Improving Rumor Spreading

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2 Multisource Rumor Spreading with Network Coding

3 Rumor Spreading Adaptiveness: Adjust Pull Frequency

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What achieves Rumor Spreading?

Deliver a message to every nodes of a group with high probability.



Nodes in blue know the message

Use cases for Rumor Spreading

- Collaborative Platforms
 - Chat, Communication Platform (eg. Rumor Monger)
 - Collaborative Editors (eg. CRATE)
- Collaborative News Feeds (could replace WebSub or Atom/RSS)
- Sensor Network
- Membership

Basically any publish/subscribe mechanism

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Ask for the message: a push-pull protocol

It is costly to reach every nodes with high probability. Adding a pull phase can help last nodes retrieve the message.



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With P being the payload of a message, how to know if it's a new message? How to discover missing messages?

• Add a unique identifier *I_i* to every messages

$$(A) \xrightarrow{\text{send } I_3, P} (B) \xrightarrow{\text{send } I_4, P} (C)$$

- Add a unique identifier I_i to every messages
- Piggyback a list of known identifiers on every "send" messages

$$(A) \xrightarrow{\text{send } [I_1, I_2], I_3, P} (B) \xrightarrow{\text{send } [I_1, I_4], I_3, P} (C)$$

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B already knows one message, either I_1, P_1 or I_2, P_2 .



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• Encoding messages increase the chance of providing new information.

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$$(A) \xrightarrow{\text{send } 1P_1 + 1P_2 \text{ over } \mathsf{GF}(2)} (B)$$

- Encoding messages increase the chance of providing new information.
- It can be expressed over a Galois Field

B already knows one message, either I_1, P_1 or I_2, P_2 .

$$(A) \xrightarrow{\text{send } \varnothing, \{I_1 : 1, I_2 : 1\}, X_{P_1, P_2}} (B)$$

- Encoding messages increase the chance of providing new information.
- It can be expressed over a Galois Field
- Identifier is replaced by a list of identifiers and their associated coefficient. X_{P1,P2} = P1 ⊕ P2 in this case.

Random coefficients with independant linear combination? Use a bigger Galois Field. Example with GF(256):



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• Choose random coefficients: $\{I_1 : 16, I_2 : 3, I_3 : 21\}$

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- Choose random coefficients: $\{I_1 : 16, I_2 : 3, I_3 : 21\}$
- Compute new payload: $X = 16P_1 + 3P_2 + 21P_3$ over GF(256)

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- Choose random coefficients: $\{I_1 : 16, I_2 : 3, I_3 : 21\}$
- Compute new payload: $X = 16P_1 + 3P_2 + 21P_3$ over GF(256)
- Send the encoded message

Group messages in generations

Coding explosion:



3 x 3

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Group messages in generations

Encoding is only done between message of the same generation:



How to assign a generation to a message?

- Encode messages that are emitted in the same timeframe
- Every nodes start with a logical clock at zero
- When a node emits a message, it will assign its clock as message generation then increase its clock
- If a node forwards a message tagged with a generation bigger than its local clock, it updates its clock to match the generation it has seen

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Comparison against Pulp, a push-pull algorithm



Figure: Pulp original



Figure: Our algorithm

Why pull frequency must be adapted



Figure: Too fast



Figure: Too slow

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Can we use control theory?

We want:

- to control
 - pull usefulness (number of useful pull / number of total pull)
 - delivery delay (by monitoring known missing messages evolution)
- to take action on pull frequency

Looked at:

- Open Loop vs Closed Loop
- Binary Feedback (used in TCP Congestion Control)
 - Additive (or Multiplicative) Increase, Additive (or Multiplicative) Decrease
- Proportional-Integral-Derivative (PID) Controller

Problem: unstable environment (message rate, loss, churn, network size, latency)

- Adaptive Controller
- System Identification

Pulp adaptiveness analysis

Reminder: control usefulness and missing messages, act on pull frequency

Take a sample every $\Delta adjust$ time 2 strategies:

- Δ*missing* increased
 - Apply this formula (open loop):

$$\Delta pull = rac{\Delta adjust}{\Delta missing + useful}$$

- otherwise, apply Multiplicative Increase Multiplicative Decrease
 - if *usefulness* > 0.5, decrease $\Delta pull$
 - otherwise increase $\Delta pull$

First experimentations

Start at 1.5 msg/s, then emit 150 msg/s, then go back to 1.5 msg/s We graph usefulness (useful pull / total pull) over time



Conclusion and future works

Goal: provide a versatile Gossip-based dissemination brick by:

- Improving performances
- Improving ease of use

Future work:

- Better handle unstable environment
- Provide adaptiveness on more parameters
- Create a real world implementation
- Model problem, provide bounds, etc.