

# CRAN packages comparison

## 1 DiffusionRgqd

Uses the cumulant truncation procedure developed by Varughese (2013), whereby the transition density can be approximated over arbitrarily large transition horizons for a suitably general class of non-linear diffusion models.

Generalized quadratic diffusions (GQD) are the specific class of SDEs with quadratic drift and diffusion terms:

$$\begin{aligned} dX_t &= \mu(X_t, t)dt + \sigma(X_t, t)dW_t, \text{ where} \\ \mu(X_t, t) &= G_0(t) + G_1(t)X_t + G_2(t)X_t^2, \text{ and} \\ \sigma(X_t, t) &= Q_0(t) + Q_1(t)X_t + Q_2(t)X_t^2 \end{aligned}$$

For purposes of inference the drift and diffusion terms - and consequently the transitional density - are assumed to be dependent on a vector of parameters,  $\theta$ . For example, an Ornstein-Uhlenbeck model with SDE:

$$dX_t = \theta_1(\theta_2 - X_t) + \sqrt{\theta_3^2}dW_t \quad (1)$$

```
G0=function(t){theta[1]*theta[2]}
G1=function(t){-theta[1]}
Q0=function(t){theta[3]*theta[3]}
```

For a constant drift, diffusion SDE, with given initial condition  $X_s$ :

$$dX_t = \mu dt + \sigma dW_t \quad (2)$$

The distribution at time  $t$  of the process  $X_t$  is  $\mathcal{N}(X_t, X_s + \mu(t-s), \sigma^2(t-s))$

```
Xs <- 0 # Initial state
Xt <- seq(-3/2,3/2,1/50) # Possible future states
s <- 0 # Starting time
t <- 1 # Final time
mu <- 0.5 # Drift parameter
sigma <- 0.25 # Diffusion coefficient
```

```
library(DiffusionRgqd)
# Remove any existing coefficients:
GQD.remove()
```

```
# Define the model coefficients:
G0 <- function(t){mu}
Q0 <- function(t){sigma^2}
```

```
# Calculate the transitional density:
BM <- GQD.density(Xs,Xt,s,t)
```

- 2   **pomp: statistical inference for partially-observed Markov processes**
- 3   **Robfilter**
- 4   **Sim.DiffProc Package - FitSDE**
- 5   **HPloglik**
- 6   **abctools**