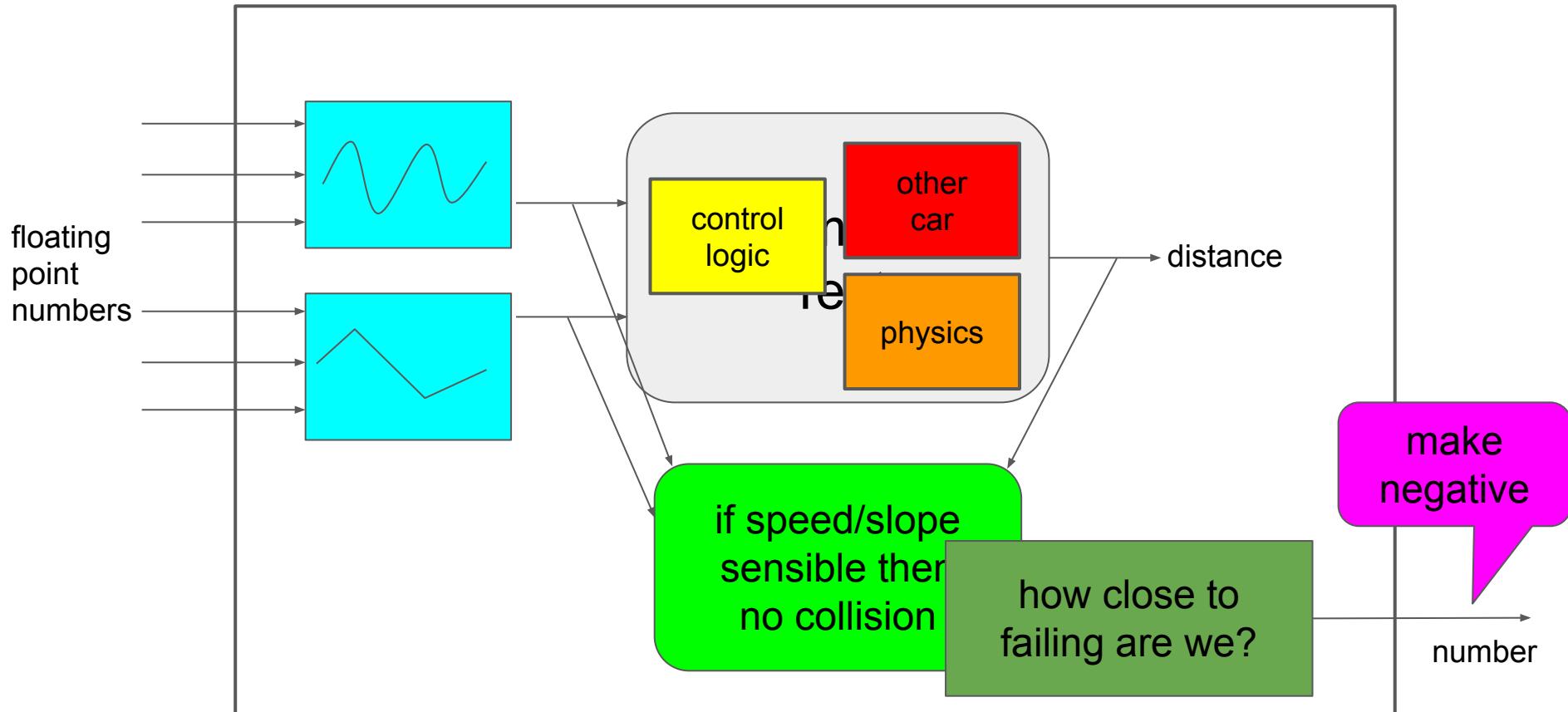


# Numerical Optimization for Generating Test Data

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Sweden

# Example: Adaptive cruise control



floating  
point  
numbers

use numerical  
minimization!

make  
negative

“falsification”

number

needs  
gradient

## Numerical Optimization Methods

- Gradient descent
- Nelder-Mead
- Swarm optimization
- ...

gravitate  
towards local  
minima

- SNOBFIT
- Bayesian optimization
- ...

expensive

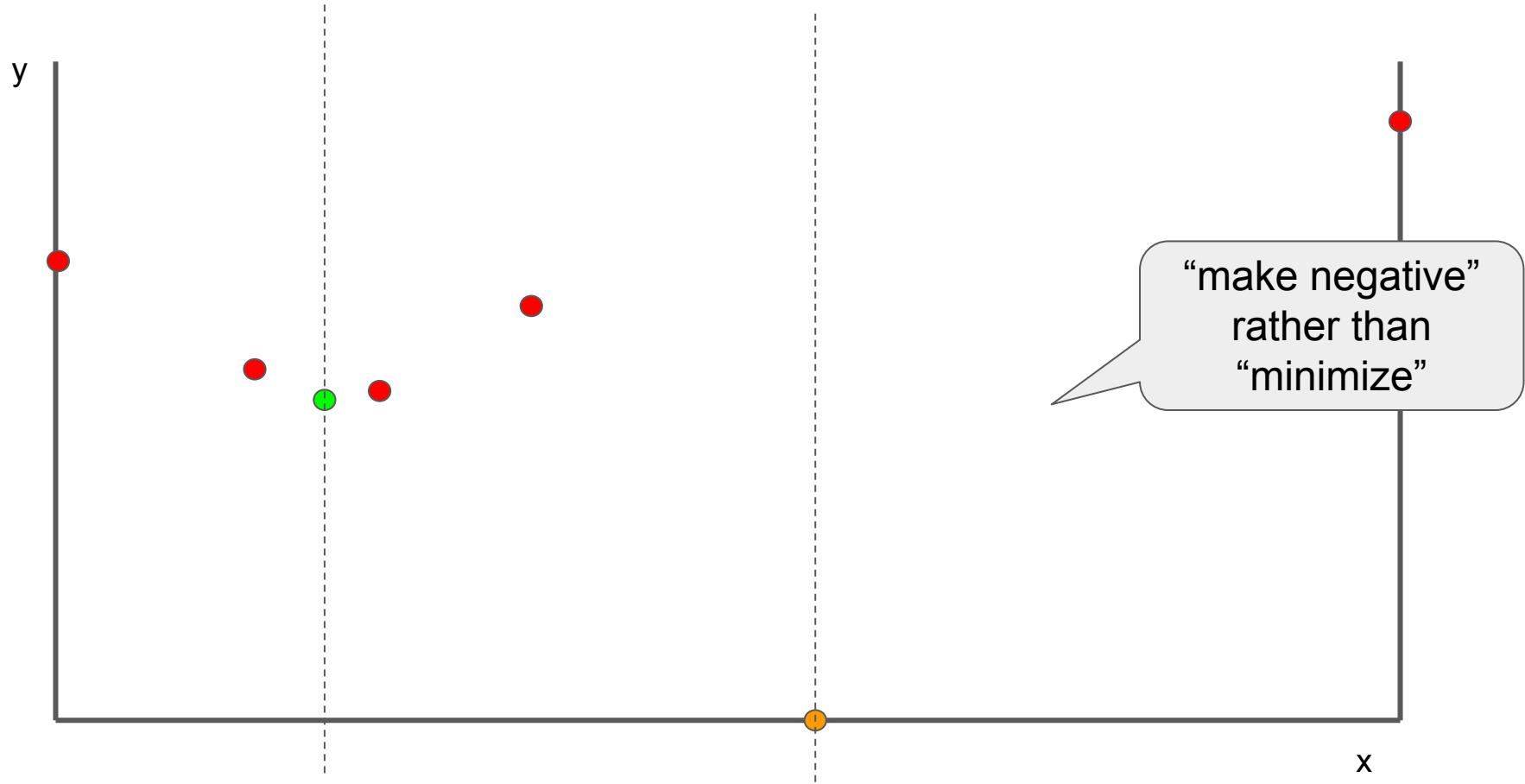
complicated

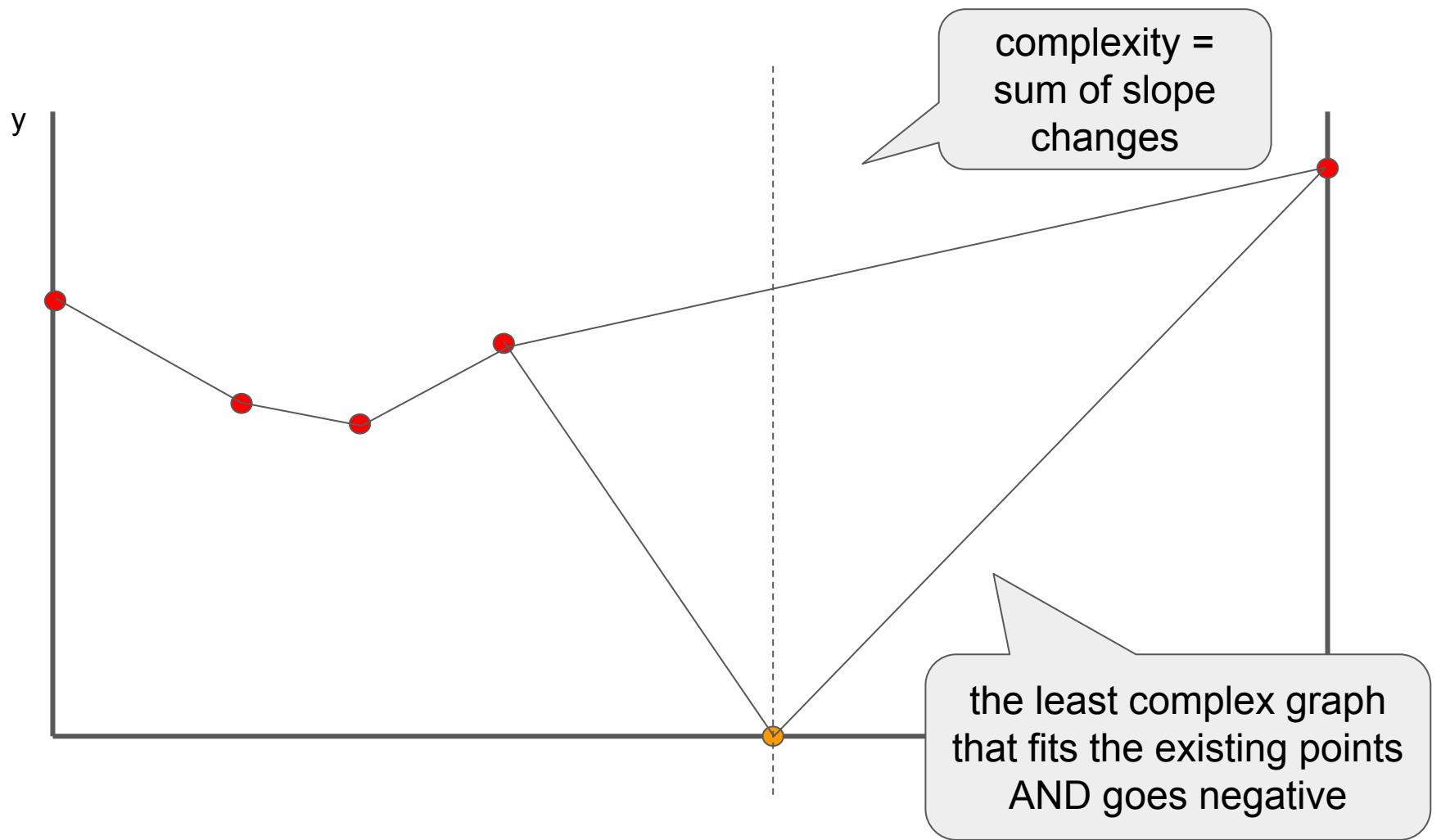
# Which aspects are important for falsification

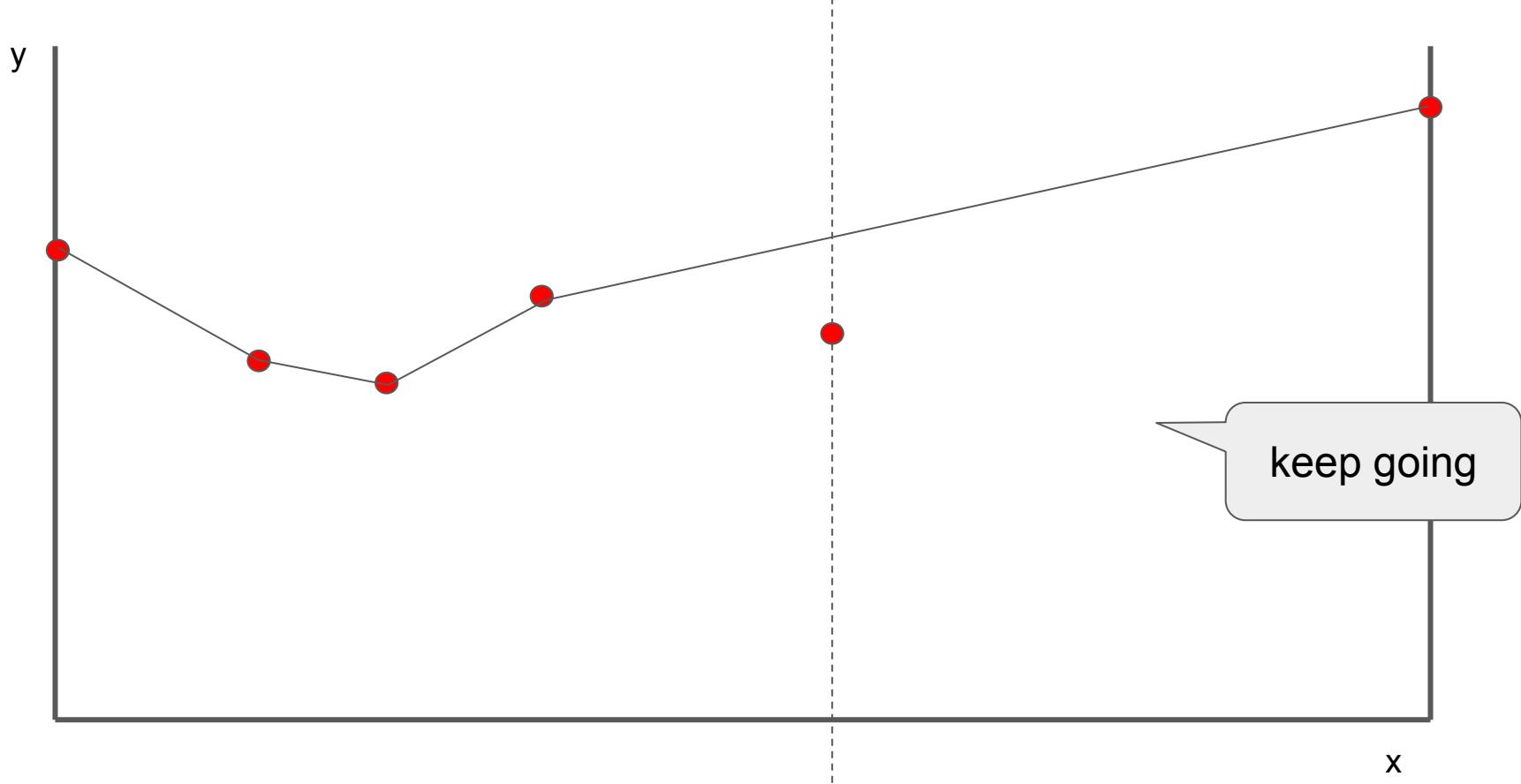
- No gradient
- Not getting stuck in local minima
  - Do not gravitate to local minima
- Having a defined “area” to search in
  - Extreme values often lead to bugs

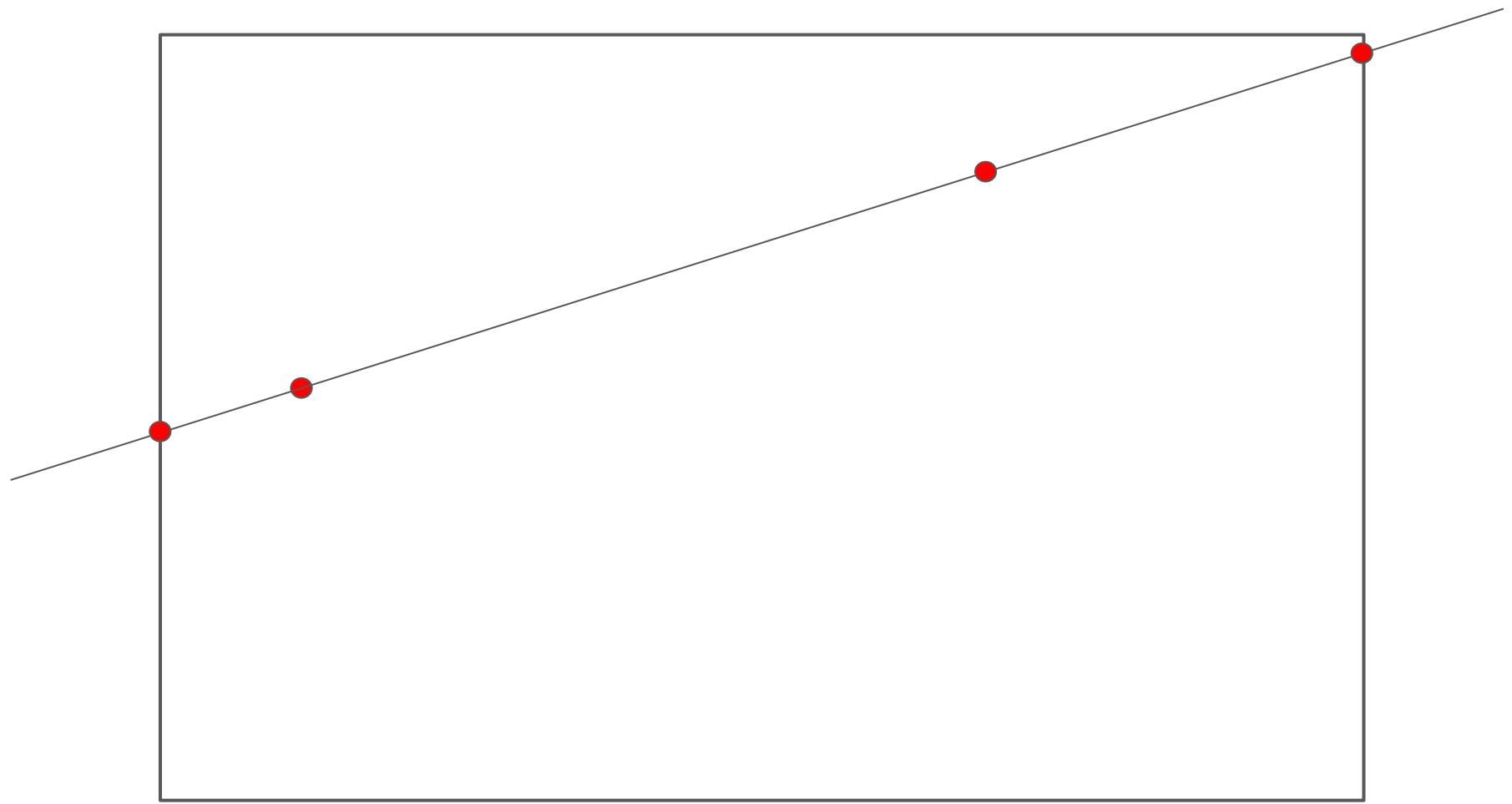
A yellow speech bubble shape with a black outline and a small black tail pointing upwards and to the left. Inside the bubble, the text "find negative results" is written in a black, sans-serif font.

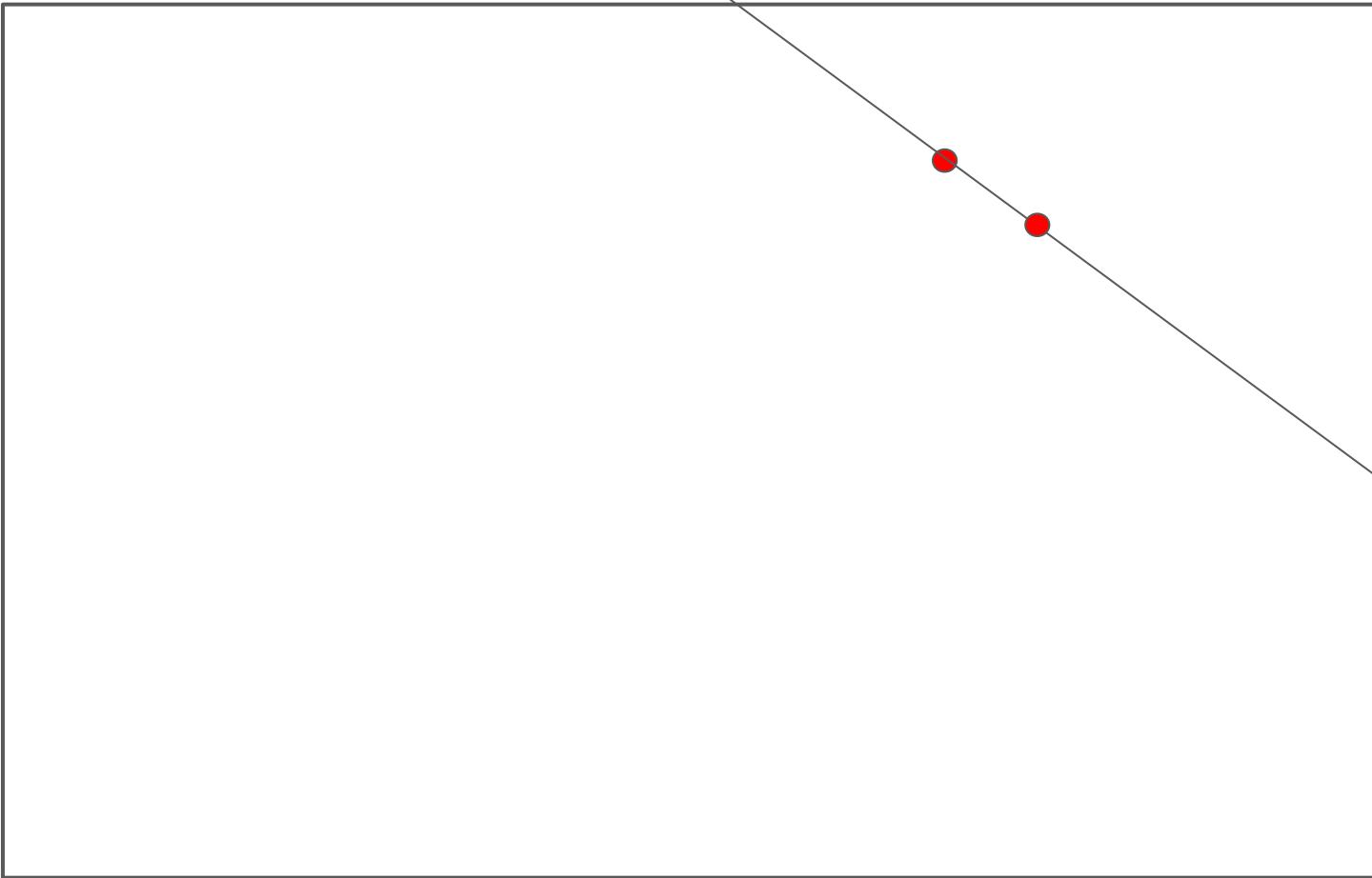
find **negative**  
results

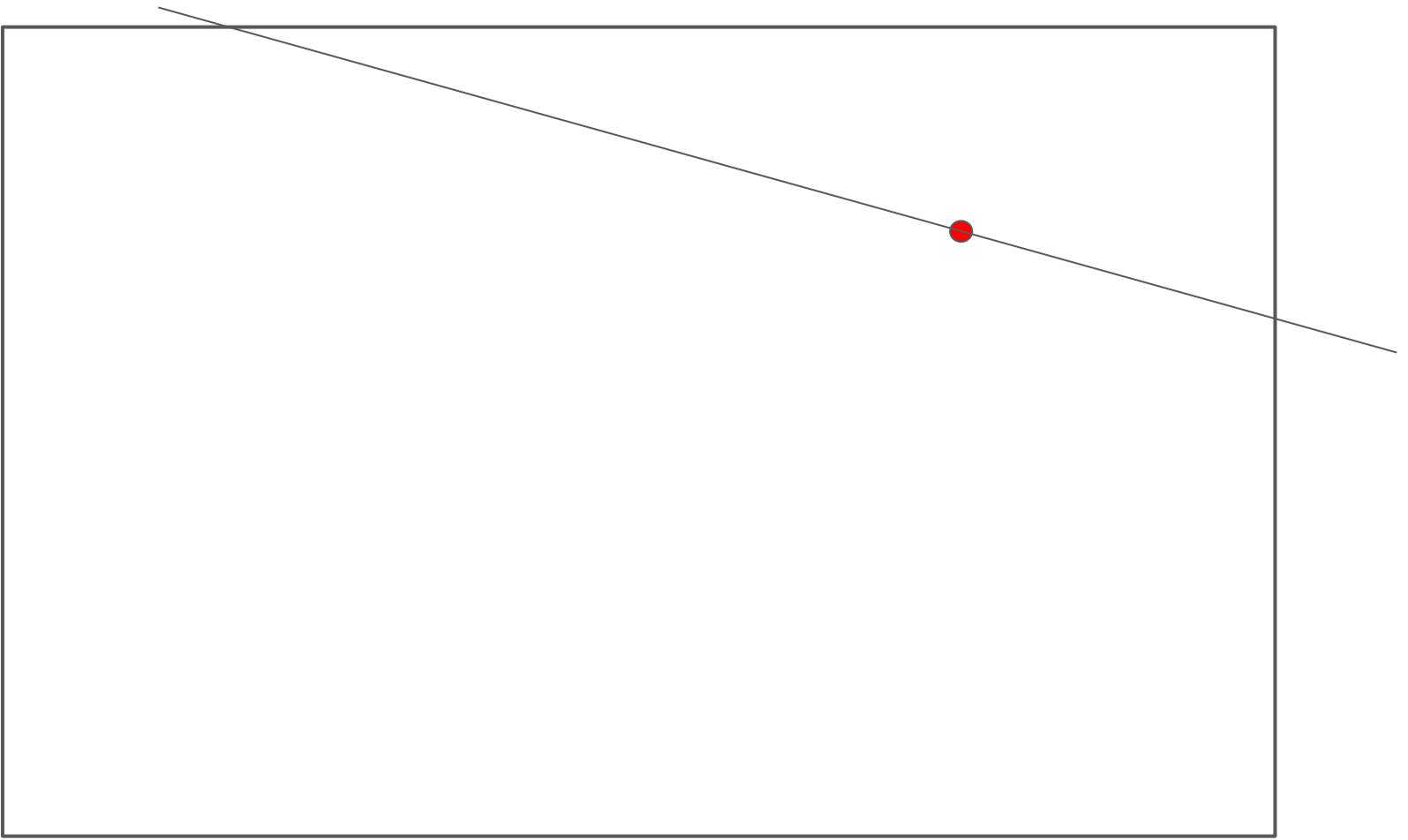






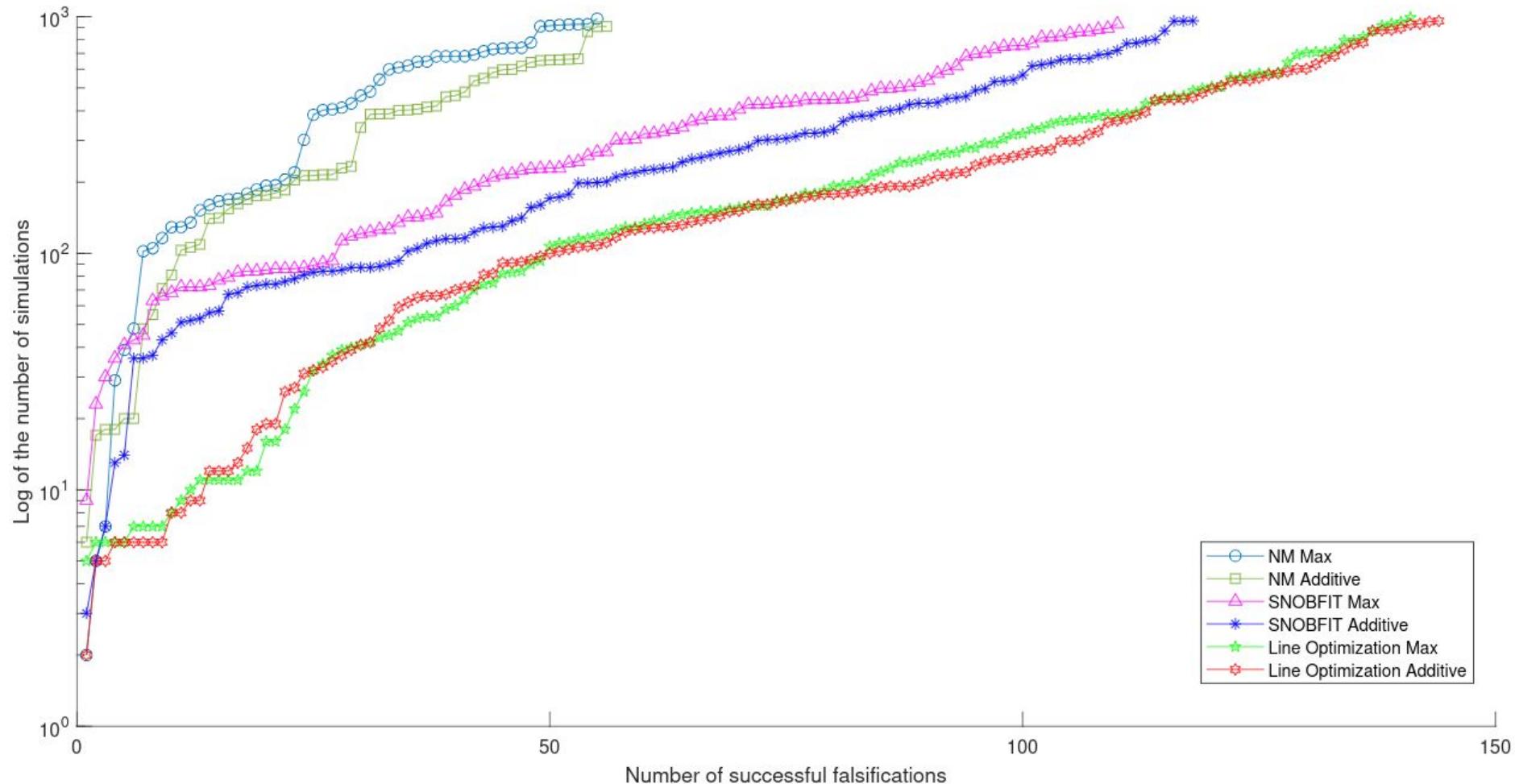




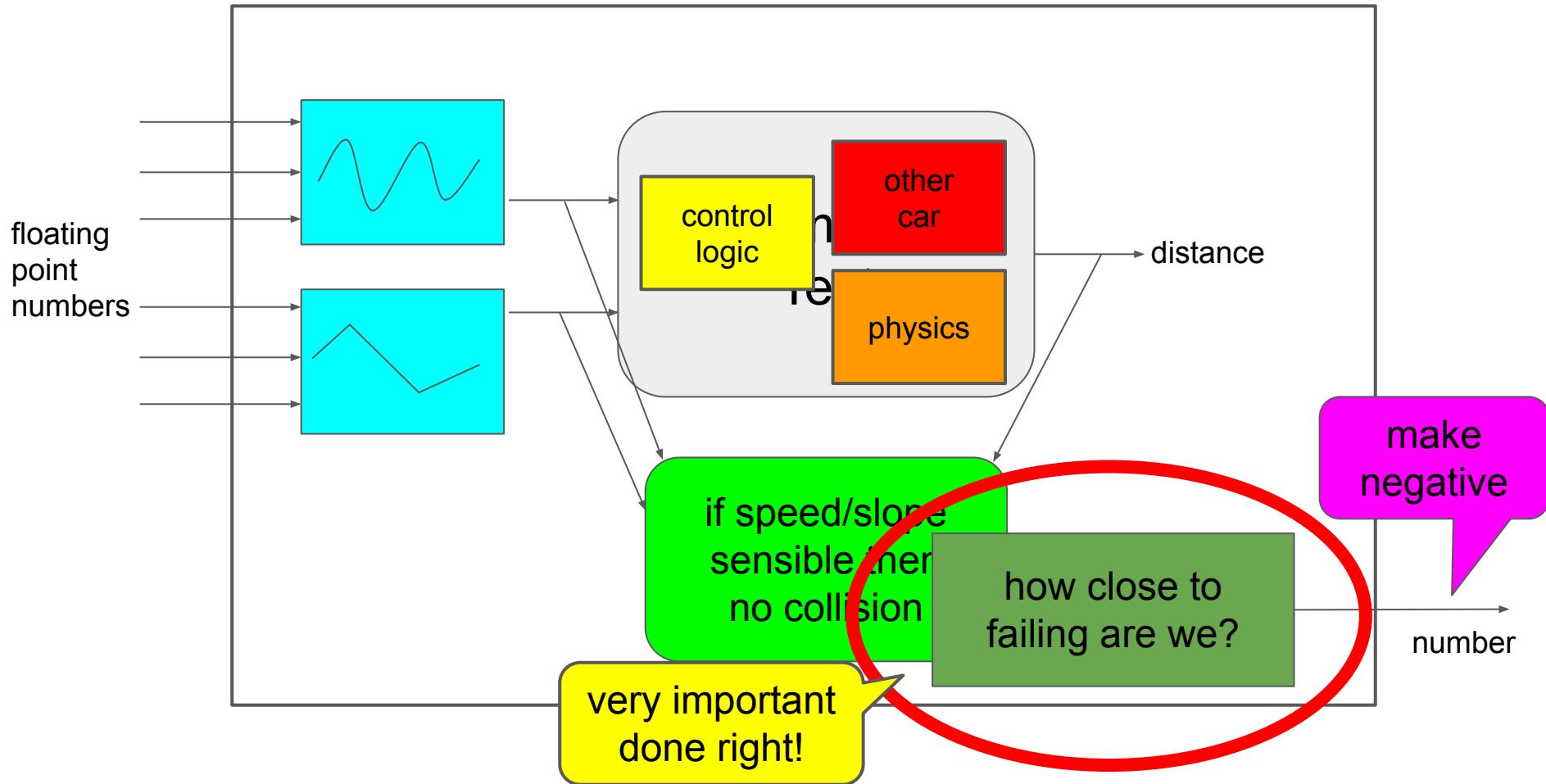


## “Line falsification”

- Simple, fast
- Exercises “extreme values” in the box
- Is not sensitive to local minima
- Is not very sensitive to #dimensions
  - Can ignore dimensions that don’t seem to matter
  - 100s of inputs



# Example: Adaptive cruise control



```
type DBool = Double {- >=0 -}
```

```
false, true :: DBool  
false = inf  
true = 0
```

```
(&&), (||) :: DBool -> DBool -> DBool  
x && y = x + y  
x || y = x `min` y
```

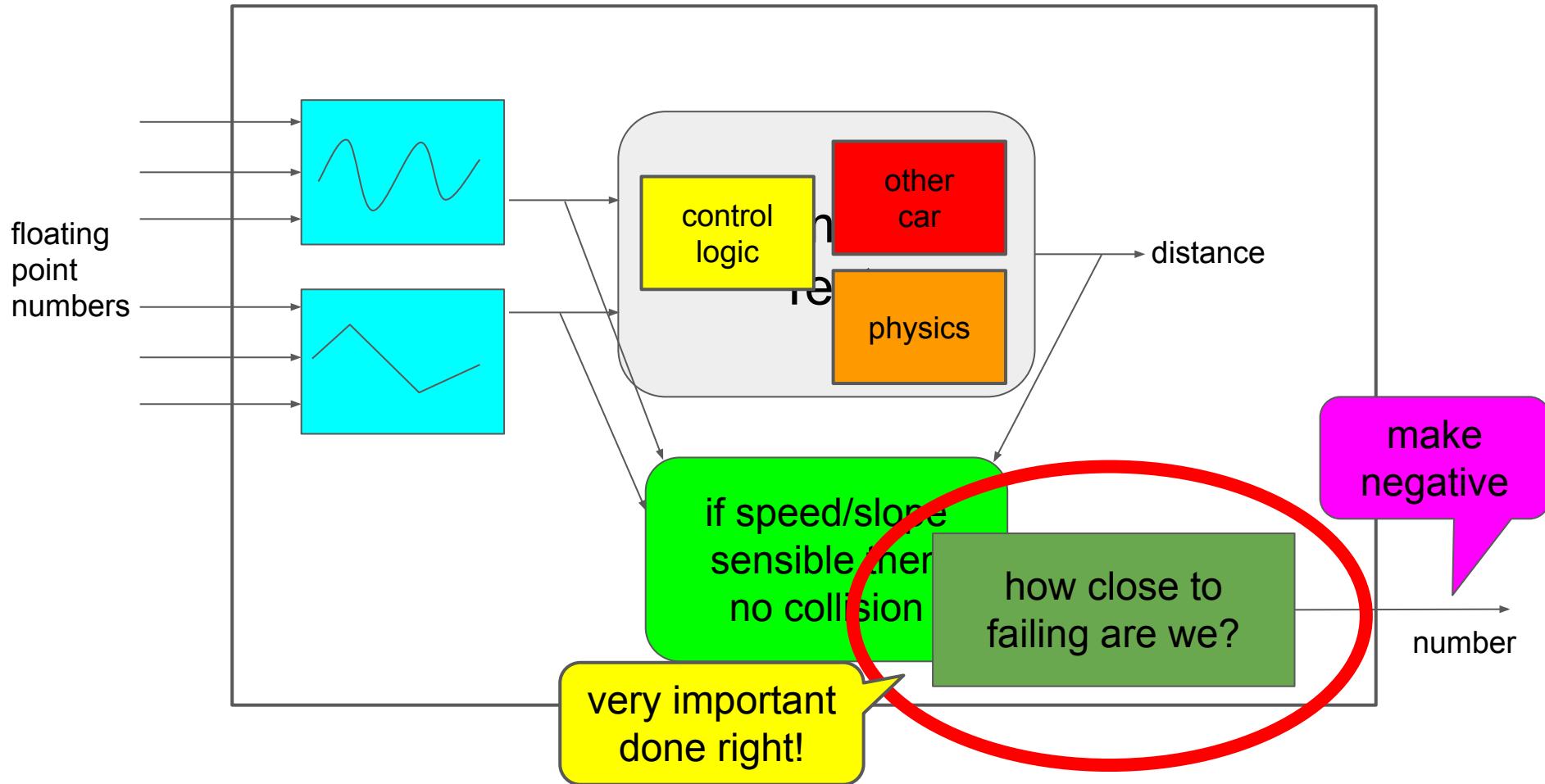
```
type DBool = Double {- >=0 -}
```

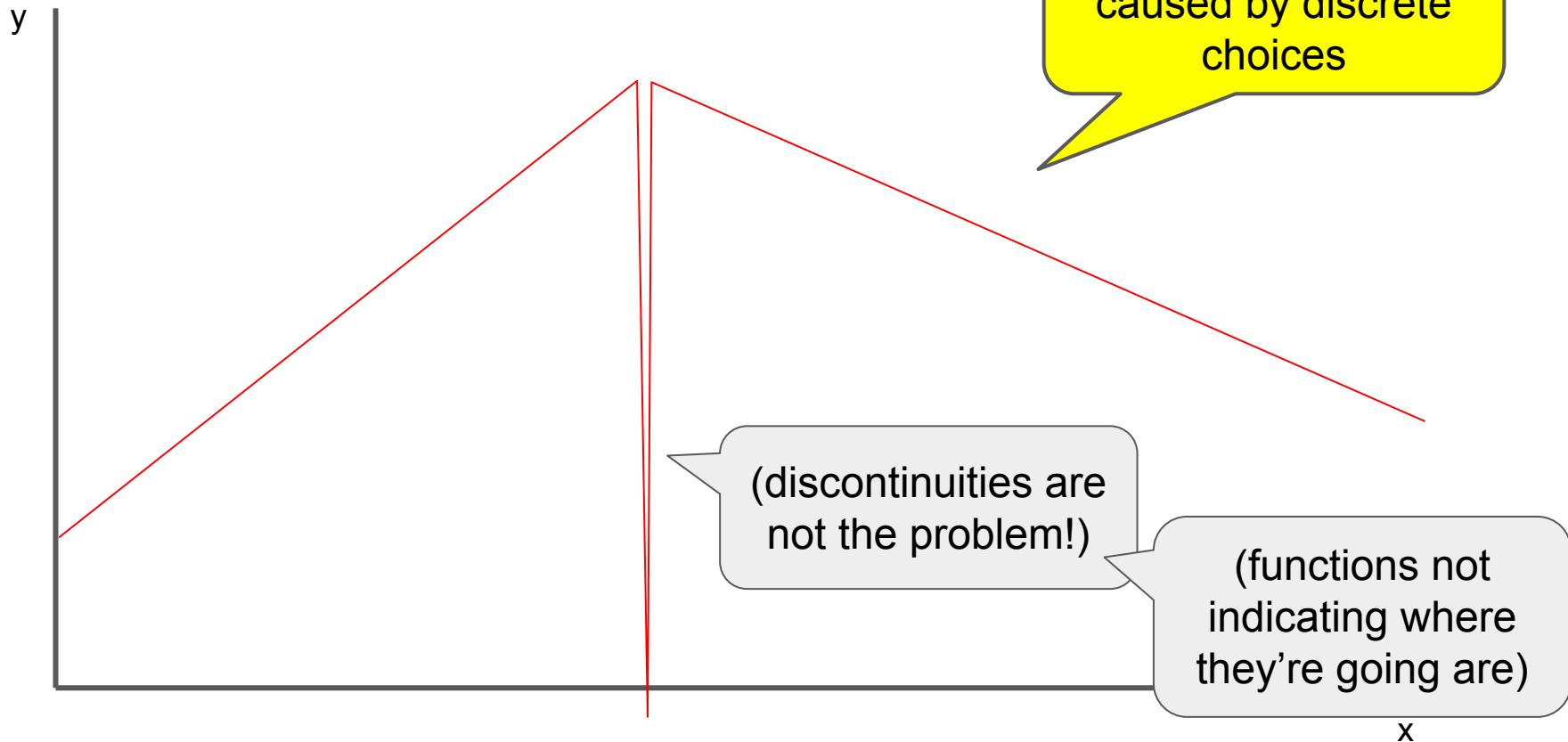
```
(>=? ) :: Double -> Double -> DBool  
x >=? y  
| x >= y      = true  
| otherwise    = y - x
```



a specification logic  
of “valued booleans”

# Example: Adaptive cruise control





type for representing  
simulation values

a table of *alternative*  
values, and *distance*  
to them

```
data Val a = Val a [(a,DBool)]
```

the actual  
value

**instance Applicative Val**

**pure :: a -> Val a**

**pure x = Val x []**

## instance Applicative Val

```
lift2 :: (a -> b -> c) ->  
        Val a -> Val b -> Val c
```

lift2 f (Val x xds) (Val y yds) = Val (f x y) zds

where

```
zds = table (  
    [ (f x y', d) | (y', d) <- yds ] ++  
    [ (f x' y, d) | (x', d) <- xds ] ++  
    [ (f x' y', d1+d2) | (x', d) <- xds  
      , (y', d) <- yds ] )
```

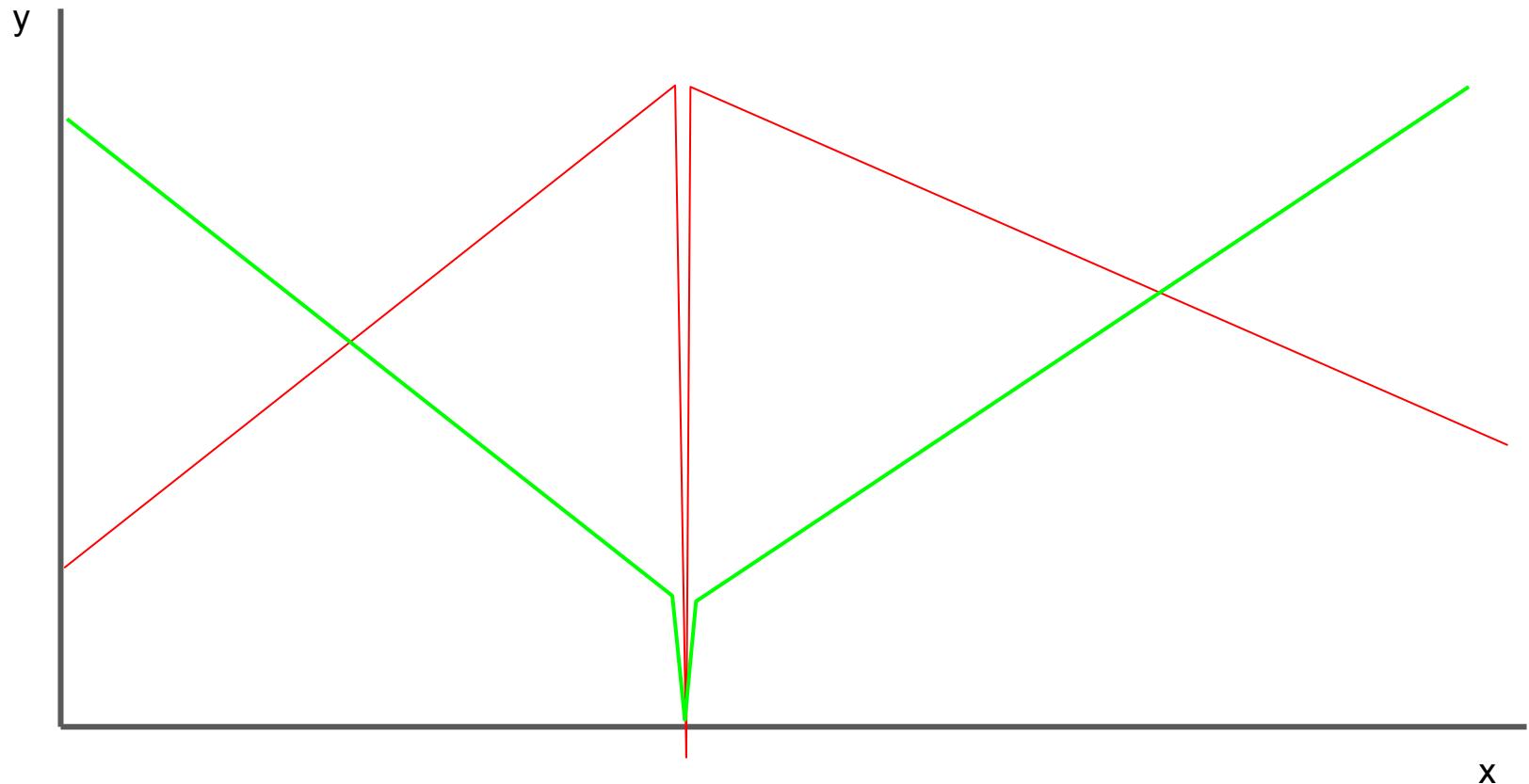
table chooses the  
minimum distance

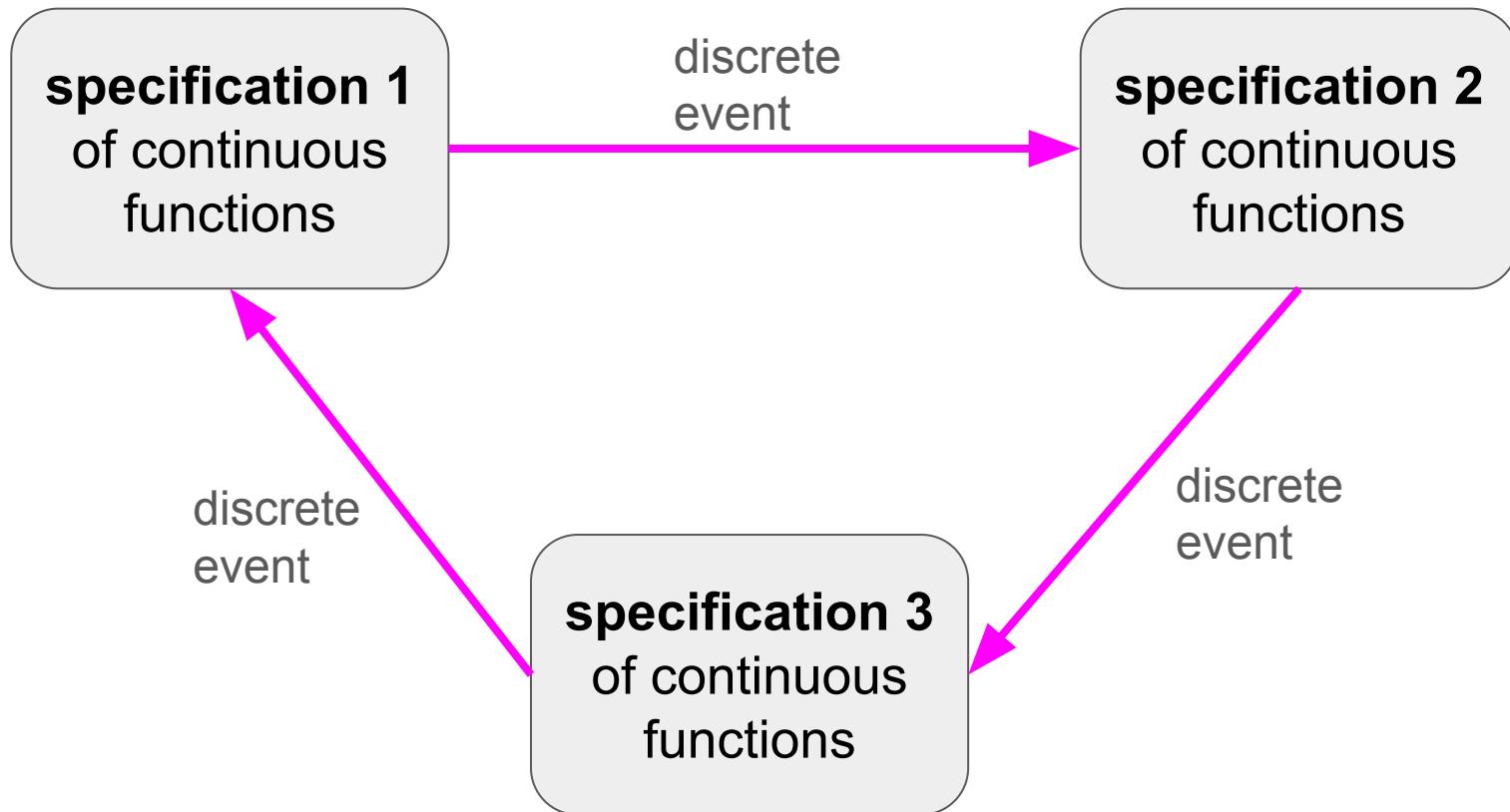
```
ifThenElse :: Val Bool ->  
           Val a -> Val a -> Val a
```

plot the distance to  
a negative result

```
f :: Val Double -> Val Double
f x = ifThenElse (x <=? 10.0)
           x
           (ifThenElse (x >=? 10.1)
             (20-x)
             (-1))
```

want to automate  
instrumentation





```
data State = Spec1 | Spec2 | Spec3
```

current state +  
**alternative states**

continuous inputs

property ok?

```
system :: Val State -> [Double] -> Val Bool  
system st xs = ... system st' xs' ...
```

# Summary

- Very effective method for finding bugs in hybrid systems
  - flexible (e.g. math / black-box)
- “Additive” valued booleans
- Our own numerical optimization method
  - good at finding negative values (vs. minimization)
- How about fully discrete systems?

compares well in typical benchmarks

used in physics!  
(lasers)

```
parse :: String -> Maybe T  
show   :: T -> String
```

```
prop_parse_show (x :: T) =  
  parse (show x) == Just x
```

(show should really  
be non-deterministic)

```
prop_show_parse (s :: String) =  
  let pres = parse s in  
    isJust pres ==>  
      s == show (fromJust pres)
```

need to check that  
we do not parse too  
much!

implies

```
prop_show_parse' (x :: T) =  
  let s = show x in  
    let pres = parse s in  
      isJust pres ==>  
        s == show (fromJust pres)
```

```
parse q d (c:s) =
  case c of
    '(' | q == 0 ->
      parse 0 (d+1) s

    '+' | q == 1 || q == 2 ->
      parse 0 d s

    ')' | q == 1 || q == 2 ->
      parse 2 (d-1) s

    _ | '0' <= c && c <= '9' && (q == 0 || q == 1) ->
      parse 1 d s

    _ ->
      False

parse q d [] =
  d == 0 && (q == 1 || q == 2)
```

```
parse q0 q1 q2 d (c:s) =
  case c of
    '(' | q0 ->
      parse True False False (d+1) s

    '+' | q1 || q2 ->
      parse True False False d s

    ')' | q1 || q2 ->
      parse False False True (d-1) s

    _ | '0' <= c && c <= '9' && (q0 || q1) ->
      parse False True False d s

    _ ->
      False

parse q0 q1 q2 d [] =
  d == 0 && (q1 || q2)
```

```
parse q0 q1 q2 d (c:s) =
  parse q0' q1' q2' d' s
where
  q0' = (c == '(' && q0) || (c == '+' && (q1 || q2))
  q1' = '0' <= c && c <= '9' && (q0 || q1)
  q2' = (c == ')') && (q1 || q2))
  d' = d + one (c == '(' && q0) - one (c == ')') && (q1 || q2))
```

```
one False = 0
```

```
one True = 1
```

```
parse q0 q1 q2 d [] =
  d == 0 && (q1 || q2)
```

```
parse :: Val Bool -> Val Bool -> Val Bool -> Val Int -> [Val Char]  
      -> Val Bool
```

```
parse q0 q1 q2 d (c:s) =  
  parse q0' q1' q2' d' s  
  where  
    q0' = ((c ==. '(') &&. q0) ||. ((c ==. '+') &&. (q1 ||. q2))  
    q1' = foldr1 (||.) [ c ==. w | w <- ['0'.. '9'] ] &&. (q0 ||. q1)  
    q2' = ((c ==. ')') &&. (q1 ||. q2))  
    d' = d + one ((c ==. '(') &&. q0) - one ((c ==. ')') &&. (q1 ||. q2))  
  
    one q | q > 0.1 = 0  
    one _ = 1  
  
parse q0 q1 q2 d [] =  
  zero d &&. (q1 ||. q2)
```

done manually but  
systematically



208: 618.606774397607 "(7,79\*5)+(9)'\*|<08P/"  
209: 615.6372863875167 "(7+8:\*5)+)9))\*)|<09P/"  
225: 613.5816522393113 "(119:12,2\*5)(0vF08J-"  
226: 611.7831431512103 "(119:12+3\*5))0vE08J-"  
237: 590.0426580236012 "(106834+2,6' 0v@4;H\*"  
244: 563.5745831260235 "(903930,3+4)+-x:<>D+"  
245: 555.7315838498512 "(903930,3+5)+.x:<>D\*"  
246: 538.8363549593371 "(901730-2+5)+-u:=>D+"  
247: 536.3424754714836 "(40+862-1+3)(-s:<>C+"  
250: 531.1760065113687 "(50,643+304+),s9=?A,"  
253: 509.40062523136993 "(82+2750311-\* ,s;9@A+"  
254: 497.4706664838121 "(52+2750100.\*+s:8@B+"  
257: 497.23938144030933 "(33,1715330)++s:8?@\*"  
258: 495.7613994906361 "(33,1715331)++t:8>B\*"  
259: 487.44509852135127 "(33,2704421))+t:7?A\*"  
273: 471.5657432105902 "(57+792784/(+.p9:@;)"  
275: 463.08959029546634 "(58+6927840(+-p8:@;)"  
278: 457.02702628397947 "(58+692686.)+.p9:?:;"  
280: 418.8725115002297 "(98+5726961)+,n8>73(" "  
303: 416.5640775970895 "(99+6418960\*(+o8<43-"  
309: 410.9369027686399 "(98+8565850+()n9>50."  
310: 408.6283272354027 "(97+7564861+()n9>50."  
...

...

1176: 2.7057640843563036 "7+6773+420+(93497+9,"  
1180: 2.645749153303157 "7+6773+420+(93497+9,"  
1183: 2.5125838864933314 "7+6773+420+(93497+9,"  
1184: 2.4164023100685768 "7+6773+420+(93497+9,"  
1185: 2.3587784555340363 "7+6773+420+(93497+9,"  
1188: 2.239741301366365 "7+6773+420+(93497+9,"  
1189: 2.226287007742286 "7+6773+420+(93497+9,"  
1190: 2.1885096509571014 "7+6773+420+(93497+9,"  
1191: 1.924813066798606 "7+6773+420+(93497+9,"  
1193: 1.8706175204849131 "7+6773+420+(93497+9,"  
1194: 1.867332592042274 "7+6773+420+(93497+9,"  
1195: 1.8038501729502485 "7+6773+420+(93497+9,"  
1196: 1.5728463911464843 "7+6773+420+(93497+9,"  
1203: 1.5092923931476605 "7+6773+420+(93497+9,"  
1205: 1.132590624093993 "7+6773+420+(93497+9,"  
1208: 1.1229937642522572 "7+6773+420+(93497+9,"  
1209: 1.119287494716147 "7+6773+420+(93497+9,"  
1215: 1.0790638124029215 "7+6773+420+(93497+9,"  
1216: 0.0 "7+6773+420+(93497+9+"

# How about symbolic evaluation (using SAT/SMT)?

- We can handle more complicated arithmetic
- We can abstract away completely over unknown functions
  - black-box
- Every run is an actual run
- State space exploration can explode
- Clearly a complement to random testing

finding bugs  
in type systems

(demo)

<https://ifc-challenge.appspot.com/>



The Information Flow Control Challenge consists of 10 challenges to leak the secret in the face of increasingly hardened information flow control mechanisms.

[Start the IFC Challenge](#)

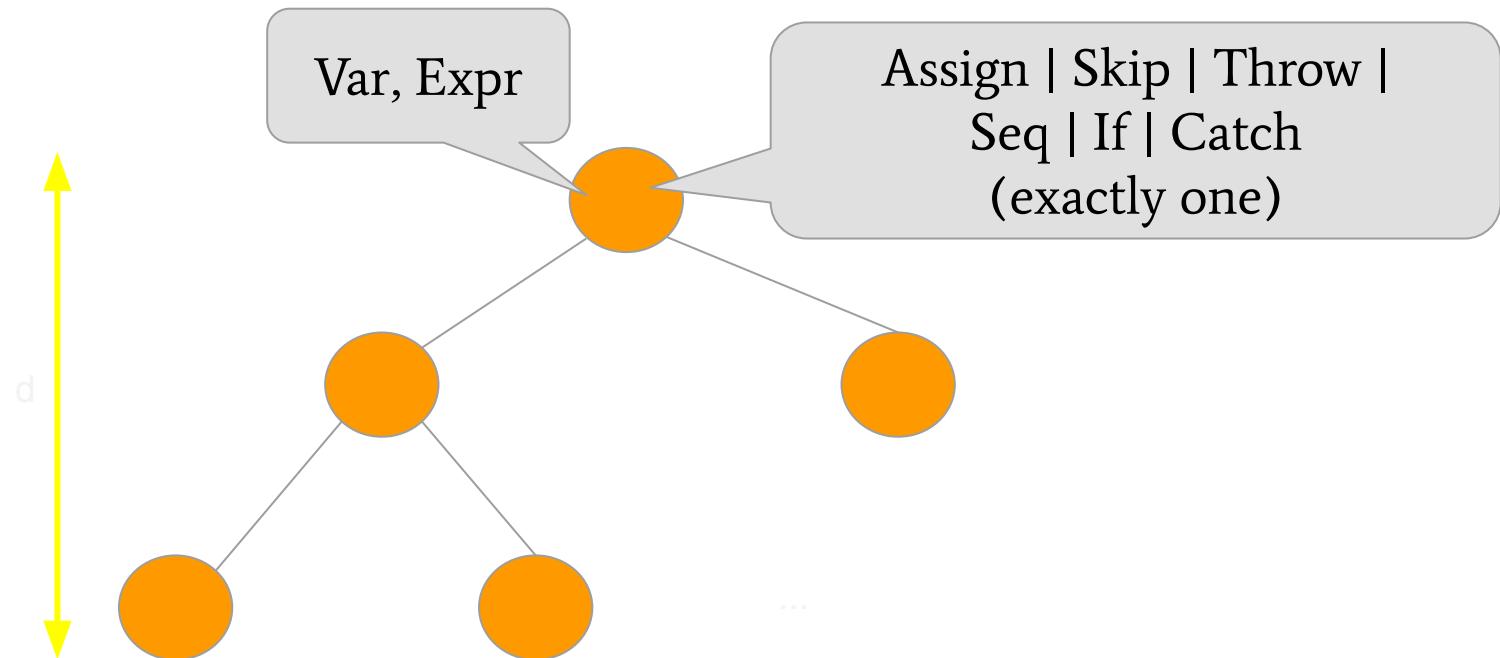
# SAT

scalable

constraints &  
variables over  
finite domains

other theories  
via SMT

# IFC Challenge in SAT (take 1)



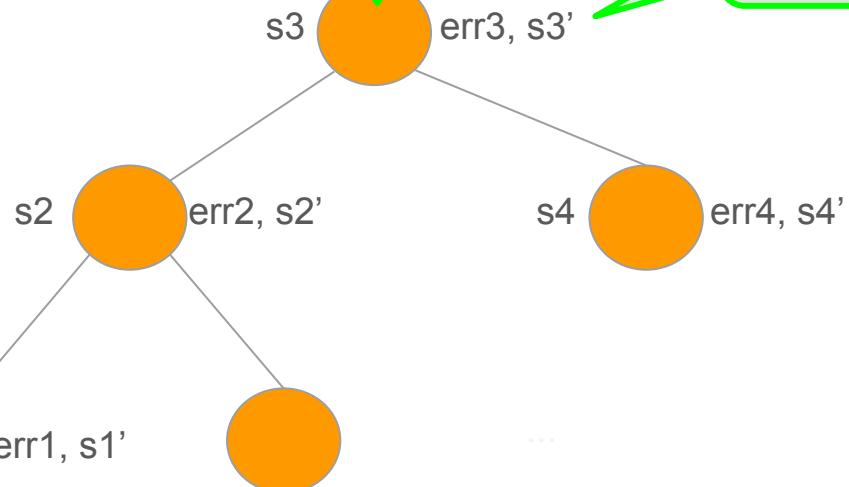
# IFC Chal

(take 1)

1 run

add constraints about  
s2,err2,s3,...,err4,s4'

say that  $s3'[l] = s3[h]$



all variables

add for all  
runs

a few milliseconds

solve

# IFC Challenge in SAT (take 1)

solver knows  
about types

solver knows  
about semantics

trees are  
**wasteful**

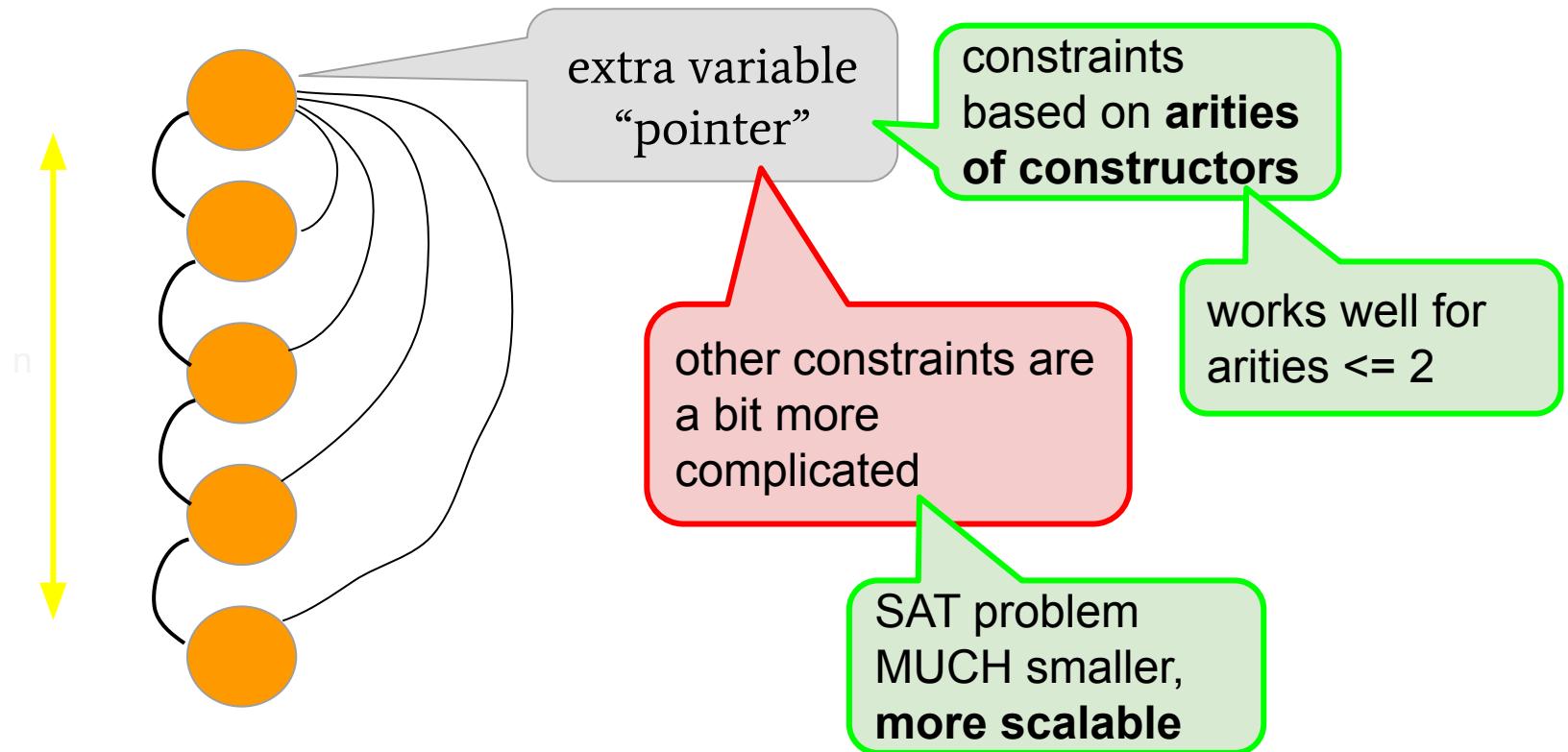
what about  
**loops /**  
**recursion?**

what about  
**non-finite**  
types?

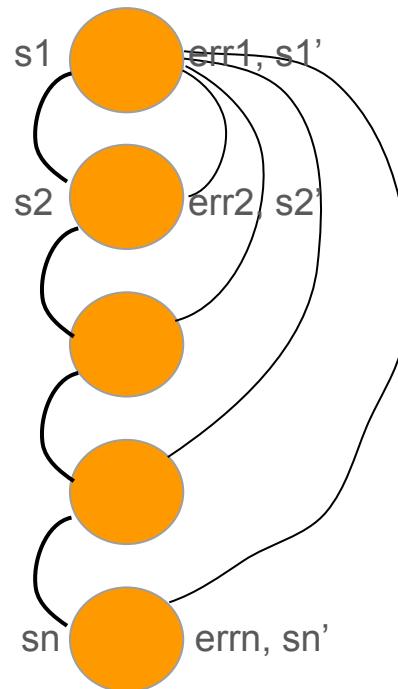
# Size-based tree encoding



# Size-based tree encoding



# Size-based tree encoding



constraints have to  
pick the right  $s_i'$

# Non-termination / loops

have a **maximum number of loop takes** for a semantic run

size of the program

number of variables

size of expressions

maximum number of function calls

enumerate these in some way

dependent on the input

# Summary

- Works OK to find bugs in **type systems**
- hand-coded
- Problems with:
  - unbounded program execution
  - more complicated data types (e.g. lists)
  - scalability

## Experiment - use Val + numerical optimization

- Similar coding of **type systems**
  - But, don't care about many details
- Still hand-coded
- Could find the same bugs as SAT, in similar running times
  - No problems with coding types, bounded running times. etc.

want to find bugs in  
**type checkers**  
(implementation)

speculative part

# Idea: instrument a type checker with Val

- Still, hand-coded
  - but systematic
- Have it compute the “distance” to when things go wrong
- Use numerical optimization

# Idea: instrument a type checker with Val

- Still, hand-coded
  - but systematic
- Have it compute the “distance” to when things go wrong
- Use numerical optimization

new project starting  
2026

NULL-pointer  
dereferencing

## Numerical optimization methods for automated bug finding in software

non-interference

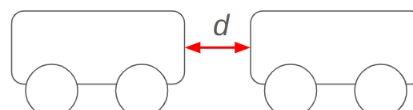
### 1 Purpose and aims

The project is about investigating how **numerical optimization methods** can be used to generate **automatic generation of test cases** in order to search for **difficult to find bugs** in software.

real-time  
constraints

**Earlier work.** The inspiration for this project comes from our earlier work on specification and testing of hybrid systems [1]. In a hybrid system, *discrete* software interacts with *continuous* physics. An example is the adaptive cruise controller in a vehicle; the discrete software reads inputs from the sensors (radar and speed) and provides control signals to the engine; the engine and the cars on the road are described by continuous physics. A possible safety property is: the distance between our car and the car in front of ours should always be more than, say, 10 meters. How can we generate test data to test a property like that?

Here is what we did: (1) Rather than expressing the safety property as a boolean statement that is merely true or false for any run of the system, we instead express the property as a real<sup>1</sup> number  $p$  that indicates how close to being false the property is. For example, for the cruise controller, we would simply use  $p=d-10$  where  $d$  is the distance between the cars in meters. If  $p>0$ , the safety



1 PhD student

working on bug  
finding using  
numerical  
optimization

1 post-doc

working on  
programming  
language semantics