



HSE

Radiation Protection

# Machine learning and transfer functions to characterise radioactive waste for particle accelerators

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# CERN

... an accelerator complex with 11 machines

During their operation, beam losses lead to the inevitable activation of material.

When machines are upgraded or dismantled, radioactive waste is generated.

# Radioactive waste

...generated at CERN is collected at a dedicated treatment center.

It is then split in categories and characterized ...

...in order to be eliminated to final repositories in France and Switzerland.



# What does it mean

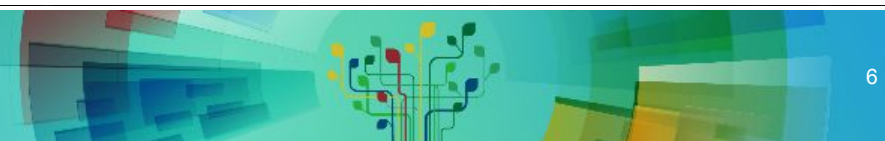
to **characterize** radioactive waste?

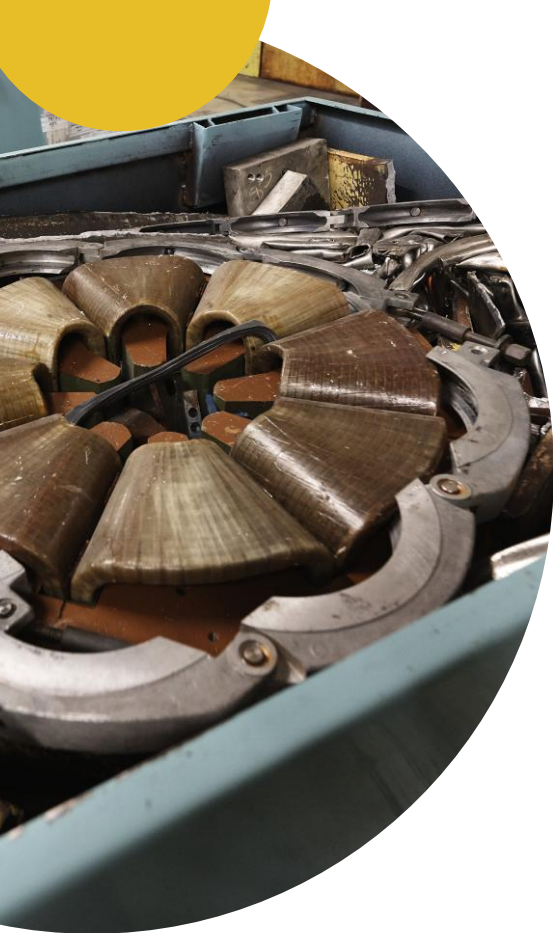


To confirm the **acceptability** of waste in the final repository...

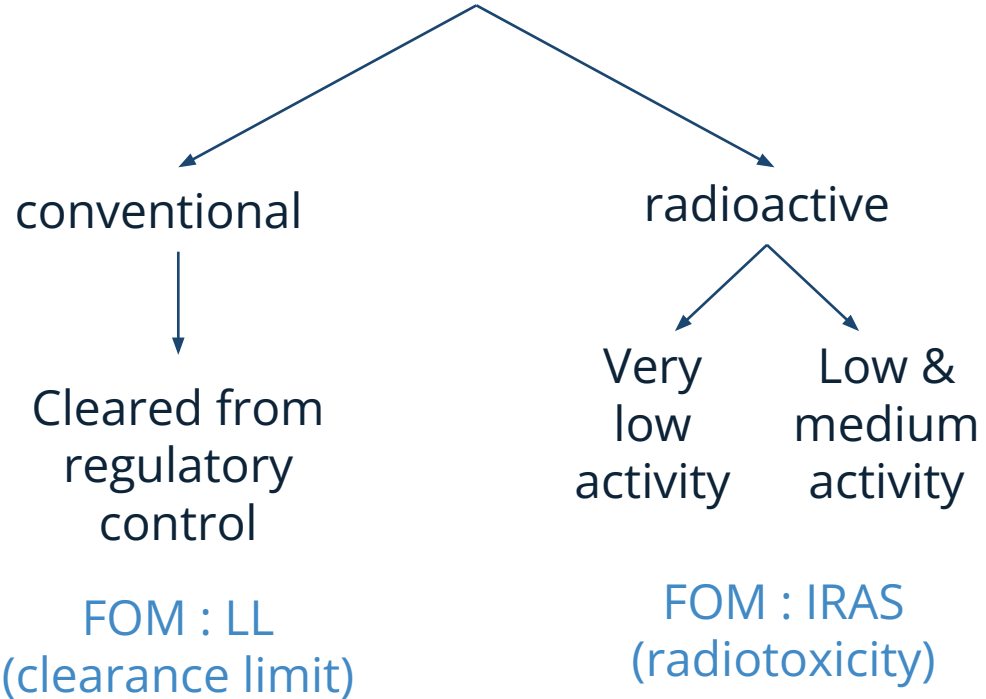
...we need to know the **nuclides** present in the waste and their **activity**

Then we can calculate a figure of merit depending on the **elimination path**





# Irradiated material



*\* More info at backup slides*

# What does it mean

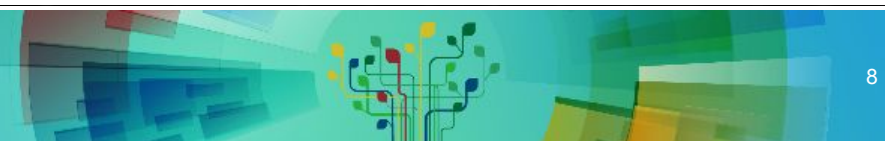
to **characterize** radioactive waste?



In order to **extract** the nuclides and their activity...

...we need to perform **measurements** on the waste.

Typical examples : **Gamma spectrometry** or chemically analyzing **samples**



# What does it mean

to **characterize** radioactive waste?



*But....*

*Is there **another** way to do it?*



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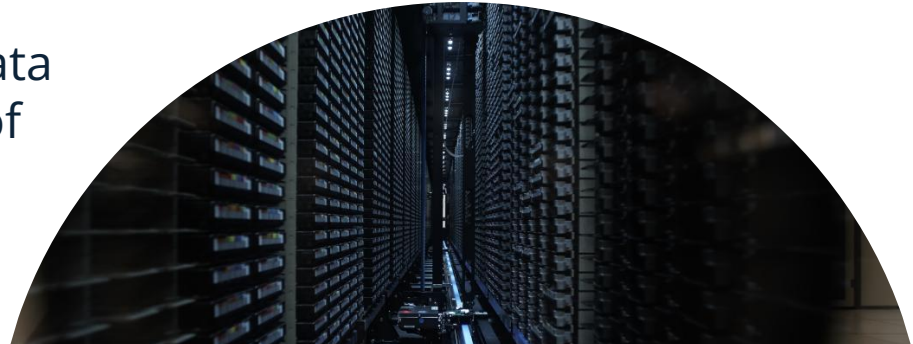
What have we achieved and what's next

# Machine learning

...when a computer mimics human behavior and **learns** from data

We “show” the computer a lot of data and it **identifies** underlying patterns and correlations

It can use models to connect data points, identify similar groups of data points, find relations between parameters, etc...



# But what about statistics?

We are used to working with statistical models, why do we need machine learning?

With statistics, we use data to set parameters so we can build a model that describes a **population**

With machine learning, we identify patterns and can make predictions about **individuals**



# So how can we employ it?

We need **data**...

...**representative** enough of our waste so the computer can find patterns...

...and apply them to create **models** and **predictions** for individual waste

# How do we create data?

At CERN we have two powerful tools



TREC: Traceability system for potentially radioactive equipment

ActiWiz: A software calculating irradiation results based on input parameters

*\* More info at backup slides*

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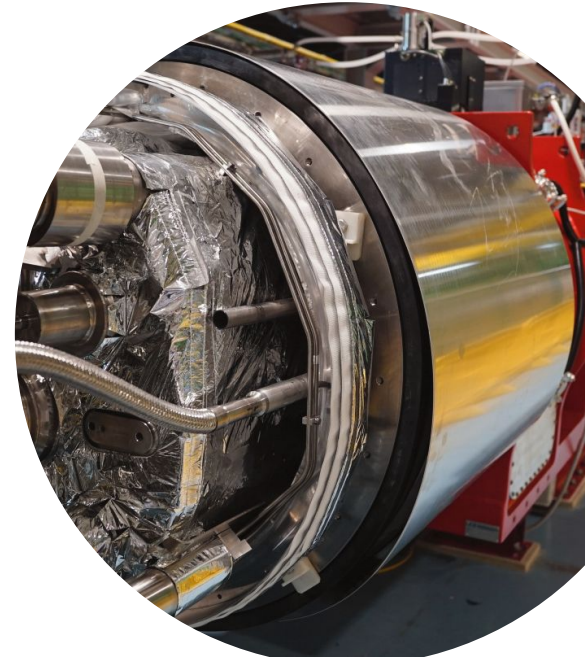
What have we achieved and what's next

# The massive items case

Some of the waste generated at CERN consists of bulky items such as magnets and shielding blocks.

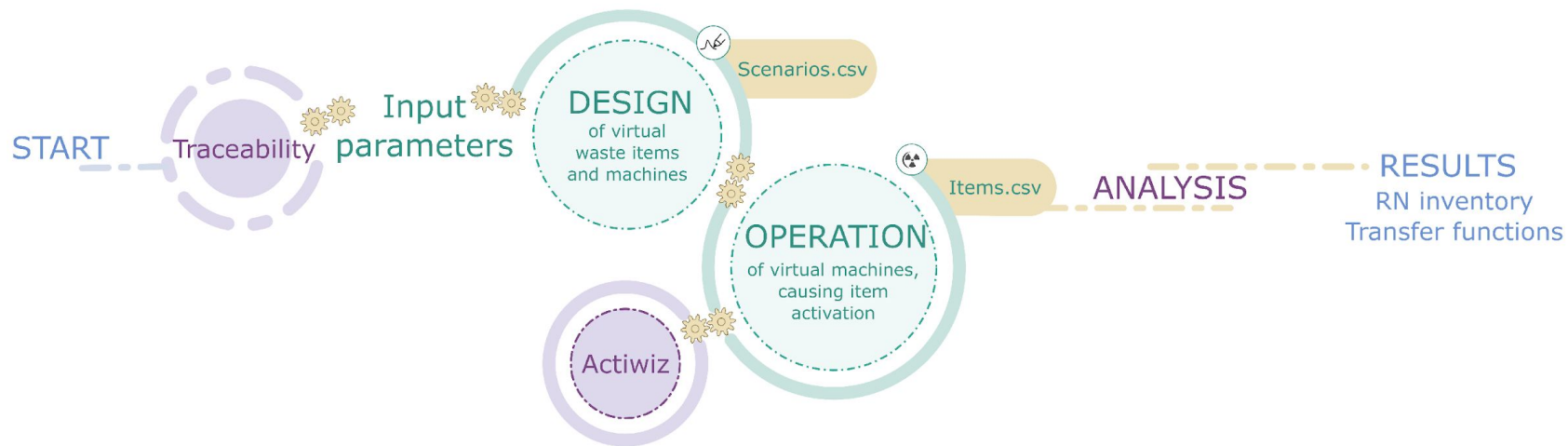
Characterization via g-spec : Technically and practically **complex**

Alternative : Characterize based on the **dose rate** instead



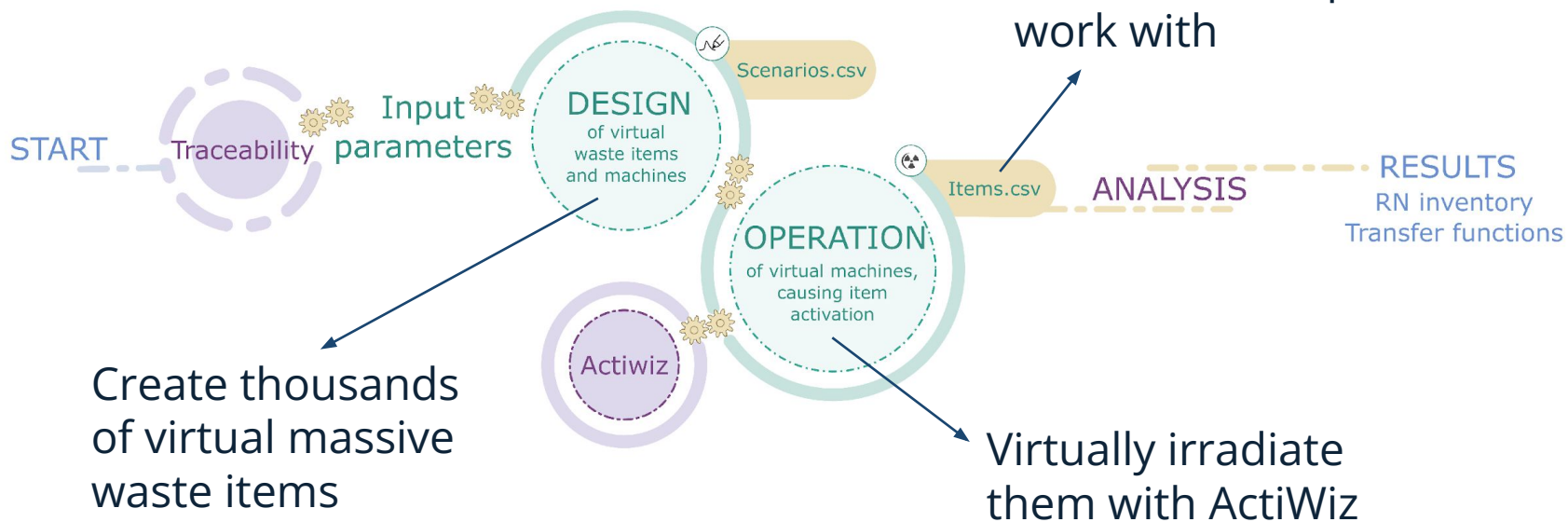
# The massive items case

Our workflow:



# The massive items case

Our workflow:



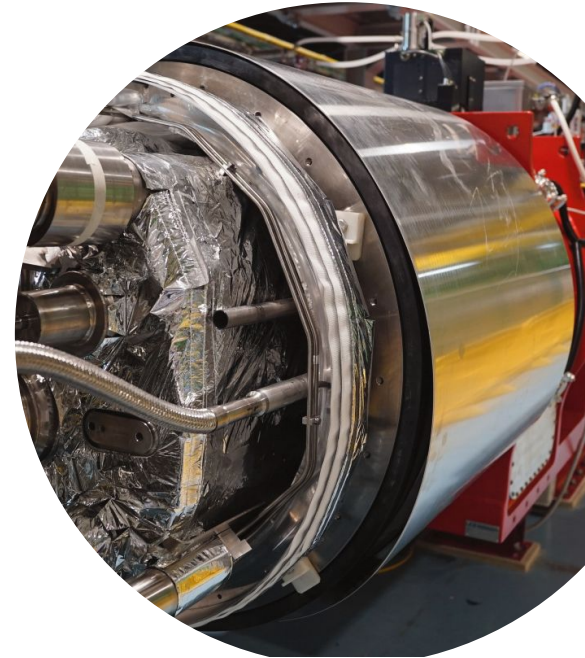
\* More info at backup slides

# The massive items case

Aim: Translate an item's dose rate to RN inventory

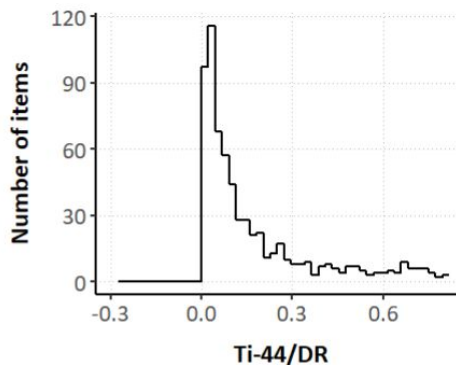
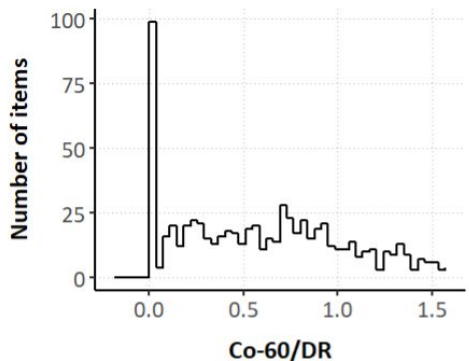
Data: Virtual waste items created with ActiWiz  
Data contain each item's dose rate and  
RN inventory

Path: Study the relationship between item dose  
rate and each nuclide's activity

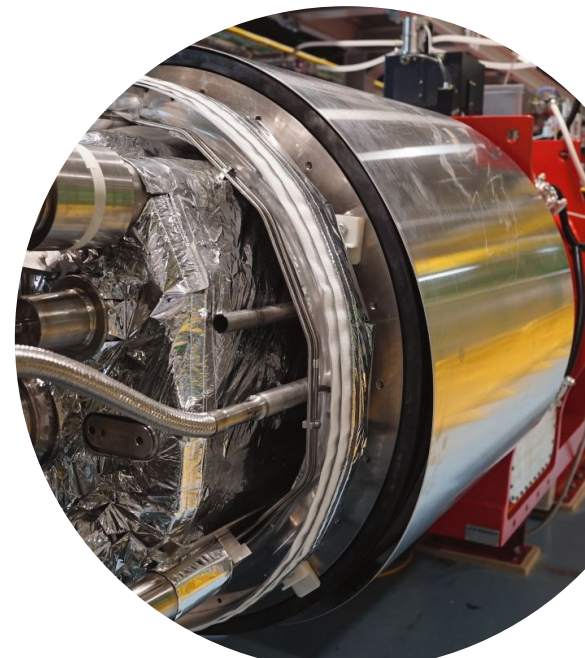


# The massive items case

Path: Study the relationship between item dose rate and each nuclide's activity

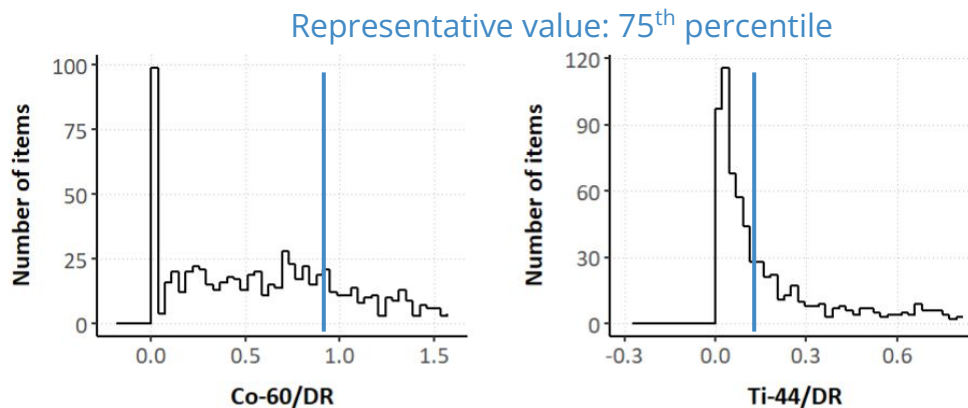


Distribution of ratio between nuclide activity and dose rate for 10000 items for two example nuclides, Co-60 and Ti-44

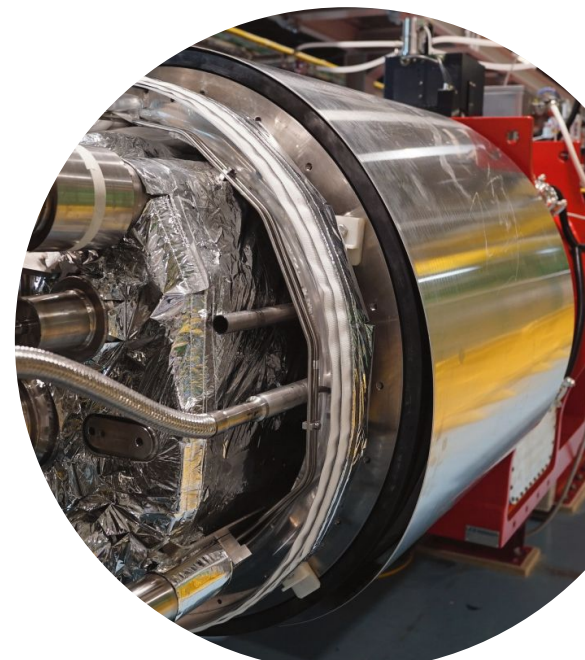


# The massive items case

Path: Study the relationship between item dose rate and each nuclide's activity

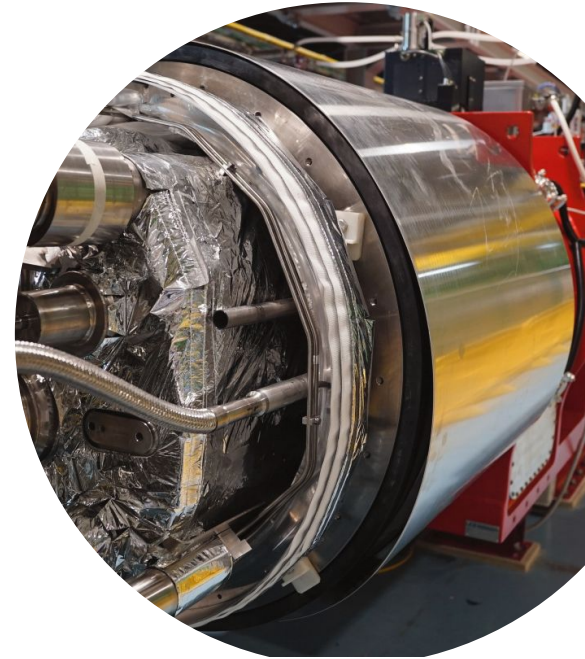


Distribution of ratio between nuclide activity and dose rate for 10000 items for two example nuclides, Co-60 and Ti-44



# The massive items case

We call this value the “transfer function” and use it to translate an item’s dose rate to radionuclide activities



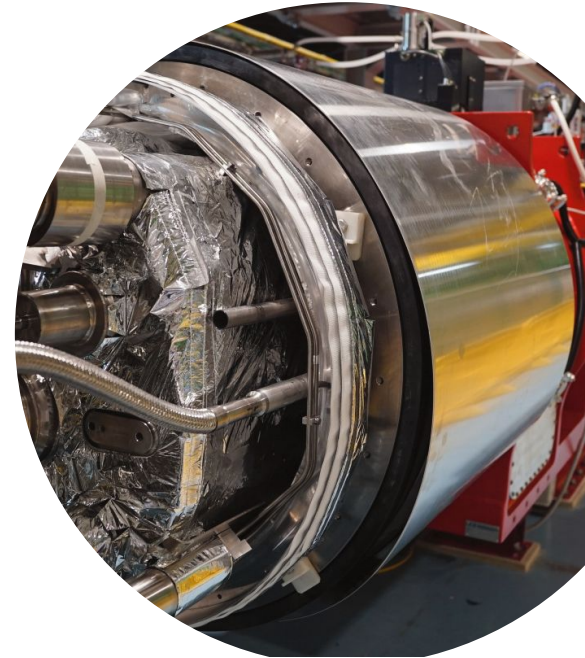
# The massive items case

We call this value the “transfer function” and use it to translate an item’s dose rate to radionuclide activities

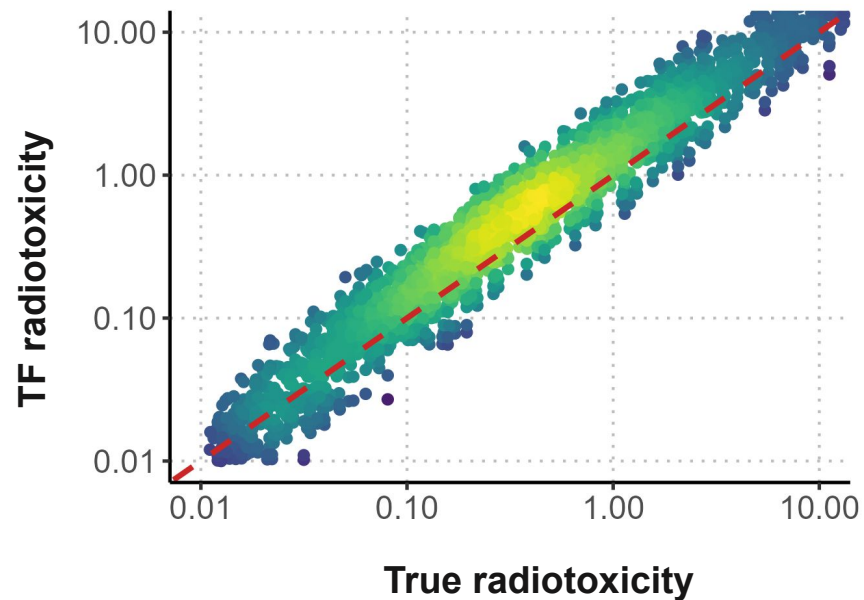
How do we know it works?

Virtual items advantage : We created them, we know their details.

We can create an independent “test” dataset and compare!



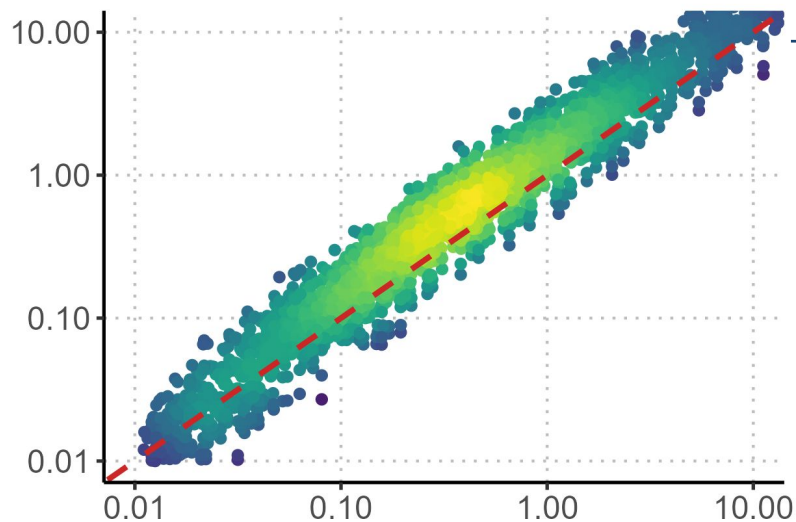
# The massive items case



# The massive items case

As we calculate it with our transfer functions

TF radiotoxicity

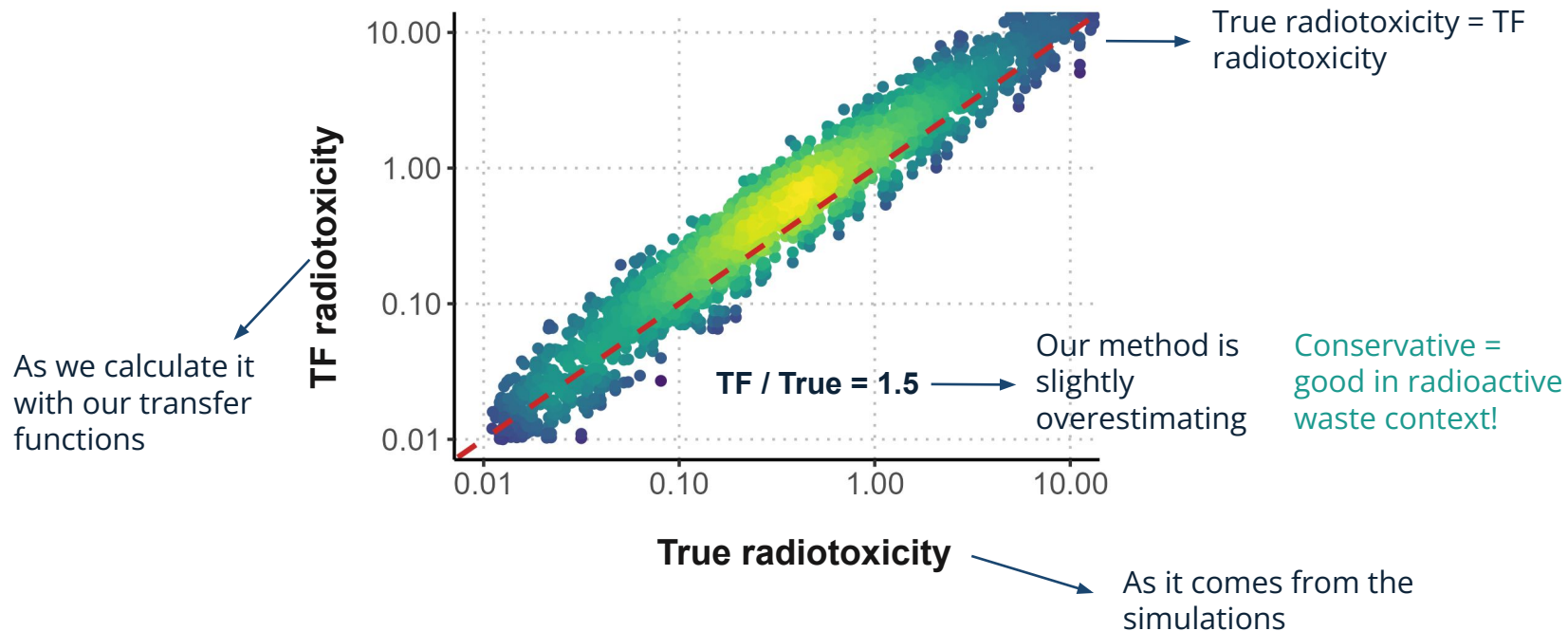


True radiotoxicity = TF radiotoxicity

True radiotoxicity

As it comes from the simulations

# The massive items case





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# What are we doing?

Launching a new methodology in radiological characterisation by employing the advantages of **machine learning**

This large-scale approach helps us spot trends with more **confidence** and reduces the influence of random variations.

In the massive items case, replacing gamma-spec measurements with dose-rate mapping **reduces** complexity



# What's next

Dive deeper into machine learning!

Use **other methods** of machine learning and computational techniques such as clustering, modelling, principal component analysis, etc. not only in the characterisation methodologies but also in **quality control**.



A nighttime photograph of the CERN building complex. On the left, a large, illuminated dome structure with a grid-like facade is the central focus. To its right, a tram with orange and white stripes is moving along a track. The background shows other modern buildings and city lights under a dark sky.

# Thank you for your attention!

Contact me:  
[maria-elisso.stamati@cern.ch](mailto:maria-elisso.stamati@cern.ch)



Find us:  
<https://radiological-characterization.web.cern.ch/>

# Backup slides



Date

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# CERN's waste

Since 2011, ~8000 m<sup>3</sup> of waste have been produced and ~11000 m<sup>3</sup> (eq. before treatment) have been disposed of.

Some waste families:

- Metals
- Inert
- Cables
- Sources / Targets
- Burnable

# CERN's regulatory framework

CERN is an **international organisation** and, as such, it has a right to establish its own rules and regulations.

CERN agrees to follow best practices in matters of radiation protection and radiation safety taking into account the legislation of its **host states**, as well European and international standards.

Their implementation is discussed between CERN and its host states (FR and CH) authorities according to a "**Tripartite Agreement**" of 2010.



# CERN's regulatory framework

Radioactive waste from CERN Facilities is **disposed** of by the host States according to the existing pathways, in conformity with their national legislation.

The waste study of the organisation takes into account the need for a **fair distribution** between the host States, according to the quantity, activity and toxicity of the waste, and aims to ensure that it is disposed of through the most technically and economically advantageous pathways.



Cleared  
Low & Medium activity



Very low activity  
Low & Medium activity : Short half-life

# Characterisation FOMs

## Clearance conditions:

- Surface contamination (CS) level: CS is defined as the sum of the non fixed activity which can be removed from a surface by wiping or washing + the fixed activity which can be removed during future use. Condition:  $CS < \text{or } =$  to the defined CS limit.
- Specific activity level: It should be lower than or equal to the clearance limit / limite de liberation (LL).
- Dose rate level: The dose rate (D) is an operational quantity used to estimate the exposure of a person to radiation. The ambient equivalent D at 10cm from the material surface should be lower than  $0.1 \mu\text{Sv/h}$ .

# Characterisation FOMs

## Radioactive waste:

- IRAS = Indice Radiologique d'Acceptabilité en Stockage, defined by the French National Agency for Radioactive Waste Management (ANDRA)

$$IRAS = \sum_i \frac{a_i}{AL_i},$$

with  $a_i$  the activity of isotope  $i$  and  $AL_i$  a coefficient expressing the radiotoxicity of isotope  $i$

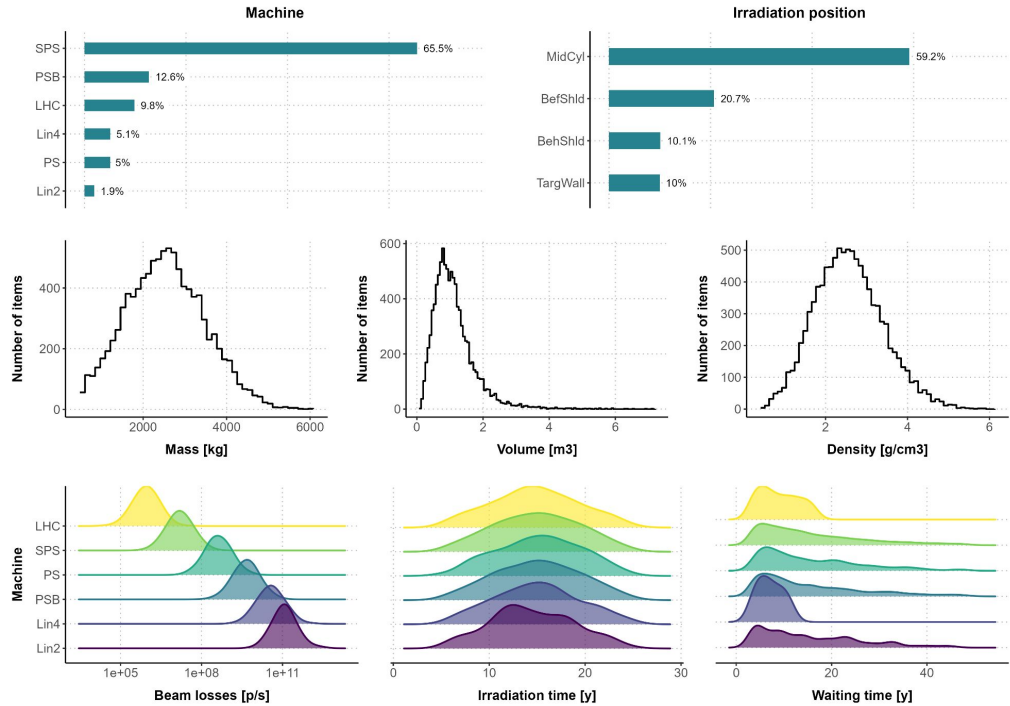
All isotopes with activities surpassing an established declaration threshold must be included in the IRAS calculation

ActiWiz is an **analytical code** developed at CERN in 2011. It is based on simulated particle fluence spectra together with a dedicated decay engine, combined with user-defined input parameters. The aim of ActiWiz is to compute quantities related to the activation of an object inside a CERN accelerator.

The basis of ActiWiz is the definition of a **scenario** that specifies the physical characteristics of an item as well as its radiological history parameters. It can then calculate the induced activity values for all radionuclides present in the item, as well as the item's average dose rate at a user-defined distance.

# Massive items parameters

In order for our virtual items to be as representative as possible, we sample their parameters from **probability distributions** based on our knowledge of CERN's radioactive waste



# Massive items parameters

To confirm that the parameters are representative, we can simulate **specific items** whose parameters are known to us thanks to TREC and compare with the general study

