



Piecewise Deterministic Markov Processes in Julia: PDMP.jl

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Abstract

Simulation of stochastic processes is an important part of modeling and inference in many biological fields, including neuroscience, epidemiology population genetics, and systems biology. We present **PDMP.jl**, a package written in Julia for the efficient simulation of stochastic processes that comprise of both a discrete and a continuous component.

Keywords: Markov processes, simulation, Julia.

1. Introduction

The True Jump Method (TJM, [Veltz \(2015\)](#)) is an algorithm for the simulation of PDMPs.

2. The PDMP.jl package

The **PDMP.jl** package is written in the Julia programming language.

The True Jump Method involves solving stiff ordinary differential equations. We use the CVODE routine available in the **Sundials.jl** package; in preliminary analyses, CVODE was approximately twice as fast as the ode23s routine, implemented in native Julia code in the **ODE.jl** package.

3. Examples

The interface to the simulations closely resembles that used in the R package **GillespieSSA** ([Pineda-Krch 2008](#)).

3.1. The Morris-Lecar model

```
using PDMP, JSON, GR
```

```
const p0 = convert(Dict{AbstractString,Float64}, JSON.parsefile("ml.json")["type II"])
const p1 = ( JSON.parsefile("ml.json"))
include("morris_lecar_variables.jl")
const p_ml = ml(p0)

function F_ml(xcdot::Vector{Float64}, xc::Vector{Float64},xd::Array{Int64},t::Float64, par
    # vector field used for the continuous variable
    #compute the current, v = xc[1]
    xcdot[1] = xd[2] / p_ml.N * (p_ml.g_Na * (p_ml.v_Na - xc[1])) +
        xd[4] / p_ml.M * (p_ml.g_K * (p_ml.v_K - xc[1])) +
        (p_ml.g_L * (p_ml.v_L - xc[1])) + p_ml.I_app
    nothing
end

function R_ml(xc::Vector{Float64},xd::Array{Int64},t::Float64, parms::Vector, sum_rate::Bo
    if sum_rate==false
        return vec([p_ml.beta_na * exp(4.0 * p_ml.gamma_na * xc[1] + 4.0 * p_ml.k_na) * xd[1],
            p_ml.beta_na * xd[2],
            p_ml.beta_k * exp(p_ml.gamma_k * xc[1] + p_ml.k_k) * xd[3],
            p_ml.beta_k * exp(-p_ml.gamma_k * xc[1] - p_ml.k_k) * xd[4]])
    else
        return (p_ml.beta_na * exp(4.0 * p_ml.gamma_na * xc[1] + 4.0 * p_ml.k_na) * xd[1] +
            p_ml.beta_na * xd[2] +
            p_ml.beta_k * exp( p_ml.gamma_k * xc[1] + p_ml.k_k) * xd[3] +
            p_ml.beta_k * exp(-p_ml.gamma_k * xc[1] - p_ml.k_k) * xd[4])
    end
end

function Delta_ml(xc::Array{Float64},xd::Array{Int64},t::Float64,parms::Vector,ind_reactio
    # this function return the jump in the continuous component
    return true
end

immutable F_type; end
call(::Type{F_type},xcd, xc, xd, t, parms) = F_ml(xcd, xc, xd, t, parms)

immutable R_type; end
call(::Type{R_type},xc, xd, t, parms, sr) = R_ml(xc, xd, t, parms, sr)

immutable DX_type; end
call(::Type{DX_type},xc, xd, t, parms, ind_reaction) = Delta_ml(xc, xd, t, parms, ind_reac

xc0 = vec([p1["v(0)"]])
```

```

xd0 = vec([Int(p0["N"]),      #Na closed
           0,                 #Na opened
           Int(p0["M"]),      #K closed
           0]),               #K opened

nu = [[-1 1 0 0];[1 -1 0 1];[0 0 -1 1];[0 0 1 -1]]
parms = vec([0.])
tf = p1["t_end"]

dummy_t = chv(6,xc0,xd0, F_ml, R_ml,(x,y,t,pr,id)->vec([0.]), nu , parms,0.0,0.01,false)
srand(123)
dummy_t = @time chv(4500,xc0,xd0, F_ml, R_ml,(x,y,t,pr,id)->vec([0.]), nu , parms,0.0,tf,f
result = PDMP.chv_optim(2,xc0,xd0,F_type,R_type,DX_type,nu,parms,0.0,tf,false)
srand(123)
result = @time PDMP.chv_optim(4500,xc0,xd0,F_type,R_type,DX_type,nu,parms,0.0,tf,false) #
println("#jumps = ", length(dummy_t.time)," ", length(result.time))
try
  println(norm(dummy_t.time-result.time))
  println("--> xc_f-xc_t = ",norm(dummy_t.xc-result.xc))
  println("--> xd_f-xd_t = ",norm(dummy_t.xd-result.xd))
end
GR.plot(result.time,result.xc[1,:],"y",result.time, 0*result.xd[3,:],title = string("#Jump

```

4. Future directions

5. Acknowledgements

References

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