#### Consistency without concurrency control in large, dynamic systems

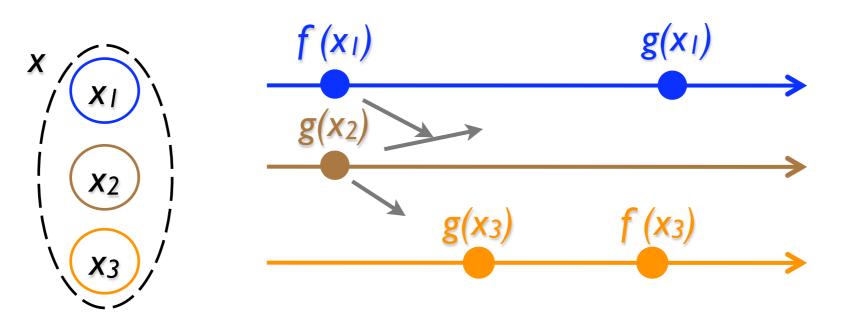
Marc Shapiro, INRIA & LIP6 Nuno Preguiça, Universidade Nova de Lisboa Mihai Leția, ENS Lyon



Two things

- no concurrency control
- large dynamic

# Consistency without concurrency control



Object x, operation f(x)

- propose f(x<sub>1</sub>)
- eventually replay f(x<sub>2</sub>), f(x<sub>3</sub>), ...
- If f || g commute: converges safely without concurrency control

#### Commutative Replicated Data Type (CRDT): Designed for commutative operations

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Not same order at 1 and 2? OK if

 concurrent f and g commute Assuming causal delivery

# A sequence CRDT

Treedoc = sequence of elements:

- insert-at-pos, delete
- Commutative when concurrent
- Minimise overhead
- Scalable

A commutative replicated data type for cooperative editing, ICDCS 2009

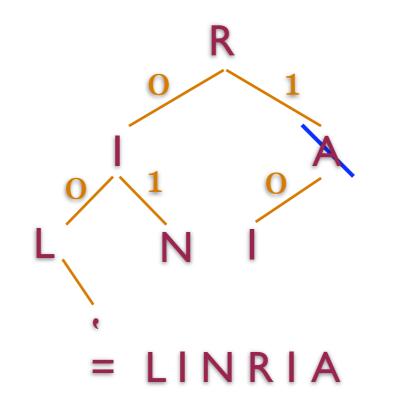
Focus today:

- Garbage collection
- vs. scale

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I will just skim the surface of the Treedoc design Refer to ICDCS paper for the details

#### Commutative updates



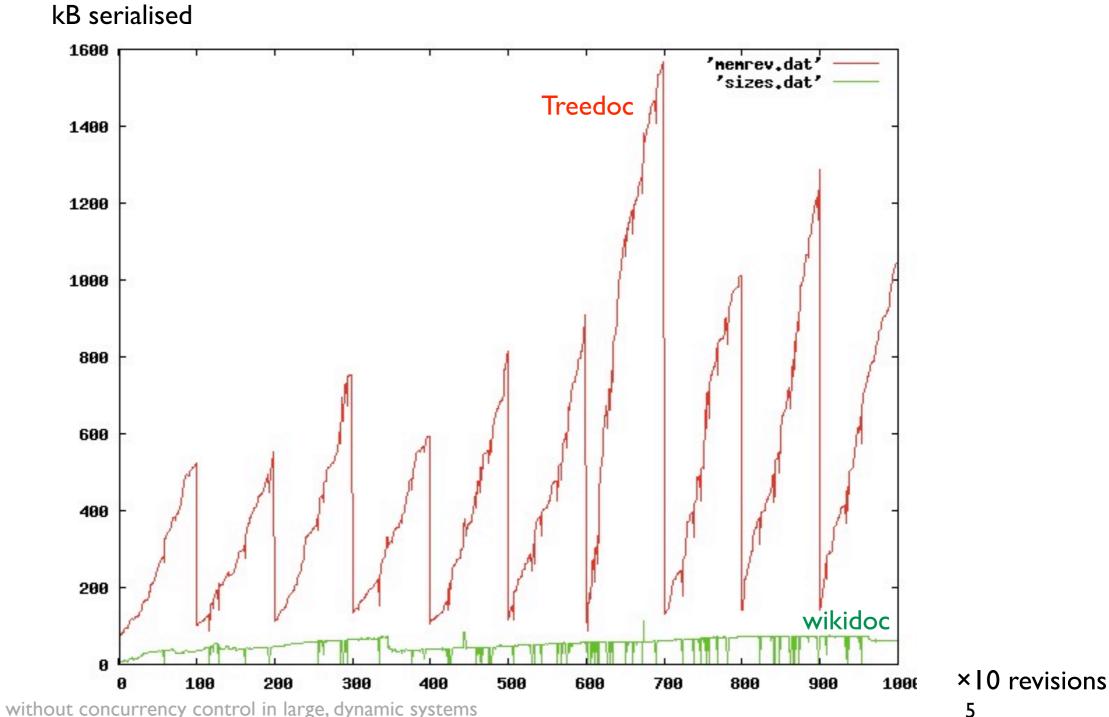
#### Naming tree: minimal, self-adjusting: logarithmic TID: path = $[0|1]^*$ Contents: infix order *insert* adds leaf $\Rightarrow$ non-destructive, TIDs don't change Delete: *tombstone*, TIDs don't change

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Thanks to non-destructive updates, immutable TIDs: concurrent updates commute Efficient: Data structures and TID lengths logarithmic \*if balanced\* Ignoring lots of details, e.g. concurrent inserts at same position (see paper)

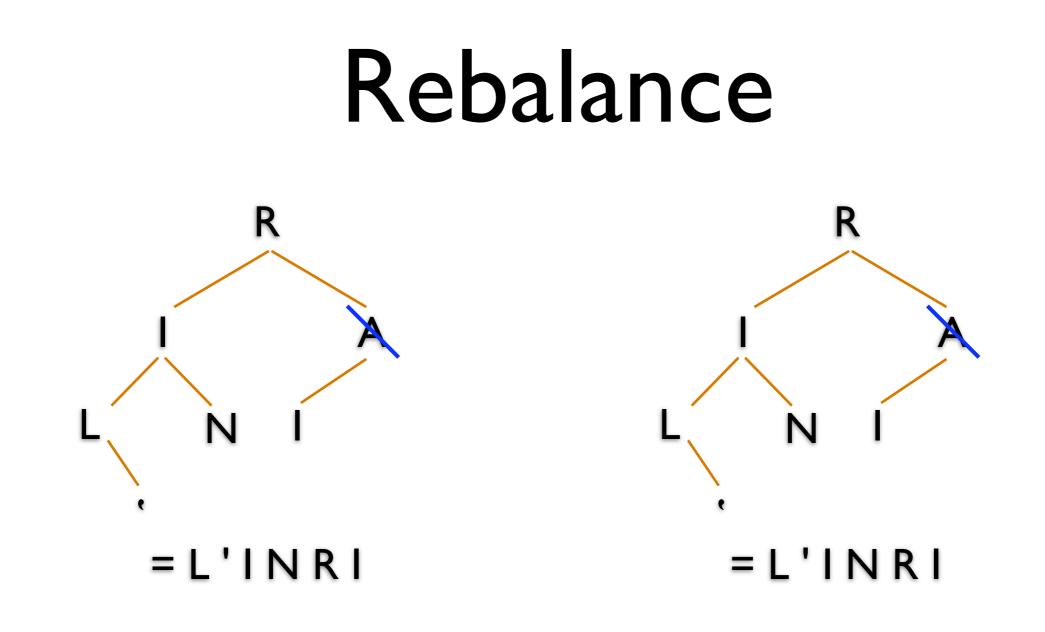
# Wikipedia GWB page: space overhead



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#### GWB: most edited page

- Edits translated into treedoc insert/deletes
- Tree unbalanced, long TIDs, lots of tombstones: not logarathmic



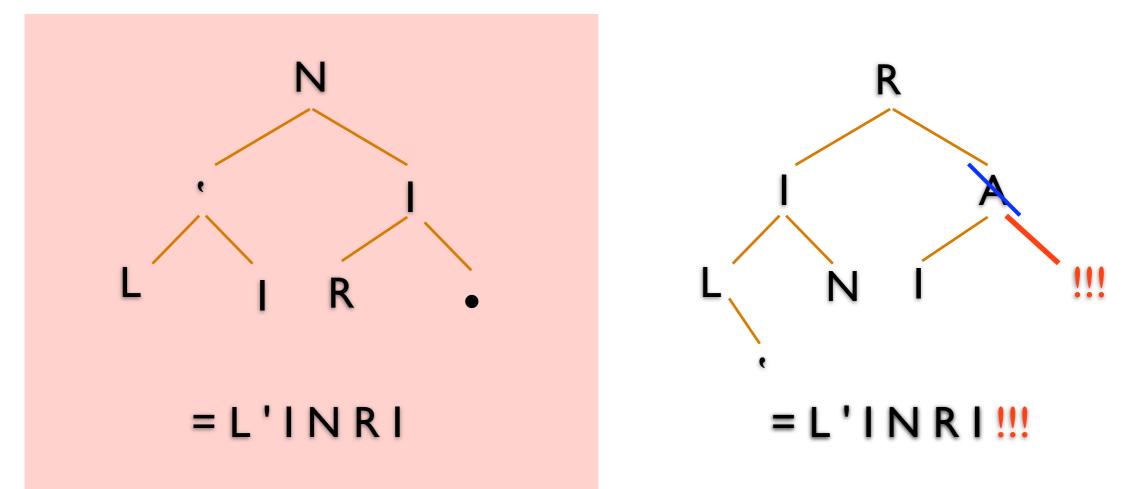
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In this example rebalancing is not spectacular. Imagine a deep unbalanced tree with lots of tombstones: large effect. Why rebalance:

- Unbalanced tree costs time, space
- Long TIDs
- Tombstone overhead

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#### Rebalance



Invalidates TIDs:

- Frame of reference = epoch
- Requires agreement
- Pervasive!
  - e.g. Vector Clocks

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TID changed: R was ε, now 10 Pervasive problem:

- asynchronous updates ==> old data structures
- see cleaning up Vector Clocks

#### (Background colour indicates epoch)

# Rebalance in large, dynamic systems

Rebalance requires consensus

Consensus requires small, stable membership

- Large communities?!
- Dynamic scenarios?!

Solution: two tiers

- Core: rebalancing (and updates)
- Nebula: updates (and rebalancing)
- Migration protocol

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Core: controls rebalancing

Core

#### Nebula

#### Group membership Small, stable *Rebalance:*

- Unanimous agreement (2-phase commit)
- All core sites in same epoch

Arbitrary membership Large, dynamic Communicate with sites in same epoch only *Catch-up* to rebalance, join core epoch

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Migrate core to nebula: just leave group



### Nebula

Group membership Small, stable *Rebalance*:

- Unanimous agreement (2-phase commit)
- All core sites in same epoch

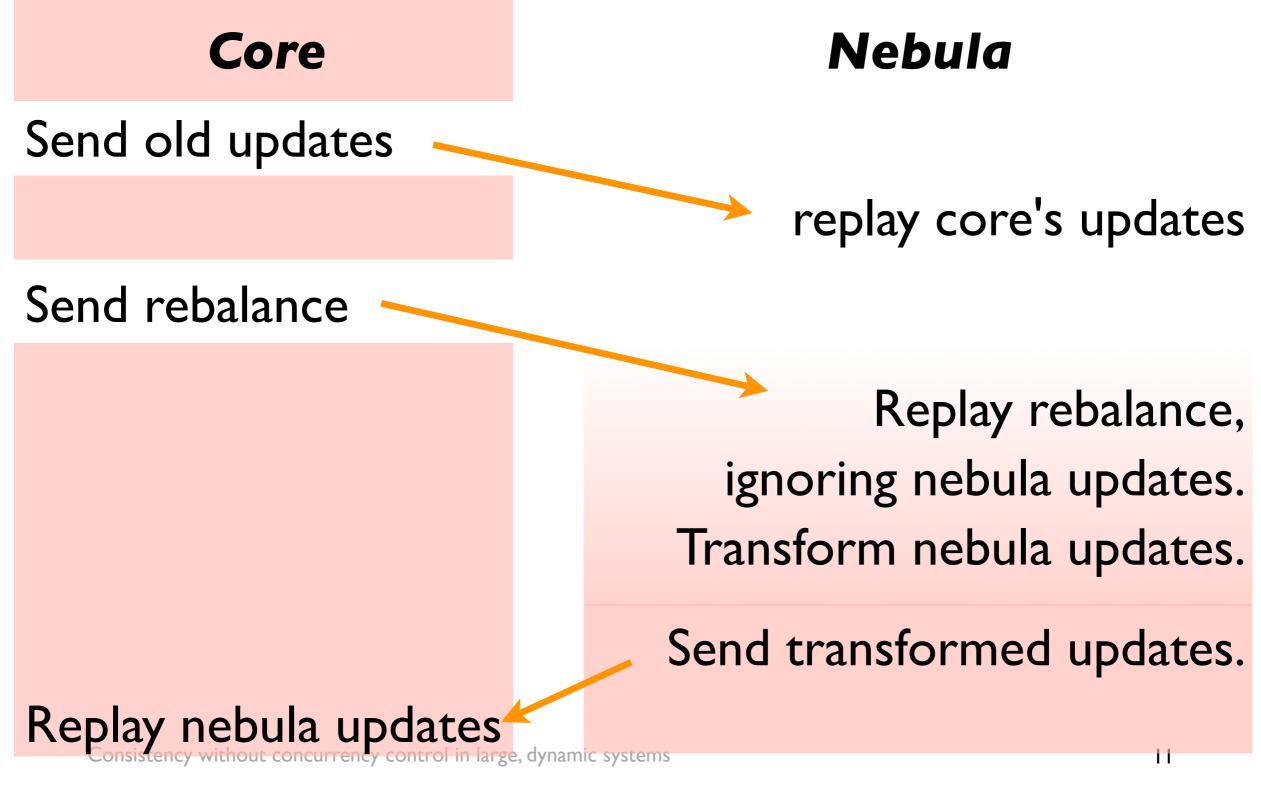
Arbitrary membership Large, dynamic Communicate with sites in same epoch only *Catch-up* to rebalance, join core epoch

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Migrate from nebula to core: migrate to core epoch + join group

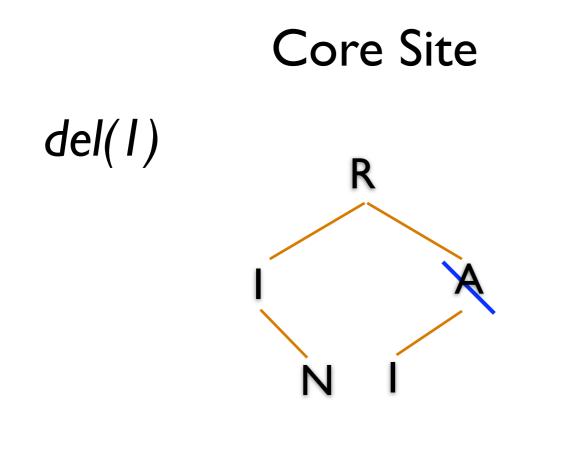
# Catch-up protocol summary

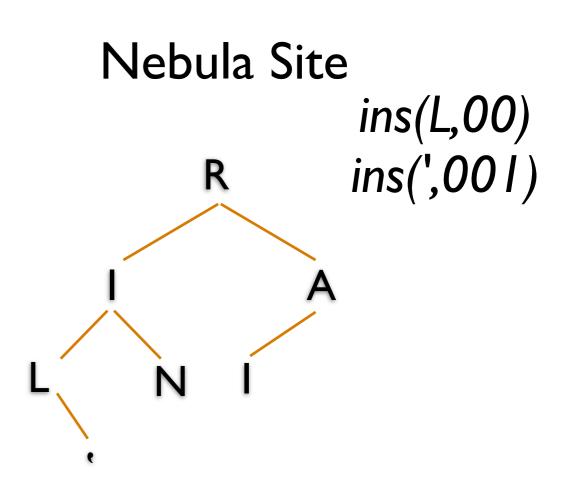


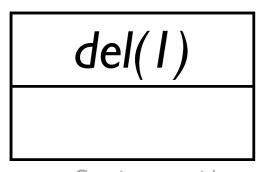
Here is the basic insight to the migration protocol

- Replay core's updates: N now in same state as C before rebalance
- Replay rebalance: \*ignoring concurrent N updates\*, has same effect as in C ==> same TIDs, same epoch
- Transform buffer: now ready to be replayed in C (in different order, but that's OK since they commute)
- N now in C state, can join the core or remain in nebula

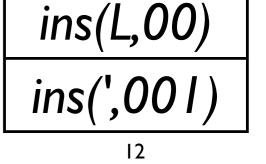
Furthermore updates are idempotent (multiple catch-ups cause no harm)





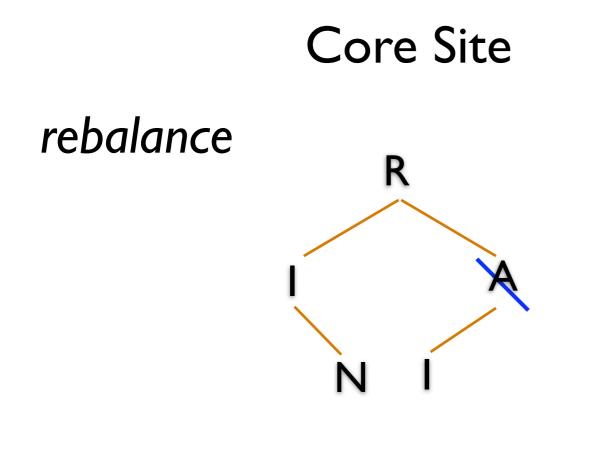


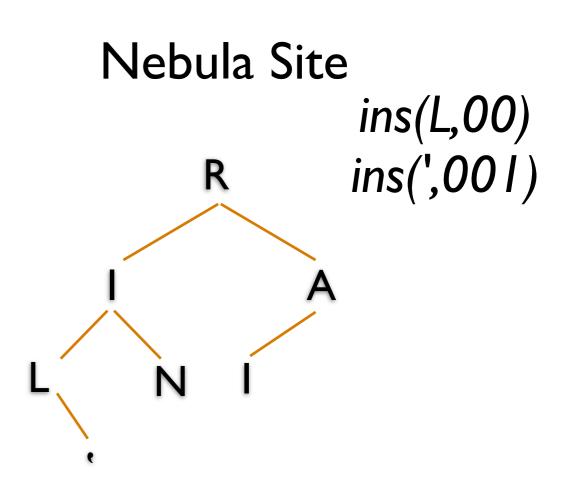
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del in old epoch old epoch rebalance starts new epoch

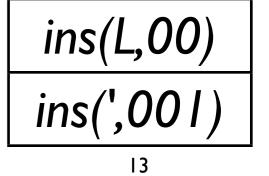
ins + ins in



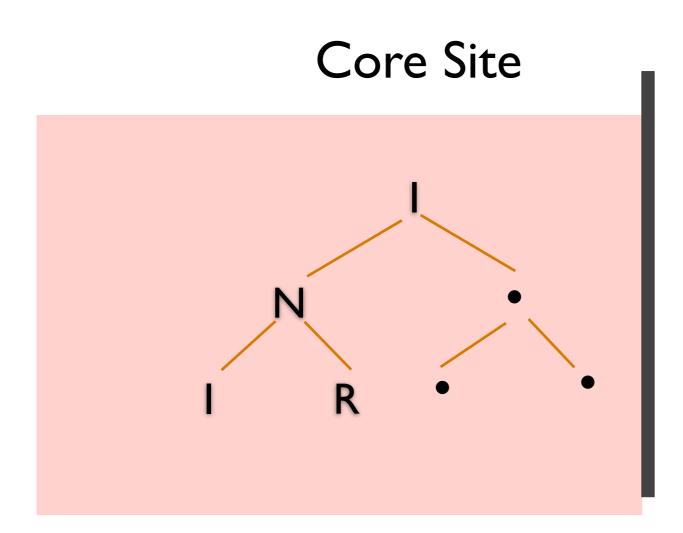


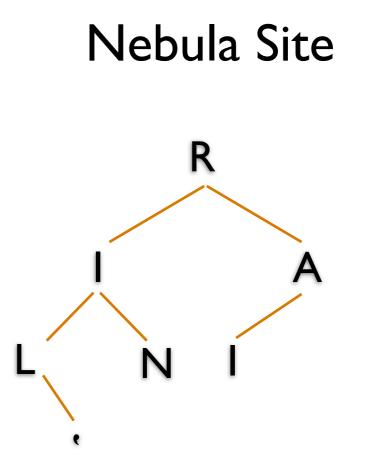


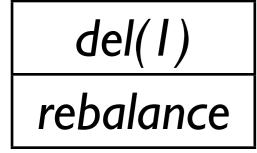
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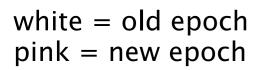
del in old epoch old epoch rebalance starts new epoch ins + ins in

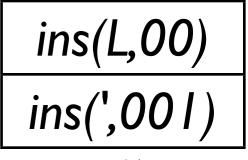




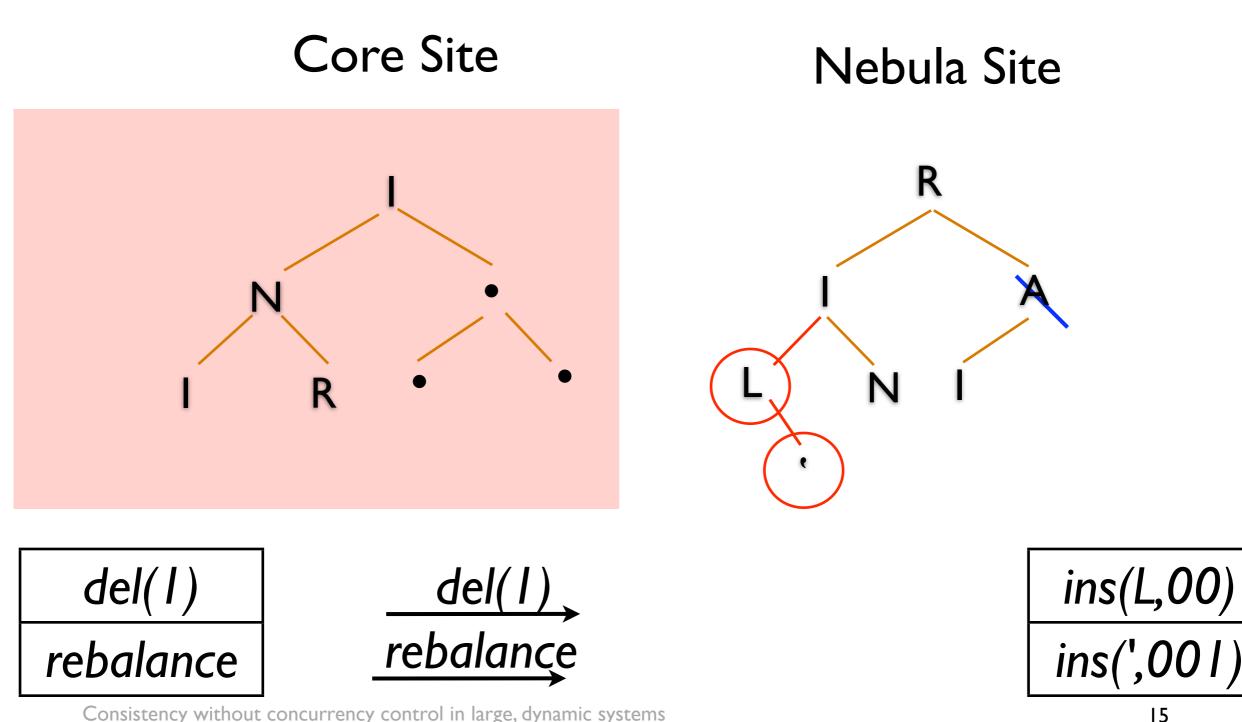


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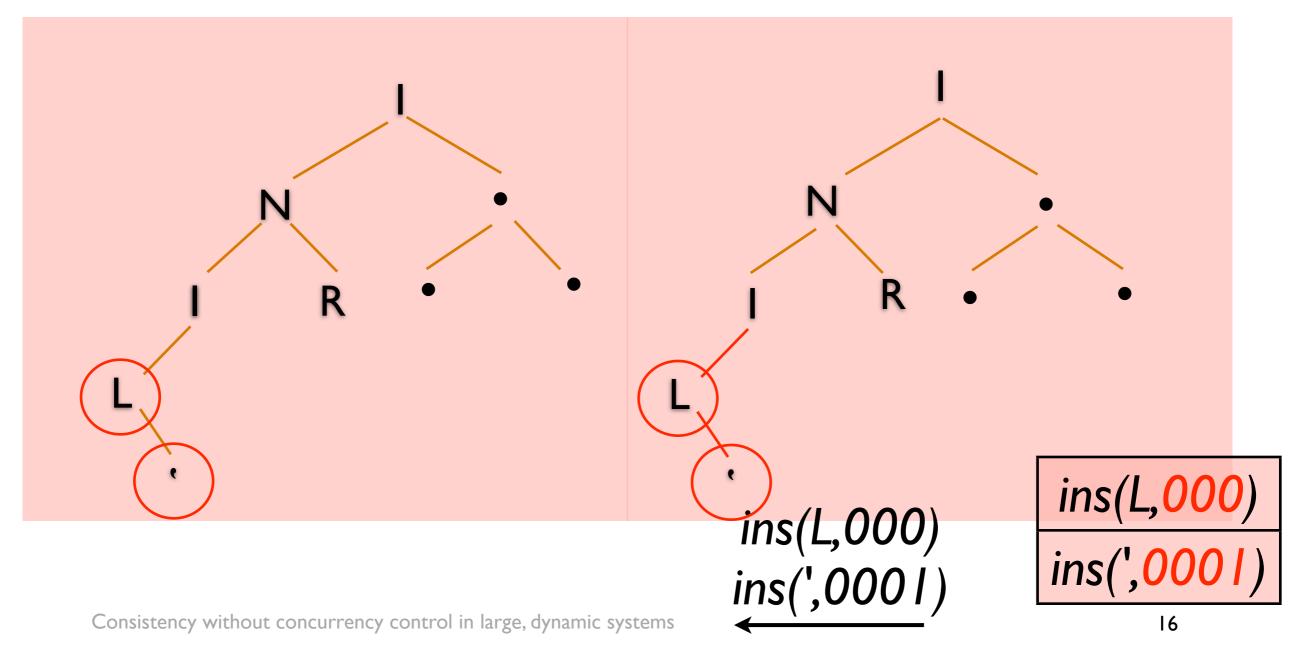
del uttered in old epoch ==> can send to S now up to date with Core send rebalance

S replays del;

S replays rebalance; intervening ins move; S now in new epoch



Nebula Site



ins arguments transformed to new epoch

Core replays ins

# Summary

#### CRDT:

- Convergence ensured
- Design for commutativity

#### GC cannot be ignored

- Requires commitment
- Pervasive issue

#### Large-scale commitment:

- Core / Nebula
- To synchronise: catch-up
  + migration

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CRDT = non-locking synchronisation in weak memory model

#### Future work

More CRDTs Understanding CRDTs: what invariants can be CRDTized Approximations of CRDTs Data types for consistent cloud computing without concurrency control

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