

# Timeout Strategies for Distributed Systems

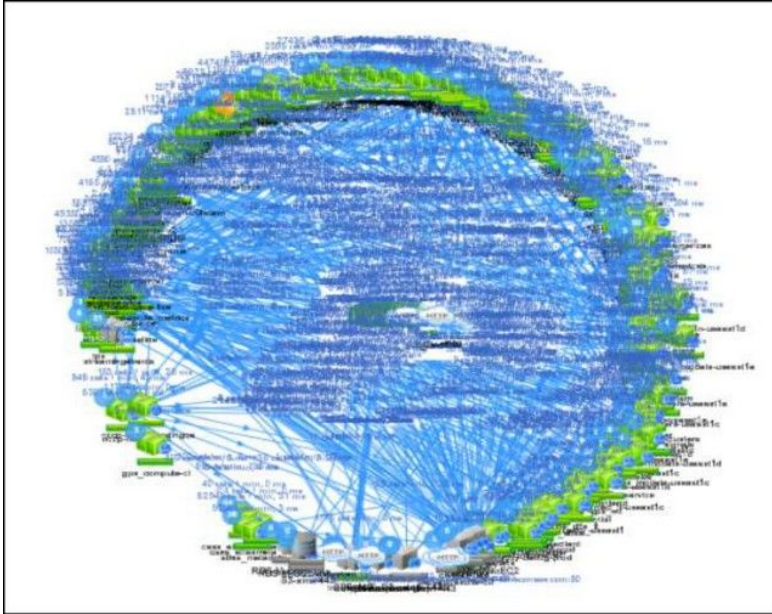
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Mentors: Lan (Max) Liu, Zhaoqi (Roy) Zhang,  
Prof. Raja Sambasivan

MIT PRIMES Spring Conference, 5/18/2025

# Distributed systems are powerful but complex

Netflix's distributed systems



**Benefits:**

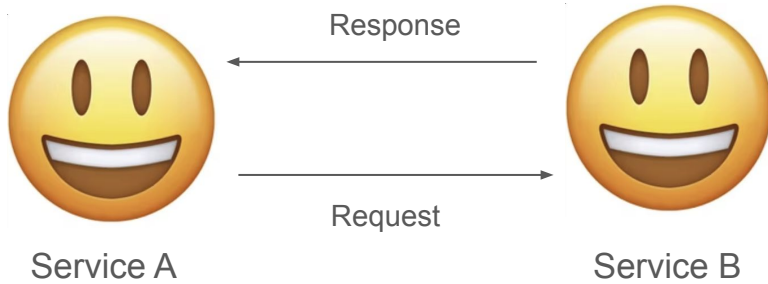
Scaling

Performance

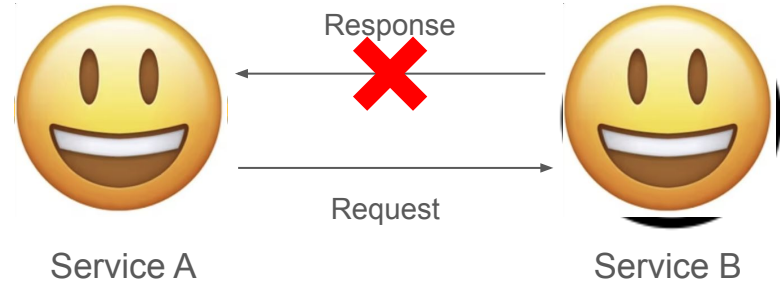
**Challenge:**

Debugging and failure-tolerance

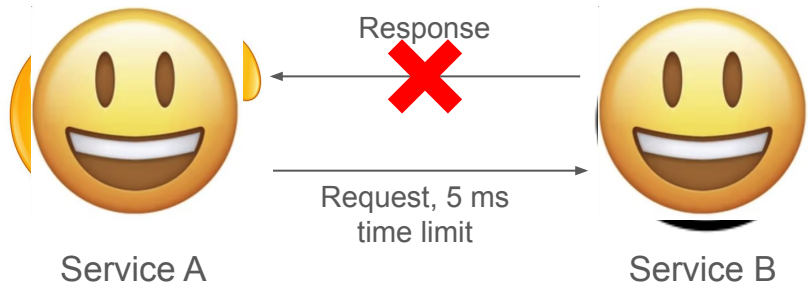
# Why are timeouts important for distributed systems?



Normal Request



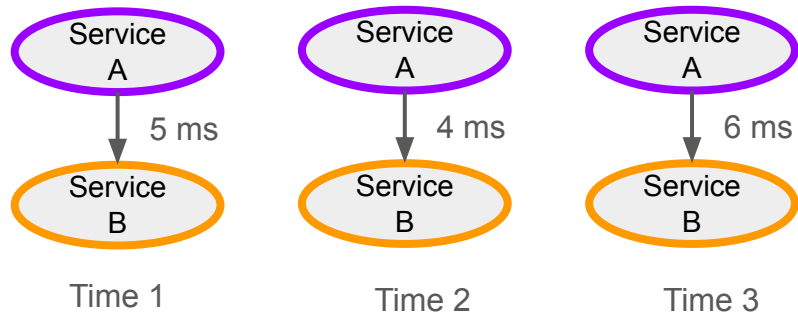
Failed Request



Request with a timeout

# Challenges with setting timeouts

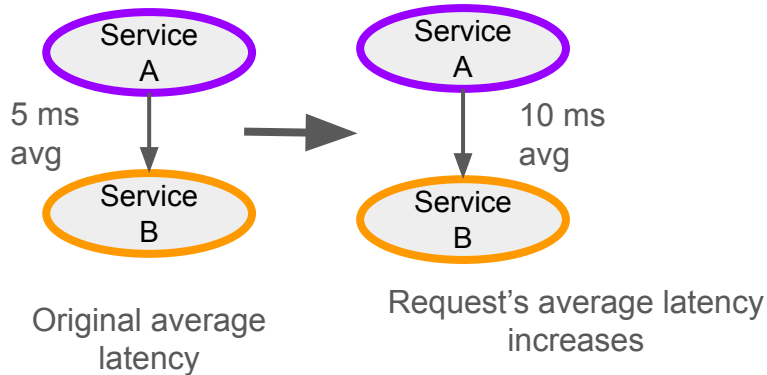
## Nonconstant latency



Other ways latency can change:

- Cache hit or miss.
- Systems can get overloaded.

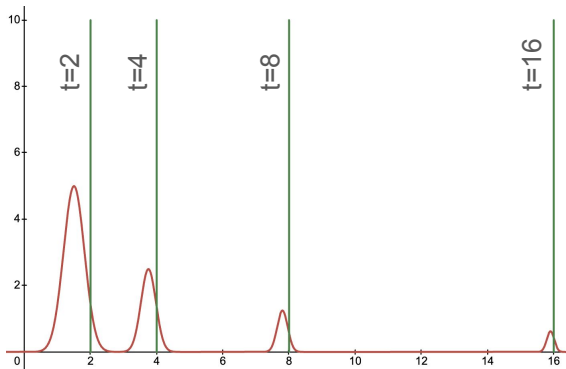
## Developers can change services' implementation



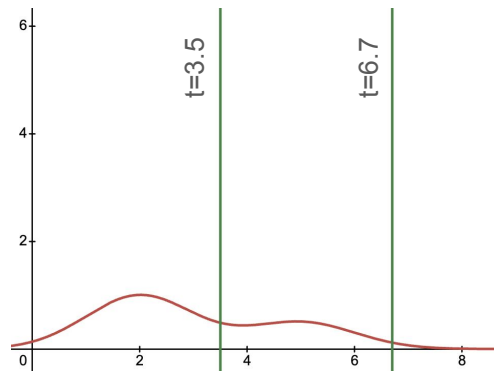
# Optimal Timeouts

# Previous work: Why ALTO is optimal

- **ALTO: Analyzer of Latency for Timeout Optimization**
- 40% faster than industry standard in worst-case
- Send “diagnostic” requests, those without timeout values, to monitor the real state of each service



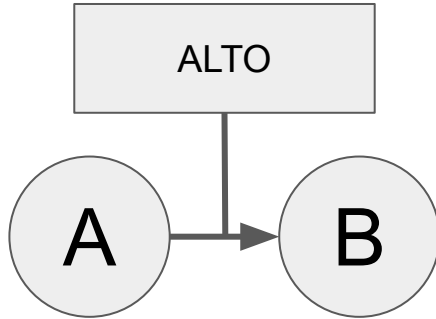
Distribution for which exponential backoff is optimal



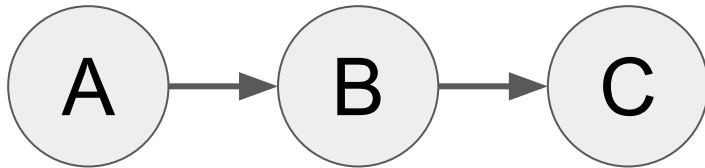
Arbitrary distribution and optimal timeouts  
computed by ALTO

# Current work: Research Goals

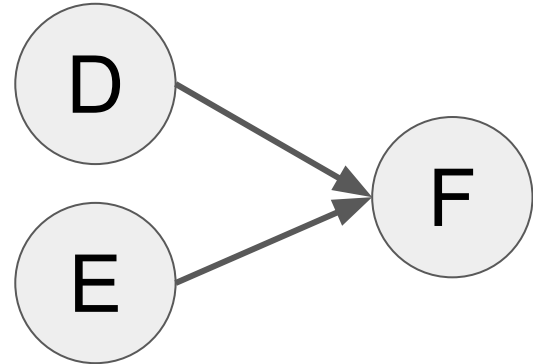
## 1. Determining optimal timeout values



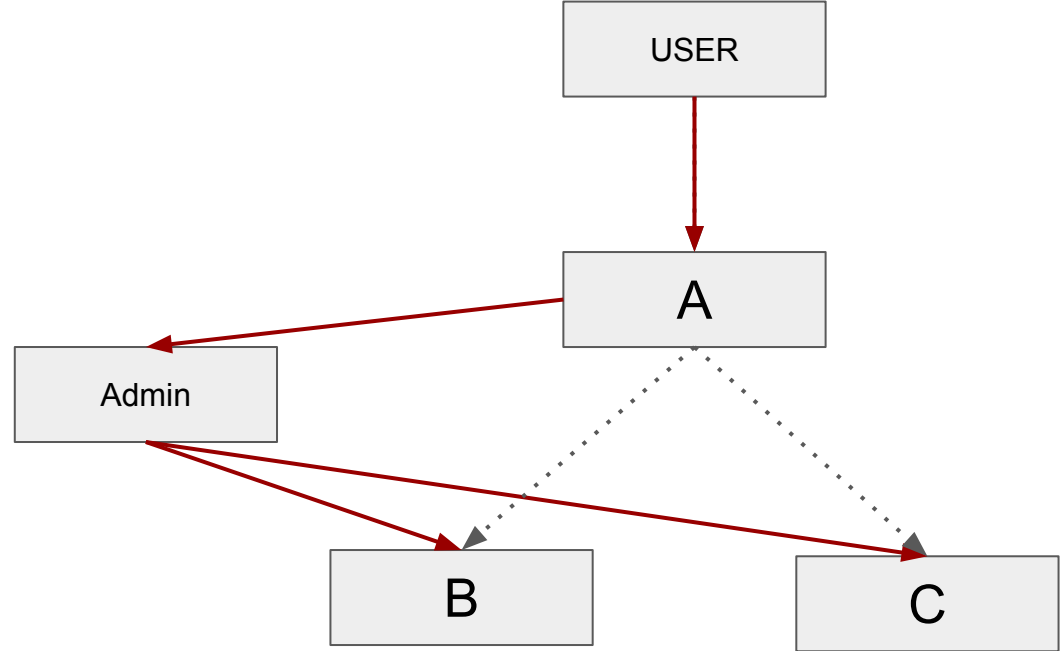
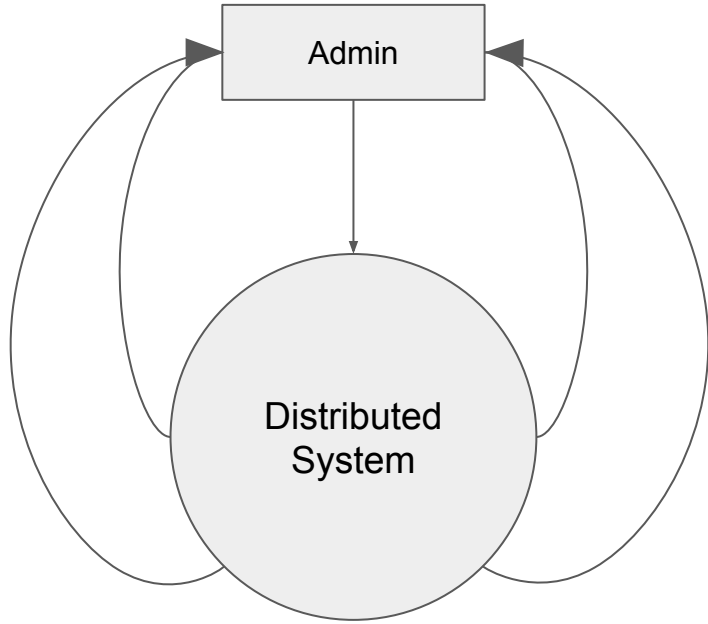
## 2. Fast timeout adaptability



## 3. Efficiently computing timeout values across the entire distributed system



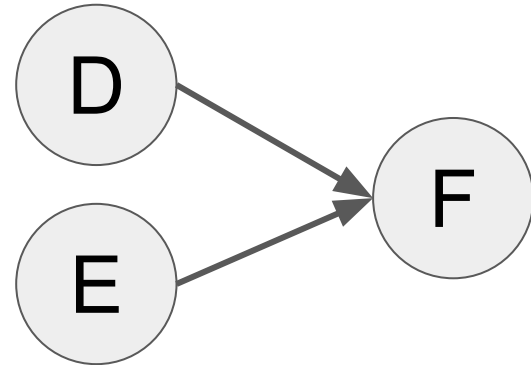
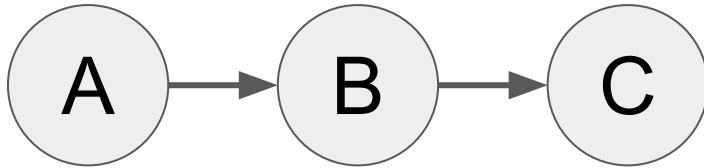
# Global implementation of ALTO in our testbed



# Computing timeouts globally & adapting to new situations

1. Find distributions from all services
2. Compute timeouts to each service based on distributions
3. Determine contribution from each service

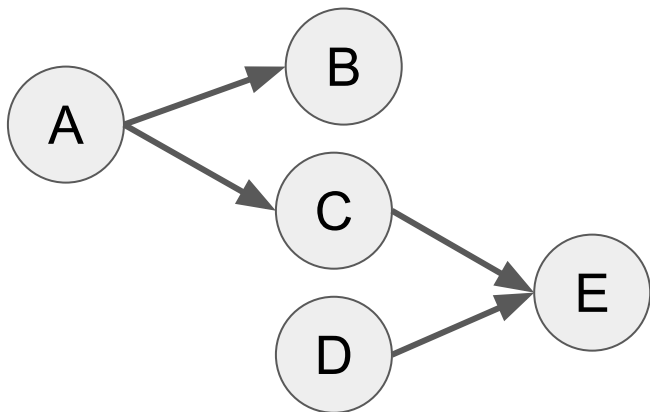
$$C(N) = D(N) - \sum_{n \in N.V} D(n)$$





# Proposed evaluation

- Evaluate whether the global approach is faster-adapting and more resource-efficient than the local approach
- Consider:
  - The number of timed out requests (measure of lag in adapting timeout values)
  - The closeness of the timeout value to latency

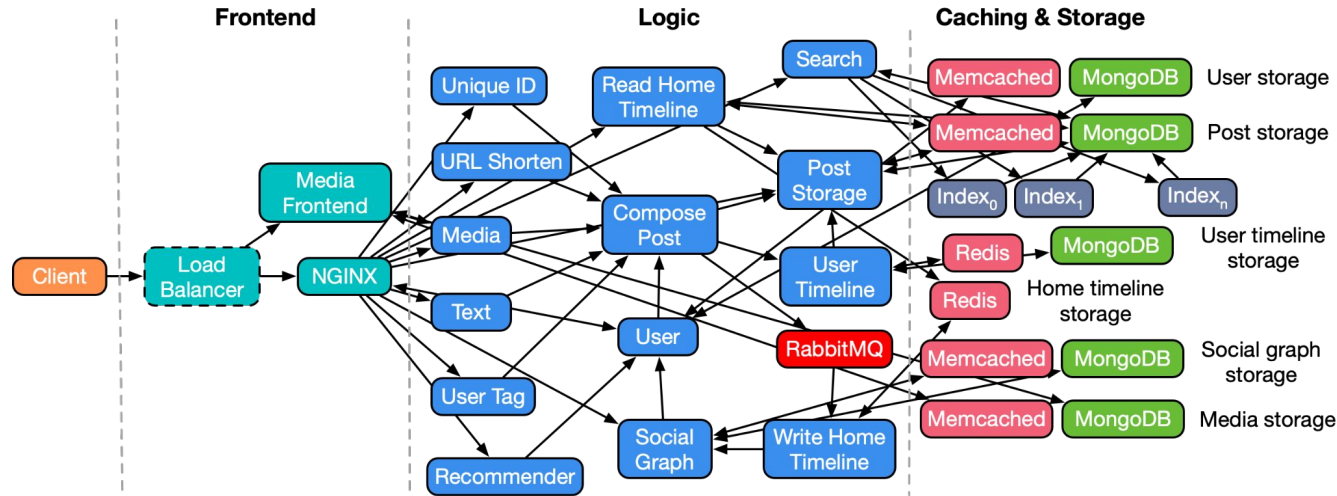


## Methodology









- Use my custom application
- Inject latency increases/decreases
- See how long services take to adapt

# Future Work

- Implement ALTO global in Social Network, a larger and industry-standardized distributed system



# Conclusions

Timeout Algorithm Goal	Exponential Backoff (Industry standard)	ALTO (previous work)	ALTO, Global (current work)
Optimal timeout values			
Efficient computation			
Fast adaptability			

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My family and friends



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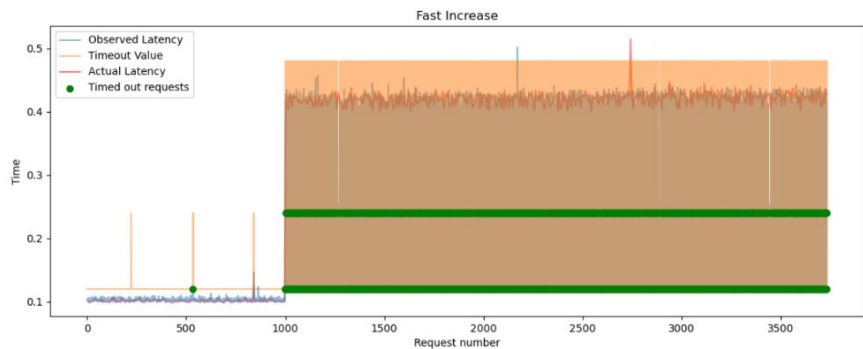


# References

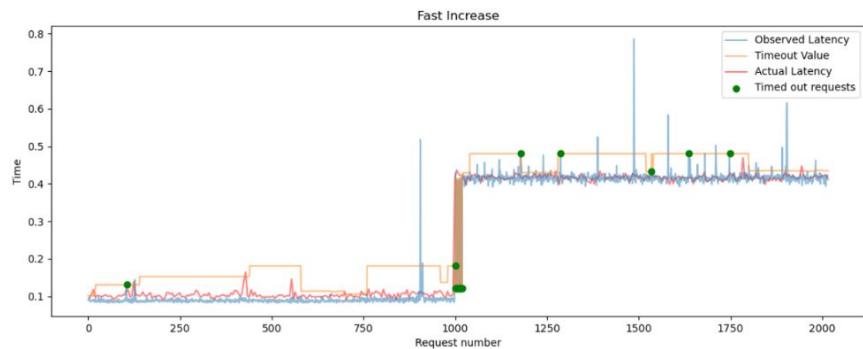
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# Backup Slides

# Our previous work, ALTO performs significantly better than industry standard



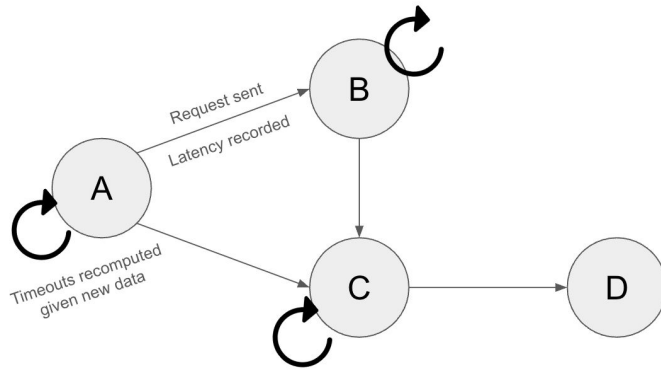
Exponential Backoff



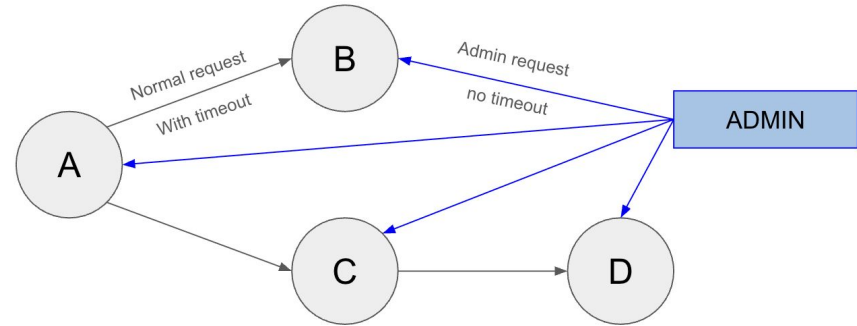
ALTO

# Local vs Global timeout algorithms

**Local** algorithms operate strictly between two services





**Global** algorithms have data about the entire distributed system





# When is a timeout **optimal**?

- As systems evolve, timeouts change.
- An **optimal timeout** is a timeout that results in the minimal possible average amount of time before a response is received.
  - Too short  wasting work since we have to reissue requests
  - Too long  wasting time when request should have been discarded
- We **continuously update** the timeout values to adapt accordingly.

# Increasing timeout values allow for precise hedging



Normal timeouts



Sequential timeouts

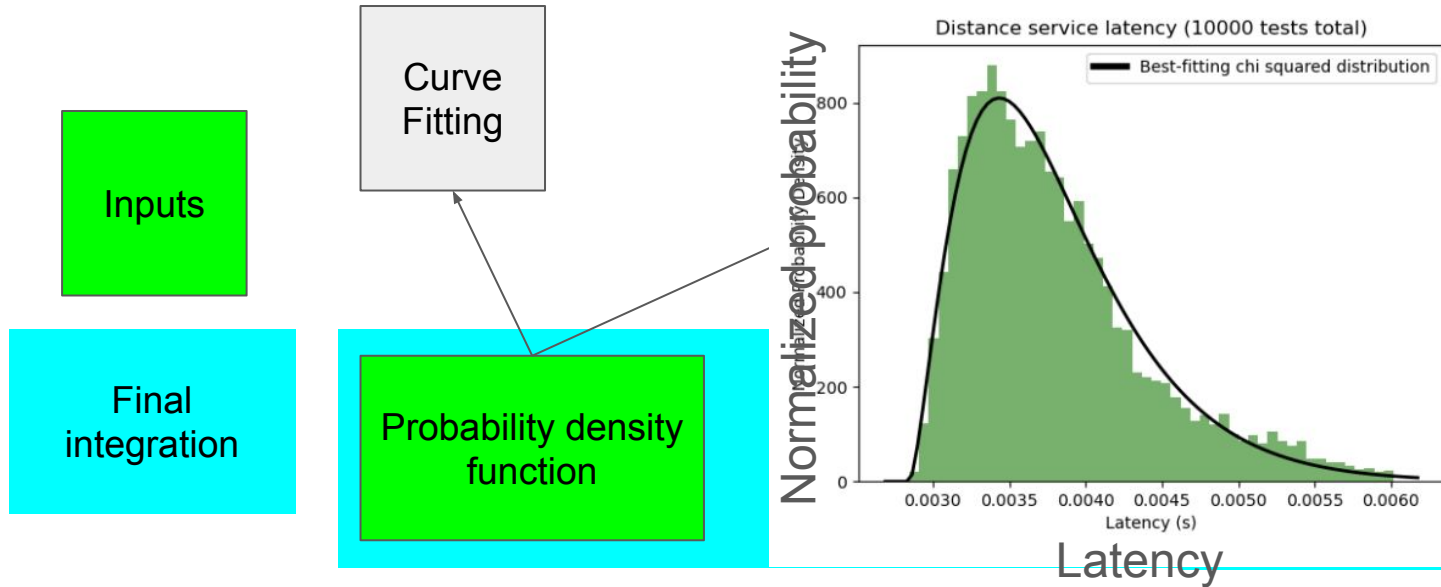
## Failure conditions

- Temporary increase
- System failure

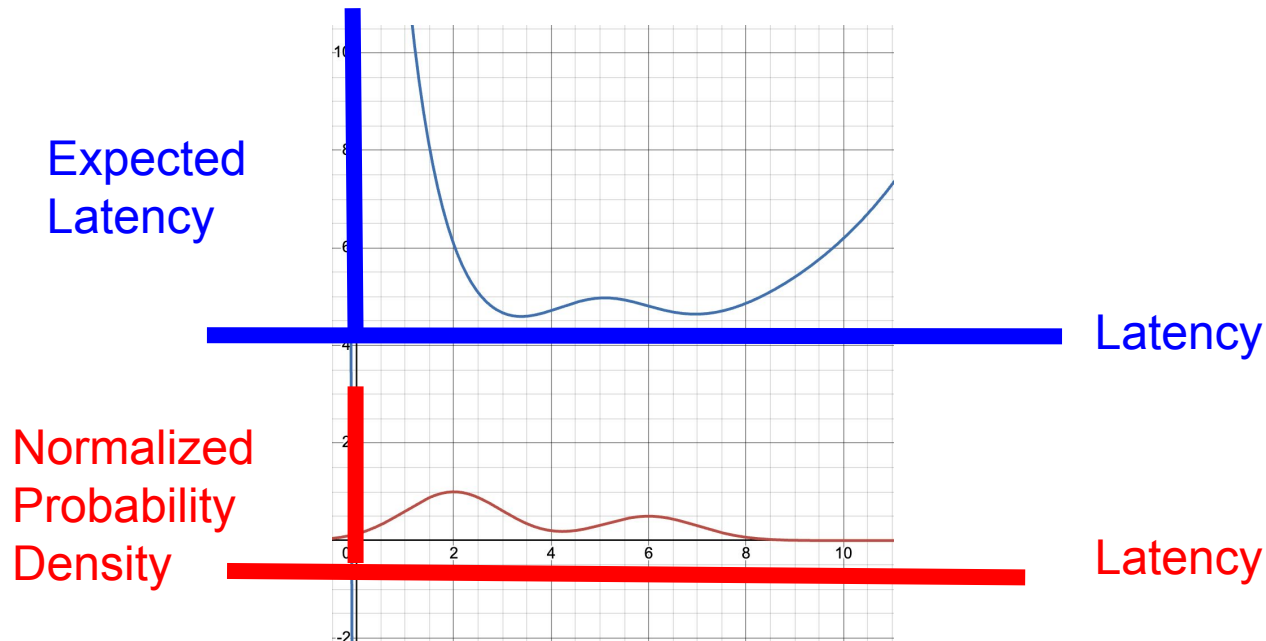
## Increase timeout value

- Sensitivity reduction
- Failure confirmation

# Mathematical Model for Latency



# Math behind mathematical model (1)



## Math behind mathematical model (2)

$$E(t) = \boxed{\frac{\int_t^\infty f(x)dx}{\int_0^\infty f(x)dx} (t + E(t))} + \boxed{\frac{\int_0^t f(x)dx}{\int_0^\infty f(x)dx} \cdot \frac{\int_0^t xf(x)dx}{\int_0^t f(x)dx}} + \boxed{g(t)}$$

Case where we time out      Case where the request is successful      Cost function

# Derivation

$$E(t) = \frac{\int_t^\infty f(x)dx}{\int_0^\infty f(x)dx} (t + E(t)) + \frac{\int_0^t f(x)dx}{\int_0^\infty f(x)dx} \cdot \frac{\int_0^t x f(x)dx}{\int_0^t f(x)dx} + g(t)$$

$$E(t) = \frac{\int_t^\infty f(x)dx}{\int_0^\infty f(x)dx} t + \frac{\int_t^\infty f(x)dx}{\int_0^\infty f(x)dx} E(t) + \frac{\int_0^t f(x)dx}{\int_0^\infty f(x)dx} \cdot \frac{\int_0^t x f(x)dx}{\int_0^t f(x)dx} + g(t)$$

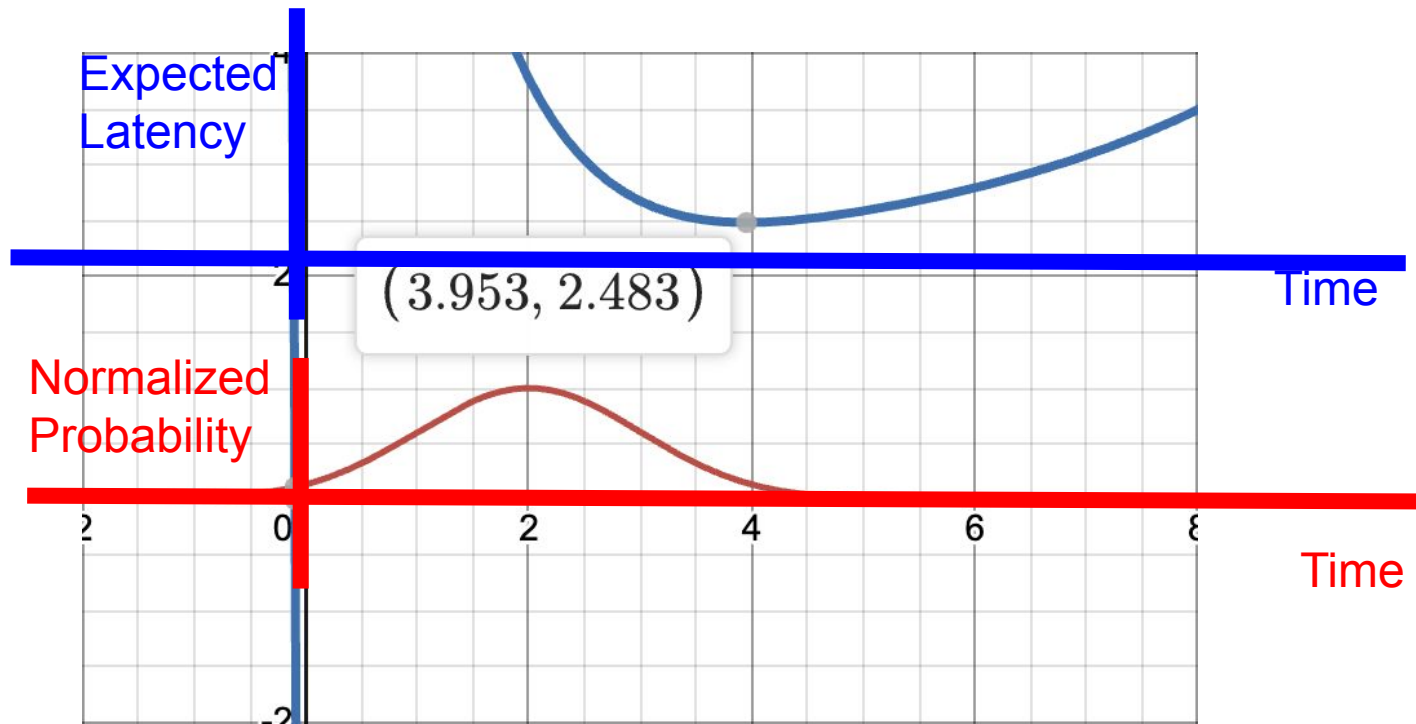
$$E(t) \left( 1 - \frac{\int_t^\infty f(x)dx}{\int_0^\infty f(x)dx} \right) = \frac{\int_t^\infty f(x)dx}{\int_0^\infty f(x)dx} t + \frac{\int_0^t f(x)dx}{\int_0^\infty f(x)dx} \cdot \frac{\int_0^t x f(x)dx}{\int_0^t f(x)dx} + g(t)$$

$$E(t) \left( \frac{\int_0^t f(x)dx}{\int_0^\infty f(x)dx} \right) = \frac{\int_t^\infty f(x)dx}{\int_0^\infty f(x)dx} t + \frac{\int_0^t f(x)dx}{\int_0^\infty f(x)dx} \cdot \frac{\int_0^t x f(x)dx}{\int_0^t f(x)dx} + g(t)$$

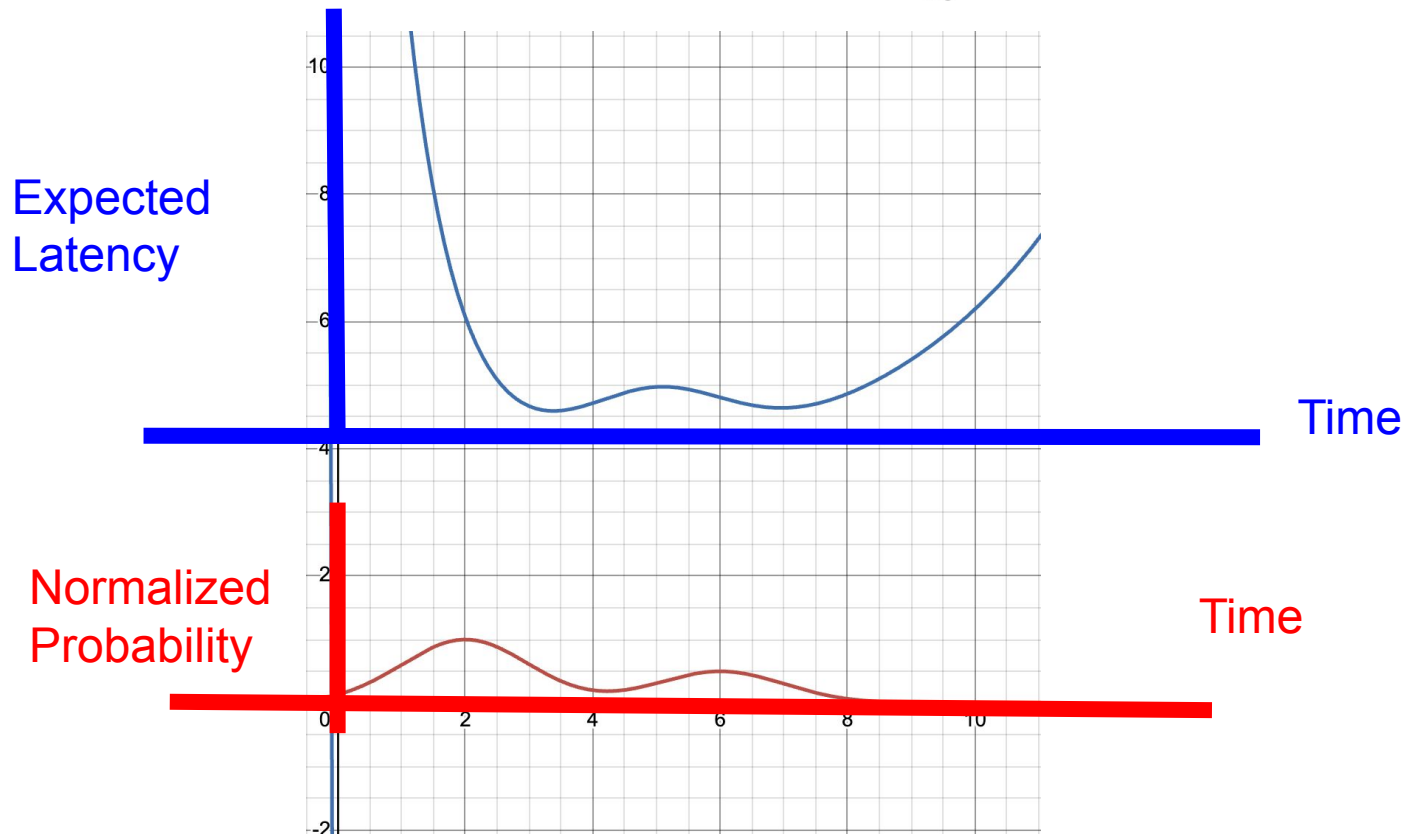
$$E(t) \int_0^t f(x)dx = \int_t^\infty f(x)dx \cdot t + \int_0^t f(x)dx \cdot \frac{\int_0^t x f(x)dx}{\int_0^t f(x)dx} + g(t) \int_0^\infty f(x)dx$$

$$E(t) = \frac{t \int_t^\infty f(x)dx + \int_0^t x f(x)dx + g(t) \int_0^\infty f(x)dx}{\int_0^t f(x)dx}$$

Performance (  $f(x) = e^{-\frac{(x-2)^2}{2}}$  ,  $g(x) = \frac{e^{\left(\frac{x}{3}\right)}}{10}$  )

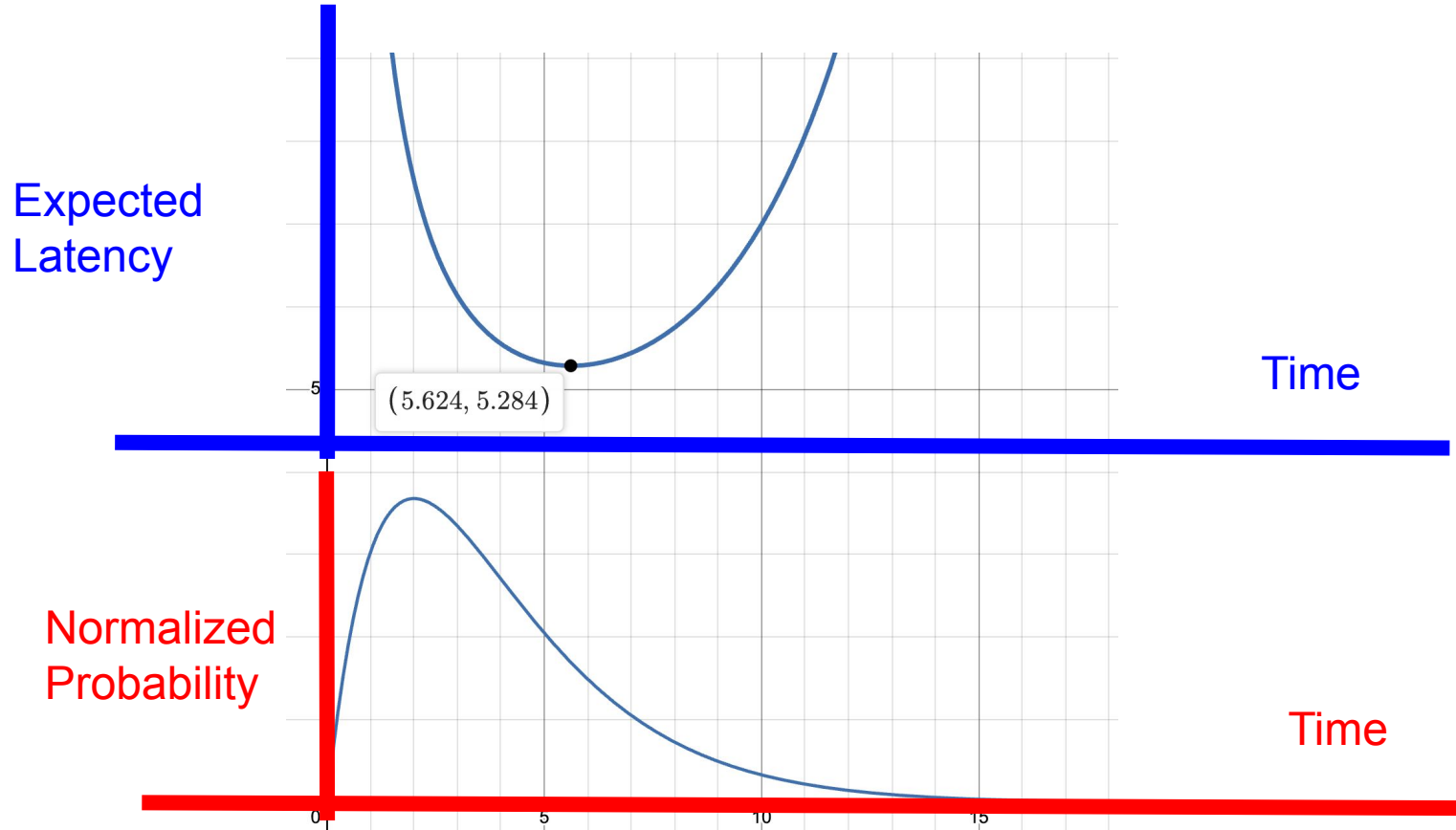


Performance (  $f(x) = e^{-\frac{(x-2)^2}{2}} + \frac{e^{-\frac{(x-6)^2}{2}}}{2}$  ,  $g(x) = \frac{e^{\left(\frac{x}{3}\right)}}{10}$  )





Performance (chi squared,  $g(x) = \frac{e^{\left(\frac{x}{3}\right)}}{10}$  )



# Extensions to a sequence of timeouts

Let's define a sequence of timeouts  $\{t_n\}_{n \geq 0}$

We can modify the original equation in the following way:

$$E_n = \frac{\int_{t_n}^{\infty} f(x) dx}{\int_0^{\infty} f(x) dx} (t_n + E_{n+1}) + \frac{\int_0^{t_n} f(x) dx}{\int_0^{\infty} f(x) dx} \cdot \frac{\int_0^{t_n} x f(x) dx}{\int_0^{t_n} f(x) dx} + g(t_n)$$

From now on, let  $E_n = a_n + b_n E_{n+1}$  where

$$a_n = \frac{\int_0^{t_n} x f(x) dx + t_n \int_{t_n}^{\infty} f(x) dx}{\int_0^{\infty} f(x) dx} + g(t_n)$$

And

$$b_n = \frac{\int_{t_n}^{\infty} f(x) dx}{\int_0^{\infty} f(x) dx}$$

(these are both functions of only  $t_n$ ) 26

(Derivation for  $a_n$ )

$$a_n = t_n \frac{\int_{t_n}^{\infty} f(x) dx}{\int_0^{\infty} f(x) dx} + \frac{\int_0^{t_n} f(x) dx}{\int_0^{\infty} f(x) dx} \cdot \frac{\int_0^{t_n} x f(x) dx}{\int_0^{t_n} f(x) dx} + g(t_n)$$

$$a_n = t_n \frac{\int_{t_n}^{\infty} f(x) dx}{\int_0^{\infty} f(x) dx} + \frac{\int_0^{t_n} x f(x) dx}{\int_0^{\infty} f(x) dx} + g(t_n)$$

$$a_n = \frac{\int_0^{t_n} x f(x) dx + t_n \int_{t_n}^{\infty} f(x) dx}{\int_0^{\infty} f(x) dx} + g(t_n)$$

# Extensions to a sequence of timeouts

The problem with the current equation is that it never ends.

$$E_n = a_n + b_n E_{n+1}$$

So, let's define a “terminal” timeout  $t_z$  : it's the last timeout in the list, and the timeout will never increase past it.

$$E_z = a_z + b_z E_z$$

$$E_z = \frac{a_z}{1 - b_z}$$

We now have the piecewise function

$$E_n = \begin{cases} a_n + b_n E_{n+1} & n < z \\ \frac{a_z}{1 - b_z} & n \geq z \end{cases}$$

# Extensions to a sequence of timeouts

How do we actually compute the timeouts?

$E_0$  is a function of every timeout, and we need to minimize it over  $t_0, t_1, t_2, \dots, t_z$

$$E_0 = a_0 + b_0 \left( a_1 + b_1 \left( a_2 + b_2 \left( \dots \frac{a_z}{1 - b_z} \dots \right) \right) \right)$$

$$a_n = \frac{\int_0^{t_n} x f(x) dx + t_n \int_{t_n}^{\infty} f(x) dx}{\int_0^{\infty} f(x) dx} + g(t_n) \qquad b_n = \frac{\int_{t_n}^{\infty} f(x) dx}{\int_0^{\infty} f(x) dx}$$

# Finding minima

We require that  $\nabla E_0 = \mathbf{0}$ .

$$\frac{\partial E_0}{\partial t_0} = \frac{\partial E_0}{\partial t_1} = \frac{\partial E_0}{\partial t_2} = \dots = \frac{\partial E_0}{\partial t_{z-1}} = \frac{\partial E_0}{\partial t_z} = 0$$

$$\frac{\partial E_0}{\partial t_0} = \frac{\partial (a_0 + b_0 E_1)}{\partial t_0} = \frac{da_0}{dt_0} + b_0 \frac{\partial E_1}{\partial t_0} + E_1 \frac{db_0}{dt_0} = \frac{da_0}{dt_0} + E_1 \frac{db_0}{dt_0}$$

$$\frac{\partial E_0}{\partial t_i} = \frac{\partial (a_0 + b_0 (a_1 + b_1 (a_2 + b_2 (\dots a_i + b_i E_{i+1} \dots))))}{\partial t_i}$$

$$\frac{\partial E_0}{\partial t_i} = \frac{\partial (b_0 b_1 \dots b_{i-1} (a_i + b_i E_{i+1}))}{\partial t_i}$$

$$\frac{\partial E_0}{\partial t_i} = b_0 b_1 \dots b_{i-1} \frac{\partial (a_i + b_i E_{i+1})}{\partial t_i}$$

$$\frac{\partial E_0}{\partial t_i} = b_0 b_1 \dots b_{i-1} \left( \frac{da_i}{dt_i} + E_{i+1} \frac{db_i}{dt_i} \right)$$

## Finding minima

Thus,

$$\frac{\partial E_0}{\partial t_i} = 0 \implies \frac{da_i}{dt_i} + E_{i+1} \frac{db_i}{dt_i} = 0,$$

As  $b_i$  are strictly positive.

# Finding minima

Well, what are  $\frac{da_i}{dt_i}$  and  $\frac{db_i}{dt_i}$ ?

$$\frac{da_i}{dt_i} = \frac{d}{dt_i} \left( \frac{\int_0^{t_i} x f(x) dx + t_i \int_{t_i}^{\infty} f(x) dx}{\int_0^{\infty} f(x) dx} + g(t_i) \right)$$

$$\frac{da_i}{dt_i} = \frac{t_i f(t_i) + t_i (-f(t_i)) + \int_{t_i}^{\infty} f(x) dx}{\int_0^{\infty} f(x) dx} + g'(t_i)$$

$$\frac{da_i}{dt_i} = \frac{\int_{t_i}^{\infty} f(x) dx}{\int_0^{\infty} f(x) dx} + g'(t_i)$$

$$\frac{db_i}{dt_i} = \frac{d}{dt_i} \left( \frac{\int_{t_i}^{\infty} f(x) dx}{\int_0^{\infty} f(x) dx} \right) = \frac{-f(t_i)}{\int_0^{\infty} f(x) dx}$$



## Finding minima

So,

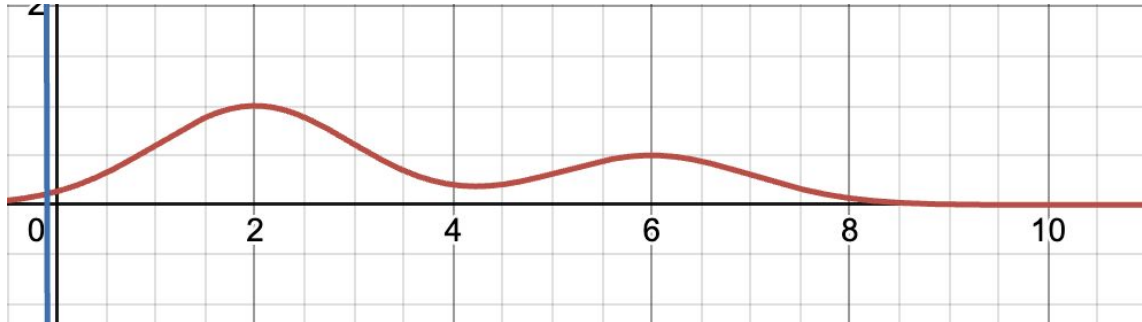
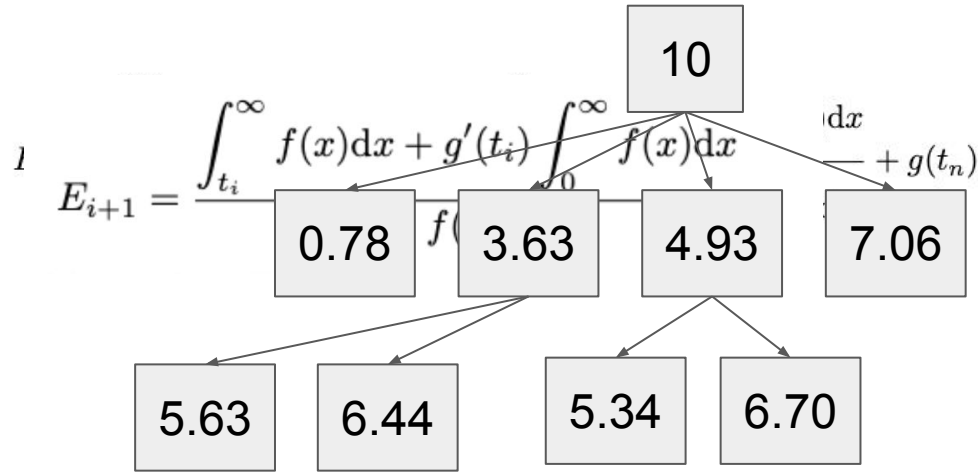
$$\frac{\int_{t_i}^{\infty} f(x)dx}{\int_0^{\infty} f(x)dx} + g'(t_i) + E_{i+1} \left( \frac{-f(t_i)}{\int_0^{\infty} f(x)dx} \right) = 0.$$

$$\int_{t_i}^{\infty} f(x)dx + g'(t_i) \int_0^{\infty} f(x)dx - f(t_i)E_{i+1} = 0$$

$$E_{i+1} = \frac{\int_{t_i}^{\infty} f(x)dx + g'(t_i) \int_0^{\infty} f(x)dx}{f(t_i)}$$

As we know  $t_z$ , and have that  $E_z = \frac{a_z}{1-b_z}$ , we can thus work our way backwards to determine all  $t_i$ .

# Math behind mathematical model (3)

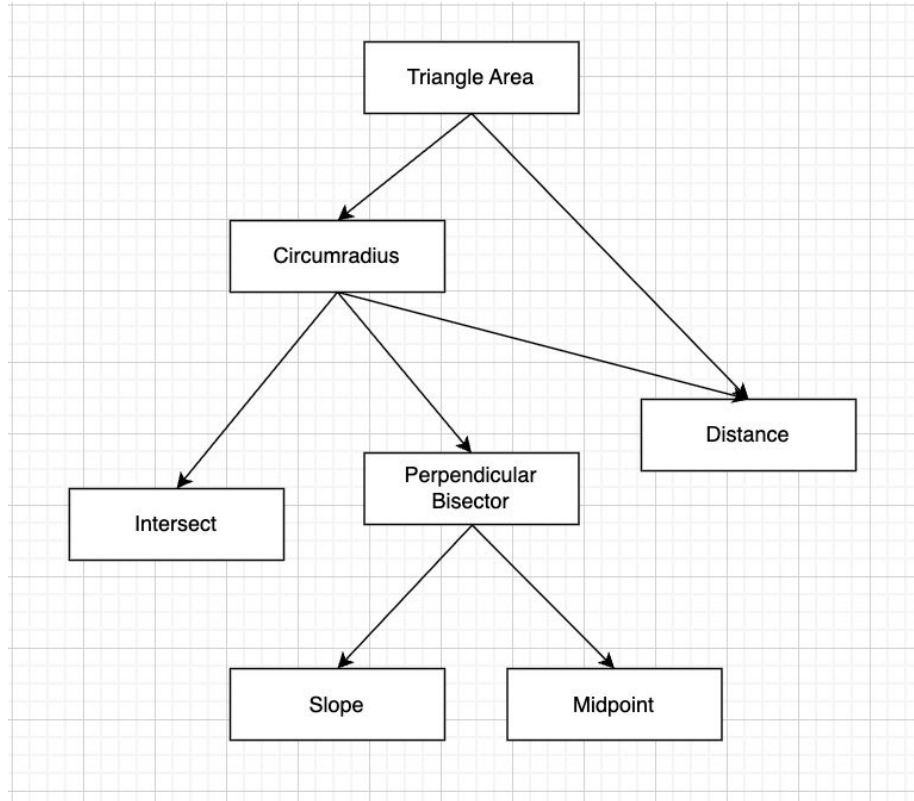


$$E_{i+1} = \frac{\int_{t_i}^{\infty} f(x)dx + g'(t_i) \int_0^{\infty} f(x)dx}{f(t_i)}$$

```

0.78
  10
  ---> 6.33
3.63
  5.63
    10
    ---> 3.53
  6.44
    10
    ---> 3.49
  10
  ---> 3.84
4.93
  5.34
    10
    ---> 3.89
  6.7
    10
    ---> 3.84
  10
  ---> 4.16
7.06
  10
  ---> 3.93
10
---> 5.61
    
```

# Testbed Architecture



# Global Architecture

