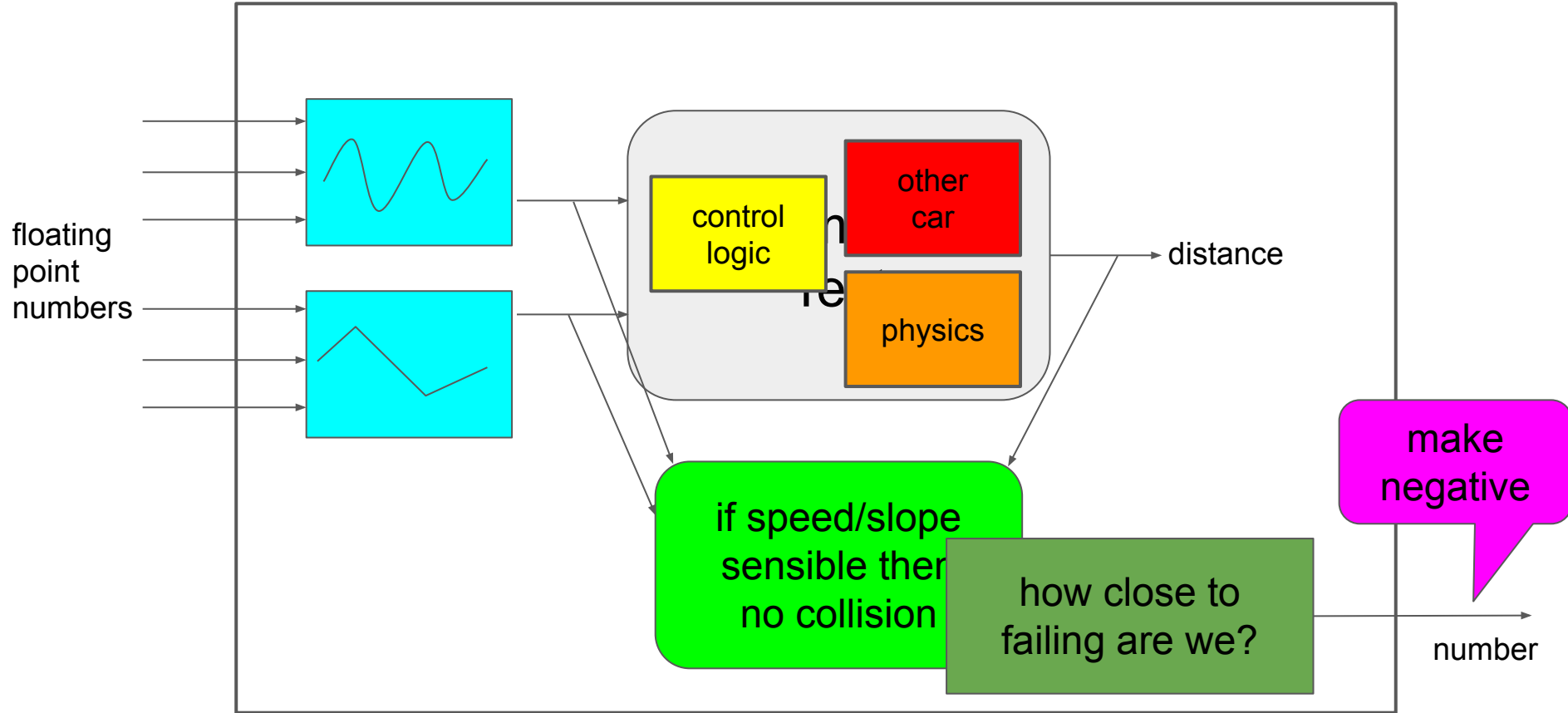


Numerical Optimization for Generating Test Data

Koen Claessen
Chalmers University of Technology
Sweden

Example: Adaptive cruise control



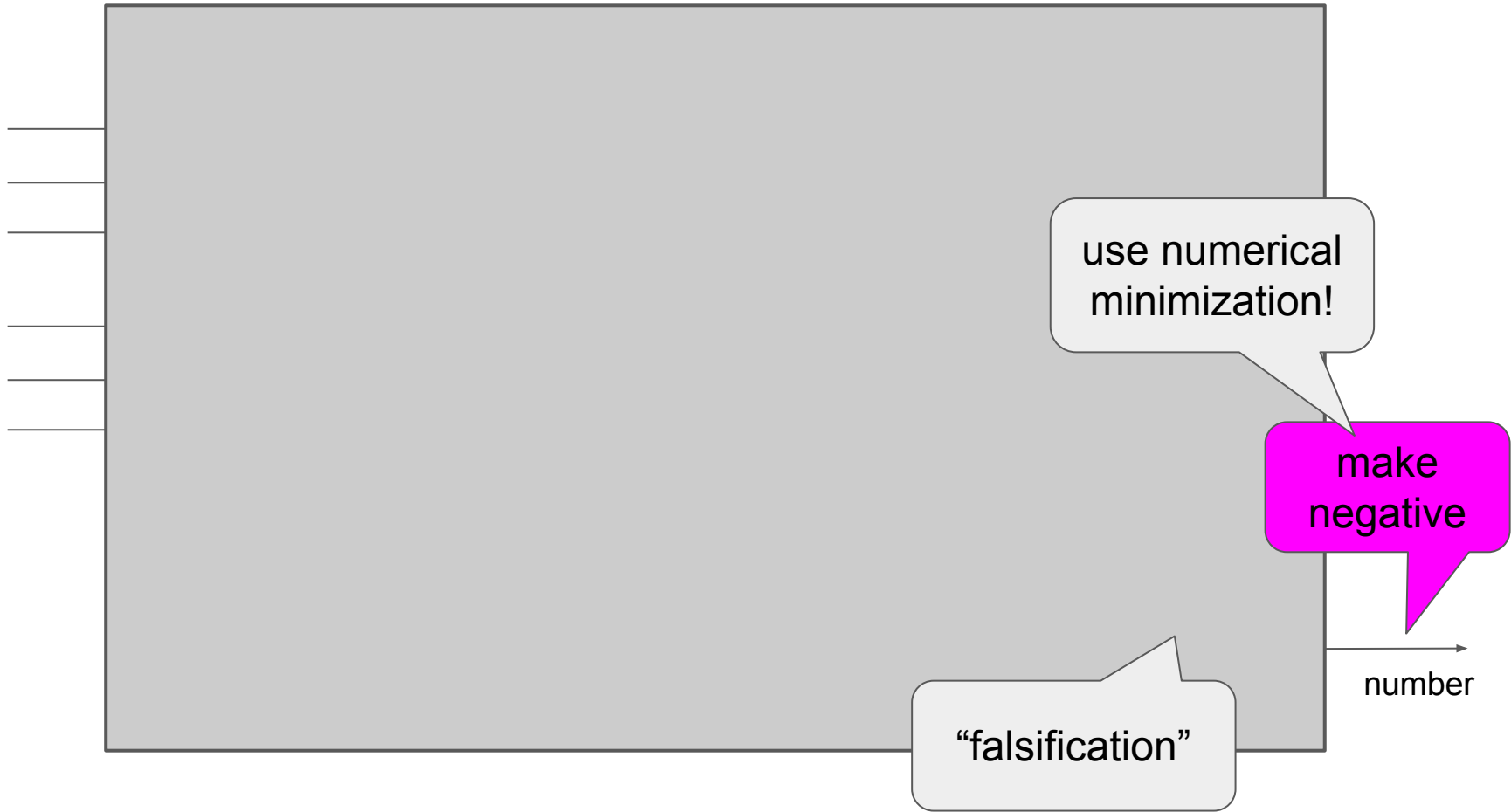
floating
point
numbers

use numerical
minimization!

make
negative

“falsification”

number



needs
gradient

Numerical Optimization Methods

complicated

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

- SNOBFIT
- Bayesian optimization
- ...

expensive

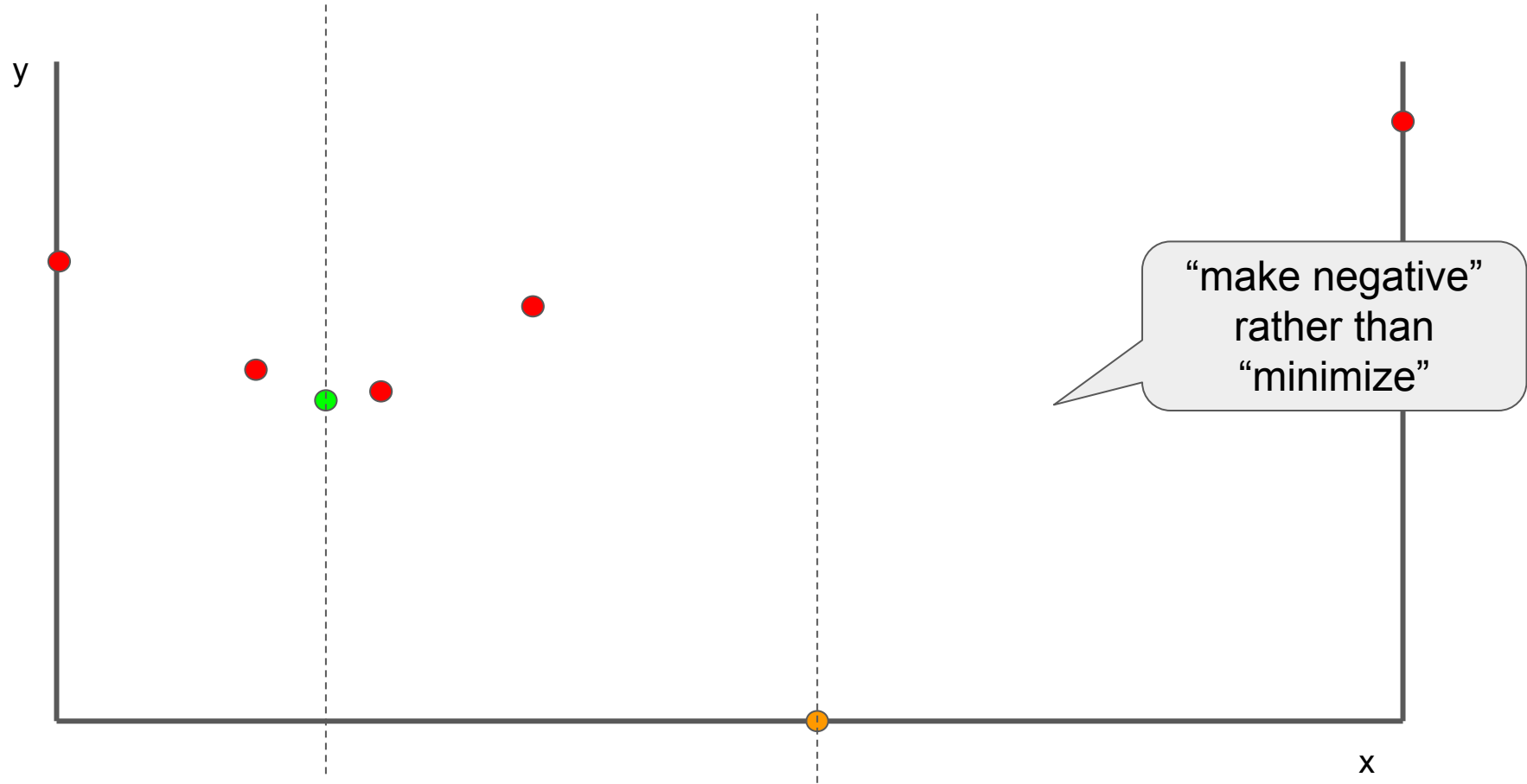
gravitate
towards local
minima

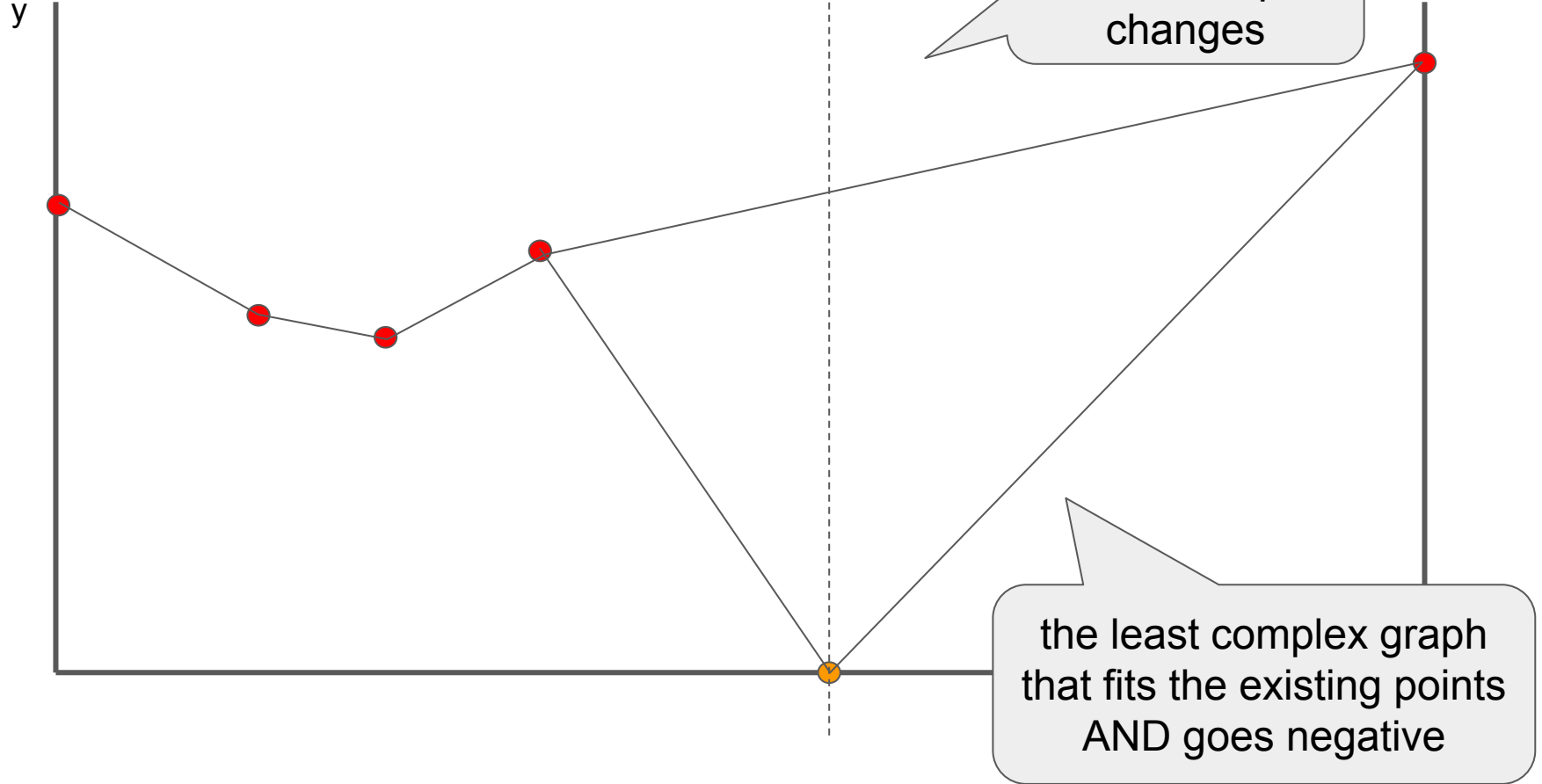
Which aspects are important for falsification

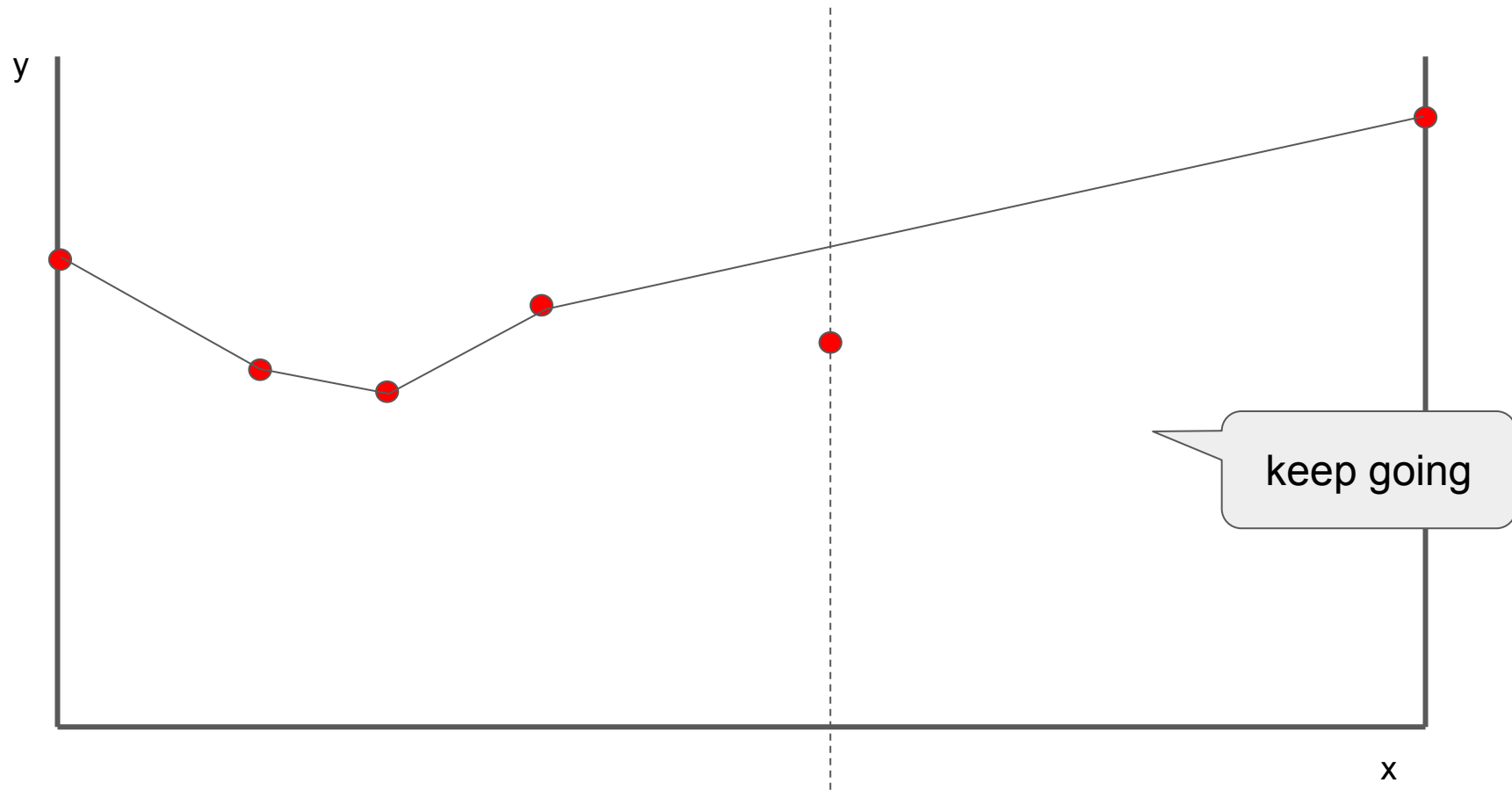
- No gradient
- Not getting stuck in local minima
 - Do not gravitate to local minima
- Having a defined “area” to search in
 - Extreme values often lead to bugs

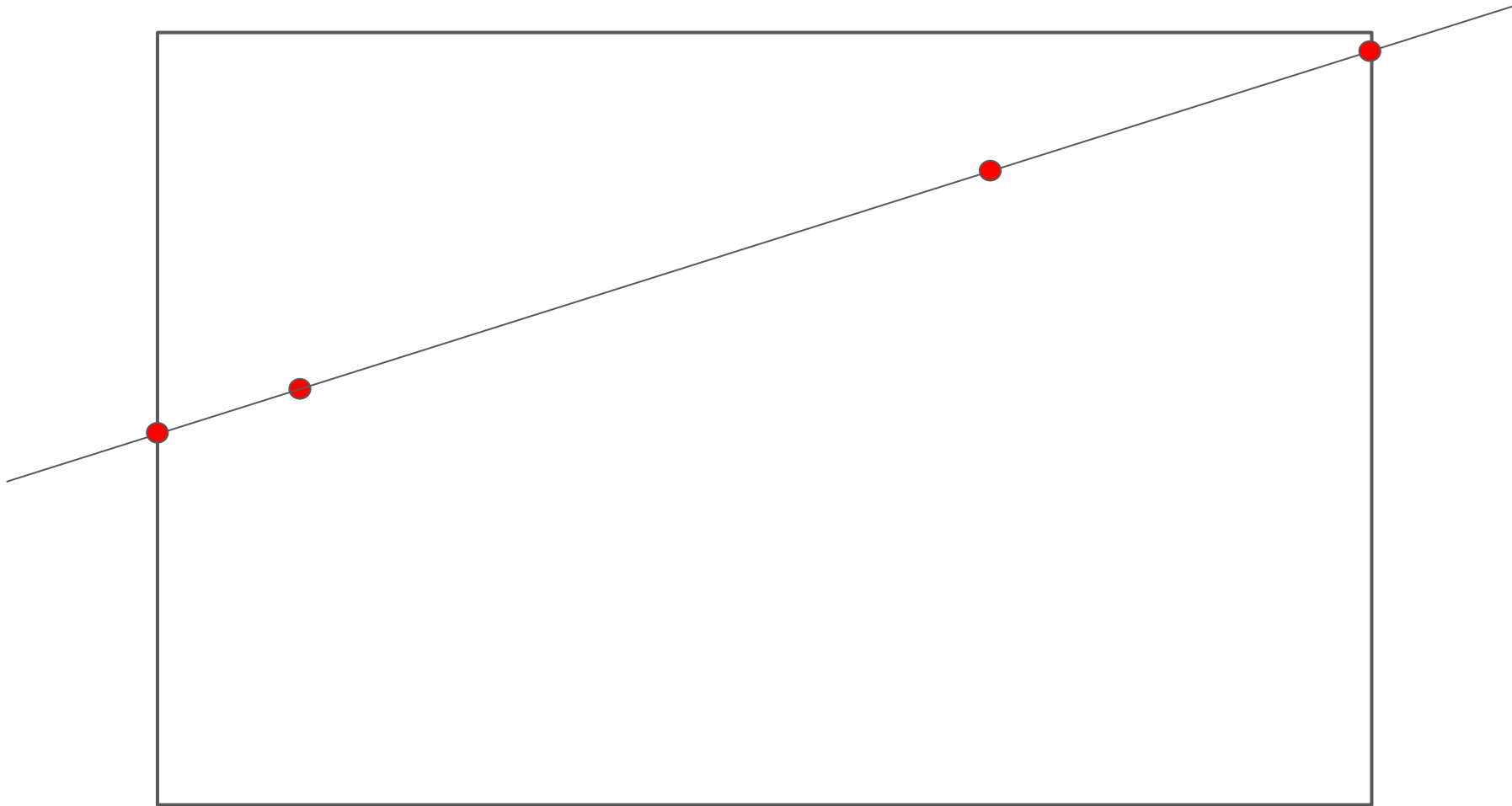


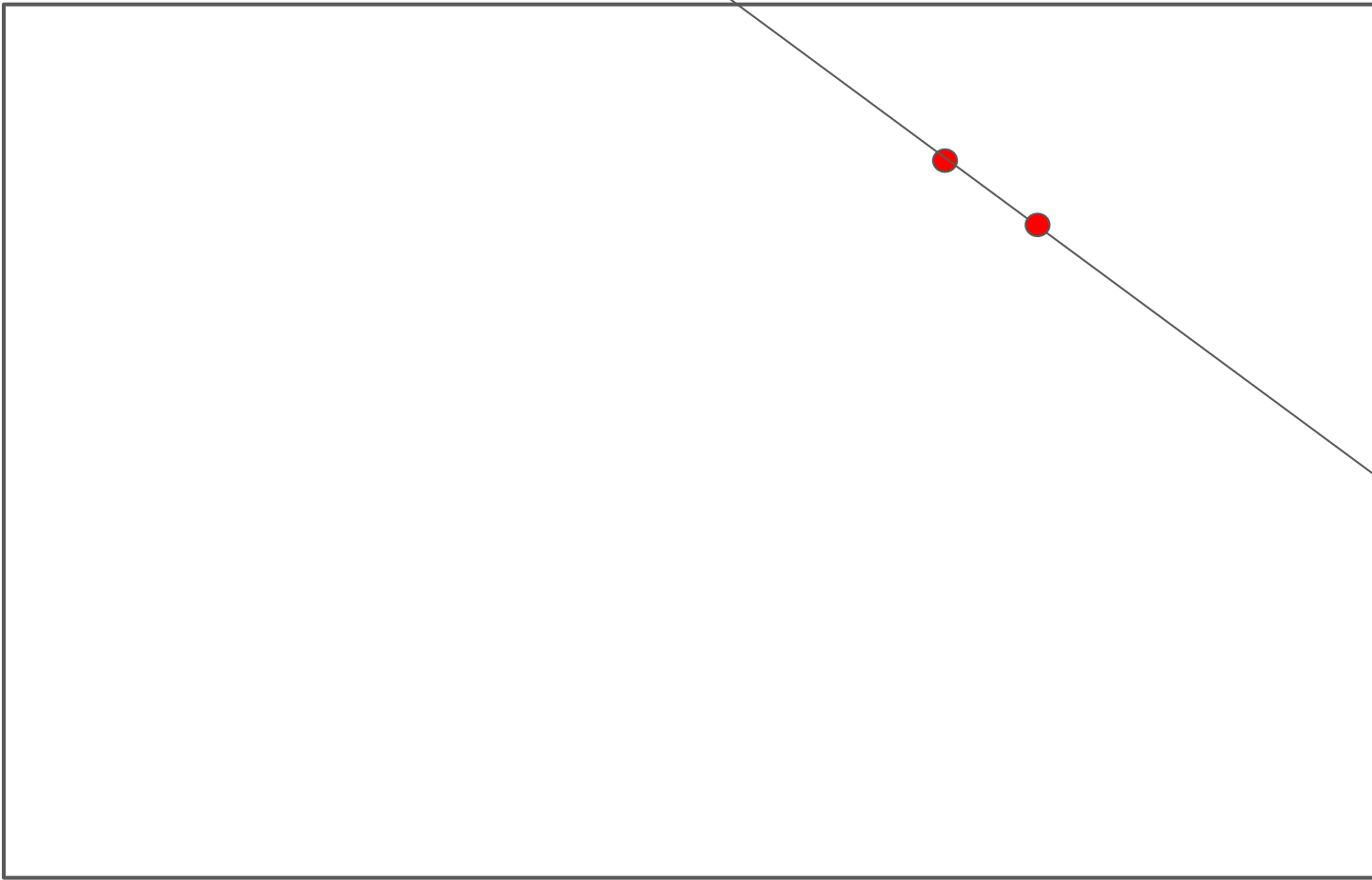
find **negative**
results

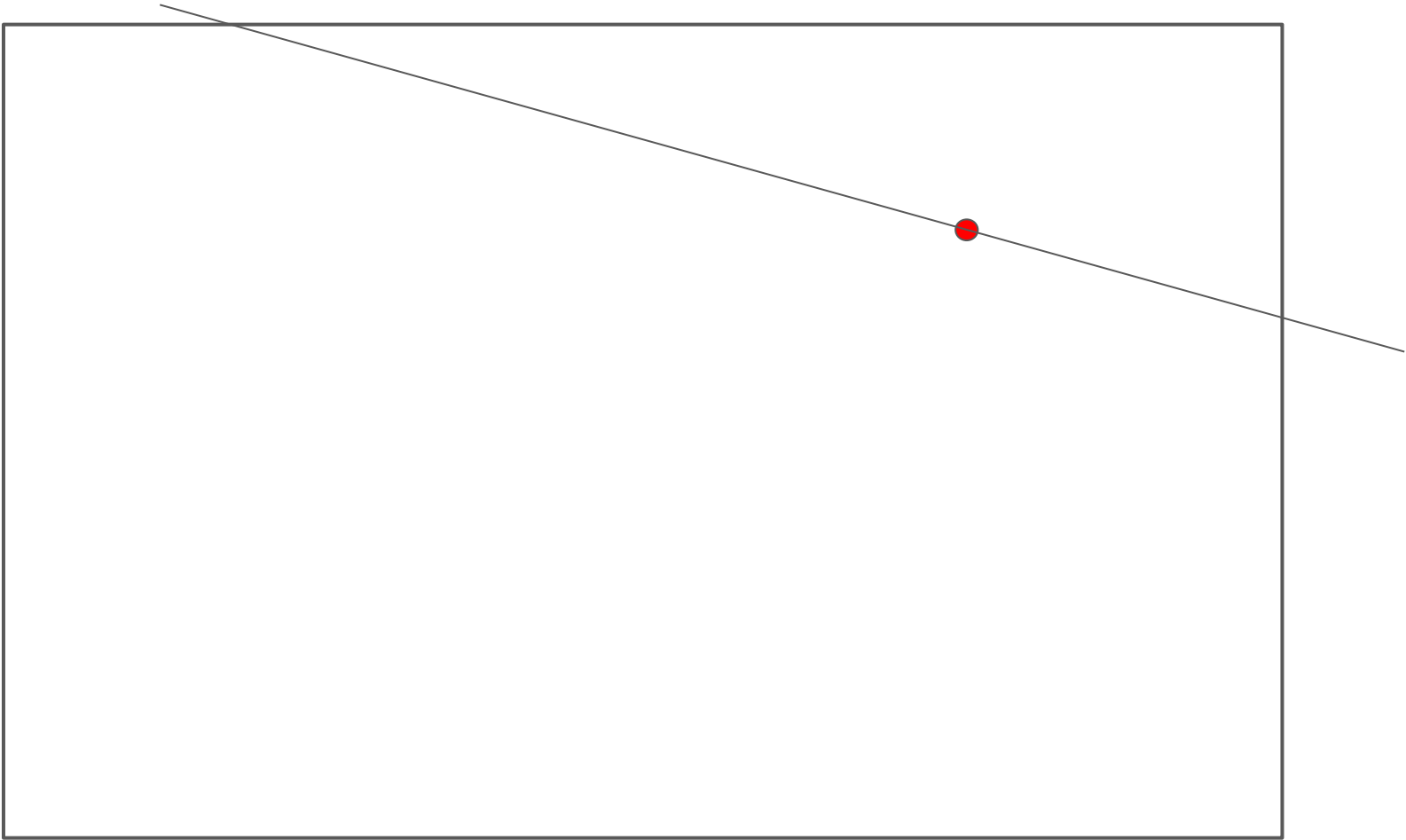






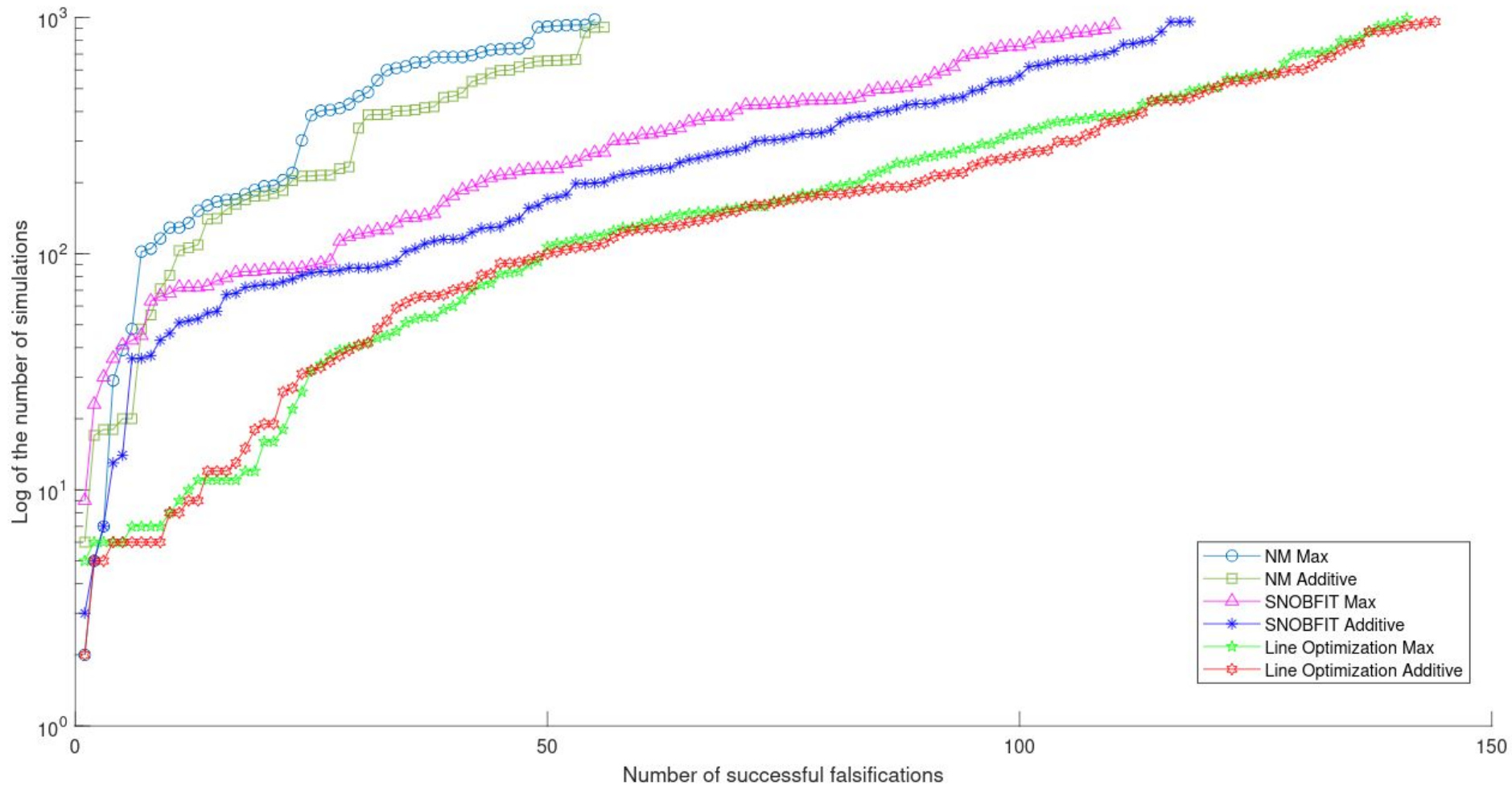




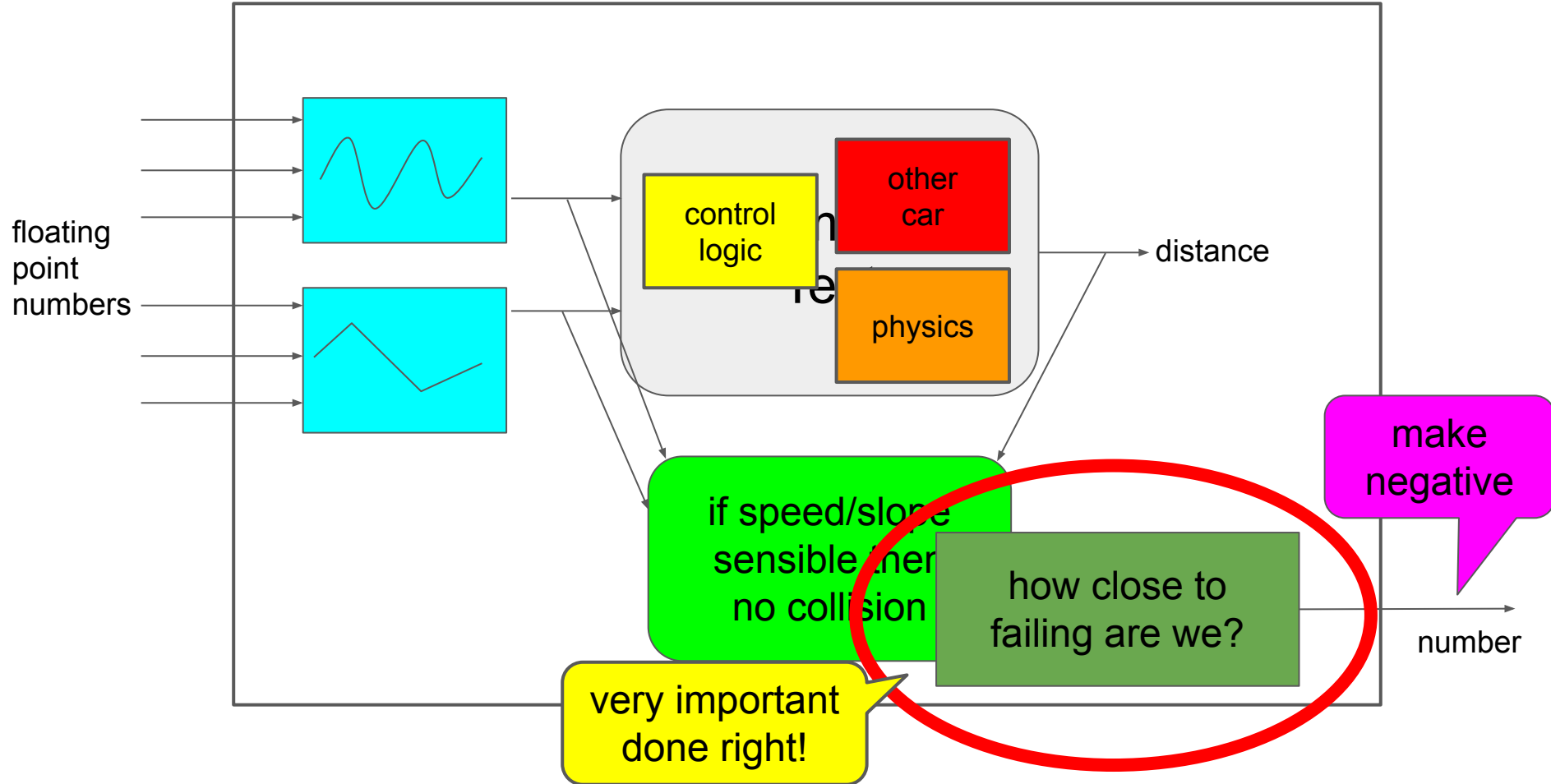


“Line falsification”

- 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
```

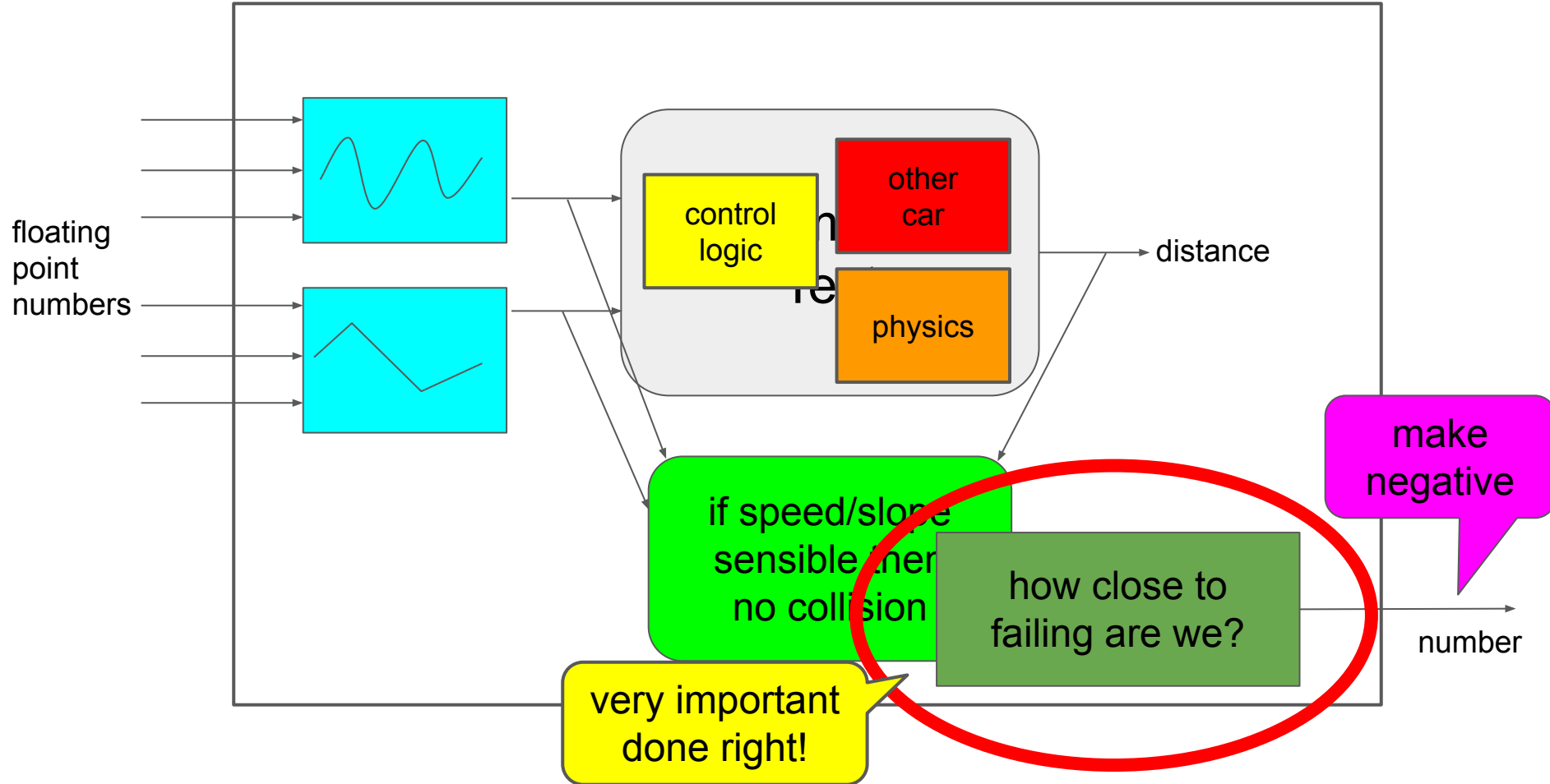
```
(&&), (||) :: DBool -> DBool -> DBool  
x && y = x + y  
x || y = x `min` y
```

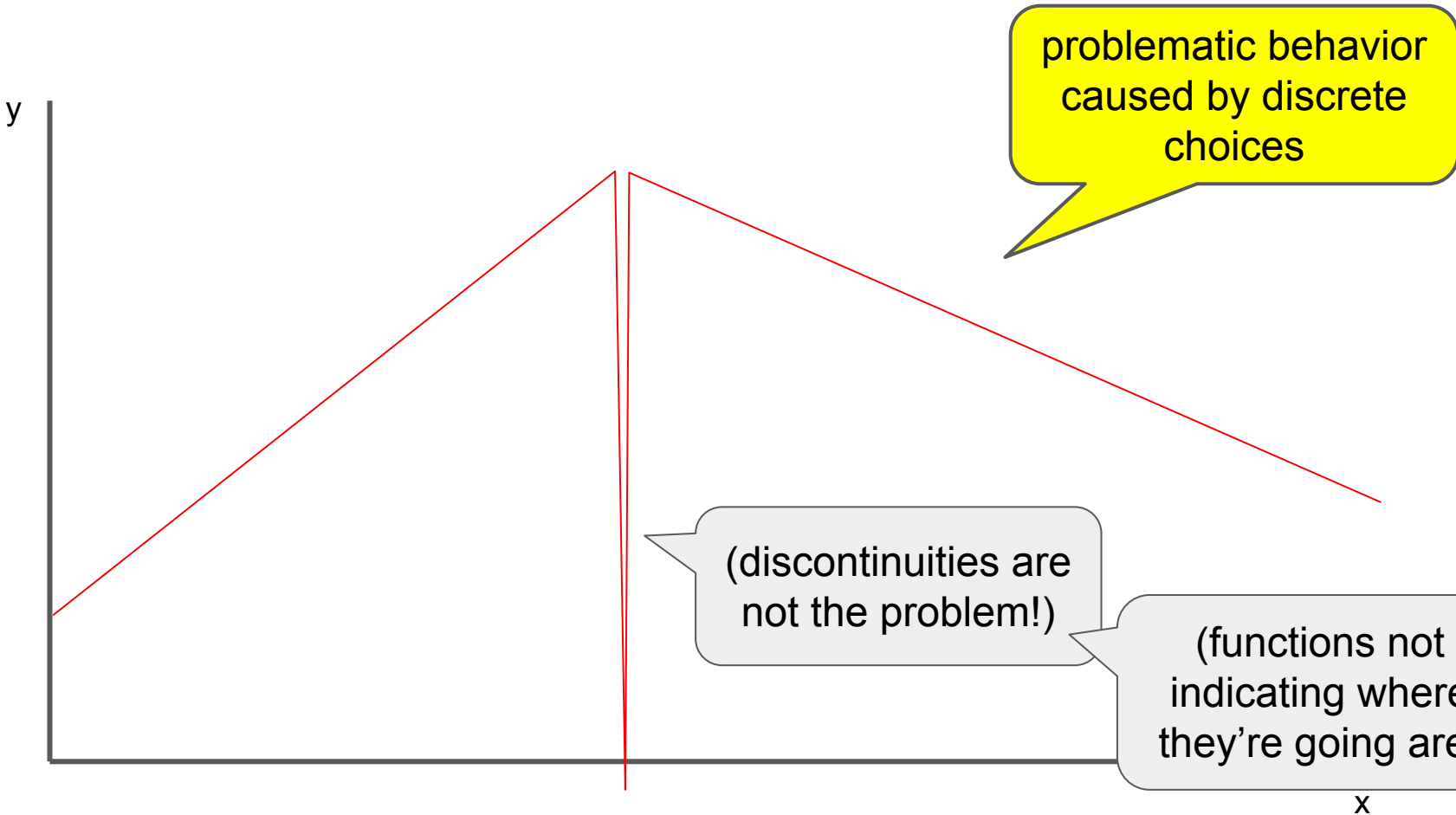
```
type DBool = Double {- >=0 -}
```

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

a specification logic
of “valued booleans”

Example: Adaptive cruise control





problematic behavior
caused by discrete
choices

(discontinuities are
not the problem!)

(functions not
indicating where
they're going are)

x

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 ] )
```

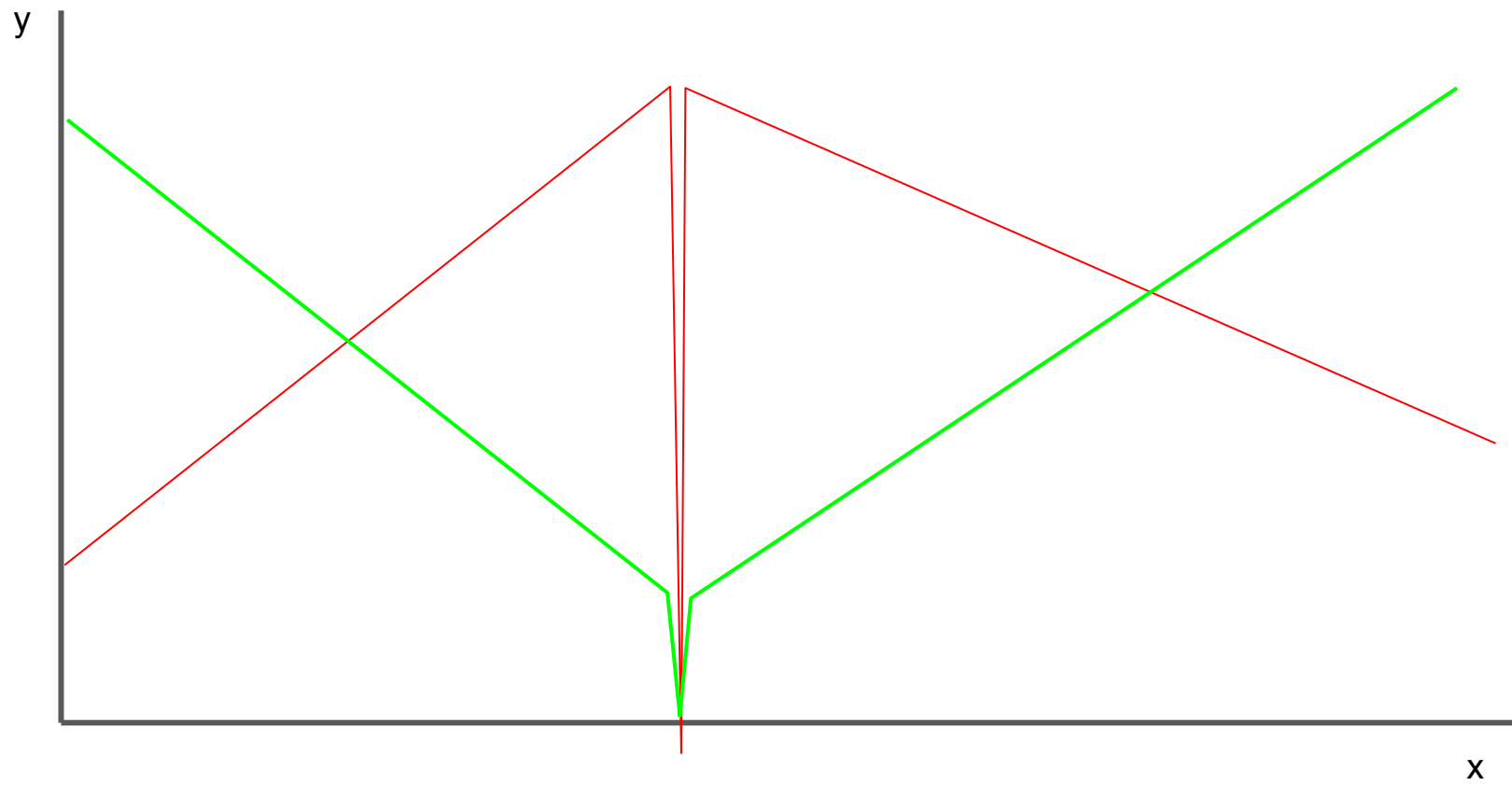
table chooses the
minimum distance

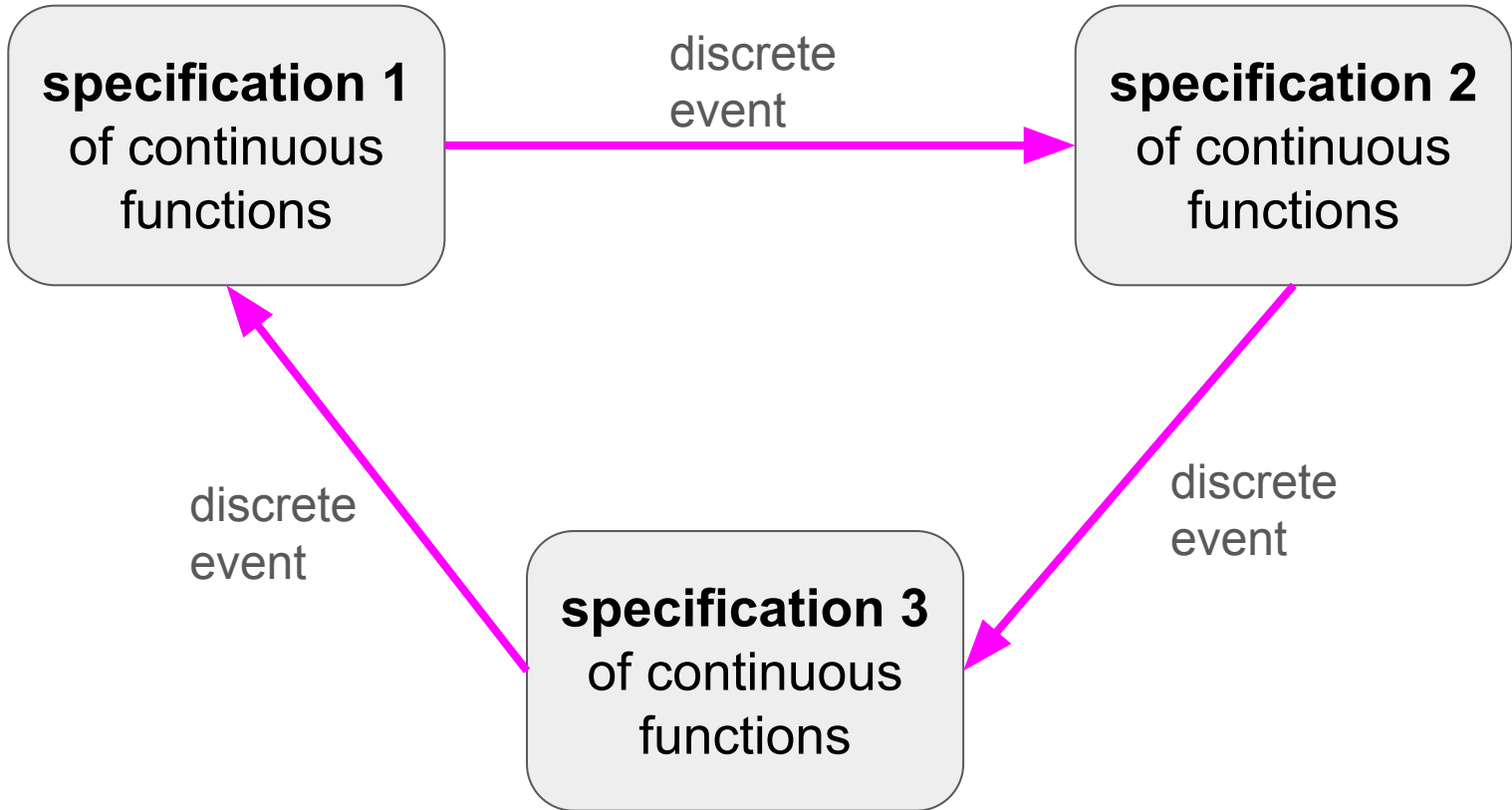
```
ifThenElse :: Val Bool ->  
            Val a  -> Val a  -> Val a
```

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)

implies

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

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

**need to check that
we do not parse too
much!**

```
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

[illegible]

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](#)

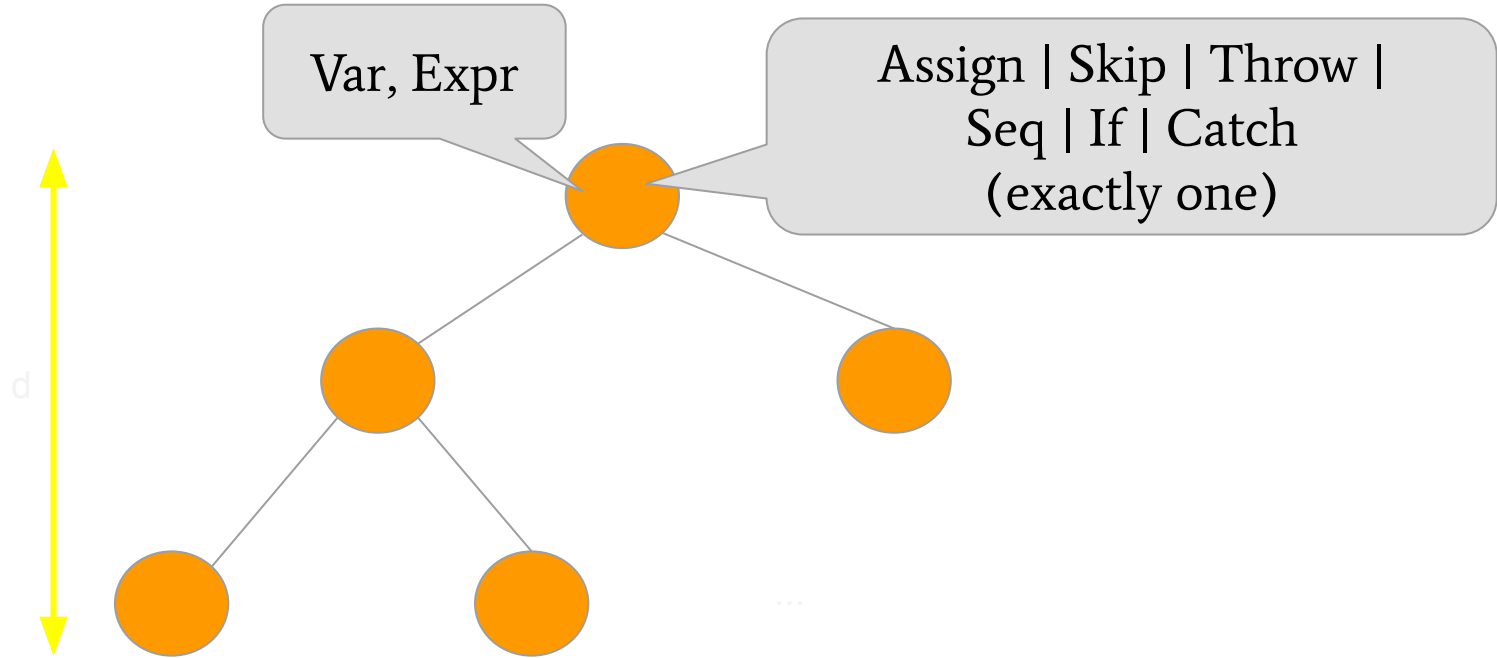
scalable

constraints &
variables over
finite domains

SAT

other theories
via SMT

IFC Challenge in SAT (take 1)



IFC Chal (take 1)

1 run

add constraints about $s2, err2, s3, \dots, err4, s4'$

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

$s3$ $err3, s3'$

$s2$ $err2, s2'$

$s4$ $err4, s4'$

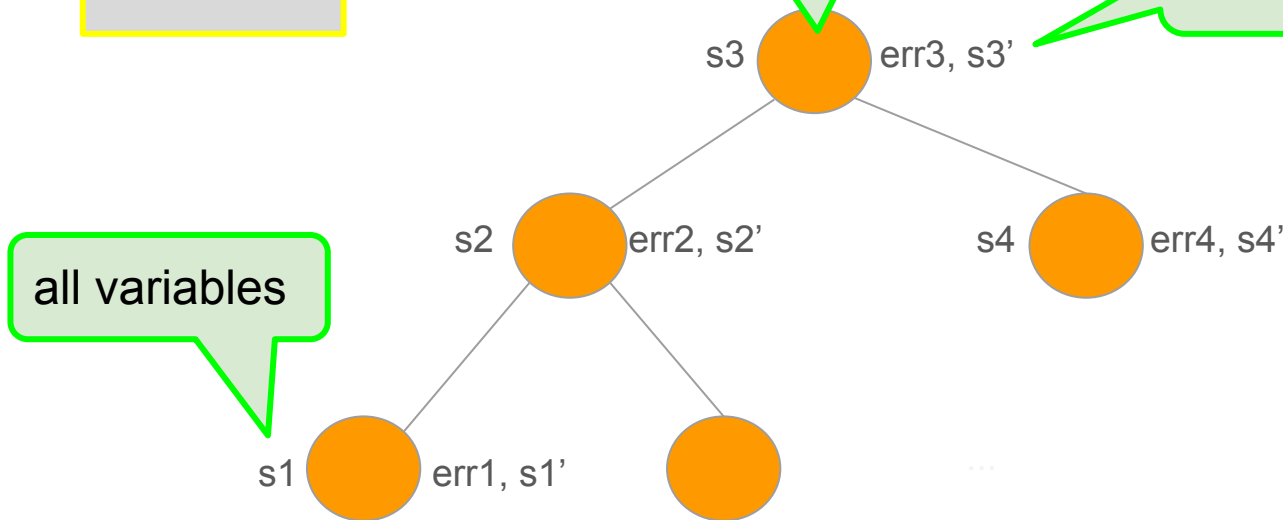
all variables

$s1$ $err1, s1'$

add for all runs

a few milliseconds

solve



IFC Challenge in SAT (take 1)

solver knows
about types

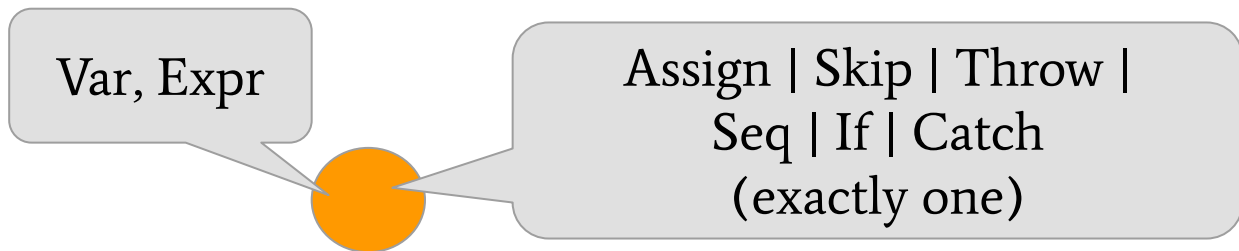
solver knows
about semantics

trees are
wasteful

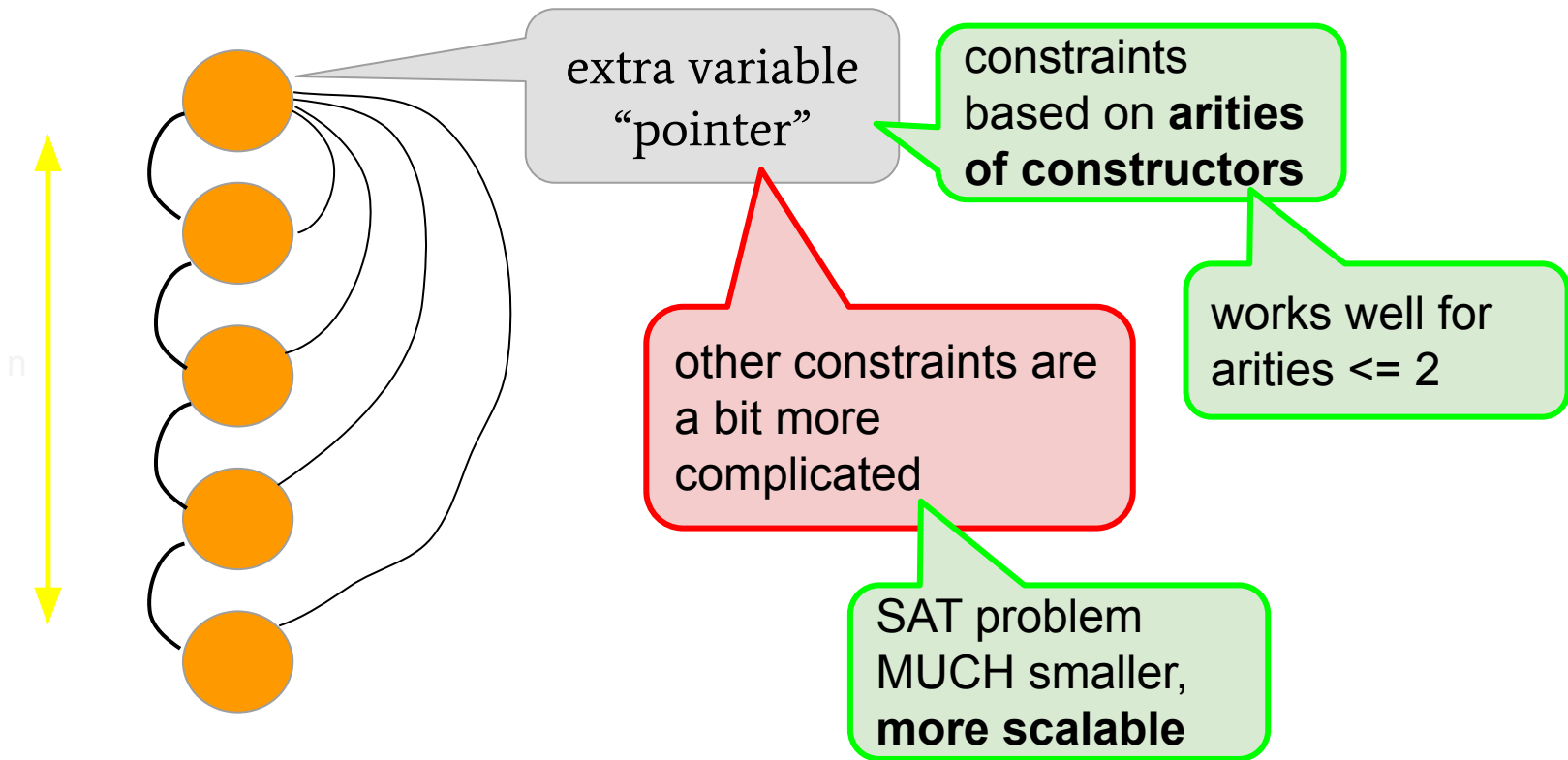
what about
**loops /
recursion?**

what about
non-finite
types?

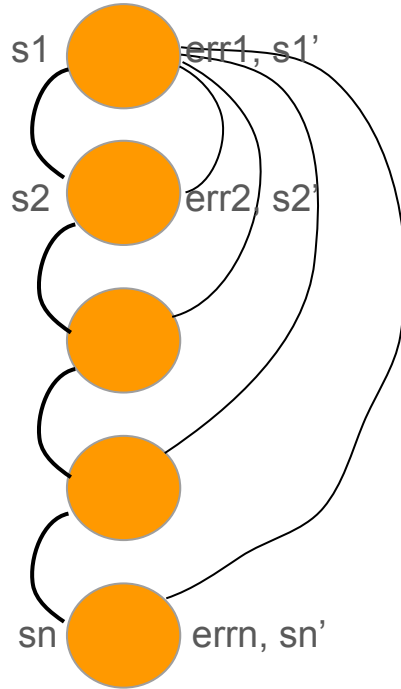
Size-based tree encoding



Size-based tree encoding

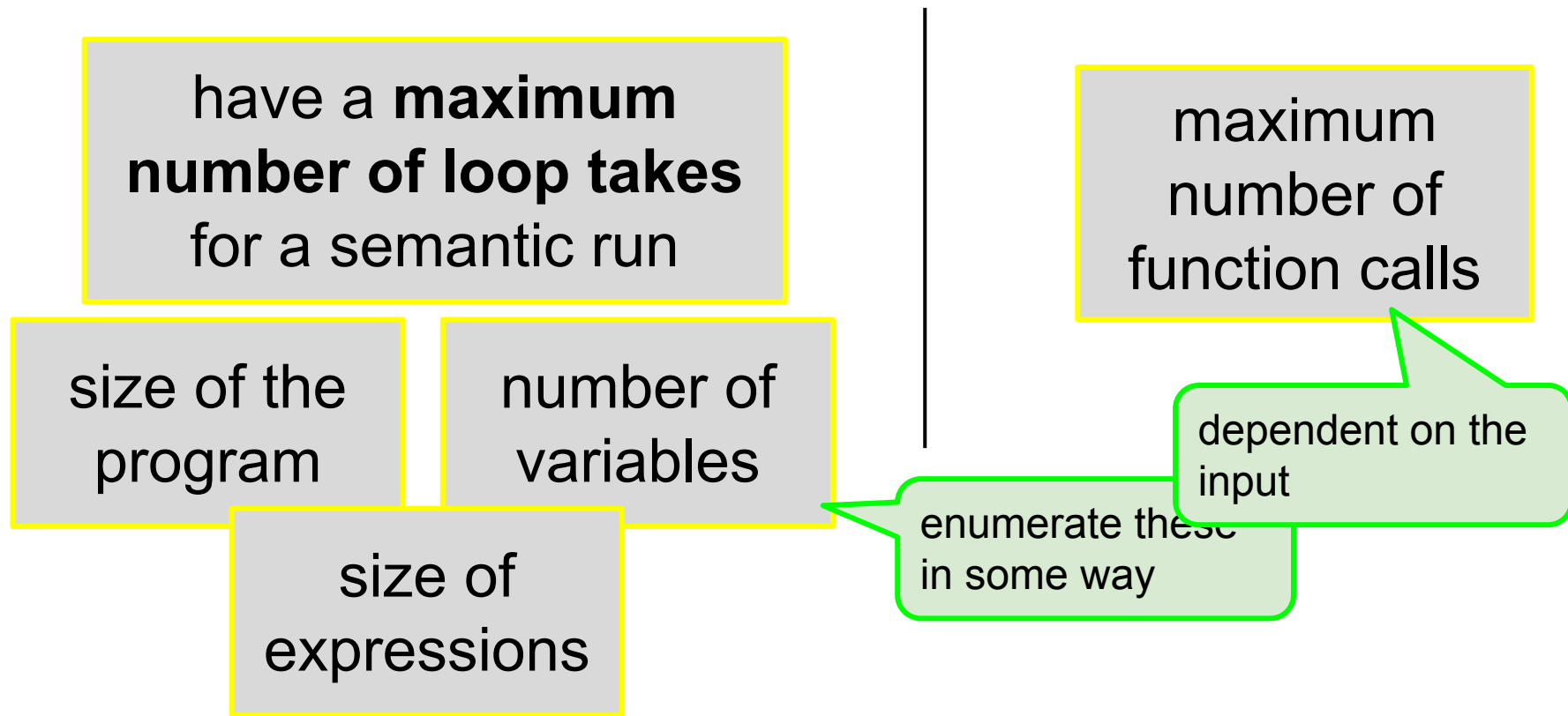


Size-based tree encoding



constraints have to
pick the right s_i'

Non-termination / loops



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

want to find bugs in
type checkers
(implementation)

- 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.

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

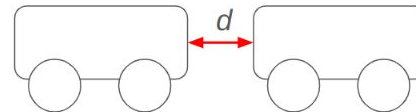
This project is about investigating how **numerical optimization methods** can be used to **automatically generate test cases** in order to search for **difficult to find bugs** in software.

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 real¹ 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