

# File Fragmentation over an Unreliable Network

## Bachelor Thesis

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September 7, 2009



- Imagine yourself playing an action game



- There is always a risk of getting caught, which increases by time

- If you get caught, you have to start over where you last saved
- But saving very frequently would cost too much time

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# Basic model

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# Large files are likely to be corrupted

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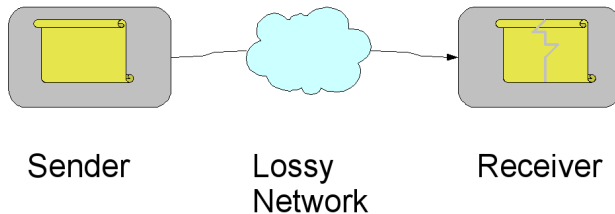
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# Basic model

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- Smaller fragments are less likely to be corrupted

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# Fragmentation prevents corruption

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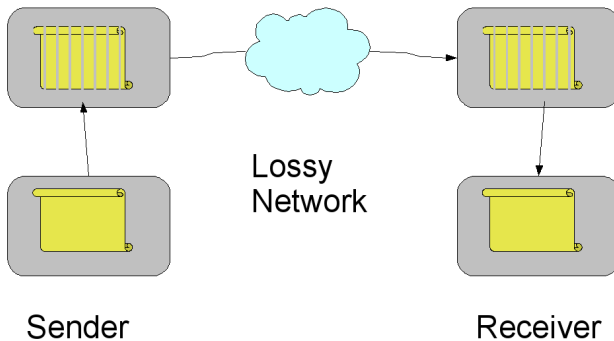
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- Packets however, carry an overhead
- This makes too small packet sizes inefficient



- What is the optimal fragment size?

# Mathematical model

- $x_i + \phi$  denotes the packet size ( $x_i$  data size,  $\phi$  overhead size)
- Transmission will succeed if  $A_i \geq x_i + \phi$  ( $A_i$  random variable with distribution  $f$ )

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# Expected transmission time

$$J = \sum_{i=0}^{M-1} \frac{x_i + \phi}{\bar{F}(x_i + \phi)} = \sum_{i=0}^{M-1} h(x_i) \quad \bar{F}(x) = P(A_i > x) \quad (1)$$

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# Minimizing J

The KKT condition for (1) is

$$h'(x_i) = \frac{1}{\bar{F}(x_i + \phi)} + (x_i + \phi) \frac{f(x_i + \phi)}{(\bar{F}(x_i + \phi))^2} = \lambda$$

$$\sum_{i=0}^{M-1} x_i = L \quad (2)$$

## Lemma

*If the failure rate (failure risk)  $f(x)/\bar{F}(x)$  is increasing:*

- *The optimization over  $x_i$  is convex*
- *The solution is  $x_i^* = x^*$  for all  $i$  (packet sizes are equal)*

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- Seek minimizer  $M^*$  by inserting  $x^* = L/M$  into (1):

$$J = M \frac{\frac{L}{M} + \phi}{\bar{F}\left(\frac{L}{M} + \phi\right)} = L \frac{\frac{L}{M} + \phi}{\frac{L}{M} \bar{F}\left(\frac{L}{M} + \phi\right)} := L \cdot g\left(\frac{L}{M}\right) \quad (3)$$

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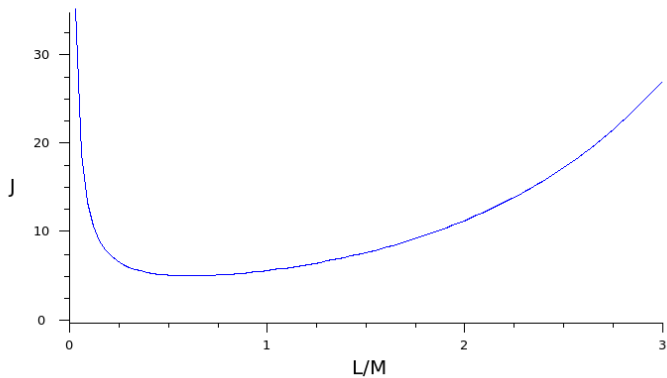


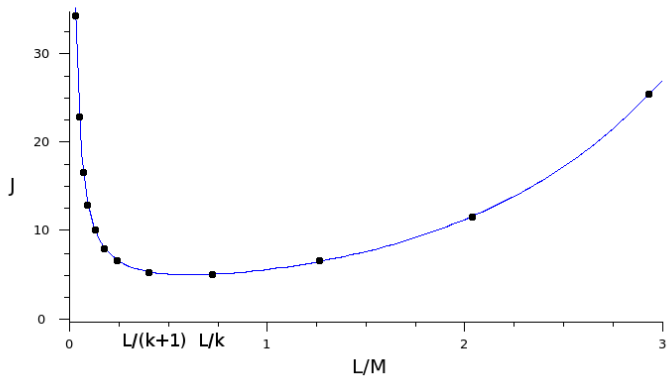
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# Bounding $x^*$

## Definition

$$a := \operatorname{argmin}_x (g(x))$$

## Theorem

If  $L > a$ :

$$\frac{a}{2} \leq \frac{a}{1 + a/L} \leq x^*(L) \leq \min \left( 2a, \frac{a}{1 - a/L} \right)$$

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- Constant fragmentation is optimal
- The fragment sizes are unique
- If the file size and distribution  $f$  are known, the file can be pre-fragmented
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