## Reproducible Research for OMNeT++ Based on Python and Pweave

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#### Outline

- Reproducible Research
- Python and Pweave
- Reproducible Research for OMNeT++
- Example: OMNeT++ FIFO Simulation

## Reproducible Research

#### Reproducible Research

- Reproducible research is a key to any scientific method and ensures repeating an experiment and the results of its analysis in any place with any person.
- A study can be truly reproducible when it satisfies at least the following three criteria:
  - All experimental methods are fully reported.
  - All data and files used for the analysis are (publicly) available.
  - The process of analyzing raw data is well reported and preserved.
- Reproducible research is to ensure
  - Same data + Same script = Same results

#### Why Do We Need Reproducible Research: Two Examples

- LIGO Gravitational Wave Detection
- Schön scandal Molecular Computing

#### LIGO - Gravitational Wave Detection

- The <u>Laser Interferometer</u>
   <u>Gravitational-Wave Observatory</u>
   (<u>LIGO</u>) is a large-scale physics
   experiment and observatory to
   detect cosmic gravitational
   waves.
  - The detection of gravitational wave was reported in *Physical Review Letters* in Feb. 2016, together with ipython notebook with analysis code and data.



```
In [9]: # We need to suppress the high frequencies with some bandpassing:
bb, ab = butter(4, [20.*2./fs], 300.*2./fs], btype='band')
strain_HI_whitenbp = filtrilt(bb, ab, strain_HI_whiten)
strain_LI_whitenbp = filtrilt(bb, ab, strain_LI_whiten)
NR_HI_whitenbp = filtrilt(bb, ab, NR_HI_whiten)

# plot the data after whitening:
# first, shift LI by 7 ms, and invert. See the GWI50914 detection paper
strain_LI_shift = -mp.roll(strain_LI_whitenbp,int(0.007*fs))

plt.figure()
plt.plot(time-tevent,strain_LI_whitenbp,'r',label='HI strain')
plt.plot(time-tevent,strain_LI_shift,'g',label='HI strain')
plt.plot(time-tevent,strain_LI_whitenbp,'r',label='HI strain')
plt.plot(Whitme+0.002,NR_HI_whitenbp,'r',label='matched NR waveform')
plt.slim([-0.1,0.05])
plt.ylim([-4,41])
plt.slabel('time (s) since '*str(tevent))
plt.slabel('whitented strain')
plt.legend(loc-'lover left')
plt.savefig('GWI50914_strain_whitened_png')

Advanced LIGO WHITENED strain data near GWI50914')
plt.savefig('GWI50914_strain_whitened_png')

Advanced LIGO WHITENED strain data near GWI50914

### Advanced LIGO WHITENED strain data near GWI50914
```

#### Schön Scandal - Molecular Computing

- No records found for his groundbreaking experimental results, including lab notebook, experimental samples and data, hard disk drives.
- During the investigation, he kept repeating

"I clearly observed them in the Lab but ..."



**Bell Labs launches inquiry into** 

Python and Pweave

#### R/Sweave to Python/Pweave

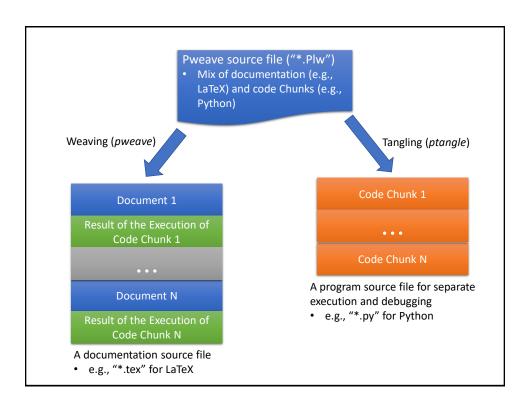
- Until recently, R was the language of choice for statistical processing and data analysis.
  - Still, R has the largest code base for a wide variety of statistical and graphical techniques.
- Like *ipython* (now *jupyter*), R provides a nice tool called *Sweave* (now replaced by *knitr*) to weave documentation and the results of the execution of R code chunks into one source file for integrated documentation.
- Python one of the most popular languages in scientific computing, including artificial intelligence & machine learning — recently takes over R in statistical processing and data analysis as well.
  - Thanks to <u>pandas</u> implementing DataFrame object similar to R and <u>Pweave</u>, python can replace R for most statistical and data analysis tasks, while retaining its many advantages over R (i.e., fully-featured programming language with easy syntax and higher speed).

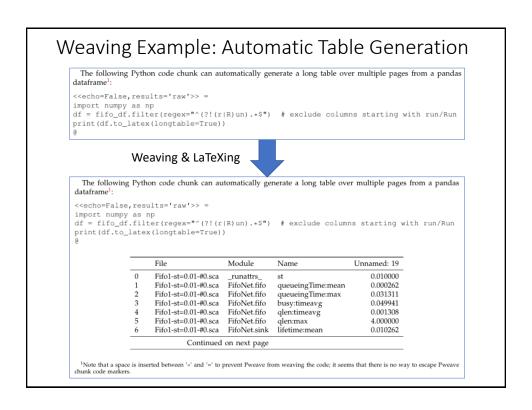
```
### customize
.old <- theme_set(theme_bw())
.pt_size <- 3.5

### generate summary plots for reference architecture with N=1
.rf_Nl.data <- paste(.rf_Nl.wd, paste(.rf_Nl.base, "data", sep="."), sep="/")
.df <- read.csv(.rf_Nl.data, header=TRUE)

## .df <- .df[order(.df$N], .df$n, .df$dr, .df$fsp, .df$repetition), ] # order data frame
.df <- sort df(.df, vars=c("N", "n", "dr', "br", "repetition")) # sort data frame
.rf_Nl.df <- ddply(.df, c(.(n), .(dr)), function(df) {return(GetMeansAndCiWidths(df))})
.rf_Nl.plots <- list()
for (.i in 1:7) {
    .df <- subset(.rf_Nl.df, select = c(1, 2, (.i*2+1):((.i+1)*2)))
    names(.df)[3:4] <- c("mean", "ci.width")
    .limits <- aes(ymin = mean - ci.width, ymax = mean +ci.width)
    .p <- ggplot(data=.df, aes(group=dr, colour=factor(dr), x=n, y=mean)) + geom_line() + scal
    *.width)))
    .p <- .p + xlab("Number of Users per ONU (n)") + ylab(.labels.measure[.i])
    ## .p <- .p + geom_point(aes(group=dr, colour=factor(dr), x=n, y=mean), size=.pt_size)
    .p <- .p + geom_point(aes(group=dr, colour=factor(dr), x=n, y=mean), size=.pt_size) + scale
    .p <- .p + geom_point(aes(group=dr, colour=factor(dr), x=n, y=mean), size=.pt_size) + scale
    .p <- .p + geom_point(aes(group=dr, colour=factor(dr), x=n, y=mean), size=.pt_size) + scale
    .p <- .p + geom_point(aes(group=dr, colour=factor(dr), x=n, y=mean), size=.pt_size) + scale
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    .p <- .p + geom_point(aes(group=dr, colour=factor(dr), x=n, y=mean), size=.pt_size) + scale
    .p <- .p + geom_point(aes(group=dr, colour=factor(dr), x=n, y=mean), size=.pt_size) + scale
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    .p <- .p + geom_point(aes(group=dr, colour=factor(dr), x=n, y=mean), size=.pt_size) + scale
    .p <- .p + geom_point(aes(group=dr, colour=factor(dr), x=n, y=mean), size=.pt
```

## Snippets of R Source Code and Sweave File for LaTeX





# Reproducible Research for OMNeT++

#### How to Deal with Simulation Input Files

- Include them the document.
  - OK for small simulations
- Use a snapshot of the whole configurations.
  - e.g., git commit hashes

```
Commit 8763336792727184467271854467208180675601

Author: Kyeong Soo (Joseph) Kim (kyeongsoo.kim@gmail.com)
Oate: Mon Feb 27 08:59:31 2017 +00006

Add ini file.

Commit 816 st distributed Witmort ANY MARMANT. See the file
// Illennee for details on this and other legal matters.

// Einple quessing natwork; generator * FITO * sink.

Date: Mon Feb 27 08:59:30 (Joseph) Kim (kyeongsoo.kim@gmail.com)
Date: Mon Feb 27 08:55:07 2017 +0000

Remove README.rst; only the markdown version of README

submodulen

gen Source (s
parameters)
// FITO * FITO *
parameters)
// Simple quessing parameters
// Simple quessing natwork; generator * FITO * sink.

parameters)
// Simple quessing natwork; generator * FITO * sink.

Commit $76533679e275543fea8c4f3730cf894da774c88e

Author: Kyeong Soo (Joseph) Kim (kyeongsoo.kim@gmail.com)
Date: Sun Oct 2 17:32:02 2016 +0000

Change simulation time.

Listing 1: 'FifoNet.ned' for FIFO sample model.
```

# How to Guarantee Match Between Input Files and Output Data

- Online generation of results
  - · Include simulation execution code within a document
    - Refer to the provided sample Pweave file.
  - OK for smaller simulations, but not for larger simulations.
- Use a snapshot of the whole configurations and data
  - · e.g., git commit hashes
  - Version controlling output data together with source code and input configuration files, however, may greatly increase the size of a repository.

#### How to Present and Analyze Output Data

- Unstacking of stacked DataFrame
  - Use *pivot* function (see the example shown here).
- Aggregated processing of measurement data over independent variables
  - Use *pivot\_table* function.
  - Useful for the calculation of mean and confidence intervals over multiple iterations.
- Online calculation of confidence intervals
  - Confidence intervals (CIs) can be calculated by assigning a custom function for CI to aggfunc parameter of pivot\_table function.
  - Now pandas support error bars in its own plot functions.

```
In [1]: df
Out[1]:
        date variable
                         value
               A 0.469112
0 2000-01-03
  2000-01-04
                   A -0.282863
  2000-01-05
                   A -1.509059
3 2000-01-03
                  B -1.135632
4 2000-01-04
                   B 1,212112
5 2000-01-05
                   B -0.173215
6 2000-01-03
                   C 0.119209
   2000-01-04
                   C -1.044236
  2000-01-05
                   C -0.861849
  2000-01-03
                   D -2.104569
10 2000-01-04
                   D -0.494929
11 2000-01-05
                   D 1.071804
```

```
In [3]: df.pivot(index='date', columns='variable', values='value')
Out[3]:
variable A B C D
date
2000-01-03 0.469112 -1.135632 0.119209 -2.104569
2000-01-04 -0.282863 1.212112 -1.044236 -0.494929
2000-01-05 -1.509059 -0.173215 -0.861849 1.071804
```

Demo: OMNeT++ FIFO Simulation