right]}{\mathbb{E} \left[|\mathbf{V}|^{\frac{\nu_2}{2}} \right]} \right) \quad (2.32)$$

$$\approx -\frac{\nu_1 + \nu_2}{2} \mathbf{D}_p^\top \text{vec}(\Omega_t^{-1}) - \frac{1}{2} \mathbf{D}_p^\top \text{vec}(\Sigma_t^{-1}) + \frac{1}{2} \mathbf{D}_p^\top (\Omega_t^{-1} \otimes \Omega_t^{-1}) \text{vec} \left(\frac{(\bar{\Sigma} + \frac{\bar{\Omega}}{\nu_1}) \left| \mathbf{I} + \frac{\bar{\Sigma}^{-1} \bar{\Omega}}{\nu_1} \right|^{\frac{\nu_2}{2}} (\nu_1 + \nu_2)}{e^{\text{tr}(-\frac{1}{2} \bar{\Sigma}^{-1} \bar{\Omega})} {}_1F_1 \left((\nu_1 + \nu_2)/2, \frac{1}{2} \nu_1, \frac{1}{2} \bar{\Sigma}^{-1} \bar{\Omega} \right)} \right),$$

see Appendix for last step

where the expectation in (2.32) are w.r.t. $\mathbf{V} \sim$ noncentral Wishart, since

$$\exp \left(-\frac{1}{2} \text{tr}(\Omega_t^{-1} \mathbf{V}) \right) |\mathbf{V}|^{\frac{\nu_1 - p - 1}{2}} {}_0F_1 \left(\frac{\nu_1}{2}; \frac{1}{4} \frac{\nu_1}{\nu_2 - p - 1} [\Sigma_t R C_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t]^{-1} \mathbf{V} \right)$$

is the kernel of the p.d.f. of the noncentral Wishart distribution with covariance matrix

$$\bar{\Sigma} = \Omega_t,$$

non-centrality matrix

$$\bar{\Omega} = \frac{\nu_1}{\nu_2 - p - 1} \Omega_t [\Sigma_t R C_t^{-1} \Sigma_t + \frac{\nu_1}{\nu_2 - p - 1} \Sigma_t]^{-1} \Omega_t$$

and

$$\mathbf{E}[\mathbf{V}] = \nu_1 \bar{\Sigma} + \bar{\Omega} \equiv \mathbf{Q}.$$

3.3 Total Score

Finally, the score matrix is calculated by

$$\nabla_{\Sigma_t} = \frac{\partial \mathcal{L}_t}{\partial \Sigma_t} = \frac{\partial \mathcal{L}_{t;1}}{\partial \Sigma_t} + \frac{\partial \mathcal{L}_{t;2}}{\partial \Sigma_t}, \quad (2.33)$$

and

$$\nabla_{\Omega_t} = \frac{\partial \mathcal{L}_t}{\partial \Omega_t} = \frac{\partial \mathcal{L}_{t;1}}{\partial \Omega_t} + \frac{\partial \mathcal{L}_{t;2}}{\partial \Omega_t}. \quad (2.34)$$

We then scale ∇_{Σ_t} as in Opschoor et al.,

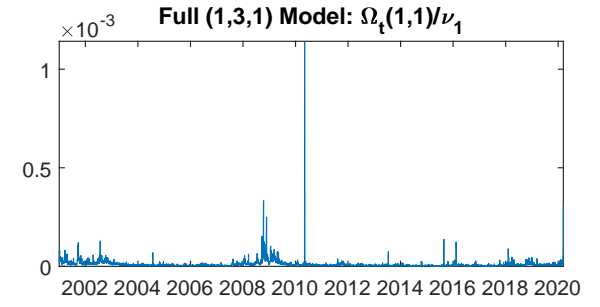
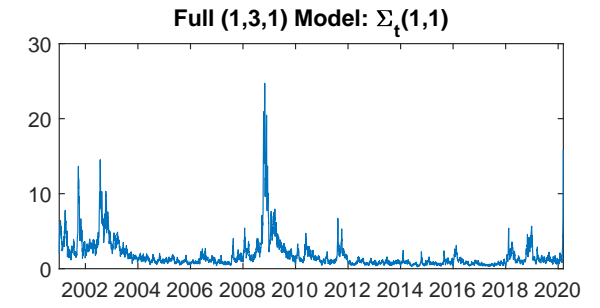
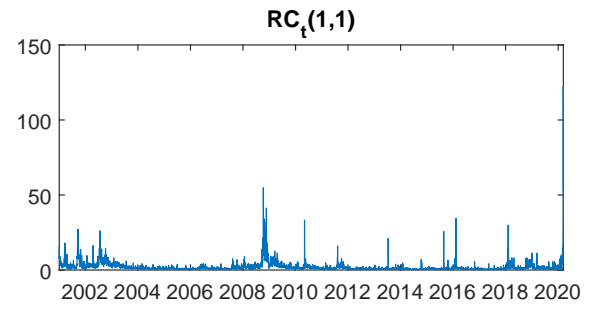
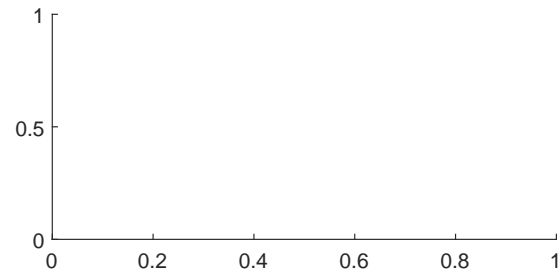
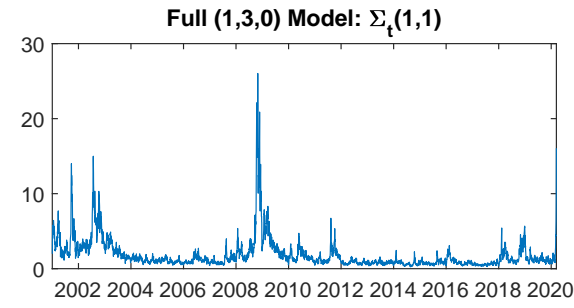
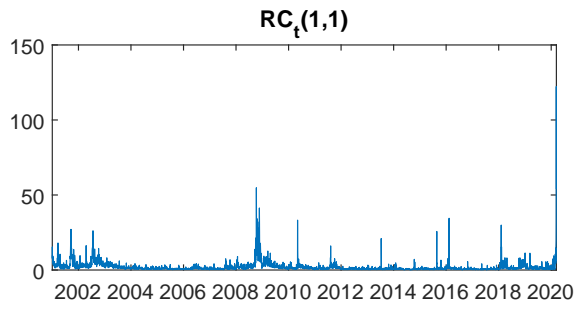
$$\mathbf{S}_{\Sigma_t} = \nabla_{\Sigma_t} \quad (2.35)$$

4 Empirical Application

$r = 0$ is the baseline Opschoor t-F model. Data is $p = 5$ (Assets: BA/HD/JPM/PFE/PG as in Opschoor Table 4 Panel A.1) from Jan 2001 to Mar 2020 for $T \approx 4500$.

(p, q, r)	$\dim(\theta)$	$\mathcal{L}(\hat{\theta})$	BIC	$\sum_{i=1}^q \beta_i$	NaN	e_{11}^*	e_{21}^*	e_{31}^*	e_{41}^*	e_{51}^*	e_{22}^*	e_{32}^*	e_{42}^*	e_{52}^*	e_{33}^*	e_{43}^*	e_{53}^*	e_{44}^*	e_{54}^*	e_{55}^*
t-noncentral-spiked-F scalar (p,q,r) model. Estimation with intercept targeting.																				
(1, 1, 0)	20	-20788	41746	0.988																
(1, 1, 1)	25	-20712	41636	0.987																
(1, 1, 2)	30	-20669	41593	0.987																
(1, 2, 0)	21	-20519	41215	0.988																
(1, 2, 1)	26	-20451	41123	0.988																
(1, 2, 2)	31	-20403	41069	0.987																
(1, 3, 0)	22	-20400	40987	0.988																
(1, 3, 1)	27	-20334	40898	0.989																
(1, 3, 2)	32	-20289	40849	0.988																
(2, 1, 0)	21	-20430	41039	0.991																
(2, 1, 1)	26	-20364	40949	0.991																
(2, 1, 2)	31	-20317	40898	0.991																
(2, 2, 0)	22	-20113	40412	0.999																
(2, 2, 1)	27	-20031	40292	0.999																
(2, 2, 2)	32	-20007	40286	0.999																
(2, 3, 0)	23	-20062	40318	1																
(2, 3, 1)	28	-19981	40200	1																
(2, 3, 2)	33	-19952	40184	1																
(3, 1, 0)	22	-20320	40827	0.993																
(3, 1, 1)	27	-20252	40732	0.993																
(3, 1, 2)	32	-20210	40692	0.993																
(3, 2, 0)	23	-20055	40306	1																
(3, 2, 1)	28	-19975	40187	1																
(3, 2, 2)	33	-19945	40170	1																
(3, 3, 0)	24	-20055	40314	1																
(3, 3, 1)	29	-19975	40195	1																
(3, 3, 2)	34	-19942	40172	1																
t-noncentral-F full (p,q,r) model. Full Likelihood Optimization.																				

Fit



5 Appendix

Let's derive $E[\mathbf{V}|\mathbf{V}|^h]$ if $\mathbf{V} \sim W(\mathbf{\Sigma}, \nu)$.

First, note that if $\mathbf{S} \sim W(\nu + 2h, \mathbf{\Sigma})$, then we know that

$$E[\mathbf{S}] = \int_{\mathbf{S} > \mathbf{0}} \mathbf{S} \frac{|\mathbf{S}|^{\frac{1}{2}((\nu+2h)p-1)} e^{-\frac{1}{2}\text{tr}((\mathbf{\Sigma})^{-1}\mathbf{S})}}{2^{((\nu+2h)p)/2} |\mathbf{\Sigma}|^{(\nu+2h)/2} \Gamma_p((\nu+2h)/2)} d\mathbf{S} = (\nu+2h)\mathbf{\Sigma}. \quad (2.36)$$

Thus

$$\int_{\mathbf{S} > \mathbf{0}} \mathbf{S} |\mathbf{S}|^{\frac{1}{2}(\nu+2h-p-1)} e^{-\frac{1}{2}\text{tr}((\mathbf{\Sigma})^{-1}\mathbf{S})} d\mathbf{S} = (\nu+2h)\mathbf{\Sigma} 2^{((\nu+2h)p)/2} |\mathbf{\Sigma}|^{(\nu+2h)/2} \Gamma_p((\nu+2h)/2). \quad (2.37)$$

Using (2.37) we can now derive $E[\mathbf{V}|\mathbf{V}|^h]$ if $\mathbf{V} \sim W(\mathbf{\Sigma}, \nu)$,

$$E[\mathbf{V}|\mathbf{V}|^h] = \int_{\mathbf{V} > \mathbf{0}} \mathbf{V} |\mathbf{V}|^h \frac{|\mathbf{V}|^{\frac{1}{2}(\nu-p-1)} e^{-\frac{1}{2}\text{tr}((\mathbf{\Sigma})^{-1}\mathbf{V})}}{2^{(\nu p)/2} |\mathbf{\Sigma}|^{\nu/2} \Gamma_p(\nu/2)} d\mathbf{V} \quad (2.38)$$

$$= \frac{1}{2^{(\nu p)/2} |\mathbf{\Sigma}|^{\nu/2} \Gamma_p(\nu/2)} \int_{\mathbf{V} > \mathbf{0}} \mathbf{V} |\mathbf{V}|^{\frac{1}{2}(\nu+2h-p-1)} e^{-\frac{1}{2}\text{tr}((\mathbf{\Sigma})^{-1}\mathbf{V})} d\mathbf{V} \quad (2.39)$$

$$\stackrel{2.37}{=} \frac{1}{2^{(\nu p)/2} |\mathbf{\Sigma}|^{\nu/2} \Gamma_p(\nu/2)} (\nu+2h)\mathbf{\Sigma} 2^{((\nu+2h)p)/2} |\mathbf{\Sigma}|^{(\nu+2h)/2} \Gamma_p((\nu+2h)/2) \quad (2.40)$$

$$= \mathbf{\Sigma} |\mathbf{\Sigma}|^h 2^{hp} (\nu+2h) \frac{\Gamma_p((\nu+2h)/2)}{\Gamma_p(\nu/2)}. \quad (2.41)$$

Now, if $\mathbf{V} \sim \text{noncentral-}W(\mathbf{\Sigma}, \mathbf{\Omega}, \nu)$

$$\begin{aligned} \frac{E[\mathbf{V}|\mathbf{V}|^{\frac{\nu_2}{2}}]}{E[|\mathbf{V}|^{\frac{\nu_2}{2}}]} &= \frac{E[\mathbf{V}|\mathbf{V}|^{\frac{\nu_2}{2}}]}{\frac{2^{(\nu_2/2)p} \Gamma_p((\nu_1+\nu_2)/2) |\mathbf{\Sigma}|^{\frac{\nu_2}{2}} e^{\text{tr}(-\frac{1}{2}\mathbf{\Sigma}^{-1}\mathbf{\Omega}_t)} {}_1F_1(\frac{1}{2}\nu_1 + \frac{\nu_2}{2}, \frac{1}{2}\nu_1, \frac{1}{2}\mathbf{\Sigma}^{-1}\mathbf{\Omega}_t)}{\Gamma_p(\nu_1/2)}}} \\ &\quad \text{Gupta and Nagar (Theorem 3.5.6, p. 119)} \\ &\approx \frac{\left(\mathbf{\Sigma} + \frac{\mathbf{\Omega}}{\nu_1}\right) \left|\mathbf{\Sigma} + \frac{\mathbf{\Omega}}{\nu_1}\right|^{\frac{\nu_2}{2}} 2^{(\nu_2/2)p} (\nu_1 + \nu_2) \frac{\Gamma_p((\nu_1+\nu_2)/2)}{\Gamma_p(\nu_1/2)}}{\frac{2^{(\nu_2/2)p} \Gamma_p((\nu_1+\nu_2)/2) |\mathbf{\Sigma}|^{\frac{\nu_2}{2}} e^{\text{tr}(-\frac{1}{2}\mathbf{\Sigma}^{-1}\mathbf{\Omega}_t)} {}_1F_1(\frac{1}{2}\nu_1 + \frac{\nu_2}{2}, \frac{1}{2}\nu_1, \frac{1}{2}\mathbf{\Sigma}^{-1}\mathbf{\Omega}_t)}{\Gamma_p(\nu_1/2)}}} \\ &\quad E[\mathbf{V}|\mathbf{V}|^{\frac{\nu_2}{2}}] \text{ is derived in (2.41) for } \mathbf{V} \sim W(\mathbf{\Sigma} + \mathbf{\Omega}/\nu, \nu) \approx \text{noncentral-}W(\mathbf{\Sigma}, \mathbf{\Omega}, \nu) \\ &= \frac{\left(\mathbf{\Sigma} + \frac{\mathbf{\Omega}}{\nu_1}\right) \left|\mathbf{I} + \frac{\mathbf{\Sigma}^{-1}\mathbf{\Omega}}{\nu_1}\right|^{\frac{\nu_2}{2}} (\nu_1 + \nu_2)}{e^{\text{tr}(-\frac{1}{2}\mathbf{\Sigma}^{-1}\mathbf{\Omega})} {}_1F_1((\nu_1 + \nu_2)/2, \frac{1}{2}\nu_1, \frac{1}{2}\mathbf{\Sigma}^{-1}\mathbf{\Omega})}. \end{aligned} \quad (2.42)$$