## **VYPR**

#### cern.ch/vypr

#### A Framework for Automated Performance Analysis of Python Programs

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## Outline

Analysis-by-Specification

A Framework for Analysis-by-Specification

Specification

Instrumentation and Monitoring

A Verdict Server

An Explanation Mode













A performance analysis technique inspired by Runtime Verification (RV).

Runtime Verification - a lightweight formal method.

General idea - specify a property that a part of a program (for us, a function) should hold, and check it at runtime.

Not formal verification; complements testing.



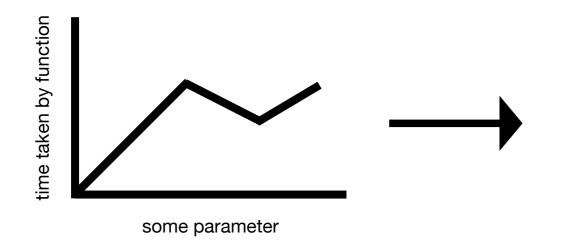




Runtime Verification for Performance Analysis.

Performance analysis on the CMS Experiment - historically a manual effort.

Analysis using natural language specifications:



When did the execution time of some part of the code exceed some t?

How often?

For which inputs?







**VYPR** 

Express performance requirements in a specification language.

Removes ambiguity inherent in natural language specifications.

We can algorithmically check for satisfaction.

We have precise structure that we can use to explain failure to satisfy specifications.







the central idea of the VyPR tool

+ some extra machinery to allow sophisticated analysis.







# VyPR as research output

The first application of Runtime Verification in High Energy Physics (as far as we know) - *J H Dawes, et al. TACAS 19.* 

Relatively language agnostic, rigorous theoretical foundation - J H Dawes, G Reger. SAC 19.

Research in the environment in which the output will be used.







## VyPR as a development tool

A tool to help developers easily analyse the performance of their Python programs.

No need to know any of the foundation mathematics.

For application to either production systems, or systems still in development.





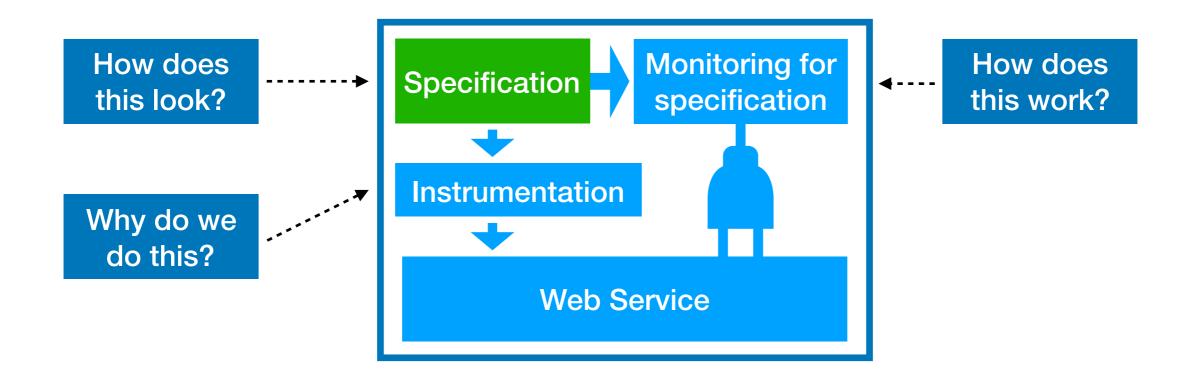


## How does VyPR look?

The developer writes a specification, adds a line to their service's code, and the rest is automatic. **VyPR2 Verdict Server Monitoring for Specification** Verdict Relational specification Registration Verdict API **Database** Instrumentation **Web-based Analysis Application** Web Service development aiming at Python shell-based Analysis Library September 2019 MANCHESTER **VYPR** 

The University of Manchester

# Service-level Monitoring









# Specification

VyPR provides its own specification language - CFTL.

Control-Flow Temporal Logic

A mathematical logic designed for expressing properties that should hold true over runs of functions in a system.

VyPR provides the PyCFTL library for expressing CFTL specifications.







```
"module description" : {
 "function name" : [
   Forall(q = changes('val')).
    Forall(t = calls('showData'), after='q').
    Check (lambda q, t : (
      If (q('val').equals(True)).then(
        t.duration().in([0, 3])
```

A full PyCFTL specification.

Let's investigate...







**VYPR** 

# **PyCFTL**

"Every time val is assigned, the next call to the function showData should take no more than 3 seconds."

Selects points of interest at runtime.

Defines the rule to check at each of these points of interest.







**VYPR** 

## In more detail

The argument q to the lambda will model *states*, so can be treated as such in the lambda body.

```
Check(lambda q : (
   q.next_call('showData').duration()._in([0, 3])
))
```

Calling *next\_call* on a state has the effect of searching forwards in time.







# More complex...

"Every time val is assigned, every future call to showData should take no more than 3 seconds."

Selects points of interest at runtime, using two selections.

One of the selections uses values from the previous as a starting point for the forward search.

```
Forall(q = changes('val')).\
Forall(t = calls('showData'), after='q').\
Check(lambda q, t : (
   t.duration()._in([0, 3])
))
```







## In more detail

The argument q to the lambda will model states, and t will model function calls.

Naturally, we can measure the duration of the function call represented by t.







# Checking for branches

What if we only want to place a constraint on a branch whose entry condition is that *val* is True?

```
Forall(q = changes('val')).\
Forall(t = calls('showData'), after='q').\
Check(lambda q, t : (
    If(q('val').equals(True)).then(
        t.duration()._in([0, 3])
    )
)
)
```

We only care about the time taken by calls to showData if val is equal to True.







**VYPR** 

# A complete specification format

```
"module description" : {
                                      Separation of module and function
                                        names makes resolution easier.
  "function name" : [
    Forall(q = changes('val')).\
    Forall(t = calls('showData'), after='q').\
    Check(lambda q, t : (
      If (q('val').equals(True)).then(
        t.duration(). in([0, 3])
                          No modification of source code.
                          Specification for entire service
                          written in one file.
```







#### **Instrumentation and Monitoring**







```
Forall(q = changes('val')).\
Forall(t = calls('showData'), after='q').\
Check(lambda q, t : (
    If(q('val').equals(True)).then(
        t.duration()._in([0, 3])
    )
))
```

How do we guarantee that we will receive enough information from the program to check this, while preserving program behaviour?







## Trace functions?

The Python interpreter gives us the nice option of a trace function.

Every event generated at runtime by the interpreter triggers a call to the trace function.

For every event, we make a decision on whether we need it.

Potentially massive overhead.





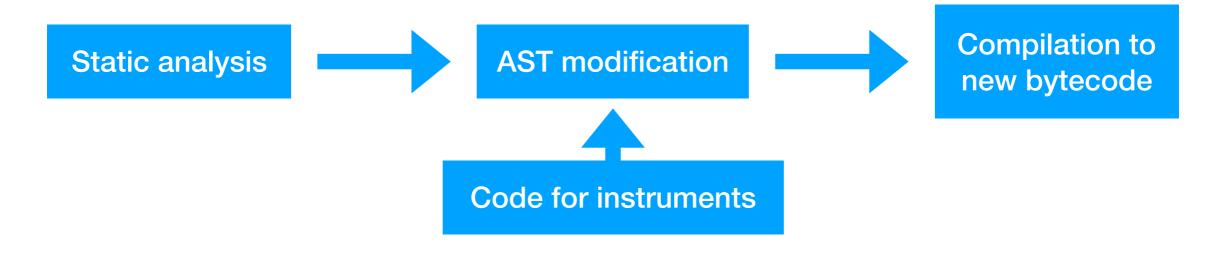


# The need for instrumentation

If a trace function (implemented in either Python or C/C++) is not an option, what do we do?

VyPR uses static analysis to perform "instrumentation".

"Instrumentation" - adding code to the program so we can check specifications.









# Monitoring

Given data from instruments, how do we check if a specification holds?

CFTL's monitoring algorithm is not straightforward... but it's efficient.

VyPR sets up a separate thread, to which instruments send their data via an intermediate consumption queue.







# Monitoring

So far, CFTL specifications are written over single function calls, hence single threads.

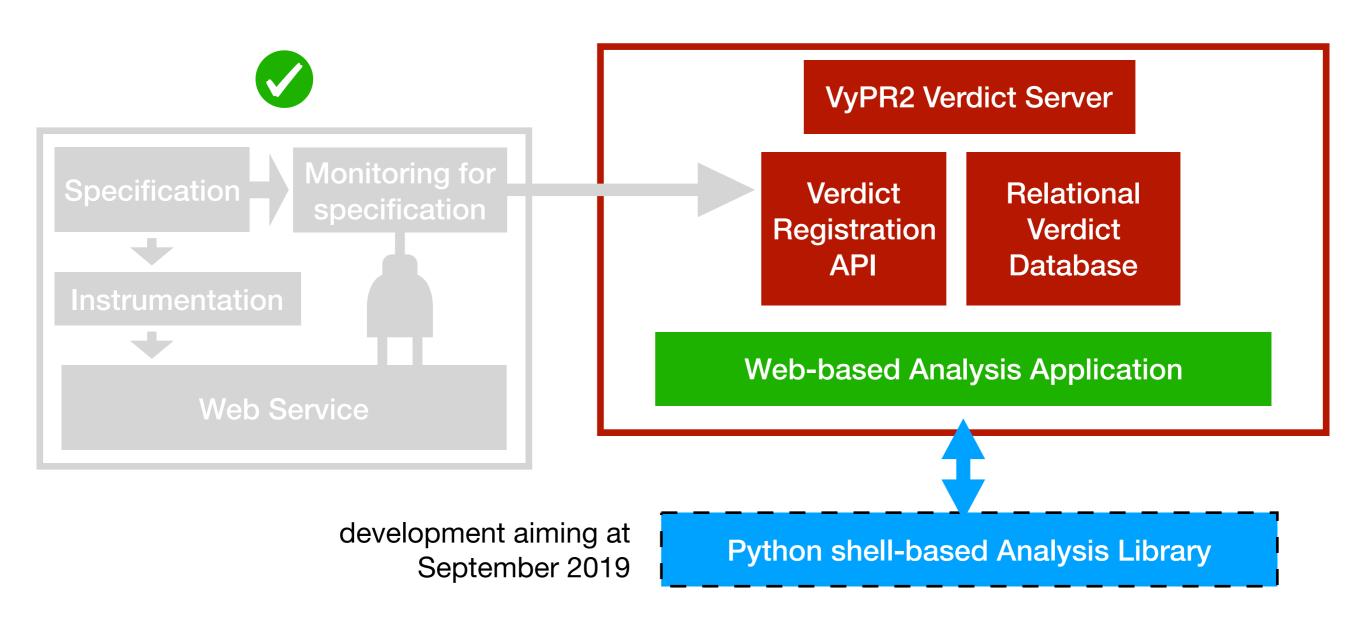
This means we send the relevant data to the verdict server once a function's execution has ended.







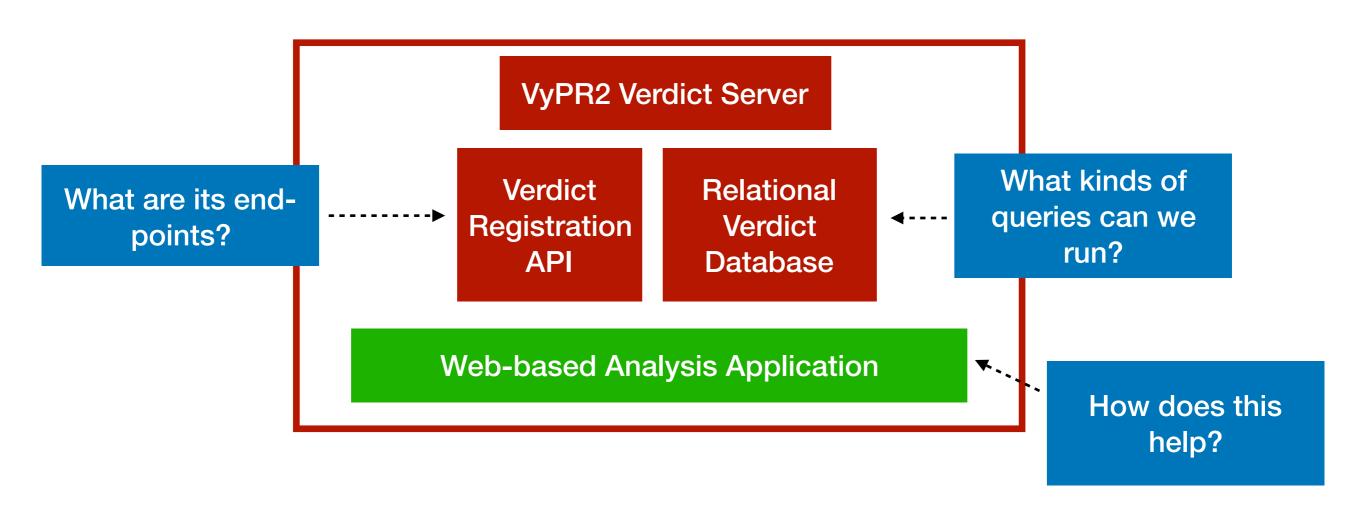
#### We almost have a complete architecture...

















## A Verdict Database

1) The top level is transactions.

```
"module description" : {
    "function name" : [

    Forall(q = changes('val')).\
    Forall(t = calls('showData'), after='q').\
    Check(lambda q, t : (
        If(q('val').equals(True)).then(
              t.duration()._in([0, 3])
        )
    ))
    ]
}
```







2) All verdict data is grouped by the

## A Verdict Database







## A Verdict Database





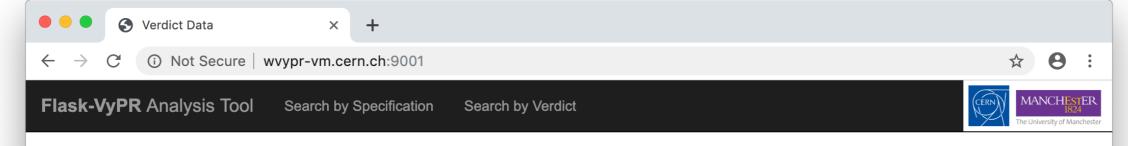


#### A Web Application for Simple Analysis









#### VyPR, CFTL and PyCFTL

Flask-VyPR is a tool for runtime verification of Python and Flask-based web services with respect to specifications written in Control Flow Temporal Logic.

#### It consists of:

- The PyCFTL library, a small library for writing CFTL specifications in Python.
- A program that performs optimal instrumentation of the web service for the given CFTL specifications.
- A library for making the web service verified at runtime.
- A server for collection of verdict data from the web service's runtime. This server provides both an API for a service under scrutiny to store its verdict data, and a front-end for users to either 1) see verdicts from certain parts of the service or 2) see which parts of the service gave a specific verdict.

Control Flow Temporal Logic (CFTL) is a specification language in the form of a temporal logic used to specify constraints over the state and time of a program.

PyCFTL is the Python binding for CFTL that allows CFTL properties to be written in Python and checked over a Python program.

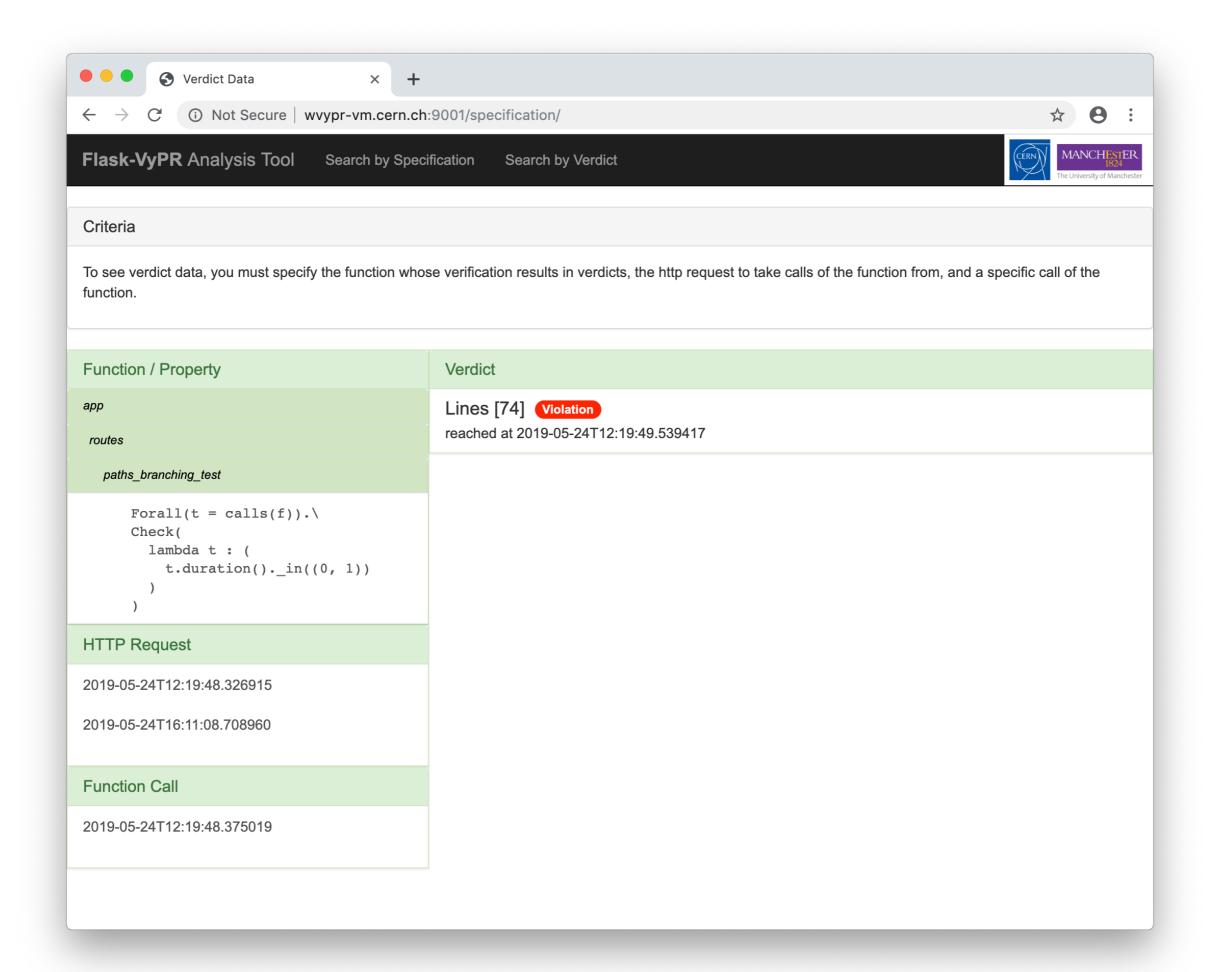
This tool shows verdict data from CFTL formulas written in PyCFTL to describe behaviour of web services.

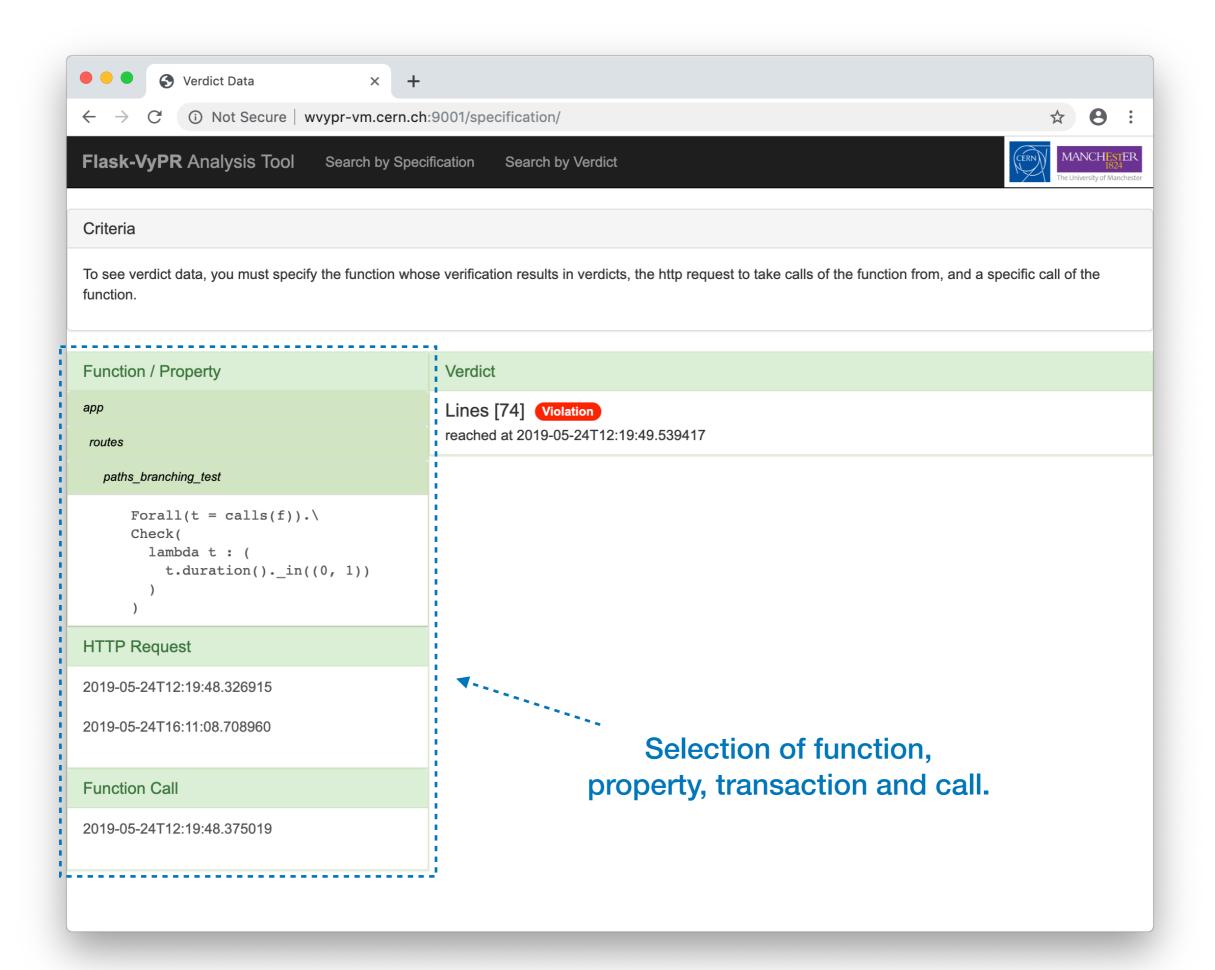
Starting from the function/property pair, you can see the verdicts reached by verifying the property over the function in a specific HTTP request.

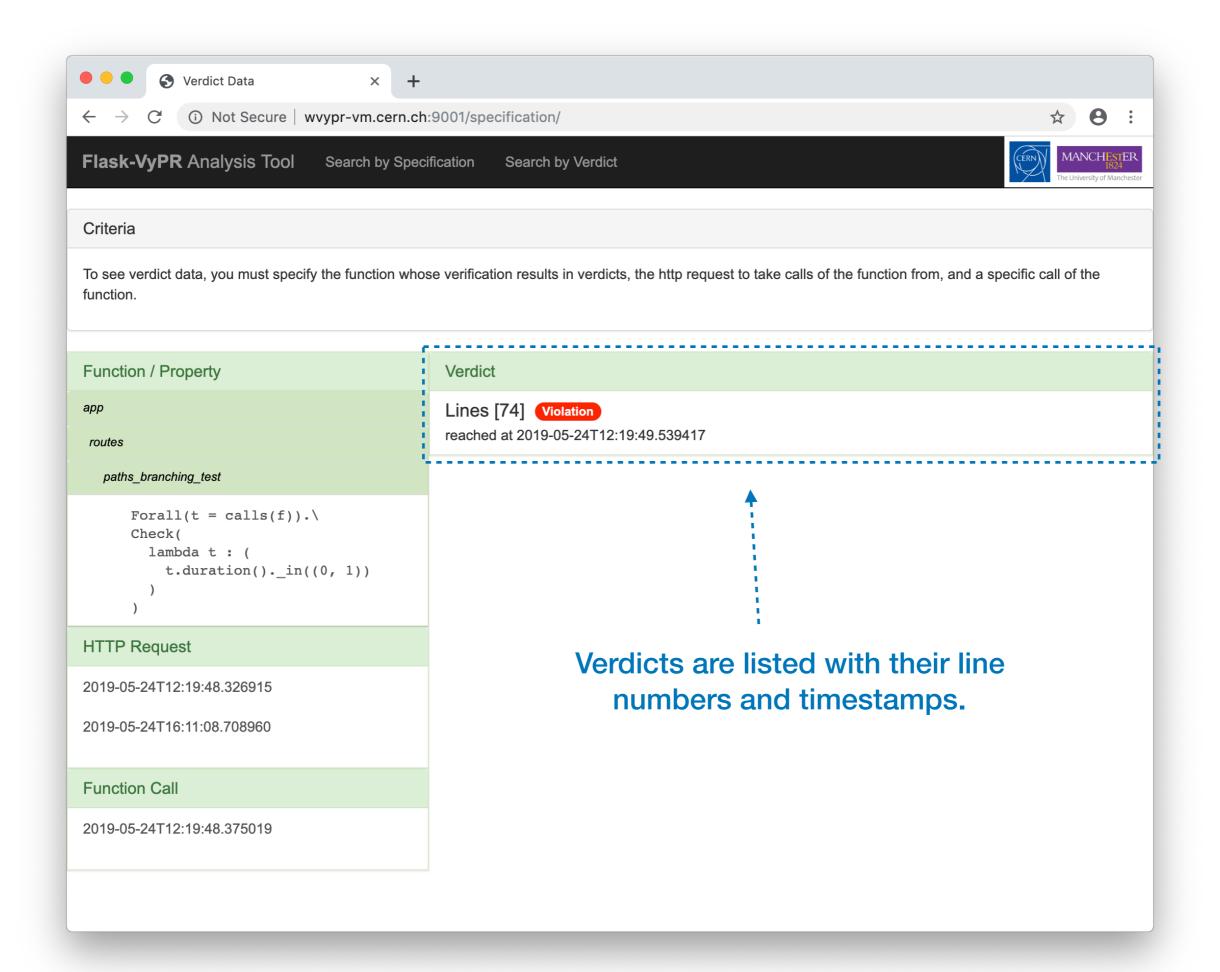
Search by Specification

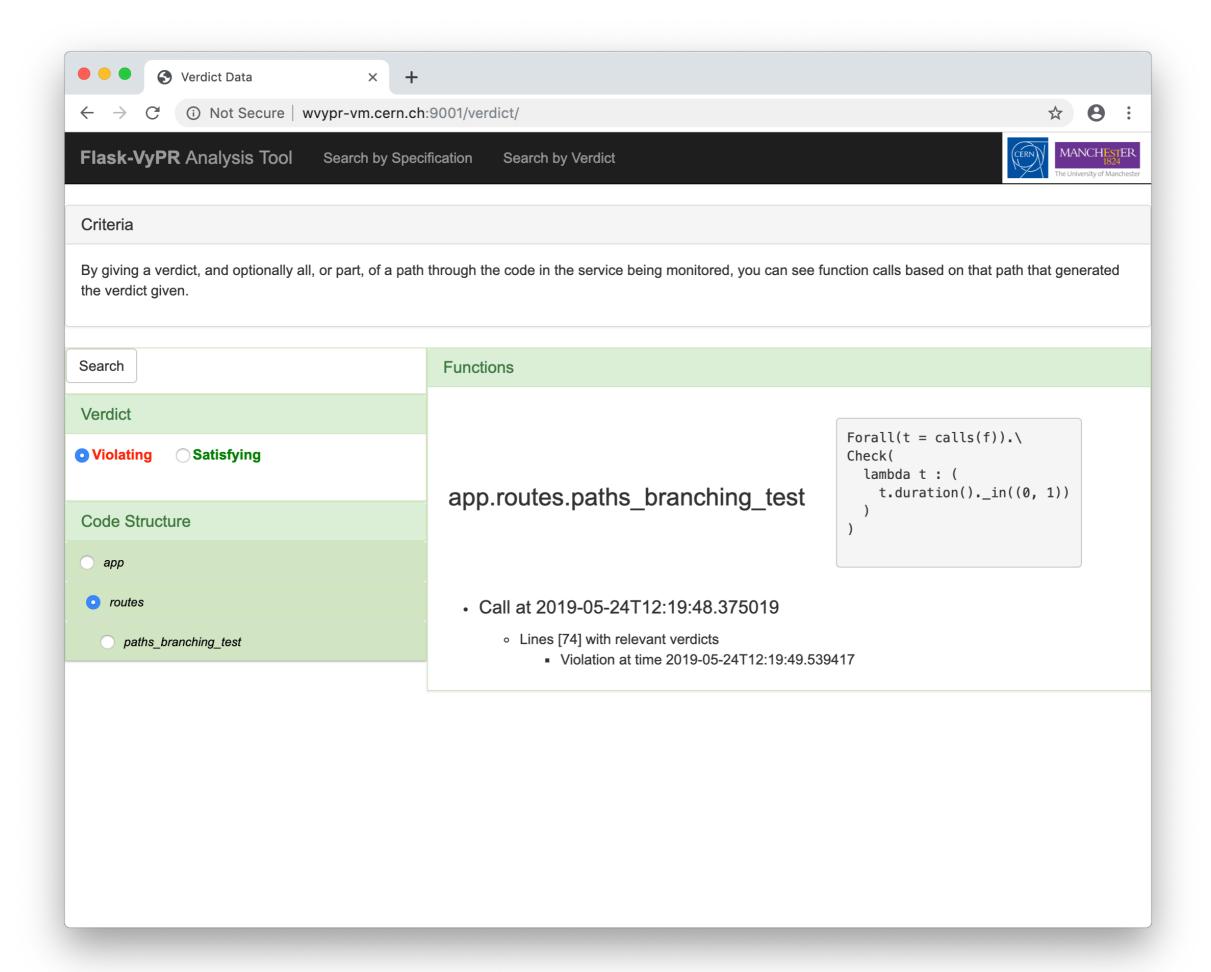
Starting from a verdict, find the code that generated that verdict at some point during runtime.

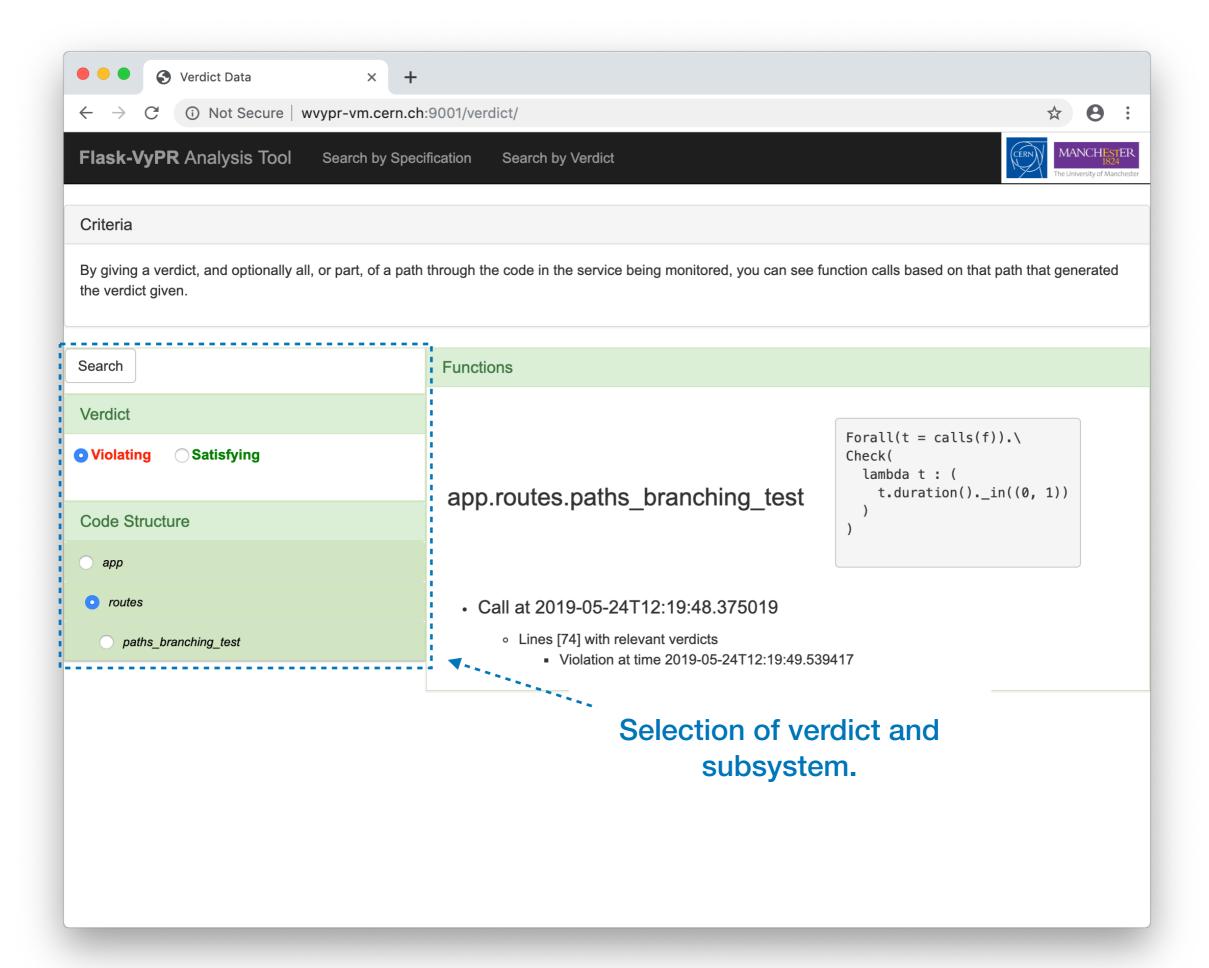
Search by Verdict

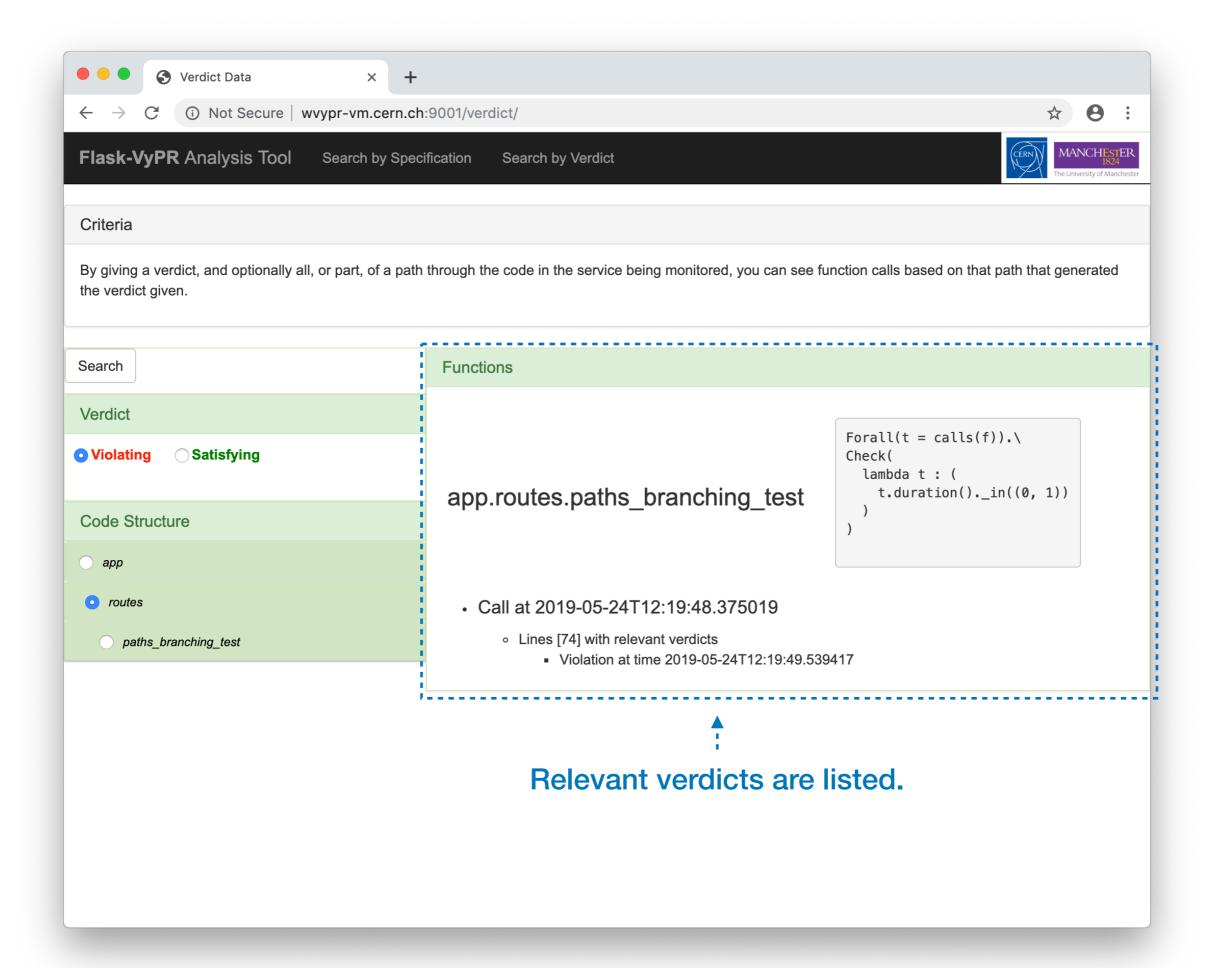




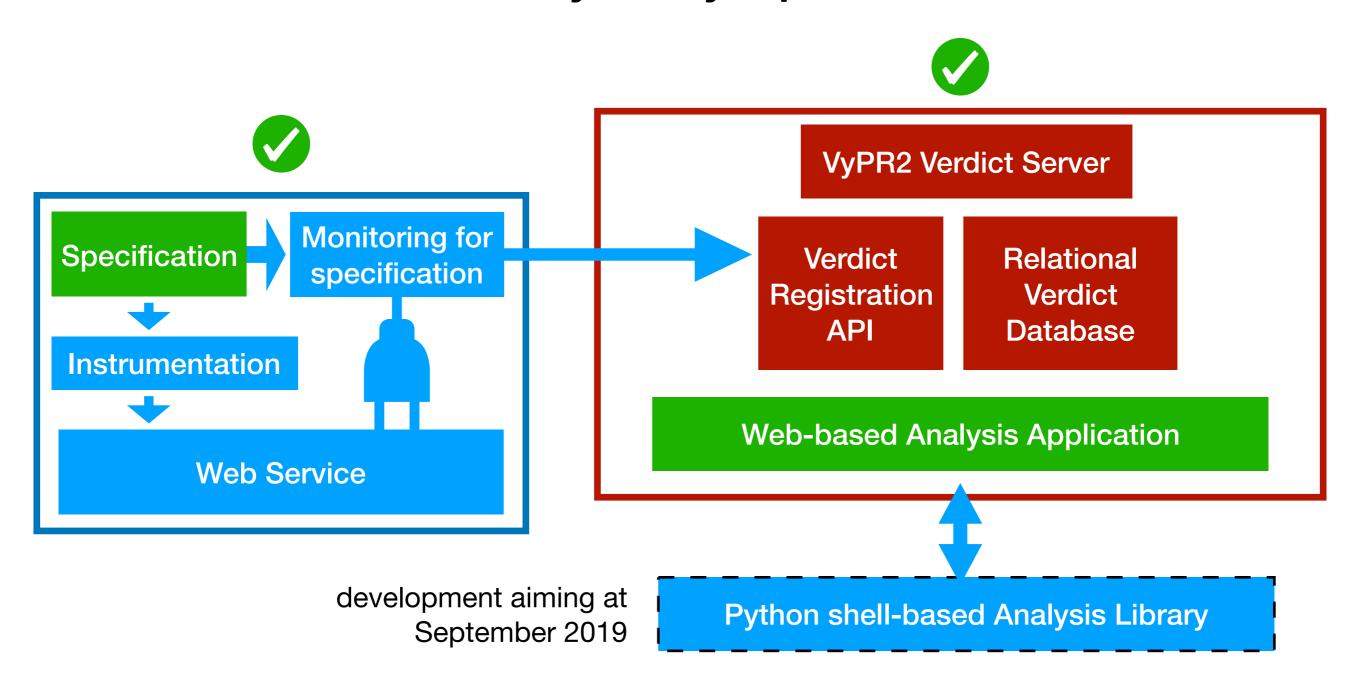








#### Architecture for Analysis-by-Specification? Check.









## How has this already helped?

VyPR has profited a lot from being developed at CERN.

The first major application - the next version of CMS' nonevent data upload service.

Worked closely with CMS' Alignment and Calibrations group - thanks to Giacomo Govi.

During replay of 6 months of uploads from LHC runs, VyPR found unexpected performance drops, while inducing little overhead. *J H Dawes, et al. TACAS 2019.* 







# How has this already helped?

#### A bit of context

"CMS non-event data uploads" - during LHC runs, nonevent data are uploaded to a central database ready for reconstruction.

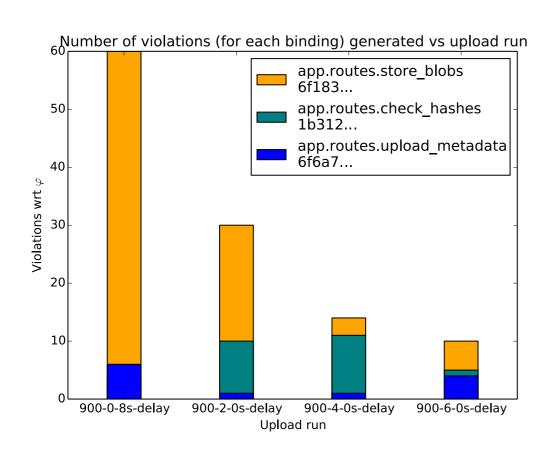
VyPR was used to analyse the service used for such uploads.







# How has this already helped?



The code covered by this part of the specification was supposed to be an optimisation.

Violations generated across 14610 uploads

The code covered by this part of the specification was supposed to be an optimisation.

Function/Property pair ID

Y axis is number of violations

Y axis is number of violations

X axis is how long we waited between uploads

X axis is pairs of functions with properties they should satisfy







### Lessons learned?

The first analysis tools we built were not powerful enough.

Analyses performed on non-event upload required custom scripts for verdict database querying.

We need an automated explanation mechanism.

How could an explanation look? What would be useful to developers?







#### An Explanation Mode for VyPR

Which parts of code led to failing observations more frequently?







```
"module" : {
    "function" : [
        Forall(t = calls('showData')).\
        Check(lambda t : (
            t.duration()._in([0, 3])
        ))
        *...
]
```

Suppose the verdict database tells us that there was a call to *showData* that failed this constraint.

Knowing the paths through code taken to reach the failing observation across multiple runs might indicate "faulty control flow".

Distinction between paths => the difference found may correlate to performance drops.

No distinction => control flow is *probably* not responsible.







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## Integration of Path Reconstruction

VyPR's architecture integrates the machinery for path reconstruction efficiently in space and time.

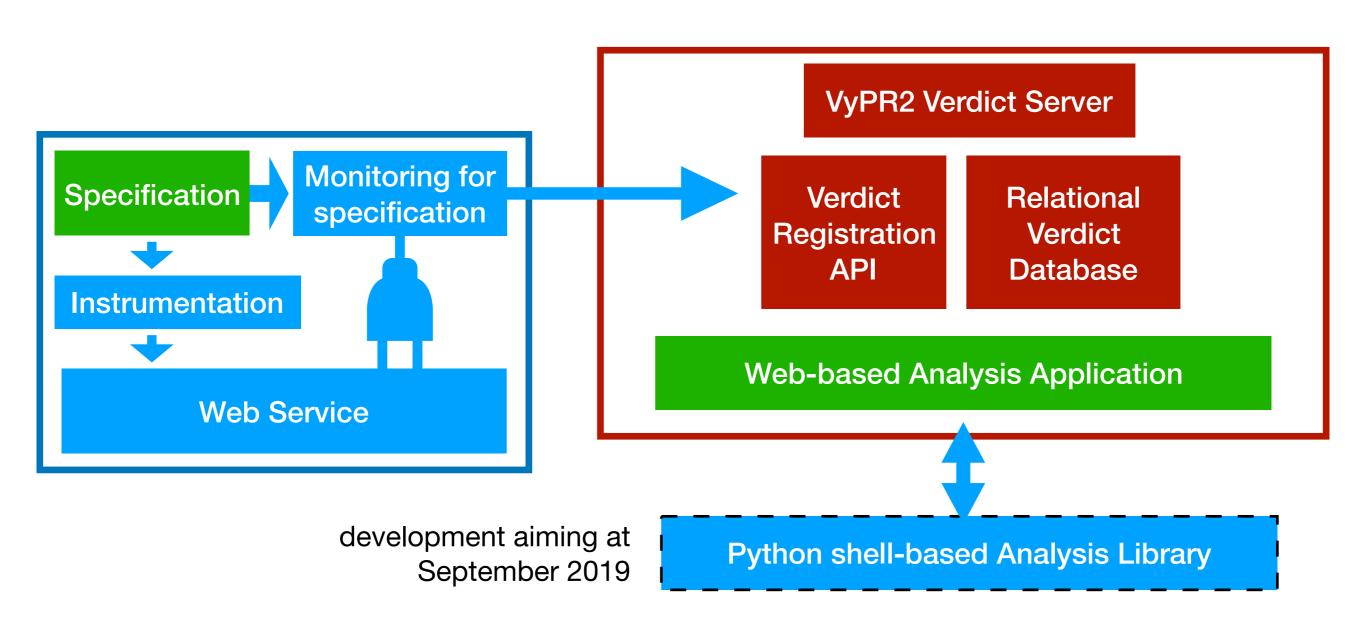
We store the minimum data needed to be able to reconstruct paths in most cases.

Path Reconstruction becomes more efficient as we observe more from the monitored system.





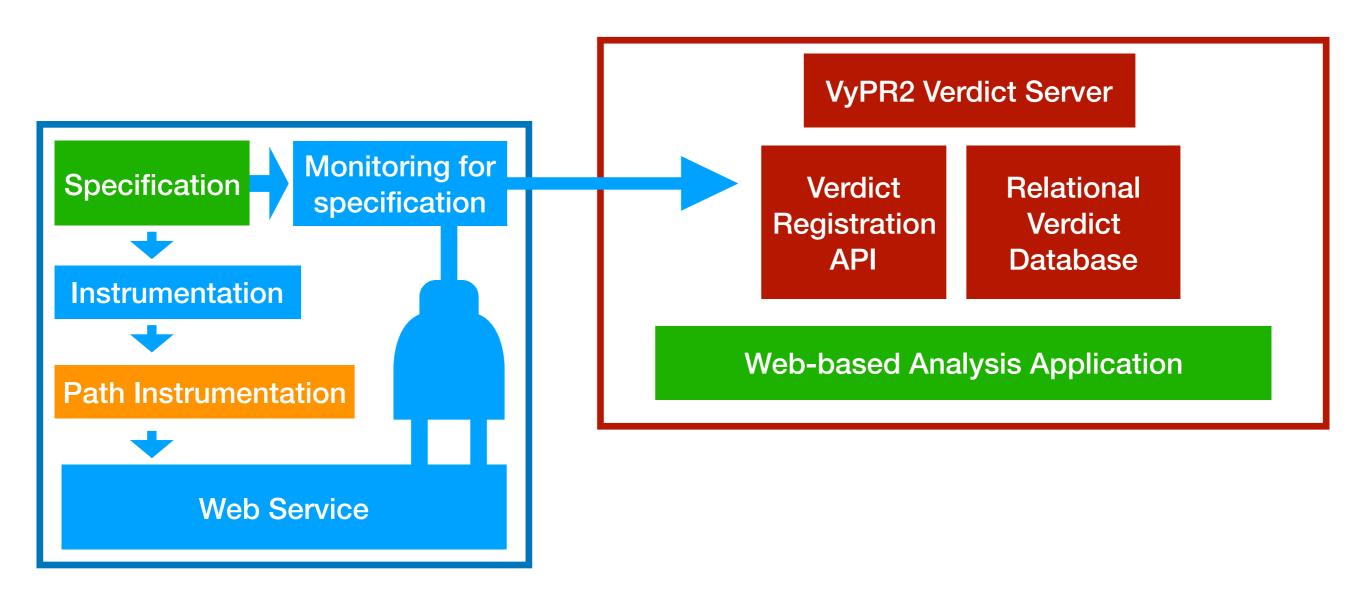








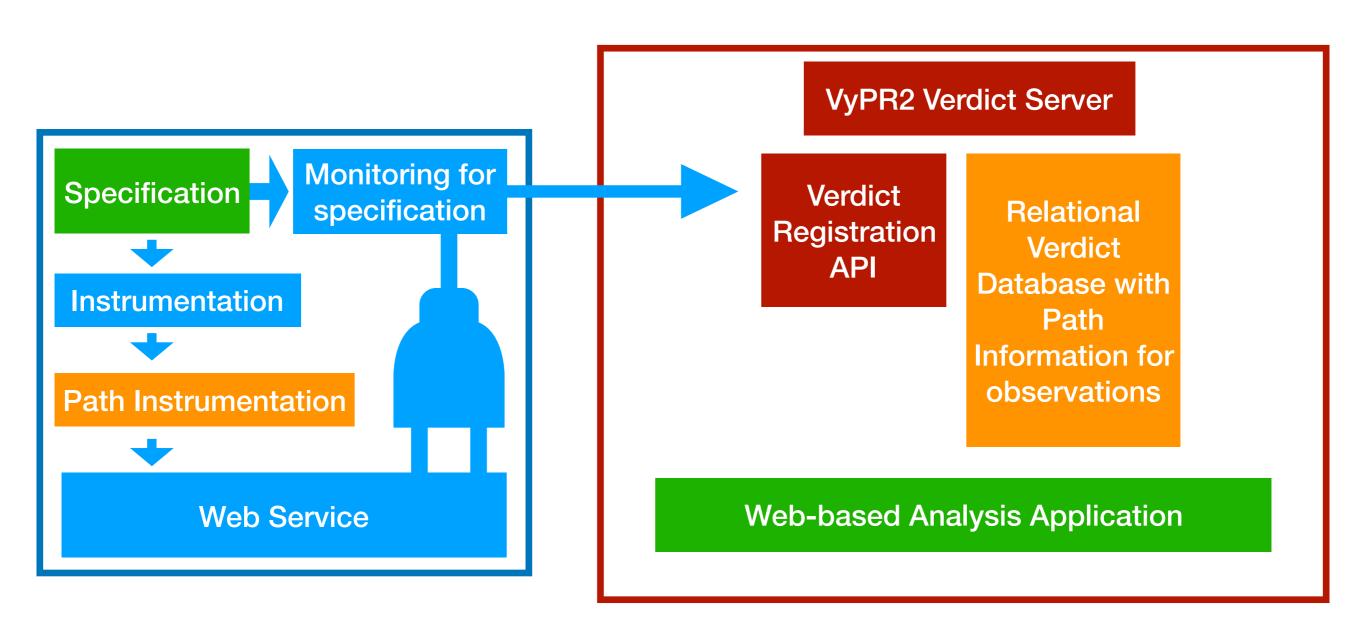








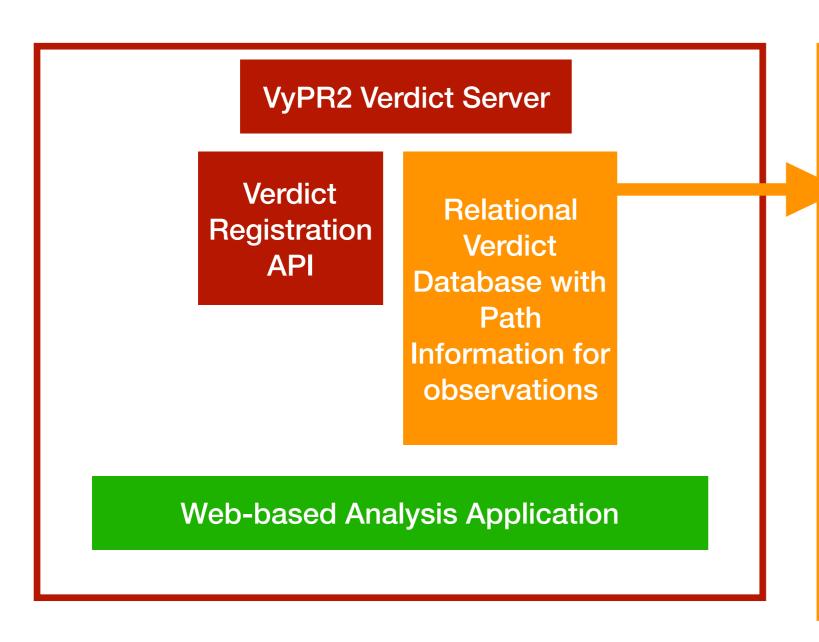


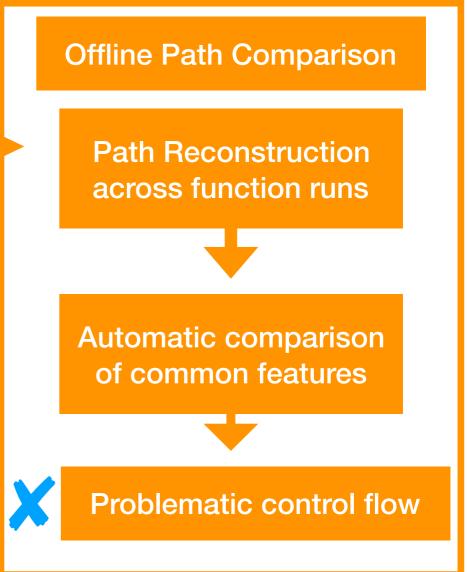
















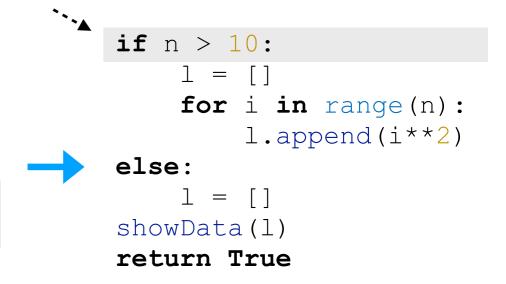




### Common Features?

Given two reconstructed paths that are similar except for some branches taken, VyPR can detect the locations of the deviations.

VyPR can determine that the conditional is the source of deviation.









# Plans for an Analysis Library

Next 6 months - development of a powerful analysis library.

Plans are for this to be a self-contained library for use in the Python shell or as CLI.

Make explanation an intuitive process for developers.







#### Conclusions

VyPR is a research tool developed in an industrial environment, and has already proven its utility.

Still more design decisions to be made.

Theory is in a good place - but the analysis tools need to be developed - will happen over the next 6 months.







During the next year of work on VyPR and inline with the sofar use-case driven development of theory and implementation:

We are looking for use cases around CERN, and further afield.

cern.ch/vypr - joshua.dawes@cern.ch

Thank you for listening!





