

# Embracing the new threat: towards automatically, self-diversifying malware

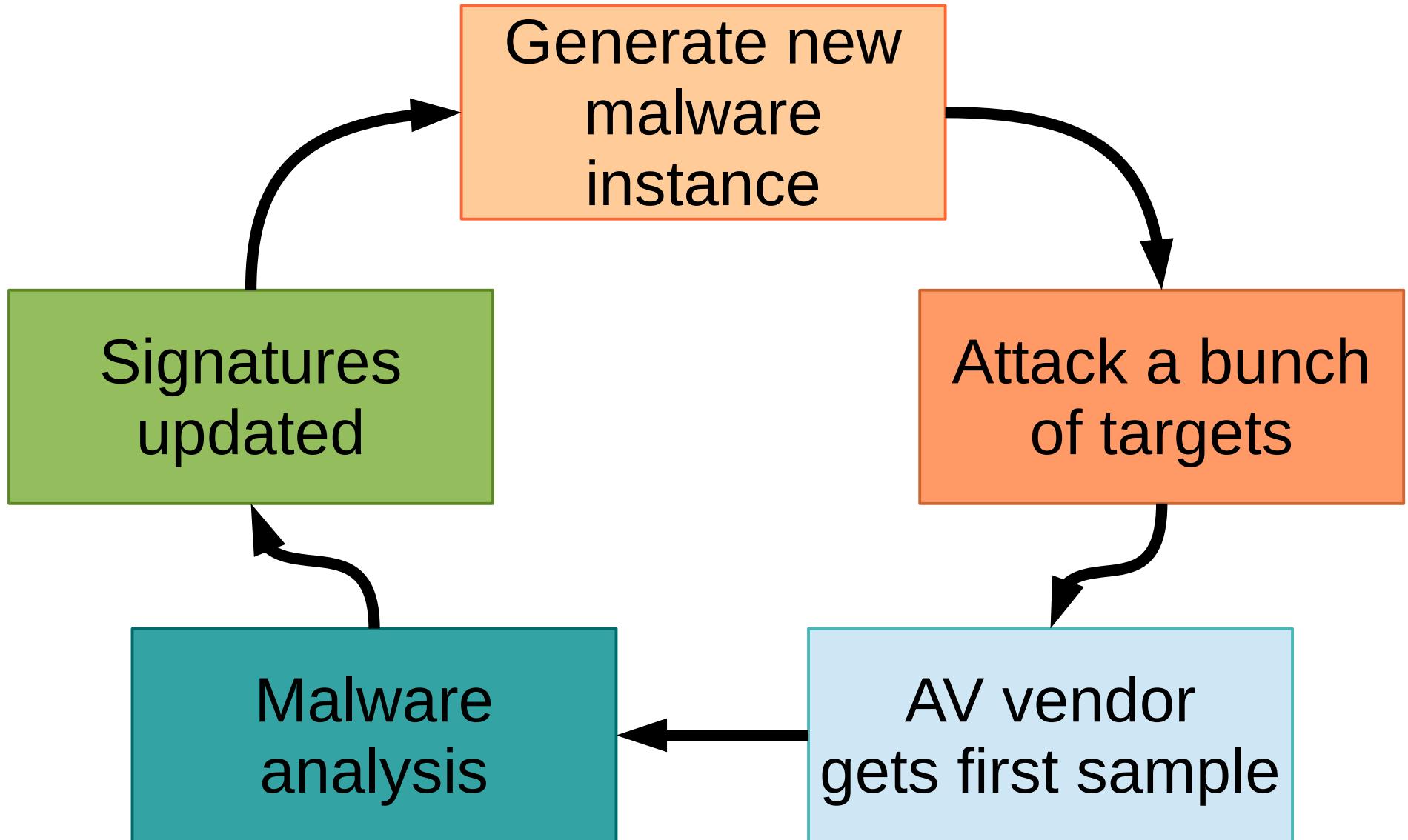


Mathias Payer <[mathias.payer@nebelwelt.net](mailto:mathias.payer@nebelwelt.net)>  
UC Berkeley and (soon) Purdue University

# Malware landscape is changing



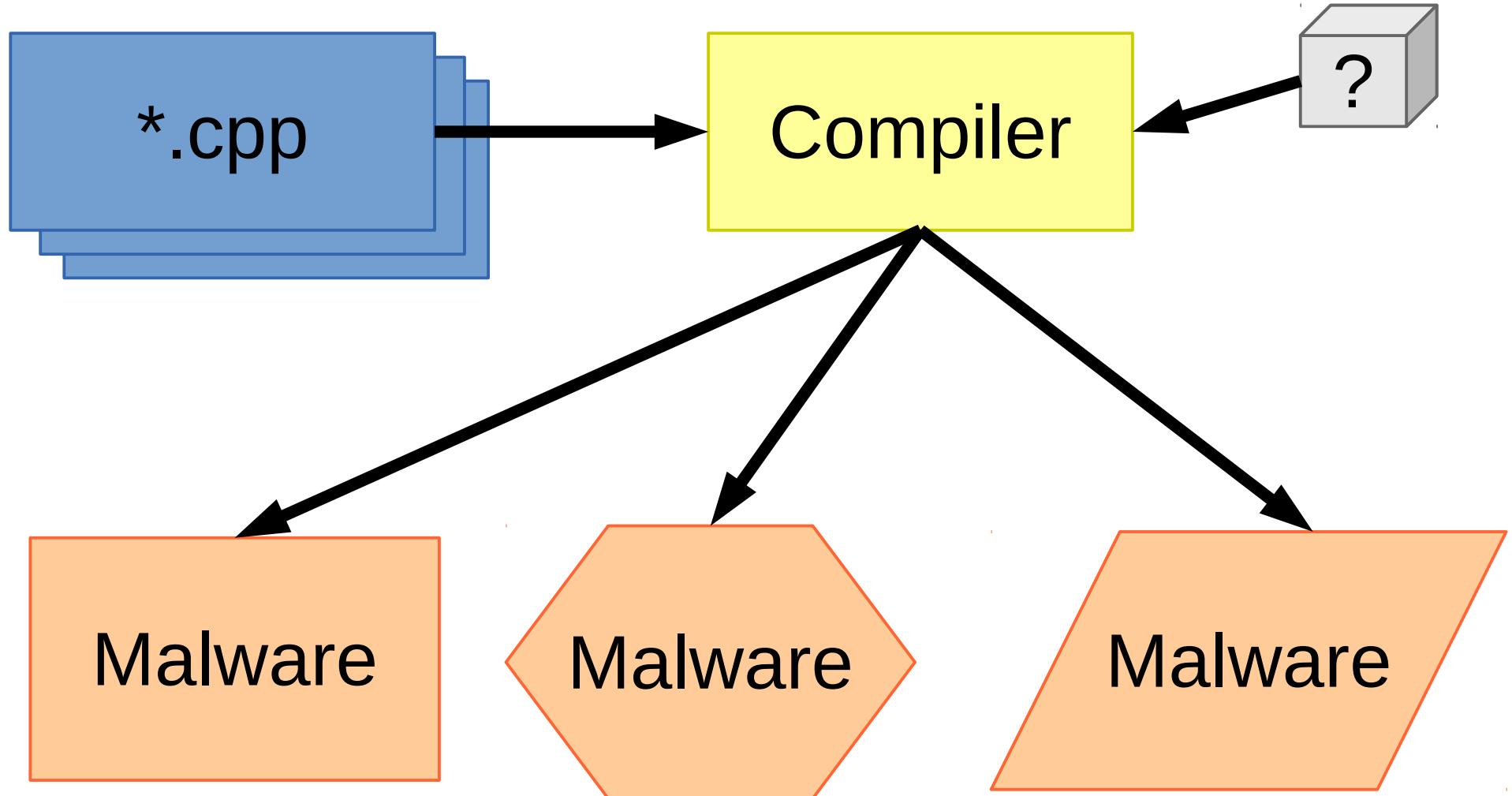
# The ongoing malware arms race



# Defense limitations

- Newly diversified samples are not detected
  - Basically a “new” attack
- New malware spreads fast
  - Time lag between analysis and updated signatures
- Can we automate this process?

# Fully automatic diversity



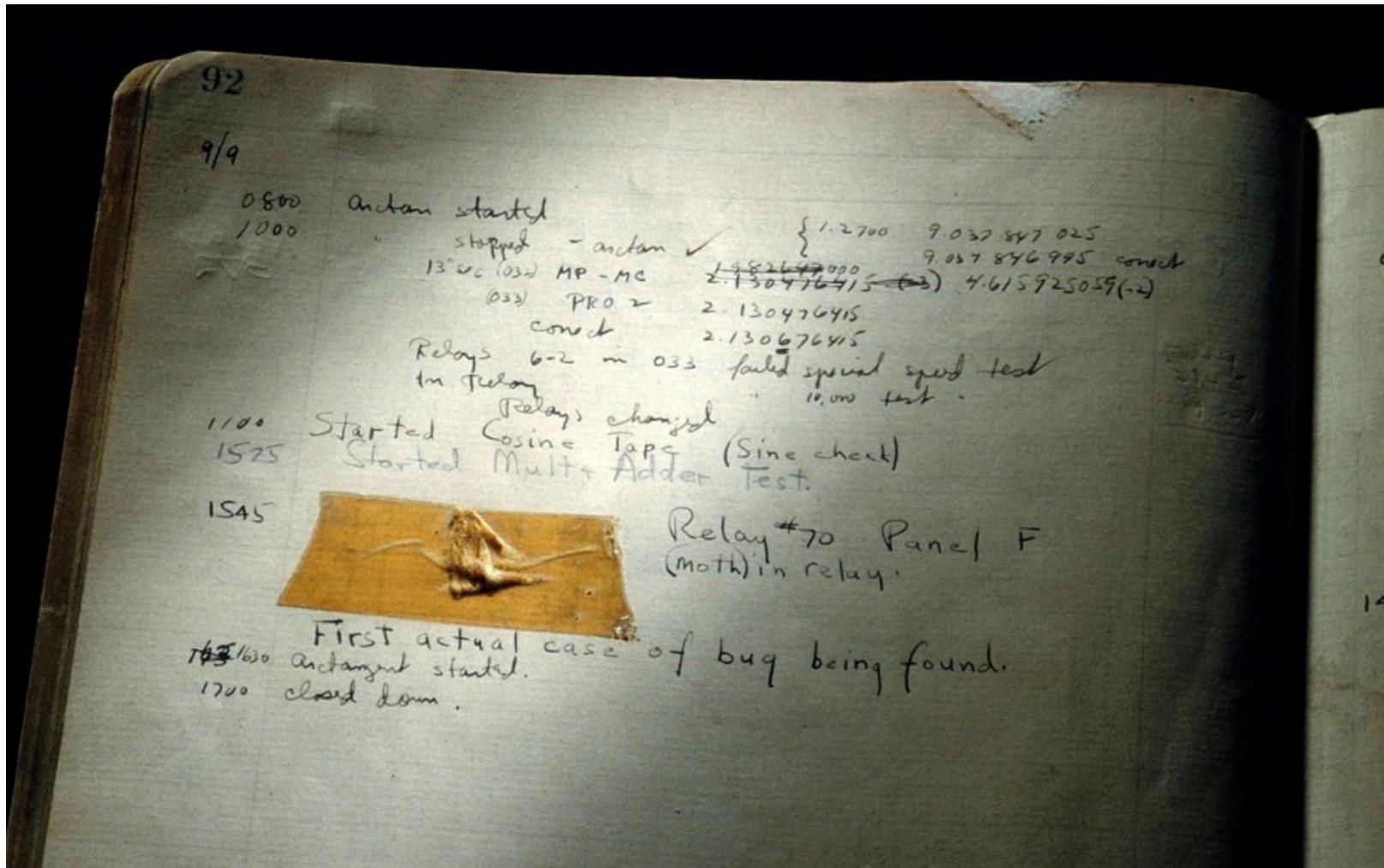
# Outline

State of the art:  
Malware detection

A new threat:  
Malware diversification

Possible mitigation:  
Better security practices

# State of the art: Malware detection



# Malware detection is limited

- Performance
  - Don't slow down a user's machine (too much)
- Precision
  - Behavioral, generic matching
- Latency
  - Time lag between spread and protection

# Detection mechanisms

Бузе



# Signature-based detection

- Compare against database of known-bad
  - Extract pattern
  - Match sequence of bytes or regular expression
- Advantages
  - Fast
  - Low false positive rate
- Disadvantages
  - Precision limited to known-bad samples

# Static analysis-based detection

- Search potentially bad patterns
  - API calls
  - System calls
- Advantages
  - Low overhead
- Disadvantages
  - False positives
  - Based on well-known heuristics

# Behavioral-based detection

- Execute “file” in a virtual machine
  - Detect modifications
- Advantages
  - Most precise
- Disadvantages
  - High overhead
  - Precision limited due to emulation detection

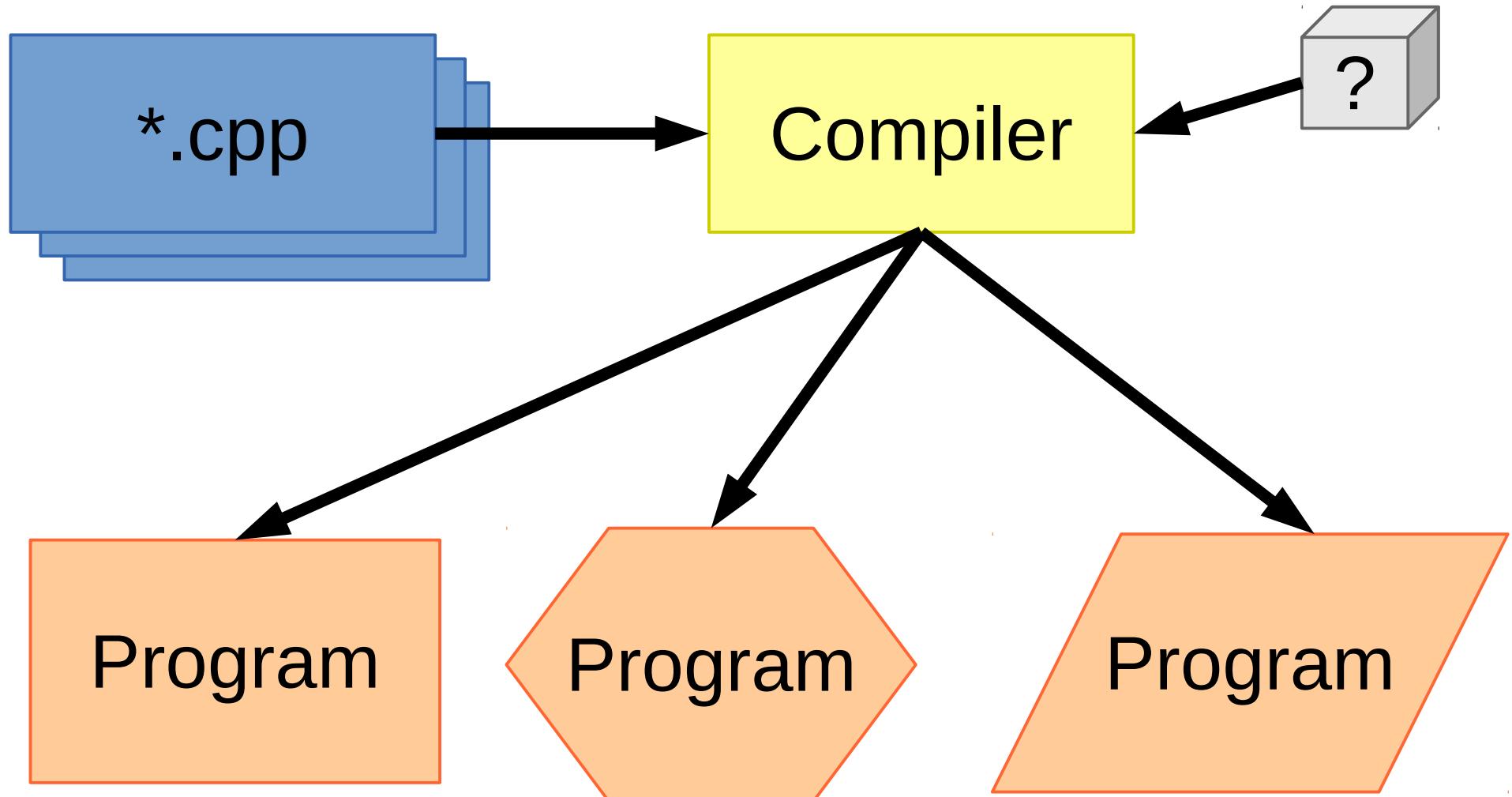
# Summary: Malware protection

- Arms race due to manual diversification
  - Signature-based techniques loose effectiveness
- Cope with limited resources
  - On the target machine, for the analysis, and to push new signatures/heuristics
- No perfect solution
  - Either false positives and/or negatives or huge performance impact

# New threat: Malware diversification



# Software diversification



# C/C++ liberties

- Data layout changes
  - Data structure layout on stack
  - Layout for heap objects (limited for structs)
- Code changes
  - Register allocation (shuffle or starve)
  - Instruction selection
  - Basic block splitting, merging, shuffling

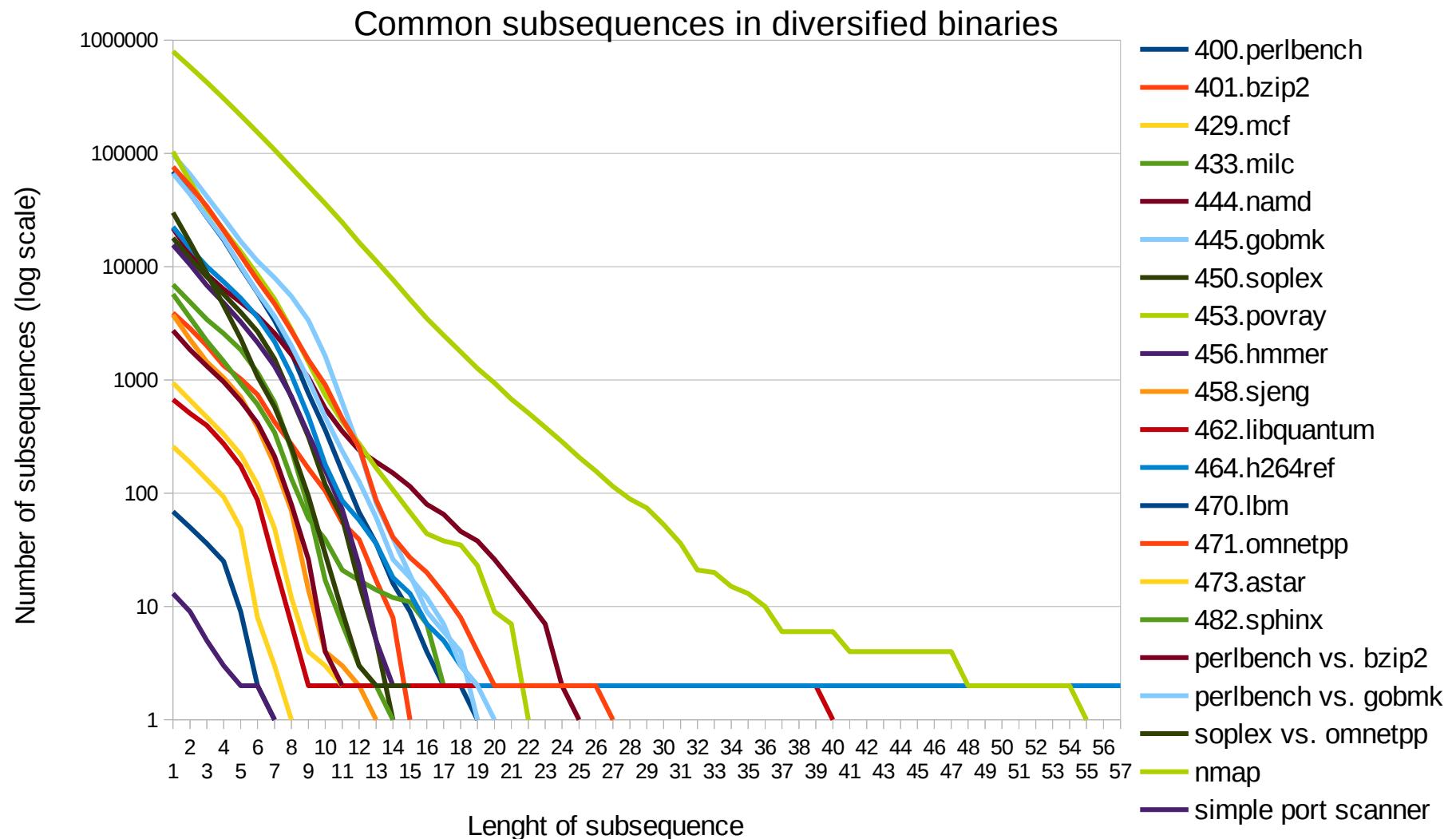
# Malware diversification

- Generate unique binaries
  - Minimize common substrings (code or data)
  - Performance overhead not an issue
- Diversify code and data layout
- Diversify static data as well

# Implementation

- Prototype built on LLVM 3.4
  - Small changes in code generator, code layouter, register allocator, stack frame layouter, some data obfuscation passes
- Input: LLVM bitcode
- Output: diversified binary
- Source: <http://github.com/gannimo/MalDiv>

# Similarity limitations



# Demo

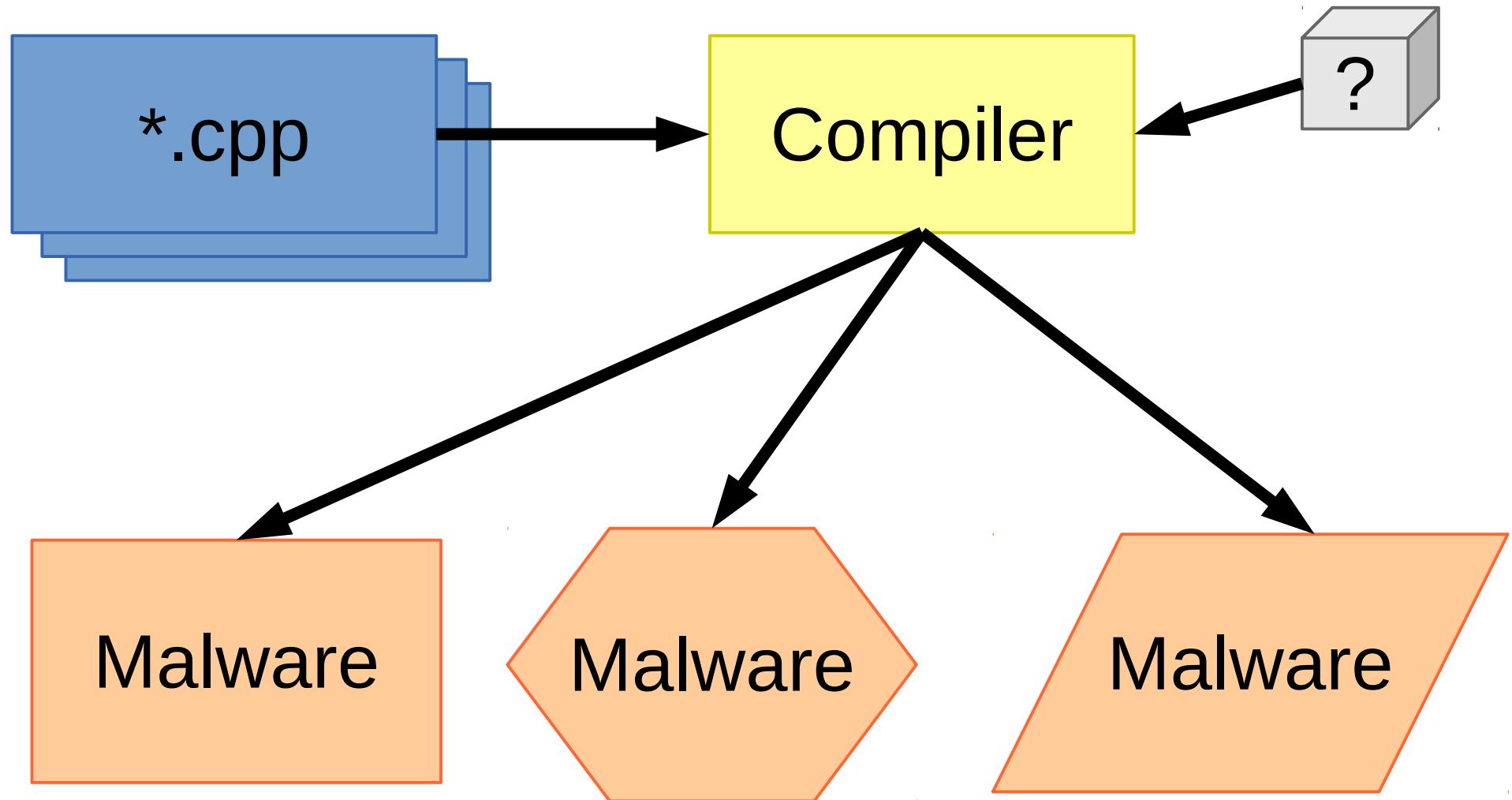
- Simple hello world
  - Let's see how far we can push this!

```
#include <stdio.h, string.h>

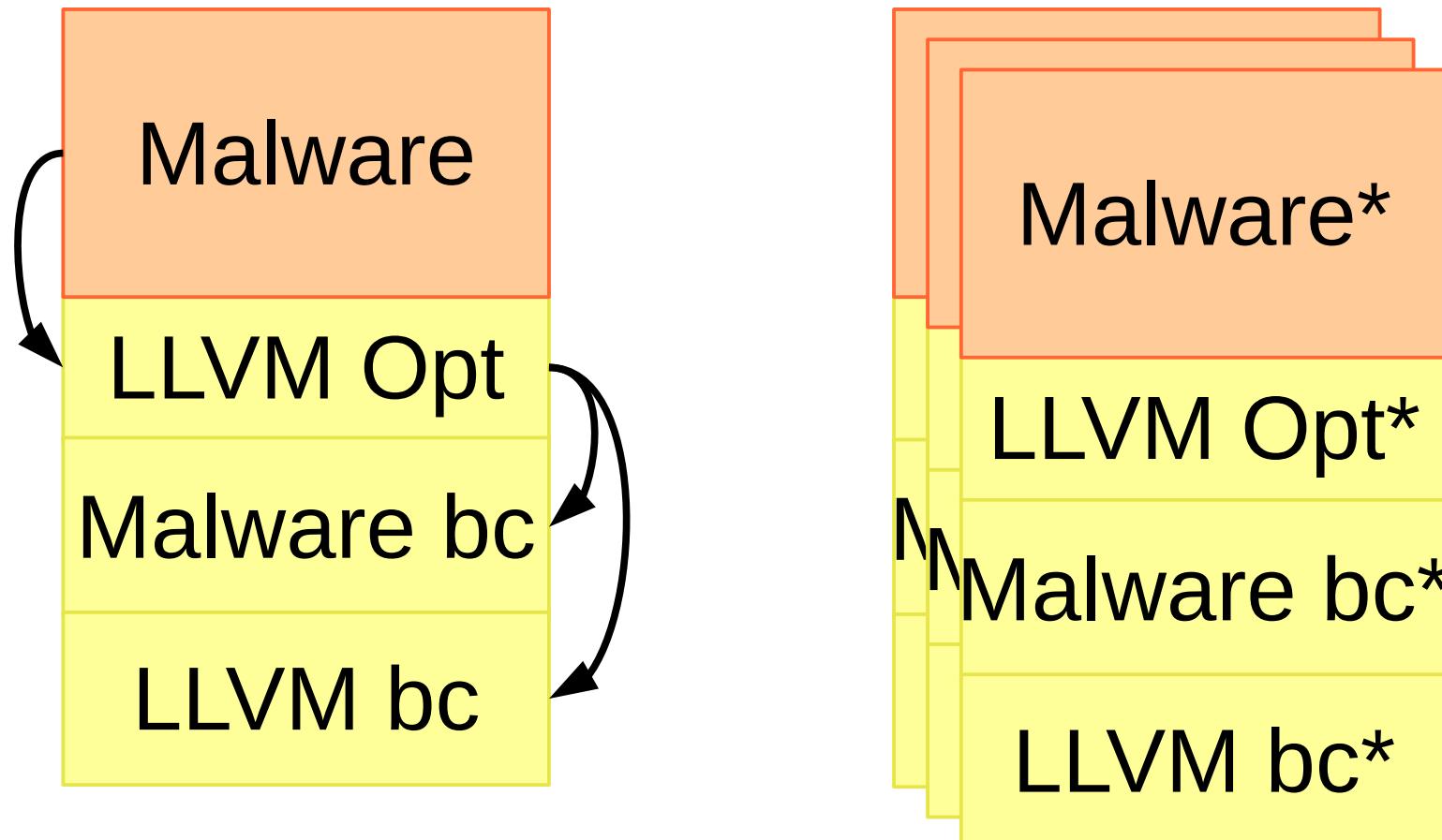
const char foo[] = "foobar";
char bar[7];

int main(int argc, char* argv[]) {
    strcpy(bar, "barfoo");
    printf("Hello World %s %s\n", foo, bar);
    printf("Arguments: %d, executable: %s\n",
           argc, argv[0]);
    return 0;
}
```

# Scenario 1: malware generator



# Scenario 2: self-diversifying MW



# Possible mitigation: Better security practices





# Mitigation

- Recover high-level semantics from code
  - Hard (and results in an arms race)
- Full behavioral analysis
  - Harder
- Prohibit initial intrusion
  - Fix broken software & educate users
  - Hardest

# Conclusion



# Conclusion

- Diversity evades malware detection
  - Fully automatic, built into compiler
  - No need for packers anymore
- Adopts to new similarity metrics
- New arms race between defenders and compiler writers
- Don't rely on simple, static similarity!

# Questions?



Mathias Payer <[mathias.payer@nebelwelt.net](mailto:mathias.payer@nebelwelt.net)>  
Project: <https://github.com/gannimo/MalDiv>  
Homepage: <https://nebelwelt.net>