

*Simple*_Xⁿ module

IN480 course [\[1\]](#)

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1 Introduction

The *Simple_x* library, named **simplexn** within the Python version of the LARCC framework, provides combinatorial algorithms for some basic functions of geometric modelling with simplicial complexes. In particular, provides the efficient creation of simplicial complexes generated by simplicial complexes of lower dimension, the production of simplicial grids of any dimension, and the extraction of facets (i.e. of $(d - 1)$ -faces) of complexes of d -simplices. [2]

1.1 Requirements

There is no need to install extra packages to run the Julia code, it is sufficient loading the built-in package *Combinatorics.jl* (i.e. `using Combinatorics`); however, in order to be able to make the graphs it is necessary to install the package *Plots.jl* [5] (i.e. `Pkg.add("Plots")`) with the preferred backend (for example `Pkg.add("GR")`) and load them (i.e. `using Plots` and `gr()`).

Four processors were used (`addprocs(4)` at startup) for all the tests, five REPL included; for the tests environment it is necessary to load `using Base.Test`.

To execute the speedup code it is required the following:

```
using Plots
Plots.scalefontsizes(0.7)
gr() # loading backend

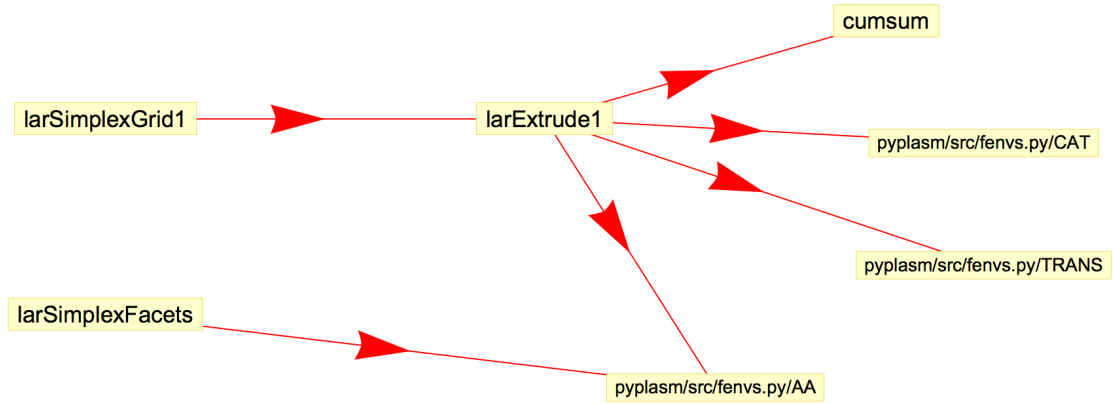
VOID = [Int64[]],[[0]] # the empty simplicial model
nt, N = 7, 5

# Compute the execution mean time
function timing(f::Function,x,n::Int64)
    t = Array{Float64}(n)
    f(x...)
    for k in 1:n
        t[k] = @elapsed f(x...)
    end
    return mean(t)
end

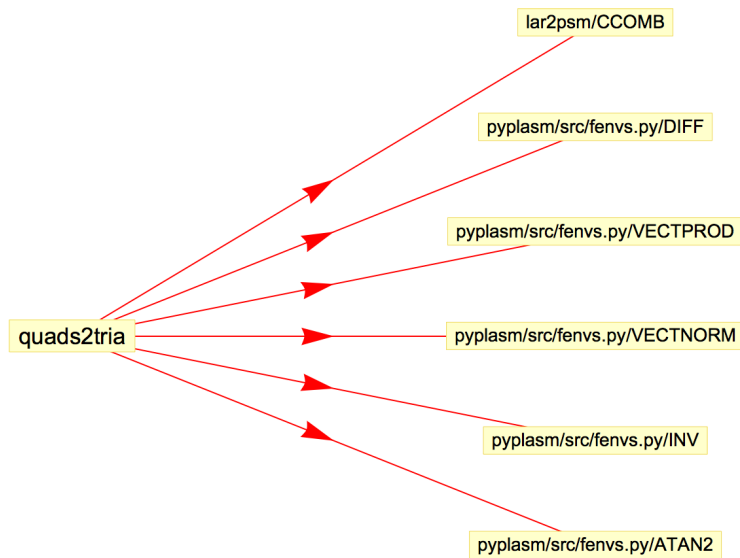
# Create the plots
function plotting(name::String,timeS::Array{Float64,1},timeP::Array{Float64,1})
    l = max(length(timeS),length(timeP))+1
    s, p, xlb, ylb, DPI = "Serial", "Parallel", "Input", "Time (seconds)", 150
    p1 = plot(timeS,label=s,title=name)
    p2 = plot(timeP,label=p)
    p3 = plot([timeS,timeP],label=[s,p])
    plot(p1,p2,p3,xlims=(1,l),xlabel=xlb,ylabel=ylb,dpi=1.5*DPI,legend=:topleft)
end
```

1.2 API

In Python the dependencies were the following (see figure 1):



(a) *larExtrude1*, *larSimplexGrid1*, and *larSimplexFacets*



(b) *quads2tria*

Figure 1: Dependencies graph

In the Julia translation all the dependencies have been removed, so now there is only `larSimplexGrid1` calling `larExtrude1`; moreover, `larSimplexFacets` uses the function `combinations` which is in the built-in package *Combinatorics.jl*.

The Julia API describes the serial functions but it is the same for the parallel ones, the only difference is the names beginning with a `p`.

```
larExtrude1{T<:Real}(model::Tuple{Array{Array{T,1},1},Array{Array{Int64,1},1}},
pattern::Array{Int64,1})
```

Description: this function generates the output model vertices in a multiple extrusion of a LAR model.

Input: `model` contains a pair (V, FV), where V is the array of input vertices, and FV is the array of d -cells (given as arrays of vertex indices) providing the input representation of a LAR cellular complex. `pattern` is an array of integers, whose absolute values provide the sizes of the ordered set of 1D (in local coords) subintervals specified by the pattern itself.

Output: it is a model representing the extrusion of the input `model`.

```
larSimplexGrid1(shape::Array{Int64,1})
```

Description: this function generates the simplicial grids of any dimension and shape.

Input: `shape` is an array of integers used to specify the shape of the created array.

Output: it is a model that contains a pair (V, FV), where V is the array of input vertices, and FV is the array of d -cells (given as arrays of vertex indices) providing the input representation of a LAR cellular complex.

```
larSimplexFacets(simplices::Array{Array{Int64,1},1})
```

Description: this function provides the extraction of non-oriented $(d - 1)$ -facets of d -dimensional simplices.

Input: `simplices` is the array of d -cells (given as arrays of vertex indices) providing the input representation of a LAR cellular complex.

Output: it is an array of d -cells of integers, i.e. the input LAR representation of the topology of a cellular complex.

```
quads2tria{T<:Real}(model::Tuple{Array{Array{T,1},1},Array{Array{Int64,1},1}})
```

Description: this function gives the conversion of a LAR boundary representation (B-Rep), i.e. a LAR model **V**, **FV** made of 2D faces, usually quads but also general polygons, into a LAR model **VERTS**, **TRIANGLES** made by triangles.

Input: `model` contains a pair (V, FV), where V is the array of input vertices, and FV is the array of d -cells (given as arrays of vertex indices) providing the input representation of a LAR cellular complex.

Output: it is a model that contains a pair (V, triangles) representing the triangulation of the input `model`.

2 Implementation

All the code in this section works with a simple copy and paste in Julia REPL; however, if a code block starts in a page and ends at the following one, it is required to pay attention at the numbers and headers of the pages.

Moreover, there are some comments, which could be useful, left on purpose in the code.

2.1 larExtrude1

This function generates the output model vertices in a multiple extrusion of a LAR model.

2.1.1 Python code

```
def larExtrude1(model,pattern):
    V, FV = model
    d, m = len(FV[0]), len(pattern)
    coords = list(cumsum([0]+(AA(ABS)(pattern))))
    offset, outcells, rangelimit = len(V), [], d*m
    for cell in FV:
        tube = [v + k*offset for k in range(m+1) for v in cell]
        cellTube = [tube[k:k+d+1] for k in range(rangelimit)]
        outcells += [reshape(cellTube, newshape=(m,d,d+1)).tolist()]

    outcells = AA(CAT)(TRANS(outcells))
    cellGroups = [group for k,group in enumerate(outcells) if pattern[k]>0]
    outVertices = [v+[z] for z in coords for v in V]
    outModel = outVertices, CAT(cellGroups)
    return outModel
```

2.1.2 Julia code - serial

During the translation a part of the code is changed to use a matrix for outcells instead of a list of lists of lists.

```
# Generation of the output model vertices in a multiple extrusion of a LAR model
function larExtrude1{T<:Real}(model::Tuple{Array{Array{T,1},1},Array{
    Array{Int64,1},1}},pattern::Array{Int64,1})
    V, FV = model
    d, m = length(FV[1]), length(pattern)
    coords = cumsum(append!([0],abs.(pattern))) # built-in function cumsum
    offset, outcells, rangelimit = length(V), Array{Int64}(m,0), d*m
    for cell in FV
        tube = [v+k*offset for k in 0:m for v in cell]
        celltube = Int64[]
        for k in 1:rangelimit
            append!(celltube,tube[k:k+d])
        end
        outcells = hcat(outcells,reshape(celltube,d*(d+1),m)')
    end
    cellGroups = Int64[]
    for k in 1:m
        if pattern[k]>0
            cellGroups = vcat(cellGroups,outcells[k,:])
        end
    end
end
```

```

        end
    end
    outVertices = [vcat(v,z) for z in coords for v in V]
    outCellGroups = Array{Int64,1}[]
    for k in 1:d+1:length(cellGroups)
        append!(outCellGroups,[cellGroups[k:k+d]])
    end
    return outVertices, outCellGroups
end

```

2.1.3 Parallel optimization

Every single `for` loop has been parallelized with the `@parallel` macro, using also `@sync` where needed.

`outVertices` is computed as soon as possible with the help of the macro `@spawn` because it is independent from the rest of the code. The `outcells` array is changed into a `SharedArray` to be available on all the processors, and the dimension is established a priori to preserve the correct order of the computation and avoid the possible random order due to the `hcat` in the `@parallel for`; same as for `cellGroups`, and in this case a `SharedArray p` has been used to remove the `if` inside the loop which would have invalidated the parallelization.

2.1.4 Julia code - parallel

```

# Generation of the output model vertices in a multiple extrusion of a LAR model
@everywhere function plarExtrude1{T<:Real}(model::Tuple{Array{Array{T,1},1},Array{
    Array{Int64,1},1}}, pattern::Array{Int64,1})
    V, FV = model
    d, m = length(FV[1]), length(pattern)
    dd1 = d*(d+1)
    coords = cumsum(append!([0],abs.(pattern))) # built-in function cumsum
    outVertices = @spawn [vcat(v,z) for z in coords for v in V]
    offset,outcells,rangelimit=length(V),SharedArray{Int64}(m,dd1*length(FV)),d*m
    @sync @parallel for j in 1:length(FV)
        tube = [v+k*offset for k in 0:m for v in FV[j]]
        celltube = Int64[]
        celltube = @sync @parallel (append!) for k in 1:rangelimit
            tube[k:k+d]
        end
        outcells[:,(j-1)*dd1+1:(j-1)*dd1+dd1] = reshape(celltube,dd1,m) '
    end
    p = convert(SharedArray,find(x->x>0,pattern))
    cellGroups = SharedArray{Int64}(length(p),size(outcells)[2])
    @sync @parallel for k in 1:length(p)
        cellGroups[k,:] = outcells[p[k],:]
    end
    cellGroupsL = vec(cellGroups')

```

```

    outCellGroups = Array{Int64,1}[]
    outCellGroups = @parallel (append!) for k in 1:d+1:length(cellGroupsL)
        [cellGroupsL[k:k+d]]
    end
    return fetch(outVertices), outCellGroups
end

```

2.1.5 Unit tests code

- *Serial Tests* -

```

@testset "larExtrude1" begin
    VOID = [Int64[]], [[0]] # the empty simplicial model
    pattern1 = [1,-1]
    model1 = larExtrude1(VOID,pattern1) # 1D
    @test typeof(model1) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
    @test length(model1[1]) == length(pattern1)+1 # num of vertices, no rep
    @test length(model1[2]) == length(filter(x->x>0,pattern1)) #num of 1D-simplices

    pattern2 = [1,1,1,-1]
    model2 = larExtrude1(VOID,pattern2) # 1D
    @test typeof(model2) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
    @test length(model2[1]) == length(pattern2)+1 # num of vertices, no rep
    @test length(model2[2]) == length(filter(x->x>0,pattern2)) #num of 1D-simplices

    pattern3 = [1,1,1,1,1,-1]
    model3 = larExtrude1(VOID,pattern3) # 1D
    @test typeof(model3) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
    @test length(model3[1]) == length(pattern3)+1 # num of vertices, no rep
    @test length(model3[2]) == length(filter(x->x>0,pattern3)) #num of 1D-simplices

    model1 = larExtrude1(model1,pattern1) # 2D
    @test typeof(model1) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
    @test length(model1[1]) == (length(pattern1)+1)^2 # num of vertices, no rep
    @test length(model1[2]) == 2*length(filter(x->x>0,pattern1))^2 #num of 2D-simplices

    model2 = larExtrude1(model2,pattern2) # 2D
    @test typeof(model2) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
    @test length(model2[1]) == (length(pattern2)+1)^2 # num of vertices, no rep
    @test length(model2[2]) == 2*length(filter(x->x>0,pattern2))^2 #num of 2D-simplices

    model3 = larExtrude1(model3,pattern3) # 2D
    @test typeof(model3) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
    @test length(model3[1]) == (length(pattern3)+1)^2 # num of vertices, no rep
    @test length(model3[2]) == 2*length(filter(x->x>0,pattern3))^2 #num of 2D-simplices

    model1 = larExtrude1(model1,pattern1) # 3D
    @test typeof(model1) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}

```



```

@test length(model1[1]) == (length(pattern1)+1)^3 # num of vertices, no rep
@test length(model1[2]) == 6*length(filter(x->x>0,pattern1))^3 #num of 3D-simplices

model2 = larExtrude1(model2,pattern2) # 3D
@test typeof(model2) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
@test length(model2[1]) == (length(pattern2)+1)^3 # num of vertices, no rep
@test length(model2[2]) == 6*length(filter(x->x>0,pattern2))^3 #num of 3D-simplices

model3 = larExtrude1(model3,pattern3) # 3D
@test typeof(model3) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
@test length(model3[1]) == (length(pattern3)+1)^3 # num of vertices, no rep
@test length(model3[2]) == 6*length(filter(x->x>0,pattern3))^3 #num of 3D-simplices

model = ([[0.0,0],[0,1]],[[1,0],[1,1]]) # 2D with float
@test typeof(model[1]) == Array{Array{Float64,1},1}
model = larExtrude1(model,pattern1)
@test typeof(model) == Tuple{Array{Array{Float64,1},1},Array{Array{Int64,1},1}}
end

```

- *Parallel Tests* -

```

@testset "plarExtrude1" begin
  VOID = [Int64[]],[[0]] # the empty simplicial model
  pattern1 = [1,-1]
  model1 = plarExtrude1(VOID,pattern1) # 1D
  @test typeof(model1) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
  @test length(model1[1]) == length(pattern1)+1 # num of vertices, no rep
  @test length(model1[2]) == length(filter(x->x>0,pattern1)) #num of 1D-simplices

  pattern2 = [1,1,1,-1]
  model2 = plarExtrude1(VOID,pattern2) # 1D
  @test typeof(model2) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
  @test length(model2[1]) == length(pattern2)+1 # num of vertices, no rep
  @test length(model2[2]) == length(filter(x->x>0,pattern2)) #num of 1D-simplices

  pattern3 = [1,1,1,1,1,-1]
  model3 = plarExtrude1(VOID,pattern3) # 1D
  @test typeof(model3) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
  @test length(model3[1]) == length(pattern3)+1 # num of vertices, no rep
  @test length(model3[2]) == length(filter(x->x>0,pattern3)) #num of 1D-simplices

  model1 = plarExtrude1(model1,pattern1) # 2D
  @test typeof(model1) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
  @test length(model1[1]) == (length(pattern1)+1)^2 # num of vertices, no rep
  @test length(model1[2]) == 2*length(filter(x->x>0,pattern1))^2 #num of 2D-simplices

  model2 = plarExtrude1(model2,pattern2) # 2D
  @test typeof(model2) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}

```

```

@test length(model2[1]) == (length(pattern2)+1)^2 # num of vertices, no rep
@test length(model2[2]) == 2*length(filter(x->x>0,pattern2))^2 #num of 2D-simplices

model3 = plarExtrude1(model3,pattern3) # 2D
@test typeof(model3) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
@test length(model3[1]) == (length(pattern3)+1)^2 # num of vertices, no rep
@test length(model3[2]) == 2*length(filter(x->x>0,pattern3))^2 #num of 2D-simplices

model1 = plarExtrude1(model1,pattern1) # 3D
@test typeof(model1) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
@test length(model1[1]) == (length(pattern1)+1)^3 # num of vertices, no rep
@test length(model1[2]) == 6*length(filter(x->x>0,pattern1))^3 #num of 3D-simplices

model2 = plarExtrude1(model2,pattern2) # 3D
@test typeof(model2) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
@test length(model2[1]) == (length(pattern2)+1)^3 # num of vertices, no rep
@test length(model2[2]) == 6*length(filter(x->x>0,pattern2))^3 #num of 3D-simplices

model3 = plarExtrude1(model3,pattern3) # 3D
@test typeof(model3) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
@test length(model3[1]) == (length(pattern3)+1)^3 # num of vertices, no rep
@test length(model3[2]) == 6*length(filter(x->x>0,pattern3))^3 #num of 3D-simplices

model = ([0.0,0],[0,1],[1,0],[1,1]) # 2D with float
@test typeof(model[1]) == Array{Array{Float64,1},1}
model = plarExtrude1(model,pattern1)
@test typeof(model) == Tuple{Array{Array{Float64,1},1},Array{Array{Int64,1},1}}
end

```

2.1.6 Speedup script code

```

timeSer = [timing(larExtrude1,[VOID,repmat([1,2,3,4],k^4)],nt) for k in 1:2*N]
timePar = [timing(plarExtrude1,[VOID,repmat([1,2,3,4],k^4)],nt) for k in 1:2*N]
plotting("larExtrude1",timeSer,timePar)

```

See figure [2](#).

2.2 larSimplexGrid1

This function generates the simplicial grids of any dimension and shape.

2.2.1 Python code

```

def larSimplexGrid1(shape):
    model = VOID
    for item in shape:
        model = larExtrude1(model,item*[1])
    return model

```

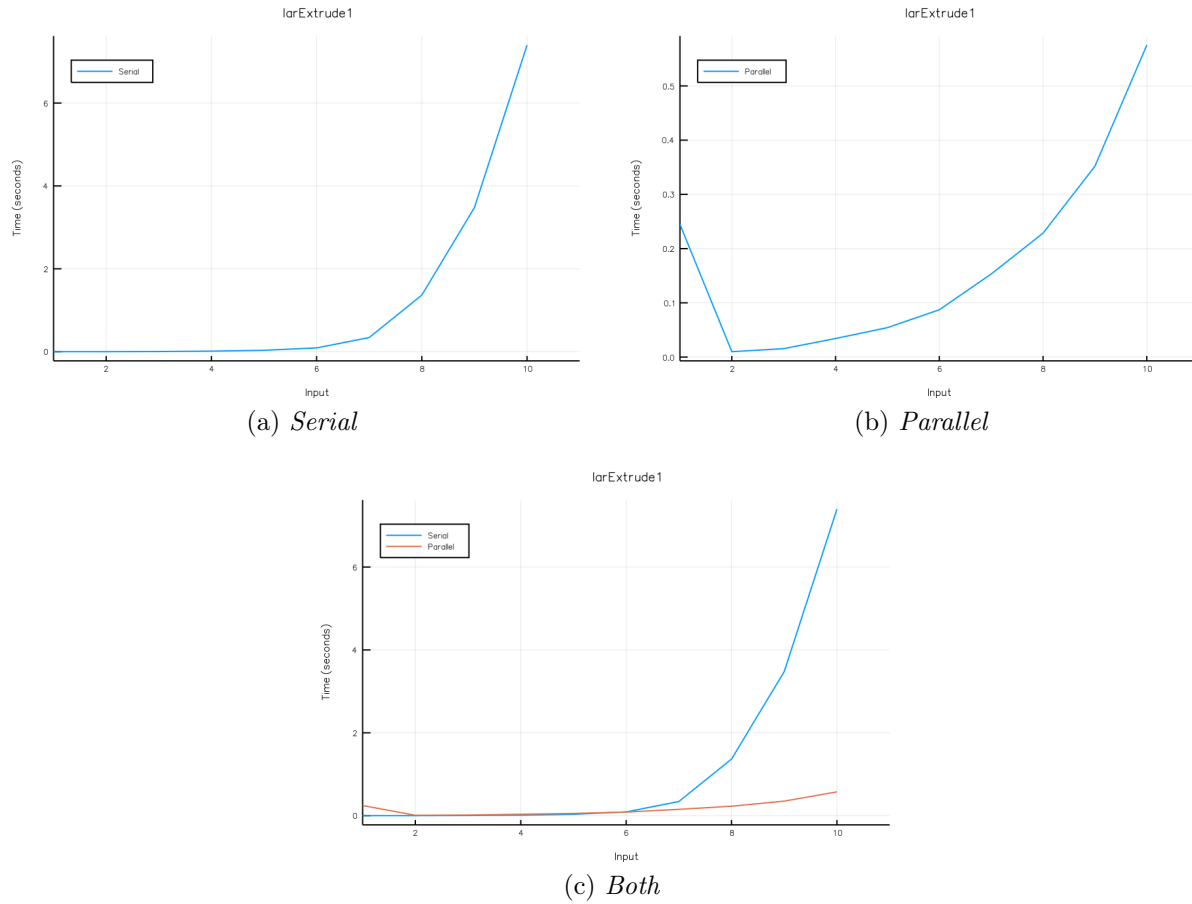


Figure 2: larExtrude1

2.2.2 Julia code - serial

```
# Generation of simplicial grids of any dimension and shape
function larSimplexGrid1(shape::Array{Int64,1})
    model = [Int64[]],[[0]] # the empty simplicial model
    for item in shape
        model = larExtrude1(model, repmat([1],item))
    end
    return model
end
```

2.2.3 Parallel optimization

It is not possible to parallelize this function because every iteration of the loop requires the `model` that is computed in the previous one. The only difference here is the addition of `@everywhere`.

2.2.4 Julia code - parallel

```
# Generation of simplicial grids of any dimension and shape
@everywhere function plarSimplexGrid1(shape::Array{Int64,1})
    model = [Int64[]],[[0]] # the empty simplicial model
    for item in shape # no parallel
        model = plarExtrude1(model, repmat([1],item))
    end
    return model
end
```

2.2.5 Unit tests code

- *Serial Tests* -

```
@testset "larSimplexGrid1" begin
    shape = [3] # 1D
    model1 = larSimplexGrid1(shape)
    @test typeof(model1) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
    @test length(model1[1]) == shape[1]+1 # num of vertices, no rep
    @test length(model1[2]) == shape[1] # num of 1D-simplices

    shape = [3,5] # 2D
    model2 = larSimplexGrid1(shape)
    @test typeof(model2) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
    # (shape[1]+1)*(shape[2]+1)
    @test length(model2[1]) == prod(shape)+sum(shape)+1 # num of vertices, no rep
    @test length(model2[2]) == 2*prod(shape) # num of 2D-simplices

    shape = [3,5,7] # 3D
    model3 = larSimplexGrid1(shape)
    @test typeof(model3) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
    # (shape[1]+1)*(shape[2]+1)*(shape[3]+1); num of vertices, no rep
    @test length(model3[1]) == prod(shape)+sum(prod.(collect(combinations(shape,2))))
        +sum(shape)+1
    @test length(model3[2]) == 6*prod(shape) # num of 3D-simplices
end
```

- *Parallel Tests* -

```
@testset "plarSimplexGrid1" begin
    shape = [3] # 1D
    model1 = plarSimplexGrid1(shape)
    @test typeof(model1) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
    @test length(model1[1]) == shape[1]+1 # num of vertices, no rep
    @test length(model1[2]) == shape[1] # num of 1D-simplices

    shape = [3,5] # 2D
```

```

model2 = plarSimplexGrid1(shape)
@test typeof(model2) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
# (shape[1]+1)*(shape[2]+1)
@test length(model2[1]) == prod(shape)+sum(shape)+1 # num of vertices, no rep
@test length(model2[2]) == 2*prod(shape) # num of 2D-simplices

shape = [3,5,7] # 3D
model3 = plarSimplexGrid1(shape)
@test typeof(model3) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
# (shape[1]+1)*(shape[2]+1)*(shape[3]+1); num of vertices, no rep
@test length(model3[1]) == prod(shape)+sum(prod.(collect(combinations(shape,2))))
    +sum(shape)+1
@test length(model3[2]) == 6*prod(shape) # num of 3D-simplices
end

```

2.2.6 Speedup script code

The code of the two versions is identical so it does not make sense to check the speedup.

2.3 larSimplexFacets

This function provides the extraction of non-oriented $(d - 1)$ -facets of d -dimensional simplices.

2.3.1 Python code

```

def larSimplexFacets(simplices):
    out = []
    d = len(simplices[0])
    for simplex in simplices:
        out += AA(sorted)([simplex[0:k]+simplex[k+1:d] for k in range(d)])
    out = set(AA(tuple)(out))
    return sorted(out)

```

2.3.2 Julia code - serial

```

# Extraction of non-oriented (d-1)-facets of d-dimensional simplices
using Combinatorics # for combinations() function

function larSimplexFacets(simplices::Array{Array{Int64,1},1})
    out = Array{Int64,1}[]
    d = length(simplices[1])
    for simplex in simplices
        append!(out,collect(combinations(simplex,d-1)))
    end
    return sort!(unique(out),lt=lexless)
end

```

2.3.3 Parallel optimization

Here, other than the usual addition of `@everywhere`, the `@parallel` was used to split the computation of the `for` among multiple processors. The `return` automatically waits the end of the computation.

2.3.4 Julia code - parallel

```
# Extraction of non-oriented (d-1)-facets of d-dimensional simplices
@everywhere using Combinatorics # for combinations() function

@everywhere function plarSimplexFacets(simplices::Array{Array{Int64,1},1})
    out = Array{Int64,1}[]
    d = length(simplices[1])
    out = @parallel (append!) for simplex in simplices
        collect(combinations(simplex,d-1))
    end
    return sort!(unique(out),lt=lexless)
end
```

2.3.5 Unit tests code

- *Serial Tests* -

```
@testset "larSimplexFacets" begin
    s = larSimplexGrid1([3])[2] # 1D
    sOut = larSimplexFacets(s)
    @test typeof(sOut) == Array{Array{Int64,1},1}
    d1 = sum([length(s[k]) for k in 1:length(s)]) / length(s)
    d2 = sum([length(sOut[k]) for k in 1:length(sOut)]) / length(sOut)
    @test d1-1 == d2 # dimension
    # length(s)*binomial(length(s[1]),(length(s[1])-1))
    @test length(sOut) <= length(s)*length(s[1]) # "<=" because no rep

    s = larSimplexGrid1([3,5])[2] # 2D
    sOut = larSimplexFacets(s)
    @test typeof(sOut) == Array{Array{Int64,1},1}
    d1 = sum([length(s[k]) for k in 1:length(s)]) / length(s)
    d2 = sum([length(sOut[k]) for k in 1:length(sOut)]) / length(sOut)
    @test d1-1 == d2 # dimension
    # length(s)*binomial(length(s[1]),(length(s[1])-1))
    @test length(sOut) <= length(s)*length(s[1]) # "<=" because no rep

    s = larSimplexGrid1([3,5,7])[2] # 3D
    sOut = larSimplexFacets(s)
    @test typeof(sOut) == Array{Array{Int64,1},1}
    d1 = sum([length(s[k]) for k in 1:length(s)]) / length(s)
    d2 = sum([length(sOut[k]) for k in 1:length(sOut)]) / length(sOut)
```

```

    @test d1-1 == d2 # dimension
    # length(s)*binomial(length(s[1]),(length(s[1])-1))
    @test length(sOut) <= length(s)*length(s[1]) # "<=" because no rep
end

```

- *Parallel Tests* -

```

@testset "plarSimplexFacets" begin
    s = plarSimplexGrid1([3])[2] # 1D
    sOut = plarSimplexFacets(s)
    @test typeof(sOut) == Array{Array{Int64,1},1}
    d1 = sum([length(s[k]) for k in 1:length(s)])/length(s)
    d2 = sum([length(sOut[k]) for k in 1:length(sOut)])/length(sOut)
    @test d1-1 == d2 # dimension
    # length(s)*binomial(length(s[1]),(length(s[1])-1))
    @test length(sOut) <= length(s)*length(s[1]) # "<=" because no rep

    s = plarSimplexGrid1([3,5])[2] # 2D
    sOut = plarSimplexFacets(s)
    @test typeof(sOut) == Array{Array{Int64,1},1}
    d1 = sum([length(s[k]) for k in 1:length(s)])/length(s)
    d2 = sum([length(sOut[k]) for k in 1:length(sOut)])/length(sOut)
    @test d1-1 == d2 # dimension
    # length(s)*binomial(length(s[1]),(length(s[1])-1))
    @test length(sOut) <= length(s)*length(s[1]) # "<=" because no rep

    s = plarSimplexGrid1([3,5,7])[2] # 3D
    sOut = plarSimplexFacets(s)
    @test typeof(sOut) == Array{Array{Int64,1},1}
    d1 = sum([length(s[k]) for k in 1:length(s)])/length(s)
    d2 = sum([length(sOut[k]) for k in 1:length(sOut)])/length(sOut)
    @test d1-1 == d2 # dimension
    # length(s)*binomial(length(s[1]),(length(s[1])-1))
    @test length(sOut) <= length(s)*length(s[1]) # "<=" because no rep
end

```

2.3.6 Speedup script code

```

simp = [collect(1:500)]
sLen = length(simp[1])
for k in 2:2*N^2
    push!(simp,simp[end]+sLen)
end
timeSer = [timing(larSimplexFacets,[simp[1:k]],nt) for k in 1:2*N^2]
timePar = [timing(plarSimplexFacets,[simp[1:k]],nt) for k in 1:2*N^2]
plotting("larSimplexFacets",timeSer,timePar)

```

See figure 3

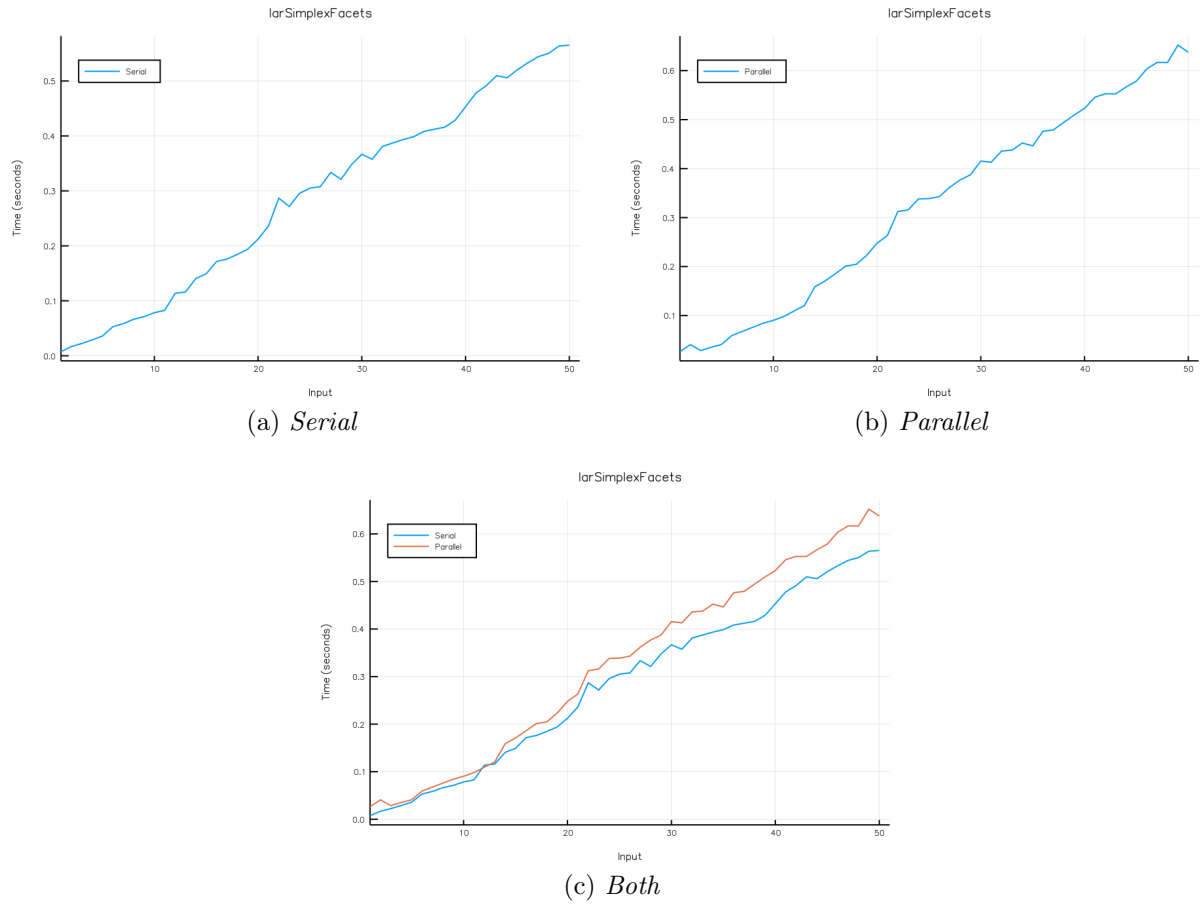


Figure 3: larSimplexFacets

2.4 quads2tria

This function gives the conversion of a LAR boundary representation (B-Rep), i.e. a LAR model **V**, **FV** made of 2D faces, usually quads but also general polygons, into a LAR model **VERTS**, **TRIANGLES** made by triangles.

2.4.1 Python code

```
def quads2tria(model):
    V,FV = model
    out = []
    nverts = len(V)-1
    for face in FV:
        centroid = CCOMB([V[v] for v in face])
        V += [centroid]
        nverts += 1

    v1, v2 = DIFF([V[face[0]],centroid]), DIFF([V[face[1]],centroid])
```



```

v3 = VECTPROD([v1,v2])
if ABS(VECTNORM(v3)) < 10**3:
    v1, v2 = DIFF([V[face[0]],centroid]), DIFF([V[face[2]],centroid])
    v3 = VECTPROD([v1,v2])
transf = mat(INV([v1,v2,v3]))
verts = [(V[v]*transf).tolist()[0][:-1] for v in face]

tcentroid = CCOMB(verts)
tverts = [DIFF([v,tcentroid]) for v in verts]
rverts = sorted([[ATAN2(vert),v] for vert,v in zip(tverts,face)])
ord = [pair[1] for pair in rverts]
ord = ord + [ord[0]]
edges = [[n,ord[k+1]] for k,n in enumerate(ord[:-1])]
triangles = [[nverts] + edge for edge in edges]
out += triangles
return V,out

```

2.4.2 Julia code - serial

During the translation the if condition was corrected from $< 10^{**3}$ to $< 1/(10^3)$.

```

# Transformation to triangles by sorting circularly the vertices of faces
function quads2tria{T<:Real}(model::Tuple{Array{Array{T,1},1},Array{
    Array{Int64,1},1}})
    V, FV = model
    if typeof(V) != Array{Array{Float64,1},1}
        V = convert(Array{Array{Float64,1},1},V)
    end
    out = Array{Int64,1}[]
    nverts = length(V)-1
    for face in FV
        arr = [V[v+1] for v in face]
        centroid = sum(arr)/length(arr)
        append!(V,[centroid])
        nverts += 1
        v1, v2 = V[face[1]+1]-centroid, V[face[2]+1]-centroid
        v3 = cross(v1,v2)
        if norm(v3) < 1/(10^3)
            v1, v2 = V[face[1]+1]-centroid, V[face[3]+1]-centroid
            v3 = cross(v1,v2)
        end
        transf = inv(hcat(v1,v2,v3)')
        verts = [(V[v+1]'*transf)'[1:end-1] for v in face]
        tcentroid = sum(verts)/length(verts)
        tverts = [v-tcentroid for v in verts]
        iterator = collect(zip(tverts,face))
        rverts = [[atan2(reverse(iterator[i][1]),iterator[i][2])) for i in

```

```

        1:length(iterator)]
    rvertsS = sort(rverts,lt=(x,y)->isless(x[1],y[1]))
    ord = [pair[2] for pair in rvertsS]
    append!(ord,ord[1])
    edges = [[i[2],ord[i[1]+1]] for i in enumerate(ord[1:end-1])]
    triangles = [prepend!(edge,nverts) for edge in edges]
    append!(out,triangles)
end
return V, out
end

```

2.4.3 Parallel optimization

The array comprehension was transformed, where possible, into a `pmap`; it is not possible to parallelize the `for` with a `@parallel` because `append!(V,[centroid])` needs to be computed before the next iteration of the loop.

2.4.4 Julia code - parallel

```

# Transformation to triangles by sorting circularly the vertices of faces
@everywhere function pquads2tria{T<:Real}(model::Tuple{Array{Array{T,1},1},Array{
    Array{Int64,1},1}})
    V, FV = model
    if typeof(V) != Array{Array{Float64,1},1}
        V = convert(Array{Array{Float64,1},1},V)
    end
    out = Array{Int64,1}[]
    nverts = length(V)-1
    for face in FV # no parallel
        arr = [V[v+1] for v in face]
        centroid = sum(arr)/length(arr)
        append!(V,[centroid])
        nverts += 1
        v1, v2 = V[face[1]+1]-centroid, V[face[2]+1]-centroid
        v3 = cross(v1,v2)
        if norm(v3) < 1/(10^3)
            v1, v2 = V[face[1]+1]-centroid, V[face[3]+1]-centroid
            v3 = cross(v1,v2)
        end
        transf = inv(hcat(v1,v2,v3)')
        verts = [(V[v+1]'*transf)'[1:end-1] for v in face]
        tcentroid = sum(verts)/length(verts)
        tverts = pmap(x->x-tcentroid,verts)
        iterator = collect(zip(tverts,face))
        rverts = [[atan2(reverse(iterator[i][1])...),iterator[i][2]] for i in
            1:length(iterator)]
        rvertsS = sort(rverts,lt=(x,y)->isless(x[1],y[1]))
        ord = [pair[2] for pair in rvertsS]
    end
end

```

```

        append!(ord,ord[1])
        edges = [[i[2],ord[i[1]+1]] for i in enumerate(ord[1:end-1])]
        triangles = pmap(x->prepend!(x,nverts),edges)
        append!(out,triangles)
    end
    return V, out
end

```

2.4.5 Unit tests code

- *Serial Tests* -

```

@testset "quads2tria" begin
    modIn = ([[0,0,0],[0,1,0],[1,0,0],[1,1,0]],[[0,1,2,3]]) # 2D
    modOut = quads2tria(modIn)
    @test typeof(modIn) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
    @test typeof(modOut) == Tuple{Array{Array{Float64,1},1},Array{Array{Int64,1},1}}
    edges = 2*sum([length(modIn[2][k]) for k in 1:length(modIn[2])])
    EulerChar = length(modOut[1])-edges+length(modOut[2]) # V - E + F
    @test EulerChar == 1

    modIn = ([[0,0,0],[0,1,0],[1,0,0],[1,1,0],[0.5,1.5,0]],[[0,1,2,3,4]]) # 2D
    modOut = quads2tria(modIn)
    @test typeof(modIn) == Tuple{Array{Array{Float64,1},1},Array{Array{Int64,1},1}}
    @test typeof(modOut) == Tuple{Array{Array{Float64,1},1},Array{Array{Int64,1},1}}
    edges = 2*sum([length(modIn[2][k]) for k in 1:length(modIn[2])])
    EulerChar = length(modOut[1])-edges+length(modOut[2]) # V - E + F
    @test EulerChar == 1

    modIn = ([[0,0,0],[0,1,0],[1,0,0],[1,1,0],[0,0,1],[0,1,1],[1,0,1],[1,1,1]],[[0,1,2,3,4,5,6,7]]) # 3D
    modOut = quads2tria(modIn)
    @test typeof(modIn) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
    @test typeof(modOut) == Tuple{Array{Array{Float64,1},1},Array{Array{Int64,1},1}}
    edges = 2*sum([length(modIn[2][k]) for k in 1:length(modIn[2])])
    EulerChar = length(modOut[1])-edges+length(modOut[2]) # V - E + F
    @test EulerChar == 1

    modIn = ([[0,0,0],[0,1,0],[1,0,0],[1,1,0],[0,0,1],[0,1,1],[1,0,1],[1,1,1],[0.5,0.5,1.5]],[[0,1,2,3,4,5,6,7,8]]) # 3D
    modOut = quads2tria(modIn)
    @test typeof(modIn) == Tuple{Array{Array{Float64,1},1},Array{Array{Int64,1},1}}
    @test typeof(modOut) == Tuple{Array{Array{Float64,1},1},Array{Array{Int64,1},1}}
    edges = 2*sum([length(modIn[2][k]) for k in 1:length(modIn[2])])
    EulerChar = length(modOut[1])-edges+length(modOut[2]) # V - E + F
    @test EulerChar == 1
end

```

- *Parallel Tests* -

```

@testset "pqquads2tria" begin
    modIn = ([0,0,0],[0,1,0],[1,0,0],[1,1,0],[0,1,2,3]) # 2D
    modOut = pqquads2tria(modIn)
    @test typeof(modIn) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
    @test typeof(modOut) == Tuple{Array{Array{Float64,1},1},Array{Array{Int64,1},1}}
    edges = 2*sum([length(modIn[2][k]) for k in 1:length(modIn[2])])
    EulerChar = length(modOut[1])-edges+length(modOut[2]) # V - E + F
    @test EulerChar == 1

    modIn = ([0,0,0],[0,1,0],[1,0,0],[1,1,0],[0.5,1.5,0],[0,1,2,3,4]) # 2D
    modOut = pqquads2tria(modIn)
    @test typeof(modIn) == Tuple{Array{Array{Float64,1},1},Array{Array{Int64,1},1}}
    @test typeof(modOut) == Tuple{Array{Array{Float64,1},1},Array{Array{Int64,1},1}}
    edges = 2*sum([length(modIn[2][k]) for k in 1:length(modIn[2])])
    EulerChar = length(modOut[1])-edges+length(modOut[2]) # V - E + F
    @test EulerChar == 1

    modIn = ([0,0,0],[0,1,0],[1,0,0],[1,1,0],[0,0,1],[0,1,1],[1,0,1],[1,1,1],
        [[0,1,2,3,4,5,6,7]]) # 3D
    modOut = pqquads2tria(modIn)
    @test typeof(modIn) == Tuple{Array{Array{Int64,1},1},Array{Array{Int64,1},1}}
    @test typeof(modOut) == Tuple{Array{Array{Float64,1},1},Array{Array{Int64,1},1}}
    edges = 2*sum([length(modIn[2][k]) for k in 1:length(modIn[2])])
    EulerChar = length(modOut[1])-edges+length(modOut[2]) # V - E + F
    @test EulerChar == 1

    modIn = ([0,0,0],[0,1,0],[1,0,0],[1,1,0],[0,0,1],[0,1,1],[1,0,1],[1,1,1],[0.5,0.5,1.5],
        [[0,1,2,3,4,5,6,7,8]]) # 3D
    modOut = pqquads2tria(modIn)
    @test typeof(modIn) == Tuple{Array{Array{Float64,1},1},Array{Array{Int64,1},1}}
    @test typeof(modOut) == Tuple{Array{Array{Float64,1},1},Array{Array{Int64,1},1}}
    edges = 2*sum([length(modIn[2][k]) for k in 1:length(modIn[2])])
    EulerChar = length(modOut[1])-edges+length(modOut[2]) # V - E + F
    @test EulerChar == 1
end

```

2.4.6 Speedup script code

```

verts = [[0,0,0],[0,1,0],[1,0,0],[1,1,0],[2,2,0],[2,3,0],[3,2,0],[3,3,0]]
quads = [[0,1,2,3],[4,5,6,7]]
len = length(quads[1])
for k in 2:2*N^2
    append!(verts,verts[end-len+1:end]+1)
    append!(quads,[quads[end]+len])
end
timeSer = [timing(quads2tria,[(verts[1:len*k],quads[1:k])],nt) for k in 1:2*N^2]
timePar = [timing(pquads2tria,[(verts[1:len*k],quads[1:k])],nt) for k in 1:2*N^2]
plotting("quads2tria",timeSer,timePar)

```

See figure 4.

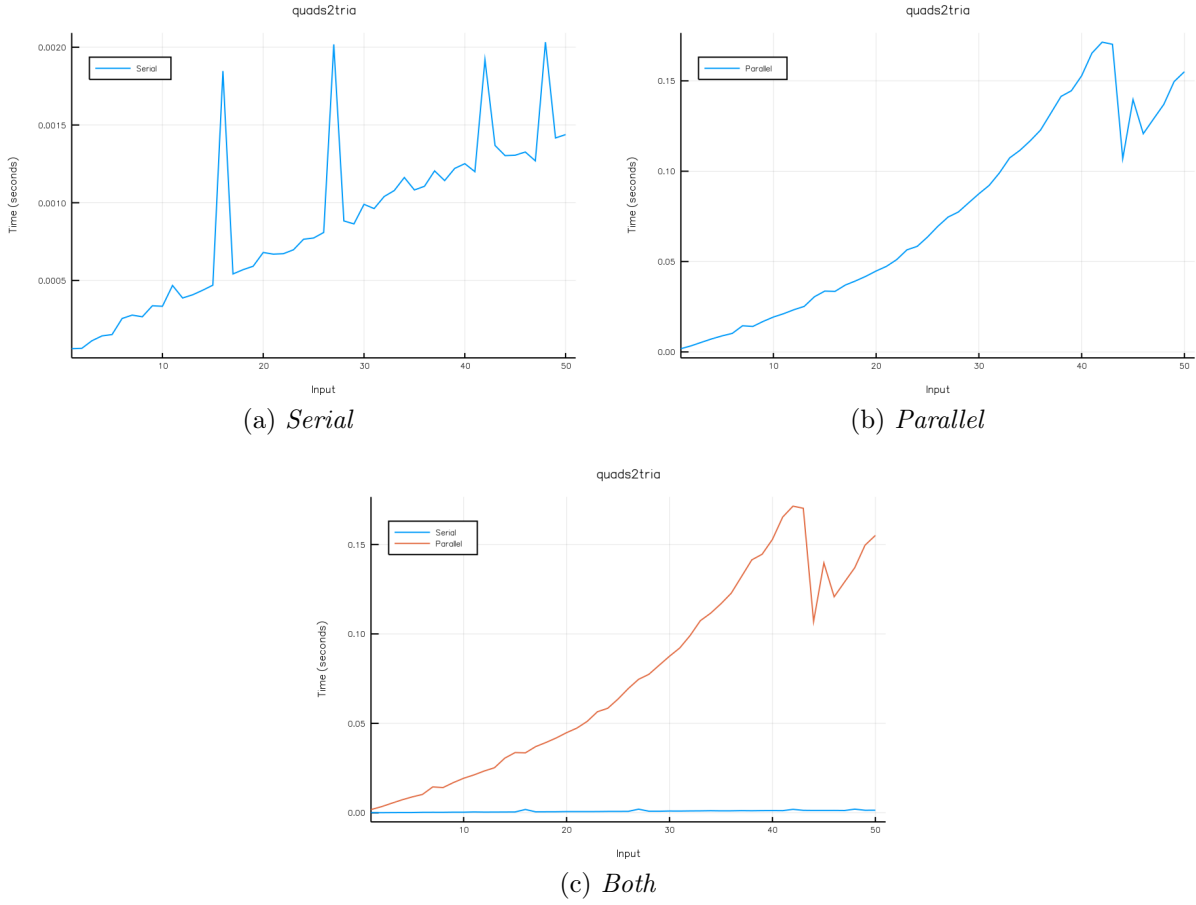


Figure 4: `quads2tria`

3 Conclusions

The `larExtrude1` function was rethought almost from scratch (to avoid the list nesting of Python and to allow the parallelization process) and in fact it is a lot faster with the increasing of the input size.

The other two parallelized functions have showed the parallel code is not faster than the serial one. A possible way to address this problem could be to rewrite all the functions using different structures and procedures to handle the data, avoiding array of arrays and similar. However, the complete lack of documentation online, official and non, for the correct use of the macros and how they specifically work makes the task quite difficult.

A Examples

The translation of the Python examples and tests is not shown in this report because the code is directly compared with the Python output, which is pretty verbose.

However, the links for the code in the GitHub repository are provided [here](#) and [here](#), respectively for the serial and the parallel version.

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References

- [1] IN480 course [web page](#).
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