NIPALS_PCA.jl G Edit on GitHub NIPALS_PCA NIPALS_PCA NIPALS_PCA A Julia package for calculating PCA and PLS using the NIPALS implementation. Both models handles missing values Installation O Running Julia REPL The package contains data structures for models and datasets Loading package Example Installation On cluster using Singularity In Julia add https://gitlab.moffitt.usf.edu:8000/Bios2Projects/NIPALS_PCA as unregistered package Tutorial using Pkg Pkg.add("https://gitlab.moffitt.usf.edu:8000/Bios2Projects/NIPALS_PCA") Running Julia REPL Julia can be started using 1. The base installation 2. From Singularity container (in progress, details pending) 3. Utilizing the downloaded folder as a local environment. To activate NIPALS_PCA as local environment cd path/to/cloned/NIPALS_PCA julia --project=. Loading package using NIPALS_PCA Example (base) ~/temp git clone https://gitlab.moffitt.usf.edu:8000/Bios2Projects/NIPALS_PCA Cloning into 'NIPALS_PCA'... Username for 'https://gitlab.moffitt.usf.edu:8000': petterhf Password for 'https://petterhf@gitlab.moffitt.usf.edu:8000': warning: redirecting to https://gitlab.moffitt.usf.edu:8000/Bios2Projects/NIPALS_PCA.git/ remote: Enumerating objects: 164, done. remote: Counting objects: 100% (164/164), done. remote: Compressing objects: 100% (102/102), done. remote: Total 164 (delta 86), reused 106 (delta 54), pack-reused 0 Receiving objects: 100% (164/164), 97.56 KiB | 1.99 MiB/s, done. Resolving deltas: 100% (86/86), done. (base) ~/temp cd NIPALS_PCA →(base) Documentation: https://docs.julialang.org Type "?" for help, "]?" for Pkg help. Version 1.5.3 (2020-11-09) Official https://julialang.org/ release julia> using NIPALS_PCA
[Info: Precompiling NIPALS_PCA [6d34e894-ae79-4b37-8bb1-324840f02c67] julia> loadIrisData() 100×9 DataFrame Row | Column1 Column2 Column3 Column4 Column5 Column6 Column7 Column8 Column9 5.1 0.222222 3.5 0.625 1.4 0.0677966 0.2 0.0416667 setosa On cluster using Singularity A prebundled singularity container for Julia can be accessed at /share/data2/applications/singularity_images/bbsrTools.sif -C, is required to utilize packages installed in the container. If this flag is not used, Julia will load packages from the users home directory -B, binds folders to be accessible to the container module load singularity/3.10 sudo singularity run -C -B ../data:/data bbsrTools.sif **Tutorial PCA** modelling From Julia REPL 1)Load package using NIPALS_PCA 2)Load dataset from .csv file to DataFrame x_df = loadIrisData() 3)Create dataset and apply normalize to mean center data xdataset = parseDataFrame(x_df) |> normalize 4)Calculate PCA model pca = calcPCA(xdataset, 3) 5)Calculate variances for model calcVariances(xdataset,pca) **PLS** normalization The PLS normalization workflow can either be run from a script or from an interactive Julia session. The default script can be find in src/scripts/plsnorm.jl Run from script julia src/scripts/plsnorm.jl \ --xfile /path/to/xmatrix.txt \ --yfile /path/to/ymatrix.txt \ --ycategorical "colname" \ --ycontinous "colname1;colname2" \ --mode calibrate \ --modelfile model.jld2 \ --outfile output_file.csv Run from interactive session using NIPALS_PCA parsed_args=Dict{String, Any}("xfile" => "/path/to/xmatrix.txt", "ycategorical" => "colname", "yfi #to calibrate calibrate_model(parsed_args) #to correct correct(parsed_args) Run on cluster using Singularity image The folder(s) used for reading and writing needs a Bind. In the example below \begin{center} b>-B ../data/:/data \begin{center} b>- data will be accessible from the root level within the container. Multiple folders can be bound. module load singularity/3.10 sudo singularity run --app plsnorm -C -B ../data/:/data bbsrTools.sif \ --xfile /data/xmatrix.txt \ --yfile /data/ymatrix.txt \ --ycategorical "colname" \ --ycontinous "colname1;colname2" \ --mode calibrate \ --modelfile model.jld2 \ --outfile output_file.csv Get help julia plsnorm.jl --help **Structures** $NIPALS_PCA.Dataset - Type$ struct Dataset • X::Array{Union{Missing, Float64},2} means::Array{Float64,1} • stdevs::Array{Float64,1} • value_columns::Array{String,1} • xmask::BitArray{2} • mv::Bool NIPALS_PCA.PCA — Type struct PCA <: NIPALS_PCA.MultivariateModel</pre> • T::DataFrames.DataFrame • P::DataFrames.DataFrame NIPALS_PCA.PLS — Type struct PLS <: NIPALS_PCA.MultivariateModel</pre> • T::DataFrames.DataFrame • P::DataFrames.DataFrame • C::DataFrames.DataFrame • W::DataFrames.DataFrame • U::DataFrames.DataFrame **Functions General functionality** NIPALS_PCA.calcPCA — Function calcPCA(dataset::Dataset, comps::Int64) Calculates a PCA model **Examples** julia> calcPCA(datset,3) NIPALS_PCA.calcPLS — Function calcPLS(xdataset::Dataset,ydataset::Dataset,comps::Int64,incsamples::Array{Int64,1} = collect(1:size(xdataset.X)[1])) Calculates a PLS model NIPALS_PCA.calcVariances — Function calcVariances(dataset::Dataset, model::PCA)::NamedTuple Calculates r2x, r2x_cum and eigenvalues for all components in PCA model **Examples** julia> r2x,r2x_cum,eigenvalues = calcPCA(datset,model) NIPALS_PCA.loadmodel — Function loadmodel(path::String)::Tuple{MultivariateModel,Array{Float64,1},Array{Float64,1}} Load PCA or PLS model from JLD2 file into a tuple containing the model, variable standard d NIPALS_PCA.savemodel — Function savemodel(model::T, dataset::Dataset, name::String) where T <: MultivariateModel Save PCA or PLS model as JLD2 file **PLS** normalization NIPALS_PCA.calibrate_model — Function calibrate_model(x::DataFrame,y::DataFrame,A::Int64, modelfile::String) Calibrates PLS model based on datatypes in DataFrame for y Columns of type CategoricalArray is handled by one-hot precedure The calibrated model is saved to specified locations NIPALS_PCA.correct — Function correct(model::T, dataset::Dataset, name::String) where T <: MultivariateModel Save PCA or PLS model as JLD2 file NIPALS_PCA.predict_xres — Function predict_xres(modelfile::String,xfile::String, outfile::String) Loads model from jld2 file, predicts using xfile and exports residual matrix into .csv file Powered by Documenter.jl and the Julia Programming Language.