AKIN KAZAKÇI

MdC / Mines ParisTech
CGS

ALEXANDRE GRAMFORT

MdC / Telecom ParisTech LTCI

BALÁZS KÉGL

DR / CNRS - UPSaclay
LAL & LRI









5 Paris-Saclay Center for Data Science

OUTLINE

- Paris-Saclay Center for Data Science
 - the data science ecosystem
- Analytics tools
 - data challenges
 - rapid analytics and model prototyping

DATA SCIENCE

Design of automated methods

to analyze massive and complex data

to extract useful information







DATA SCIENCE



BIG DATA

We are focusing on inference:

data → knowledge

Interfacing with infrastructure, security, production





Université Paris-Saclay

19 founding partners





































Université Paris-Saclay

19 fondateurs

60 000 étudiants

6 000 doctorants

15 000 étudiants en master

8 Schools

11 000 chercheurs et enseignants-chercheurs

300 laboratoires

8 000 publications /an

15 % de la recherche publique française

10 départements

+ horizontal multi-disciplinary and multi-partner initiatives ("lidex") to create cohesion











A multi-disciplinary initiative to define, structure, and manage the data science ecosystem at the Université Paris-Saclay

http://www.datascience-paris-saclay.fr/

250 researchers in 35 laboratories

Biology & bioinformatics

IBISC/ŪĒvry LRI/UPSud Hepatinov CESP/UPSud-UVSQ-Inserm IGM-I2BC/UPSud MIA/Agro MIAj-MIG/INRA LMAS/Centrale

Chemistry

EA4041/UPSud

Earth sciences

LATMOS/UVSQ GEOPS/UPSud IPSL/UVSQ LSCE/UVSQ LMD/Polytechnique

Economy

LM/ENSAE RITM/UPSud LFA/ENSAE

Neuroscience

UNICOG/Inserm U1000/Inserm NeuroSpin/CEA

Particle physics astrophysics & cosmology

LPP/Polytechnique DMPH/ONERA CosmoStat/CEA IAS/UPSud AIM/CEA LAL/UPSud

Machine learning LRI/UPSud

LTCI/Telecom CMLA/Cachan LS/ENSAE LIX/Polytechnique MIA/Agro CMA/Polytechnique LSS/Supélec CVN/Centrale LMAS/Centrale DTIM/ONERA IBISC/UEvry LIST/CEA

Visualization

INRIA LIMSI

Signal processing

LTCI/Telecom CMA/Polytechnique CVN/Centrale LSS/Supélec CMLA/Cachan LIMSI DTIM/ONERA

Statistics

LMO/UPSud LS/ENSAE LSS/Supélec CMA/Polytechnique LMAS/Centrale MIA/AgroParisTech

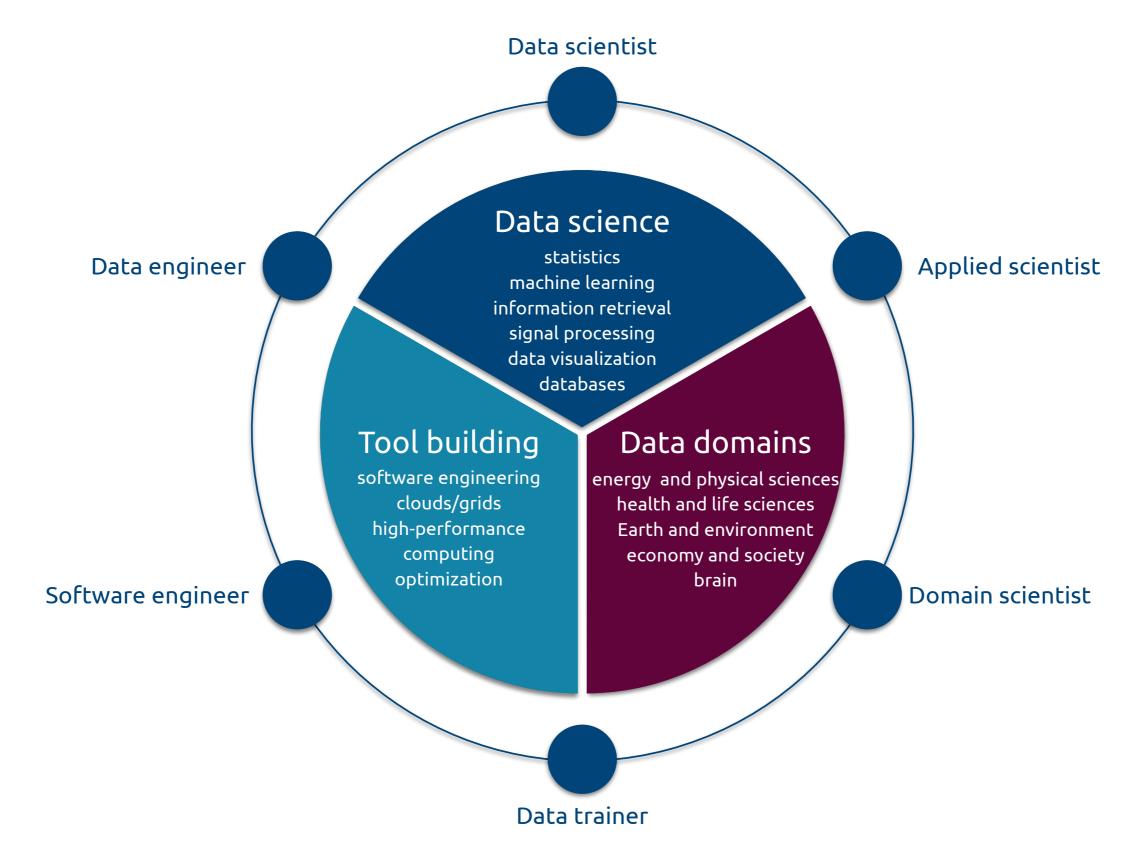








THE DATA SCIENCE ECOSYSTEM











TOOLS

We are designing and learning to manage tools

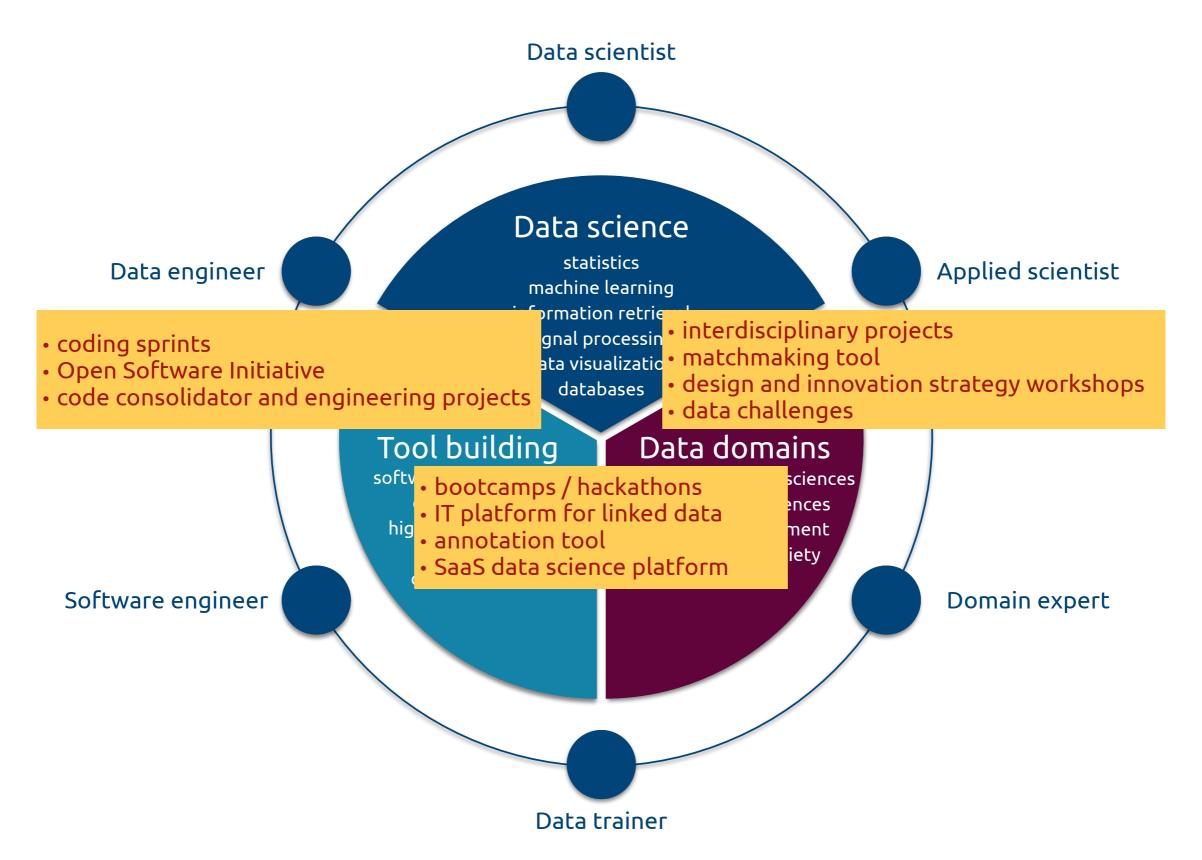
to accompany data science projects

with different needs





TOOLS: LANDSCAPE TO ECOSYSTEM











Two analytics tools

DATA CHALLENGES

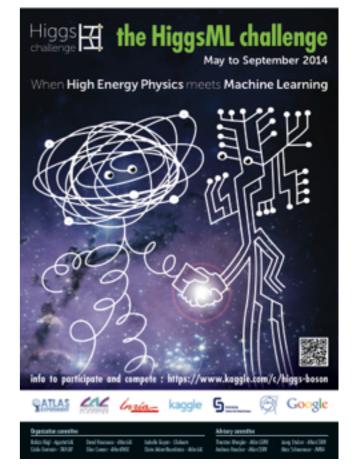






DATA CHALLENGES

- A data challenge is a recently developed unconventional dissemination and communication tool
 - a scientific or industrial data producer arrives with a well-defined problem and a corresponding annotated data set
 - defines a quantitative goal
 - makes the problem and part of the data set (the training set) public on a dedicated site
 - data science experts then take the public training data and submit solutions (predictions) for a test set with hidden annotations
 - submissions are evaluated numerically using the quantitative measure
 - contestants are listed on a leaderboard
 - after a predefined time, typically a couple of months, the final results are revealed and the winners are awarded





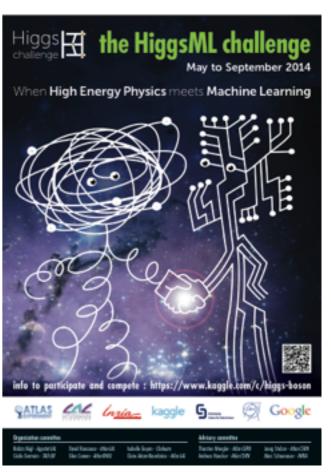




DATA CHALLENGES



- The HiggsML challenge on Kaggle
 - https://www.kaggle.com/c/higgs-boson







HUGE PUBLICITY

kaggle Customer Solutions Competitions Community ▼ Sign up Login



Completed • \$13,000 1,785 teams
Higgs Boson Machine Learning Challenge

Mon 12 May 2014 - Mon 15 Sep 2014 (21 days ago)

Dashboard

Private Leaderboard - Higgs Boson Machine Learning Challenge

This competition has completed. This leaderboard reflects the final standings.

See someone using multiple accounts? Let us know.

#	Δ1w	Team Name # model uploaded * in the money	Score @	Entries	Last Submission UTC (Best – Last Submission)
1	↑4	Gábor Melis ‡ *	3.80581	110	Sun, 14 Sep 2014 09:10:04 (-0h)
2	Į1	Tim Salimans ‡ *	3.78913	57	Mon, 15 Sep 2014 23:49:02 (-40.6d)
3	_	nhlx5haze ‡ *	3.78682	254	Mon, 15 Sep 2014 16:50:01 (-76.3d)







SIGNIFICANT IMPROVEMENT OVER THE BASELINE

#	Δ1w	Team Name ‡model uploaded * in the money		Entries	Last Submission UTC (Best - Last Submission)
1	↑4	Gábor Melis ‡ *	3.80581	1 0	Sun, 14 Sep 2014 09:10:04 (-0h)
2	‡1	Tim Salimans ‡ *		57	Mon, 15 Sep 2014 23:49:02 (-40.6d)
3	_	nhlx5haze ‡ *	3.78682	254	Mon, 15 Sep 2014 16:50:01 (-76.3d)
4	↑55	ChoKo Team 🕮	3.77526	216	Mon, 15 Sep 2014 15:21:36 (-42.1h)
5	↑23	cheng chen	3.77384	21	Mon, 15 Sep 2014 23:29:29 (-0h)
6	↓2	quantify	3.77086	8	Mon, 15 Sep 2014 16:12:48 (-7.3h)
7	↑73	Stanislav Semenov & Co (HSE Yandex)	3.76211	68	Mon, 15 Sep 2014 20:19:03
8	Į1	Luboš Motl's team 🎩	3.76050	589	Mon, 15 Sep 2014 08:38:49 (-1.6h)
9	Į1	Roberto-UCIIIM	3.75864	292	Mon, 15 Sep 2014 23:44:42 (-44d)
10	↑5	Davut & Josef 🎩	3.75838	161	Mon, 15 Sep 2014 23:24:32 (-4.5d)
990	165	sandy	3.20546	5	Fri, 29 Aug 2014 18:14:30 (-0.7h)
991	‡65	Rem.	200102	2	Mon, 16 Jun 2014 21:53:43 (-30.4h)
		simple TMVA boosted trees	3.19956		
992	165	Xiaohu SUN	51	3	Tue, 03 Jun 2014 13:14:47
993	165	Pierre Boutaud	3.19956	10	Fri, 25 Jul 2014 15:25:07 (-30d)









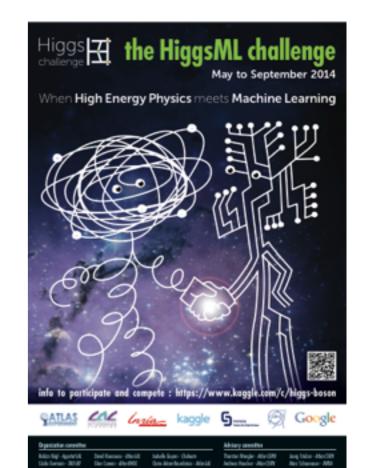
HUGE PUBLICITY

SIGNIFICANT IMPROVEMENT OVER THE BASELINE

yet partially missing the objectives

DATA CHALLENGES

- Challenges are useful for
 - generating visibility in the data science community about novel application domains
 - benchmarking in a fair way state-of-the-art techniques on well-defined problems
 - finding talented data scientists
- Limitations
 - not necessary adapted to solving complex and open-ended data science problems in realistic environments
 - no direct access to solutions and data scientist
 - emphasizes competition











We decided to design something better

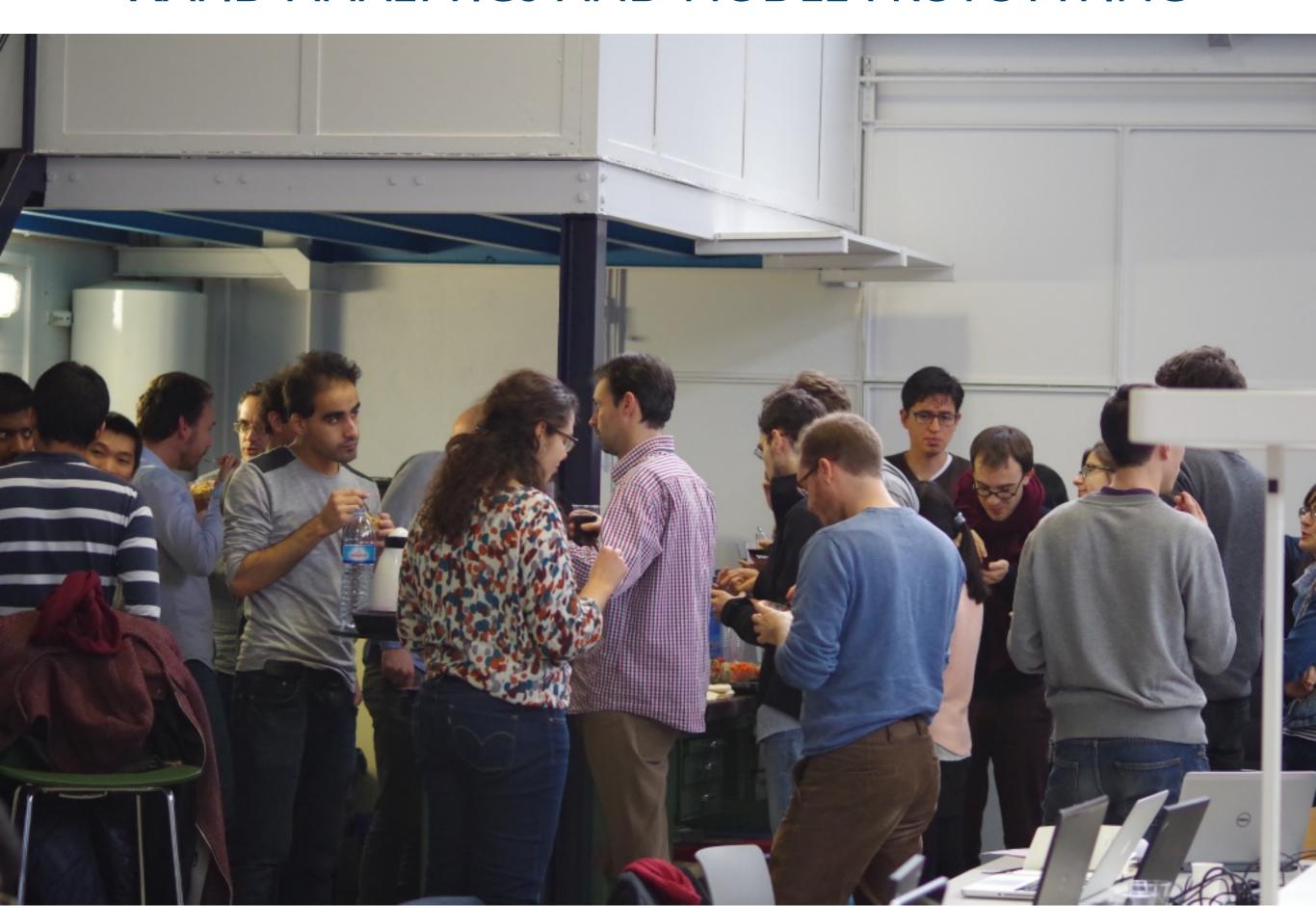
- Single-day coding sessions
 - 20-30 participants
 - preparation is similar to challenges
- Goals
 - focusing and motivating top talents
 - promoting collaboration, speed, and efficiency
 - solving (prototyping) real problems











ANALYTICS TOOLS TO PROMOTE COLLABORATION AND INNOVATION







onevm-222.lal.in2p3.fr:8080/leaderboard















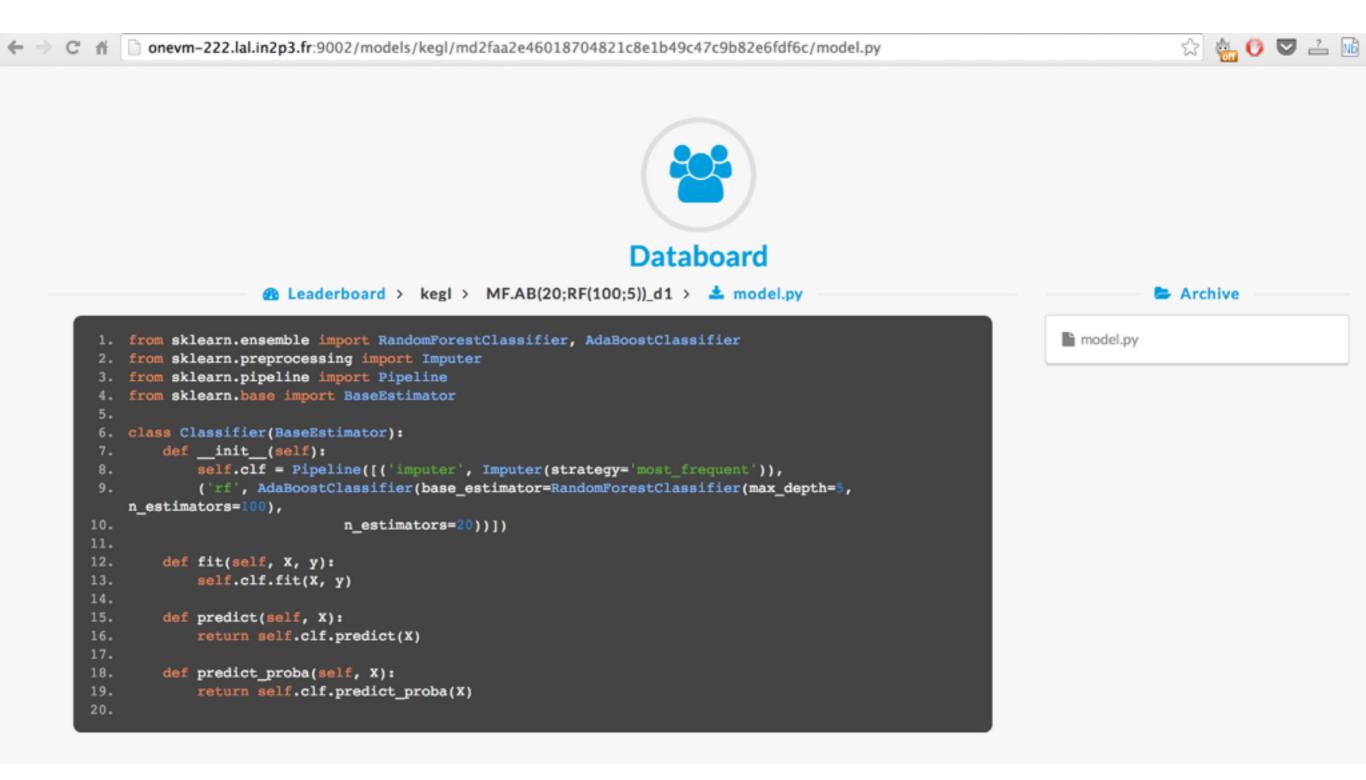
Best models

	team	model @	score
1	schwarty	AB_RF(d=3;m=300);scaler_feat_sel	0.805402
2	fauteuils	AB_RF(d=6,m=50),GridSearch	0.803887
3	fauteuils	AB_RF(d=3;m=100)	0.803705
4	Jiali_Lagree_Gregory	ADB150RF40	0.801960
5	kegl	MF.AB(20;RF(100;5))	0.798978
6	schwarty	40_percent_features_and_logistic	0.798509
7	schwarty	test	0.798509
8	schwarty	logistic	0.798509
9	schwarty	adaboot_rf_scaler_and_feature_selection	0.797638
10	kegl	R(-1).GB(1000;5;20)	0.797289
11	agramfort	logreg-on-rf-features	0.796961
12	kegl	R(-1).AB(20;RF(100;5))	0.796636
13	fauteuils	A(n=50,lr=1.5)_RF(n=50,md=6,bstp=False)	0.793990

Most contributive models

	team	model @	score
1	schwarty	AB_RF(d=3;m=300);scaler_feat_sel	32
2	schwarty	test	21
3	DelfJose	gradient_boosting	15
4	fauteuils	AB_RF(d=6,m=50),GridSearch	13
5	agramfort	logreg-on-rf-features	12
6	Jiali_Lagree_Gregory	NuSVC2	9
7	kegl	R(-1).AB(1000;5;20)	8
8	Jiali_Lagree_Gregory	extraTrees1000	8
9	Jiali_Lagree_Gregory	ADB150RF40	8
10	kegl	R(-1).GB(1000;5;20)	7
11	kegl	MF.AB(20;RF(100;5))	7
12	kegl	R(-1).AB(20;RF(100;5))	7
13	Voilavoila	randomfor_nest_16	7

ANALYTICS TOOLS TO PROMOTE COLLABORATION AND INNOVATION

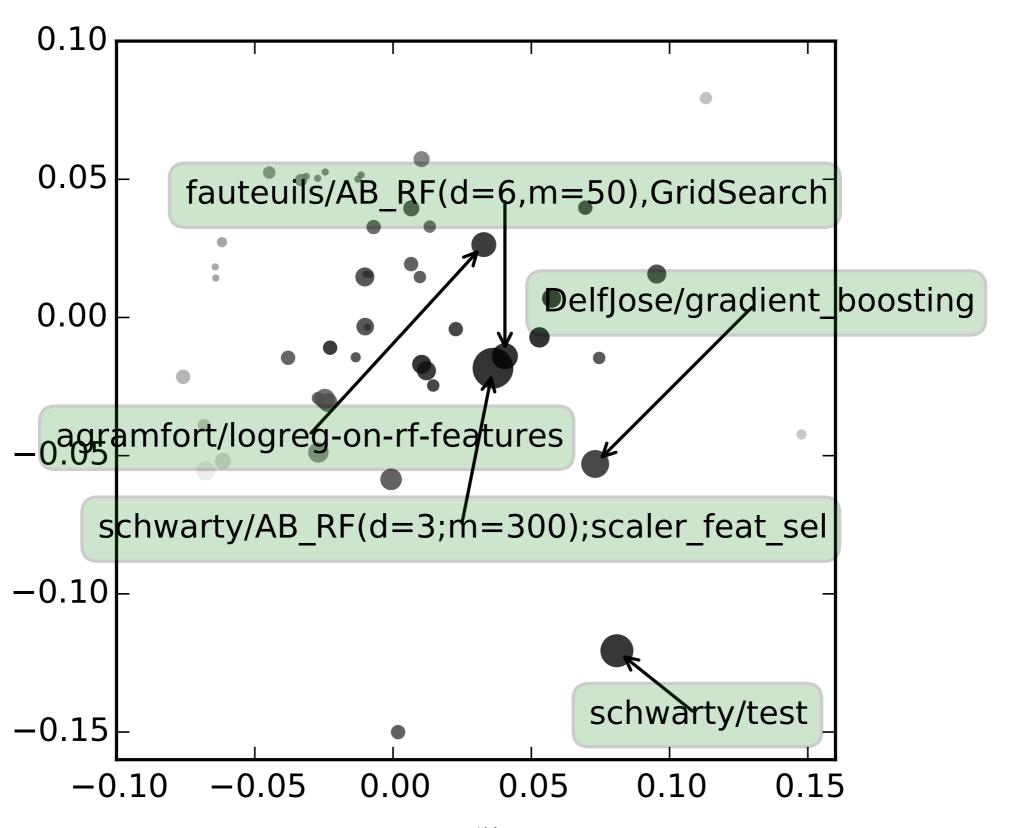








ANALYTICS TOOLS TO MONITOR PROGRESS



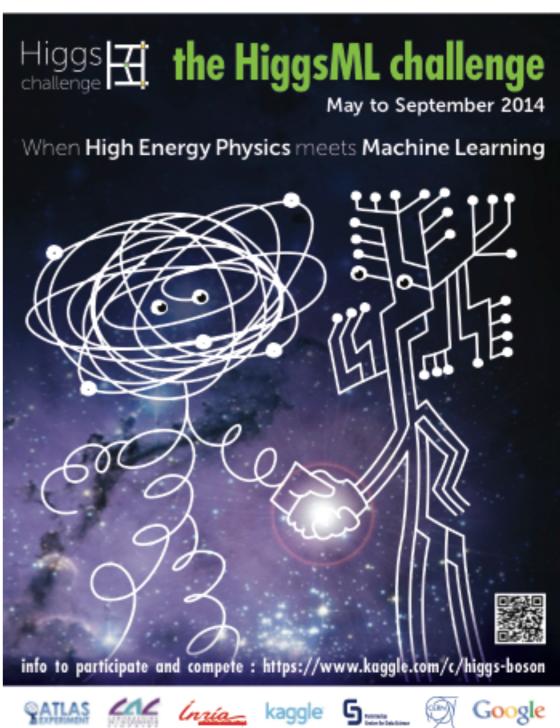
RESEARCH (BEYOND SOLVING PROBLEMS)

- Algorithm selection and hyperparameter optimization
 - studying human problem-solving
 - combining human solutions with automatic tools
 - comparing and tuning hyperparameter optimizers
 - meta-learning: embedding data sets and models, collaborative optimization





2015 Jan 15 replaying the HiggsML challenge







Organization committee





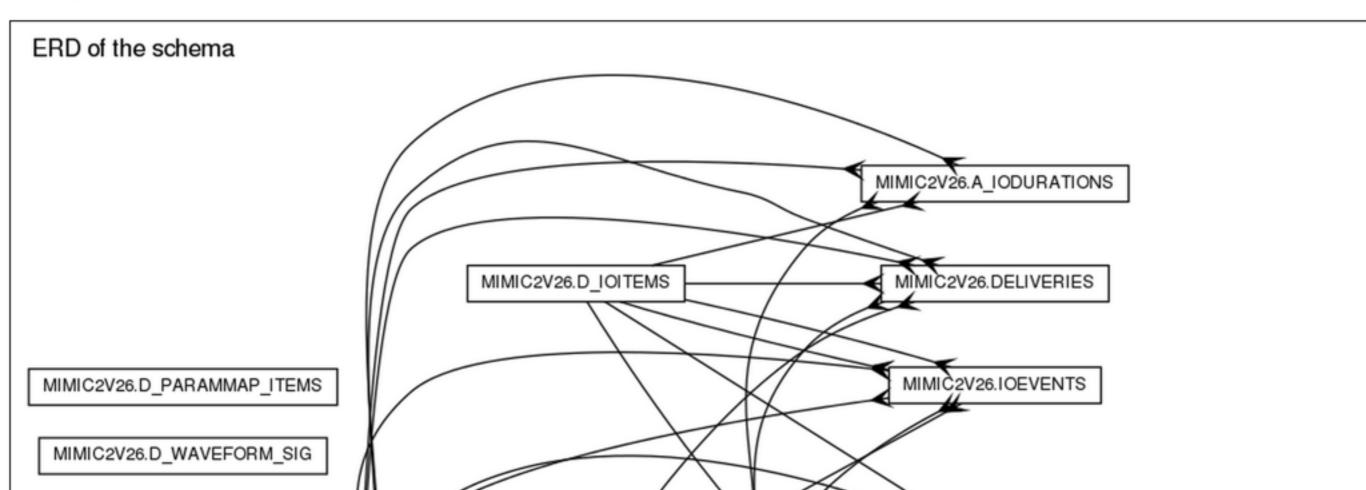


2015 Feb 9

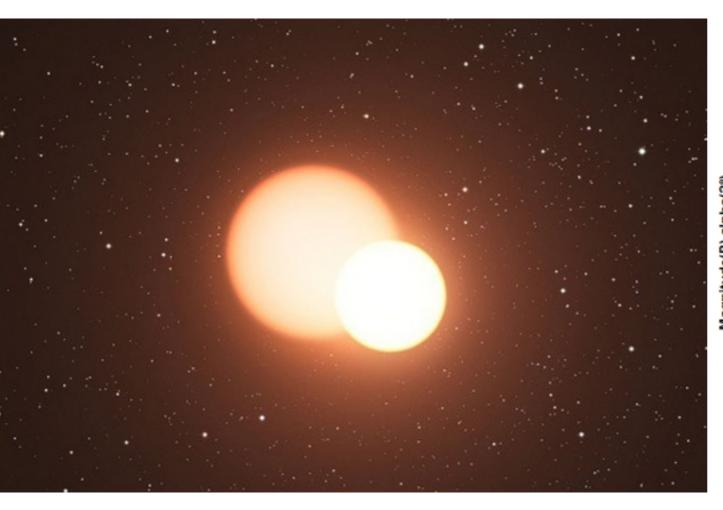
Mortality prediction in septic patients

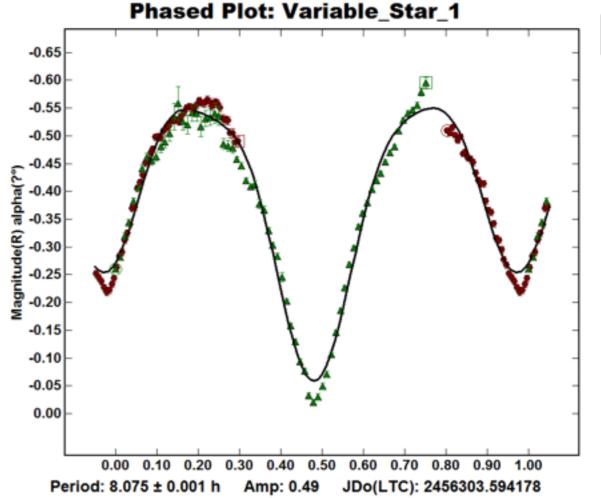
MIMIC II V2.6

Description:



2015 Apr 10 Classifying variable stars



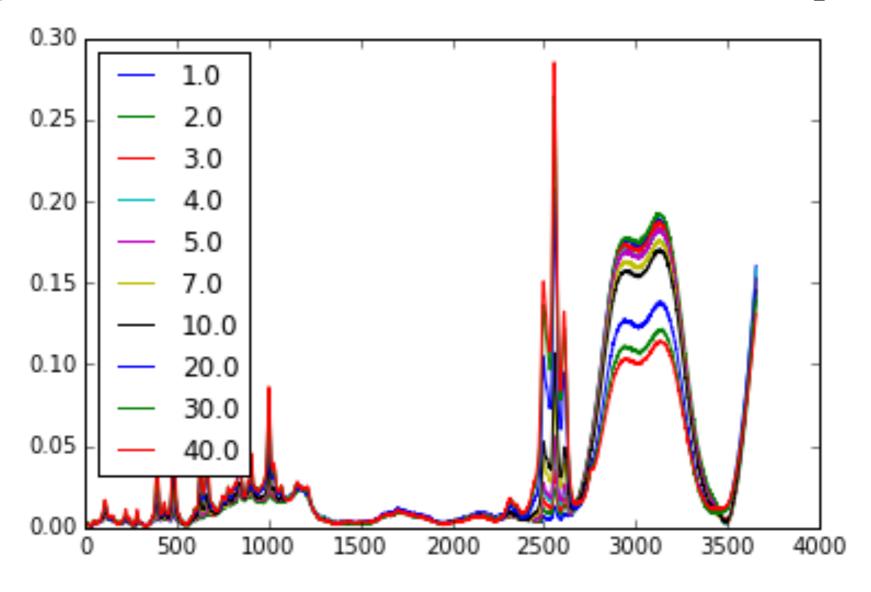






2015 May

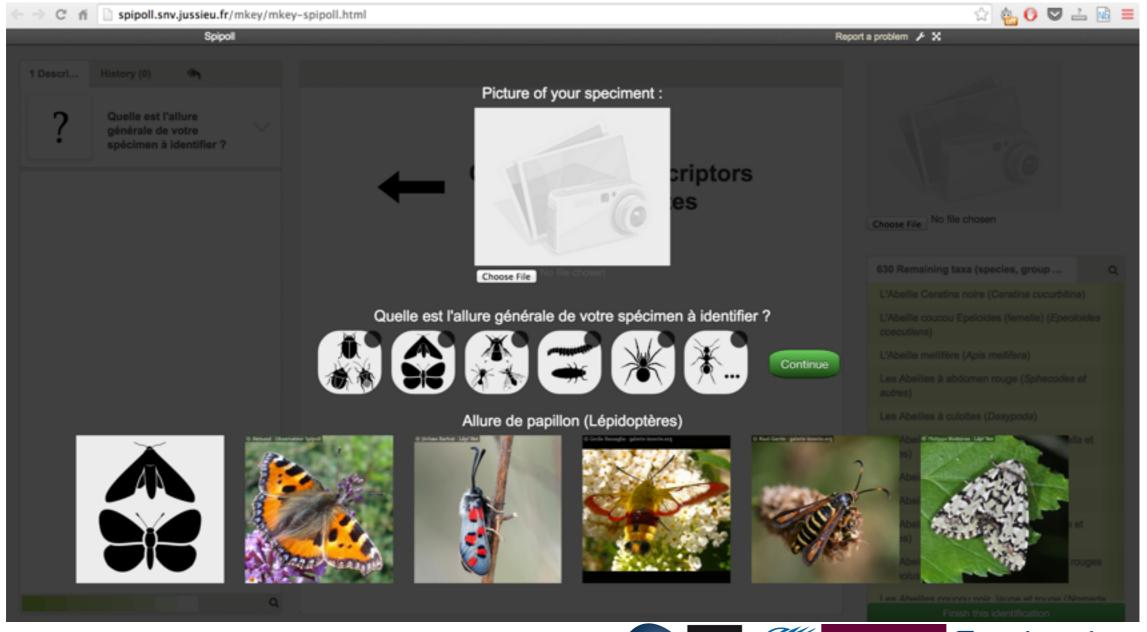
Drug identification from spectra







2015 June Insect classification









IMPLEMENTING RAMPS IN AN INDUSTRIAL CONTEXT

- Short-term, ad-hoc teams assembled for a given task
- Low-engagement consulting job
- Efficient use of scarce data scientists (i.e., your time)
- Developing and practicing marketable skills
 - fast-feedback experimentation is also useful in research
- Networking
- Management meta-tools to track your performance, to guide your training







THANK YOU!