

# 1 Problem Set 5 - Problem 1)

## 1.1 Derive GAS model

We have,

$$Y_t | \mathcal{F}_{t-1} \sim \mathcal{T}(0, \phi, \nu)$$

where the mean is 0,  $\phi$  is the scale parameter and  $\nu$  is degrees of freedom. Remembering that the  $\mathcal{T}$  distribution is defined not by std. deviation but by scale. However the following relation is true

$$\sigma_t^2 = \phi_t^2 \nu (\nu - 2), \quad \text{for } \nu > 2$$

The conditional density of  $Y_t | \mathcal{F}_{t-1}$  is given by

$$p(y_t | \mathcal{F}_{t-1}) = \frac{\Gamma\left(\frac{\nu+1}{2}\right)}{\sqrt{\pi\nu}\Gamma\left(\frac{\nu}{2}\right)\phi_t} \left[1 + \frac{y_t^2}{\nu\phi_t^2}\right]^{-\frac{\nu+1}{2}}$$

Derive a GAS model with identity scaling  $d = 0$  for the scale parameter  $\phi_t$ . Use an exponential link function to ensure  $\phi_t > 0$  for all  $t$ .

Taking the log of

$$p(y_t | \mathcal{F}_{t-1}) = \frac{\Gamma\left(\frac{\nu+1}{2}\right)}{\sqrt{\pi\nu}\Gamma\left(\frac{\nu}{2}\right)\phi_t} \left[1 + \frac{y_t^2}{\nu\phi_t^2}\right]^{-\frac{\nu+1}{2}}$$

$$\ln[p(y_t | \mathcal{F}_{t-1})] = \ln \left[ \frac{\Gamma\left(\frac{\nu+1}{2}\right)}{\sqrt{\pi\nu}\Gamma\left(\frac{\nu}{2}\right)\phi_t} \left[1 + \frac{y_t^2}{\nu\phi_t^2}\right]^{-\frac{\nu+1}{2}} \right]$$

We know this log density is proportional to. This is due to the fact that the  $\Gamma$ ,  $\nu$  and  $\pi$  can be decomposed into constants. Thus they will not have an effect on the functional form but only on the level of the process.

$$\ln[p(y_t | \mathcal{F}_{t-1})] \propto -\ln[\phi_t] - \frac{\nu+1}{2} \ln \left[1 + \frac{y_t^2}{\nu\phi_t^2}\right]$$

The score with respect to  $\phi_t$  is

$$\ln[p(y_t | \mathcal{F}_{t-1})] = -\ln[\phi_t] - \frac{\nu+1}{2} \ln \left[1 + \frac{y_t^2}{\nu\phi_t^2}\right]$$

$$\nabla_t \equiv \frac{\partial \nabla_t}{\partial \phi_t} = -\ln[\phi_t] - \frac{\nu+1}{2} \ln \left[1 + \frac{y_t^2}{\nu\phi_t^2}\right]$$

$$\nabla_t = -\frac{1}{\phi_t} - \underbrace{\frac{\nu+1}{2} \frac{1}{1 + \frac{y_t^2}{\nu\phi_t^2}} \cdot \left(-\frac{2y_t^2}{\nu\phi_t^3}\right)}_{\text{chain rule}}$$

$$\nabla_t = -\frac{1}{\phi_t} + \frac{\nu+1}{2} \frac{y_t^2}{\left(1 + \frac{y_t^2}{\nu\phi_t^2}\right) \nu\phi_t^3}$$

$$\nabla_t = -\frac{1}{\phi_t} + \frac{(\nu+1) \frac{y_t^2}{\nu\phi_t^3}}{1 + \frac{y_t^2}{\nu\phi_t^2}}$$

$$\nabla_t = -\frac{1}{\phi_t} + \frac{(\nu+1) y_t^2}{\left(1 + \frac{y_t^2}{\nu\phi_t^2}\right) \nu\phi_t^3}$$

$$\nabla_t = -\frac{1}{\phi_t} + \frac{(\nu+1) y_t^2}{\left(\nu\phi_t^3 + \frac{\nu\phi_t^3 y_t^2}{\nu\phi_t^2}\right)}$$

$$\nabla_t = -\frac{1}{\phi_t} + \frac{(\nu+1) y_t^2}{(\nu\phi_t^3 + \phi_t y_t^2)}$$

Replicating from Leo's solution

$$\begin{aligned}
&\xRightarrow[\text{multiplying } \frac{2}{2}]{} \nabla_t = -\frac{1}{\phi_t} + \frac{(\nu+1)}{2} \frac{2y_t^2}{(\nu\phi_t^3 + \phi_t y_t^2)} \\
&\nabla_t = -\frac{1}{\phi_t} + \frac{(\nu+1)}{2} \frac{2\frac{y_t^2}{\phi_t^2}}{\phi_t \left( \nu + \frac{y_t^2}{\phi_t^2} \right)} \\
&\xRightarrow[\text{defining}]{} z_t \equiv \frac{y_t}{\phi_t} \\
&\nabla_t = -\frac{1}{\phi_t} + \frac{(\nu+1)}{2} \frac{2z_t^2}{\phi_t (\nu + z_t^2)} \\
&\nabla_t = -\frac{1}{\phi_t} + \frac{(\nu+1)}{\phi_t} \frac{z_t^2}{(\nu + z_t^2)} \\
&\nabla_t = \frac{(\nu+1) z_t^2}{\phi_t (\nu + z_t^2)} - \frac{1}{\phi_t} \\
&\xRightarrow[\text{factorize}]{} \nabla_t = \frac{1}{\phi_t} \left[ \frac{(\nu+1) z_t^2}{\nu + z_t^2} - 1 \right]
\end{aligned}$$

We remember that we are to derive a GAS model with identity scaling  $d = 0$ . Thus we know that

$$d = 0 \quad \rightarrow \quad \tilde{u}_t = \frac{\partial \phi_t}{\partial \tilde{\phi}_t} \nabla_t$$

The GAS model for  $\phi_t$  is

$$\begin{aligned}
\phi_t &= \lambda \left( \tilde{\phi}_t \right) \quad \xRightarrow[\text{exponential link function}]{} \quad \phi_t = \exp \left( \tilde{\phi}_t \right) \\
\tilde{\phi}_t &= \omega + \alpha \tilde{u}_{t-1} + \beta \tilde{\phi}_{t-1}
\end{aligned}$$

Then deriving

$$\begin{aligned}
\tilde{u}_t &= \frac{\partial \phi_t}{\partial \tilde{\phi}_t} \nabla_t \\
\tilde{u}_t &= \frac{\partial}{\partial \tilde{\phi}_t} \left( \exp \left( \tilde{\phi}_t \right) \right) \frac{1}{\phi_t} \left[ \frac{(\nu+1) z_t^2}{\nu + z_t^2} - 1 \right] \\
\tilde{u}_t &= \underbrace{\exp \left( \tilde{\phi}_t \right)}_{=\phi_t} \frac{1}{\phi_t} \left[ \frac{(\nu+1) z_t^2}{\nu + z_t^2} - 1 \right] \\
\tilde{u}_t &= \frac{(\nu+1) z_t^2}{\nu + z_t^2} - 1
\end{aligned}$$

Now we can write up the process and we have the model

$$\tilde{\phi}_t = \omega + \alpha \left[ \frac{(\nu+1) z_{t-1}^2}{\nu + z_{t-1}^2} - 1 \right] + \beta \tilde{\phi}_{t-1} \quad \square$$

## 2 Problem Set 5 - Problem 2)

### 2.1 Derive unconditional mean to initialize $\phi$

We keep the process for  $\tilde{\phi}_t$  in mind

$$\tilde{\phi}_t = \omega + \alpha \underbrace{\left[ \frac{(\nu+1) z_{t-1}^2}{\nu + z_{t-1}^2} - 1 \right]}_{\tilde{u}_{t-1}} + \beta \tilde{\phi}_{t-1}$$

Using recursive substitution and unfolding the process of  $\tilde{\phi}_t$  we obtain:

$$\begin{aligned}\tilde{\phi}_t &= \frac{\omega}{1-\beta} + \alpha \sum_{s=0}^{\infty} \beta^s u_{t-s-1} \\ \mathbb{E}[\tilde{\phi}_t] &= \mathbb{E}\left[\frac{\omega}{1-\beta} + \alpha \sum_{s=0}^{\infty} \beta^s u_{t-s-1}\right] \\ \mathbb{E}[\tilde{\phi}_t] &= \frac{\omega}{1-\beta} + \underbrace{\mathbb{E}\left[\alpha \sum_{s=0}^{\infty} \beta^s u_{t-s-1}\right]}_{=0, \text{ dont know why...}} \\ \mathbb{E}[\tilde{\phi}_t] &= \frac{\omega}{1-\beta}\end{aligned}$$

This is the **unconditional mean**. We know that we have an exponential link function, whereas

$$\phi_t = \exp(\tilde{\phi}_t)$$

Now we can use these values to initialize our GAS log-likelihood function.

## 2.2 Derive starting parameters

A way of initializing  $\theta = (\omega, \alpha, \beta, \nu)$

We can compute:

$$\begin{aligned}\hat{\sigma}^2 &= \text{Var}(y_t) \\ &= \frac{1}{T} \sum_t y_t^2\end{aligned}$$

$$\begin{aligned}\sigma_t^2 &= \frac{\phi_t^2 \nu}{\nu - 2} \\ E(\sigma_t^2) &= \frac{\nu}{\nu - 2} E(\exp(2\hat{\phi}_t))\end{aligned}$$

This expectation can be derived analytically, but this is very technical. Therefore we use an approximation.

$$\begin{aligned}E(\sigma_t^2) &= \frac{\nu}{\nu - 2} E(\exp(2\hat{\phi}_t)) \\ &\approx \frac{\nu}{\nu - 2} \exp(2E(\hat{\phi}_t)) \\ &= \frac{\nu}{\nu - 2} \exp\left(\frac{2\omega}{1-\beta}\right)\end{aligned}$$

We initialize  $\omega$ :

$$\begin{aligned}\sigma^2 \frac{(\nu - 2)}{\nu} &= \exp\left(\frac{2\omega}{1-\beta}\right) \\ \ln\left(\sigma^2 \frac{(\nu - 2)}{\nu}\right) \frac{1}{2} (1-\beta) &= \omega\end{aligned}$$

(Note that  $\sigma^2$  is here the variance of  $Y$ )

So

$$\omega^{\text{initialized}} = \ln\left(\hat{\sigma}^2 \frac{(\nu - 2)}{\nu}\right) \frac{1}{2} (1-\beta)$$