

# Genhack Al Challenge Kick-Off

Wednesday, October 20th

### Agenda

5pm: Welcome and presentation of the Genhack competition!

Presented by: Luca Zanna and Jean-Marc Marty, cofounders of Outcoder.ai.

• 5:15pm: Generative Models: A Scientific Approach

Presented by: Professor Eric Moulines, Professor at Ecole Polytechnique and Member of the Académie des Sciences.

- 5:30pm: A Quick Overview of AI at BNP Paribas
- Presented by: Léa Deleris, Head of RISK Artificial Intelligence Research at BNP Paribas
- 5:45pm: Questions & Answers
- 6pm: Speed Meetings Special Q&A
- 6:30pm: End of Kick-Off







# Welcome to the GenHack AI Competition!

### What is the Genhack Al competition?

- 295 participants! wo thirds in Europe, one third outside Europe
- 15 countries
- 66 universities 🏥
- Several guilds with participants from different schools and different continents









# **Guilds**







Kurama D Team





Invicta Quant

 $e^{i\pi}$ +1=0 1st Place







War Brother's





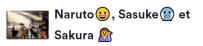




La Casa De Vinci











The Battle-Borns





Terran





























If you do not have a guild yet, use the **speed meetings at 6pm** to find guild members!









# Organisers





**Luca Zanna** Co-founder



Jean-Marc Marty
Co-founder





**Eric Moulines**Professor



**Emmanuel Gobet**Professor



Marine Saux
Project Manager



Michaël Allouche PhD



The bank for a changing world



**Léa Deleris** Head of RISK Al Research



Antoine Bezat
Head of Stress
Testing Methodologies
and Models



**Dorinel Bastide**Senior Quantitative
Analyst



Wilson Ramos
Quant/ Data Scientist



Jean-Philippe Lemor
Head of Systematic Strategies
and Hybrid Quantitative Research



Coach

# Competition **W**

#### **Context**

In the last two decades, the increasing number of shocks and financial crises has been a major issue for the financial risk management teams.

Among the wide range of exercises in this field, Stress tests have become a main guideline for the regulator in order to assess the banking system resilience against the realizations of various categories of risk (market, credit, operational, climate, etc). The main challenge is to simulate unfavorable extreme (but plausible) negative returns similar to a historical dataset.



#### The Task

This is an unsupervised learning problem: Given real data from stock market indexes that will act as a train dataset, the task is to learn a generative model that simulates synthetic stock market indexes.







# Competition Agenda

### Monday, October 25th: Start of the competition!

You will receive a training set. The training set is a matrix with 746 rows and 4 columns.

The rows are the log-return on a given day, while the columns are four stock market indices.

There is not a lot of data, so one of the challenges is related to size of data.

### First submission (not included in final ranking)

By Saturday, October 30th: You will submit your generated log-returns. You should generate 410 rows of data.

On Tuesday, November 2<sup>nd</sup>: Publication of the guild scoreboard, scored against the validation set.

### Second submission (1/3 of final ranking)

By Saturday, November 6th: You will submit your generated log-returns. You should generate 410 rows of data.

On Tuesday, November 9th: Publication of the guild scoreboard, scored against the validation set.

The validation set is shared with everybody

### Third submission (2/3 of final ranking)

**By Saturday, November 20th:** You will submit your generated log-returns. You should generate 408 rows of data. **On Thursday, November 25th:** Publication of the guild scoreboard, scored against the test set.





## How to submit the generated data?

You will submit a Zip file from the Outcoder website.

The name should include the Guild name and the date.



#### The zip file should contain:

- Data.csv => a .csv file with the generated data
- Noise.csv => the associated input noise used in the model to generate data
- Python code => the inference code for loading the model and for simulating the data
- CSV or binary => the model parameters











# Submissions Evaluation



### Guilds are judged based on two evaluation score metrics:

→ Marginals - Anderson-Darling The Anderson-Darling distance (see [1], [2] for more details) computes a weighted square difference between the hypothetical cumulative distribution function (c.d.f) F from which samples have been drawn, and the empirical c.d.f  $\hat{F}_n$  based on n observations:

$$W_n = n \int_{-\infty}^{\infty} \frac{\left(\widehat{F}_n(x) - F(x)\right)^2}{F(x)(1 - F(x))} dF(x).$$

For this challenge, let  $\widehat{F}_n^{\tau}$  be the empirical distribution function associated to  $X_1^{\tau},...,X_n^{\tau}$  and let  $X_{1,n}^{\tau} \leq \cdots \leq X_{n,n}^{\tau}$  be the order statistics for each financial ticker  $\tau = 1, \ldots, d$ . We denote  $\widetilde{u}_{i,n}^{\tau}$  the model probability of a generated variable  $\widetilde{X} = G(Z)$  for a specific ticker  $\tau$  such that

$$\widetilde{u}_{i,n}^{\tau} = \frac{1}{n+2} \left( \sum_{j=1}^{n} \mathbb{1} \left\{ X_j^{\tau} \leq \widetilde{X}_{i,n}^{\tau} \right\} + 1 \right).$$

The Anderson-Darling distance for each ticker  $\tau$  can be computed as

$$W_n^{\tau} = -n - \frac{1}{n} \sum_{i=1}^{n} (2i - 1) \left( \log(\widetilde{u}_{i,n}^{\tau}) + \log(1 - \widetilde{u}_{n-i+1,n}^{\tau}) \right). \tag{1}$$

Then, the global metric on the marginals is just the average distance for all tickers and is computed as

$$\mathcal{L}_M = \frac{1}{d} \sum_{\tau=1}^d W_n^{\tau}.$$

Ranking score For each evaluation metric, the ranking will be determined by

- 1. Computing a Z-score  $\mathcal{Z} = (x \mu)/\sigma$
- 2. Converting to points  $P = \Phi(\mathcal{Z})*1000$ , with  $\Phi$  the c.d.f. of a standard normal distribution

Dependence - Absolute Kendall error Kendall's dependence function (see [3] for more details) characterizes the dependence structure associated with a copula C and is the univariate cumulative distribution function defined by  $K_C(t) = \mathbb{P}\left(C\left(U^{(1)}, \dots, U^{(d)}\right) \leq t\right)$ for all  $t \in [0,1]$  and  $(U^{(1)}, \ldots, U^{(d)})$  a random vector with uniform margins on [0,1]. The estimation of the Kendall's dependence function is based on the pseudo-observations

$$Z_i = \frac{1}{n-1} \sum_{j \neq i}^n \mathbb{1} \left\{ X_j^1 < X_i^1, \dots, X_j^d < X_i^d \right\},$$

and we consider equivalently the ones from the model

$$\widetilde{Z}_i = \frac{1}{n-1} \sum_{j \neq i}^n \mathbb{1} \left\{ \widetilde{X}_j^1 < \widetilde{X}_i^1, \dots, \widetilde{X}_j^d < \widetilde{X}_i^d \right\}.$$

Then, the dependence metric that we define can be computed as a  $L^1$  norm

$$\mathcal{L}_D = \frac{1}{n} \sum_{i=1}^n \left| Z_{i,n} - \widetilde{Z}_{i,n} \right|,\,$$

where  $Z_{1,n} \leq \cdots \leq Z_{n,n}$  (resp.  $\widetilde{Z}_{1,n} \leq \cdots \leq Z_{n,n}$ ) are the order statistics associated with  $\{Z_1, \ldots, Z_n\}$  (resp.  $\{Z_1, \ldots, Z_n\}$ ).



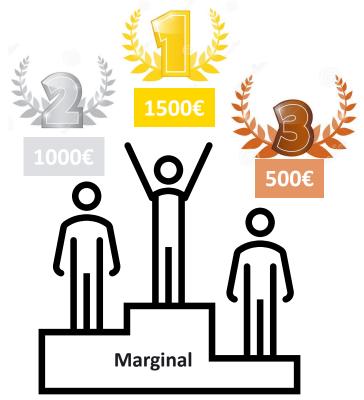


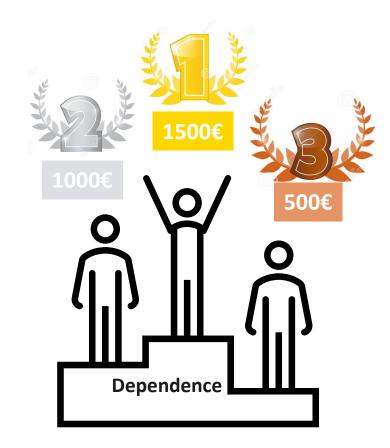


# Final Ranking and Prizes 🖳

We will have a scoreboard for the two metrics: Marginal and Dependence.

A guild can win the monetary prize for both rankings!





All winning guilds will be invited to a special session with BNP Paribas!









## Dedicated Q&A with first 5 guilds at 6pm



**FinDogz** 



Kurama D Team



Winx



**Terran** 







