

## NIPALS\_PCA

A package for calculating PCA and PLS using the NIPALS implementation. Both models handles missing values

The package contains data structures for models and datasets

### Installation

In Julia add `https://gitlab.moffitt.usf.edu:8000/Bios2Projects/NIPALS_PCA` as [unregistered package](#)

### Loading package

```
using NIPALS_PCA
```

### Tutorial

#### PCA

1. Load package

```
using NIPALS_PCA
```

1. Load dataset from .csv file to DataFrame

```
x_df = loadIrisData()
```

1. Create dataset and apply normalize to mean center data

```
xdataset = parseDataFrame(x_df) |> normalize
```

1. Calculate PCA model

```
pca = calcPCA(xdataset, 3)
```

1. Calculate variances for model

```
calcVariances(xdataset,pca)
```

### PLS modelling

#### 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
```

- `T::DataFrames.DataFrame`
- `P::DataFrames.DataFrame`

[NIPALS\\_PCA.PLS](#) — Type

```
struct PLS <: NIPALS_PCA.MultivariateModel
```

- `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; normalize::Bool = false)

Calculates a PCA model

##### Examples

```
julia> calcPCA(datset,3,normalize=true)
```

[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.correct](#) — Function

correct(model::T, dataset::Dataset, name::String) where T <: MultivariateModel

Save PCA or PLS model as JLD2 file

[NIPALS\\_PCA.calibrate\\_model](#) — Function

calibrate\_model(x::DataFramey::DataFrame,A::Int64, modelfile::String)

Calibrates PLS model based on datatypes in DataFrame for y

Columns of type CategoricalArray is handled by one-hot precedence

The calibrated model is saved to specified locations

[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