

The Case of the Fake Picasso! Preventing History Forgery with Secure Provenance



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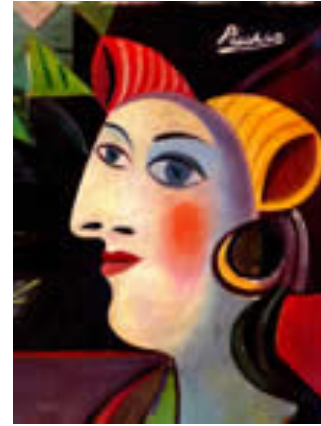
xerox



Let's play a game



Real, worth **\$101.8** million



Fake, listed at eBay,
worth nothing

Can you spot the fake **Picasso**?

So, how do art buyers authenticate art?

Among other things, they look at **provenance records**



The Museum of Modern Art

Provenance Research Project

Pablo Picasso (Spanish, 1881–1973. To France 1904.)
Painter and Model [L'artiste et son modèle], 1928

Oil on canvas, 51 1/8 x 64 1/4" (129.8 x 163 cm)
The Museum of Modern Art, New York; The Sidney and Harriet Janis Collection.

Collection work meeting criteria specified in Introduction.
644.67

Other works by this artist

Provenance:
Paul Rosenberg, Paris. Acquired from the artist in 1928 - 1933
Sidney and Harriet Janis, New York (a.k.a. Sidney Janowitz). Acquired from Rosenberg, 1933 - 1967
The Museum of Modern Art, New York; The Sidney and Harriet Janis Collection, 1967

Alternate titles:
The Painter and His Model
Le peintre et son modèle

Provenance: from Latin *provenire* 'come from', defined as

"(i) the fact of coming from some particular source or quarter; origin, derivation.

(ii) the history or pedigree of a work of art, manuscript, rare book, etc.; a record of the ultimate derivation and passage of an item through its various owners" (Oxford English Dictionary)

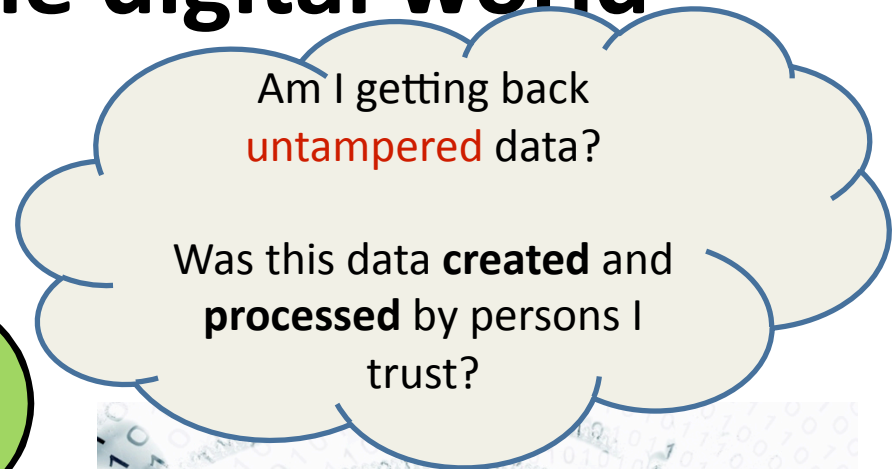
In other words, **who owned it, what was done to it, how was it transferred ...**

Widely used in arts, archives, and archeology, called the Fundamental Principle of Archival.

L'artiste et son modèle (1928), at Museum of Modern Art

<http://moma.org/collection/provenance/items/644.67.html>

Let's consider the digital world



Unlike data processing in the past, digital
data is generated, processed, and represented on digital data
To trust data we receive from others or retrieve from storage,
we need to look into the integrity of both the **present state** and
different systems and principals, **Asset Data** database/
the **past history** of data storage

What exactly is data provenance?

Definition*

- Description of the **origins** of data and the **process** by which it arrived at the database. [Buneman et al.]
- Information describing materials and **transformations** applied to derive the data. [Lanter]
- Information that helps determine the **derivation history** of a data product, starting from its original sources. [Simmhan et al.]

*Simmhan et al. A Survey of Provenance in E-Science. SIGMOD Record, 2005.

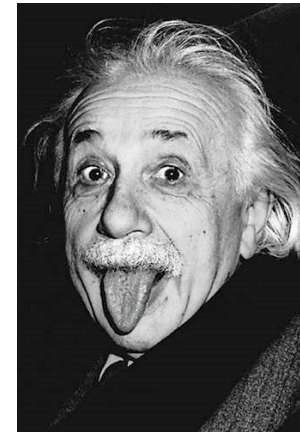
Example provenance systems

	Lanter, D. P. (LIP)	Chimera	MyGRID	CMCS	PASOA	ESSW	Tioga	Buneman, P.	Cui, Y., Widom, J. (Trio)
Applied Domain	GIS	Physics, Astronomy	Biology	Chemical Sciences	Biology	Earth Sciences	Atmospheric Science	Generic (Scientific databases)	Generic
Data Processing Framework	Command Processing	Service Oriented	Service Oriented	Service Oriented	Service Oriented	Script Based	Relational Database	Relational/Semi Structured Database	Relational Database
Application of Provenance	Informational; update stale, regenerate & compare data	Informational; Audit; Data Regeneration; Planning	Contextual Information; Re-enactment	Informational; Data Update	Informational; Re-enactment	Informational	Informational; Track errors	Annotation propagation; View Update	Information; update propagation
Data/Process Oriented	Data	Process	Process	Data	Process	Both	Data	Data	Data
Granularity	Spatial layers	Abstract datasets (Currently files)	Abstract resources having LSID	Files	Abstract parameters to Workflow	Files	Attributes in Database	Attributes & Tuples in Databases	Tuples in Database
Representation Scheme	Commands & Frames as Annotations	Virtual Data Language Annotations	XML/RDF Annotations	Dublin Core XML Annotations	Annotations	XML/RDF Annotations	Inverse Functions	Inverse Queries	Inverse queries
Semantic Information	No	No	Yes	Limited	No	No, Proposed	No	No	No
Storage Repository/ Backend	MetaDatabase	Virtual Data Catalog/ Relational DB	mIR repository/ Relational DB	SAM over WebDAV/ Relational DB	PreServ/ Relational DB, File System	Lineage Server/ Relational DB	Relational DB	N/A	Relational DB
Provenance Collection Overhead	Store User commands; solicit metadata	User defines derivations; automated WF trace	User defines service semantics; Automated WF Trace	Manual; Apps use DAV APIs, Users use portal	Manual; Actors use PReP API	Libraries assist user to generate, store provenance	User registers inverse functions	N/A	Inverse queries automatically generated
Addressed Scalability	No	Yes	No	No	No (Proposed)	No (Proposed)	Yes	N/A	No
Provenance Dissemination	Queries	Queries	Semantic browser; Lineage graph	Browser; Queries; GXL/RDF	Queries	Browser	Queries; box-and-arrows visualization	N/A	SQL/TriQL Queries

Simmhan et al., 2005

What was the common theme of all those systems?

- They were all scientific computing systems
- And scientists trust people (more or less)
- Previous research covers provenance collection, annotation, querying, and workflow, but **security** issues are **not** handled
- For provenance in untrusted environments, we need **integrity, confidentiality** and **privacy** guarantees



So, we need **provenance of provenance**, i.e. a model for **Secure Provenance**

Secure provenance means preventing “undetectable history rewriting”

- Adversaries cannot insert fake events, remove genuine events from a document’s provenance
- No one can deny history of own actions
- **Allow** fine grained preservation of privacy and confidentiality of actions
 - Users can choose which auditors can see details of their work
 - Attributes can be selectively disclosed or hidden without harming integrity check

Usage and threat model



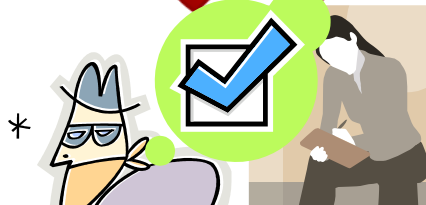
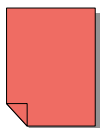
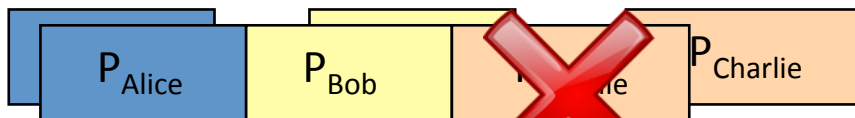
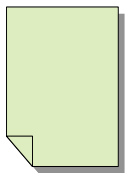
Alice



Bob



Charlie



Audrey



Marvin

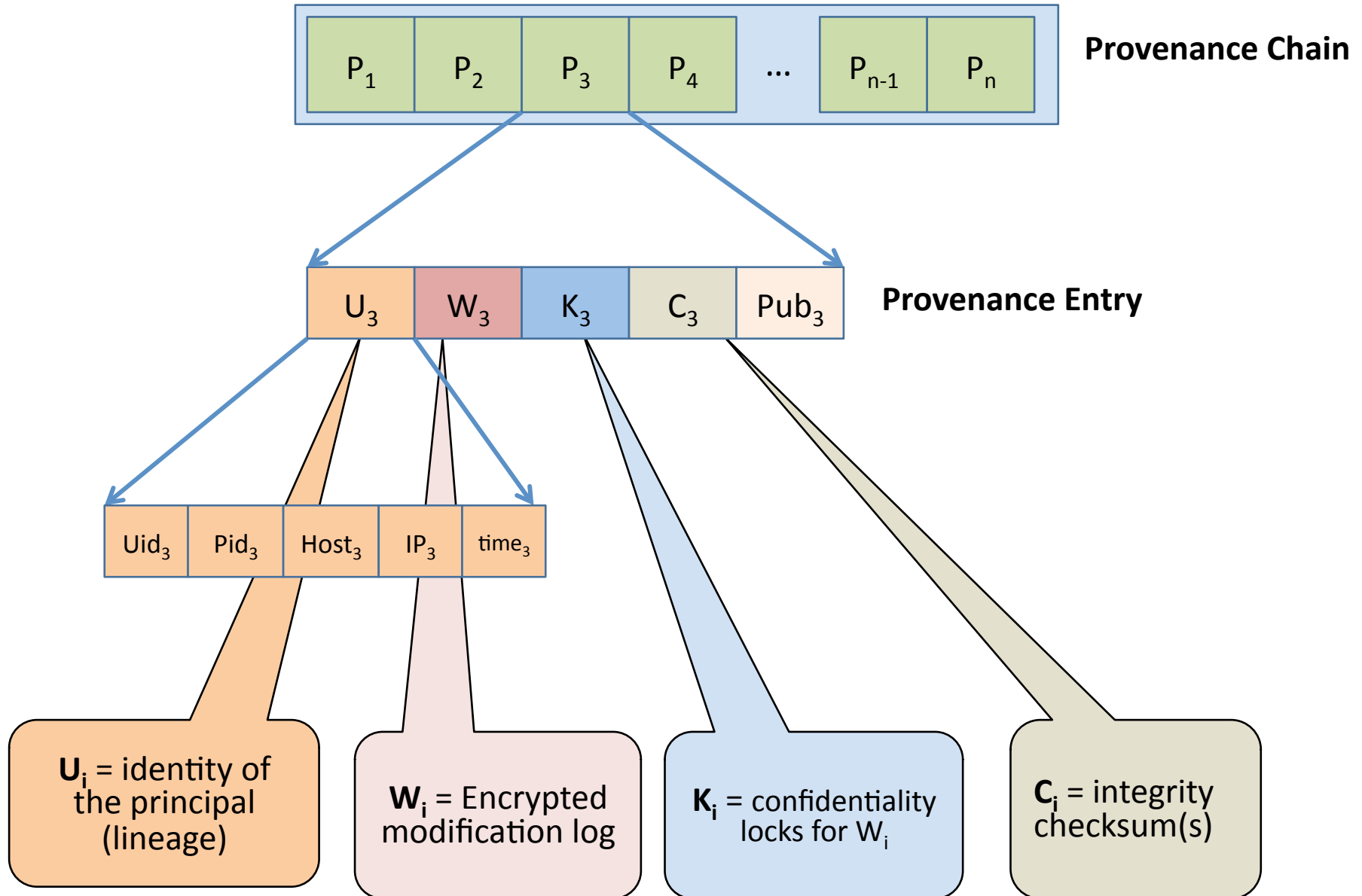
- **Users:** Edit **documents** on their machines
- **Auditors:** semi-trusted principals
 - All auditors can verify chain integrity
- **Adversaries:** insiders or outsiders who
 - Add or remove history entries
 - Collude with others to add/remove entries
 - Claim a chain belongs to another document
 - Repudiate an entry

Ragib Hasan, Radu Sion, and Marianne Winslett, "Introducing Secure Provenance: Problems and Challenges", ACM StorageSS 2007

Previous work on integrity assurances

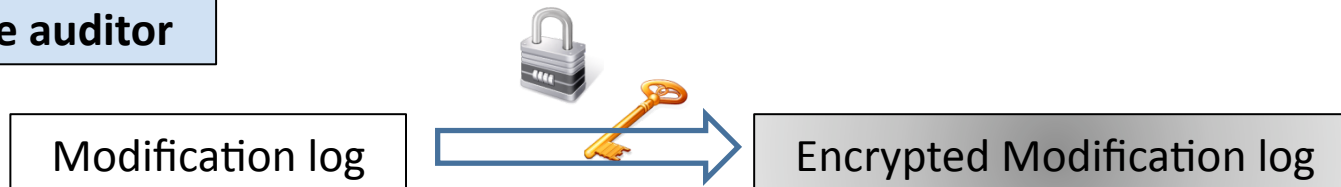
- (Logically) centralized repository (CVS, Subversion, GIT)
 - Changes to files recorded
 - Not applicable to mobile documents
- File systems with integrity assurances (SUNDR, PASIS, TCFS)
 - Provide local integrity checking
 - Do not apply to data that traverses systems
- System state entanglement (Baker 02)
 - Entangle one system's state with another, so others can serve as witness to a system's state
 - Not applicable to mobile data
- Secure audit logs / trails (Schneier and Kelsey 99), LogCrypt (Holt 2004), (Peterson et al. 2006)
 - Trusted notary certifies logs, or trusted third party given hash chain seed

Our solution: Overview



Our solution: Confidentiality

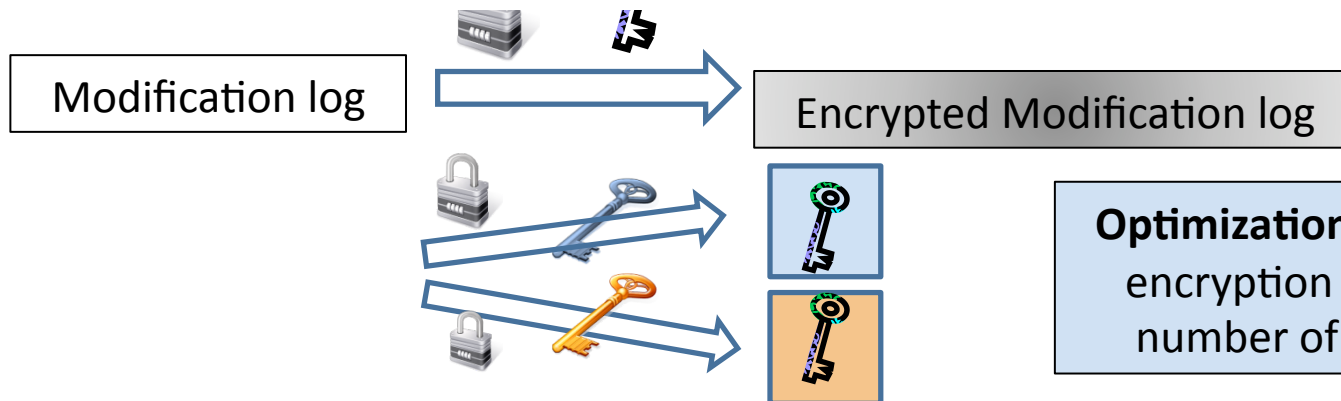
A single auditor



Multi

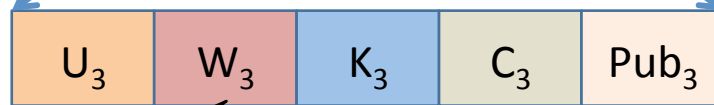
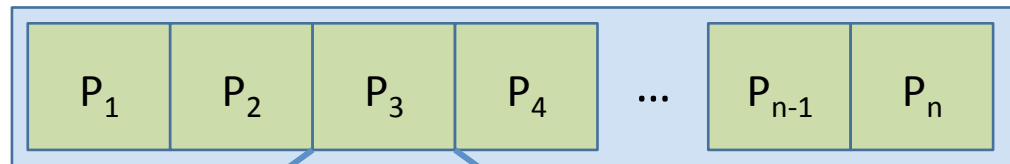
Issues

- Each user trusts a **subset** of the auditors
- Only the auditor(s) **trusted** by the user can see the user's actions on the document



Optimization: Use broadcast encryption tree to reduce number of required keys

Our solution: Confidentiality



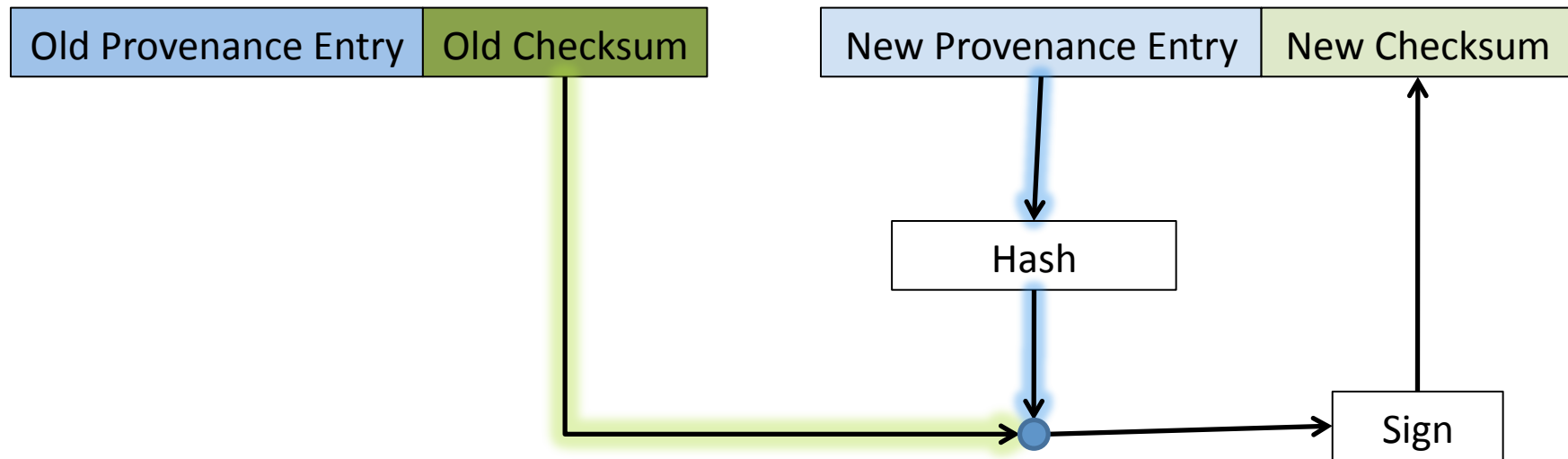
$$\mathbf{W}_i = E_{k_i} (w_i) \parallel \text{hash}(D)$$

$$\mathbf{K}_i = \{E_{k_a} (k_i)\}$$

- k_i is a secret key that authorized auditors can retrieve from the field K_i
- w_i is either the diff or the set of actions taken on the file

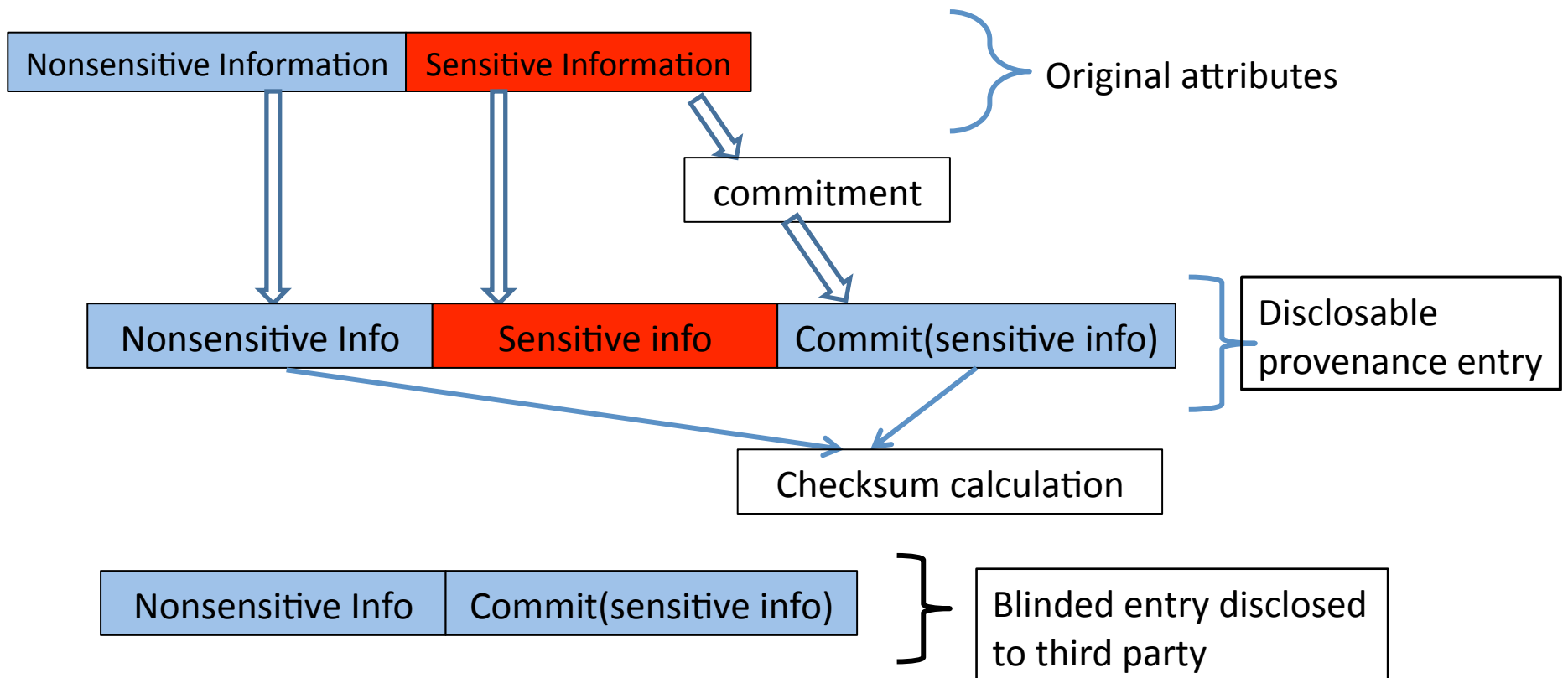
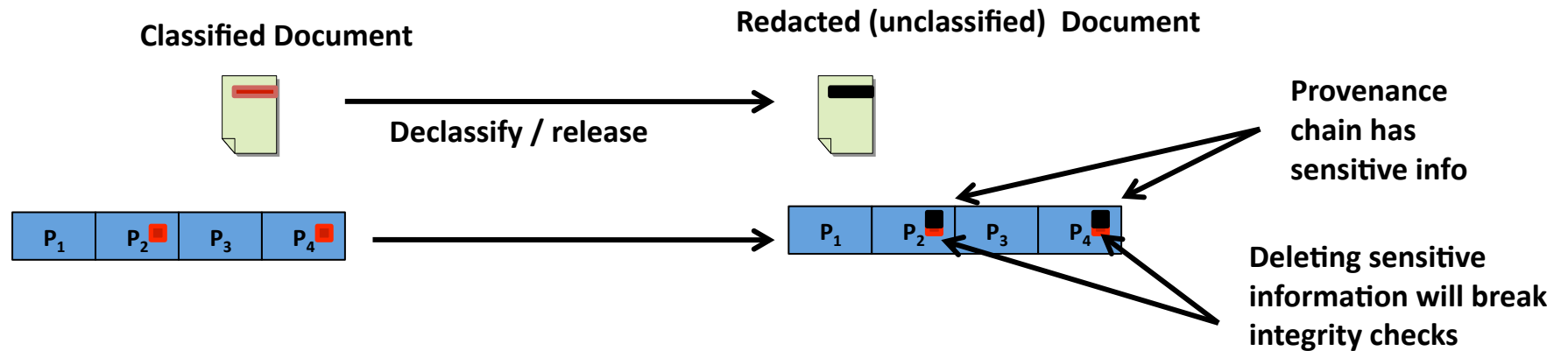
- k_a is the key of a trusted auditor

Our solution: Integrity

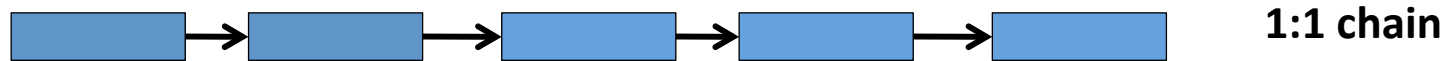


$$C_i = S_{\text{private}_i}(\text{hash}(U_i, W_i, K_i) | C_{i-1})$$

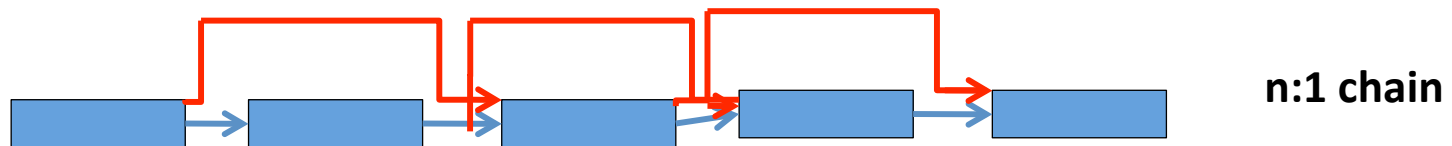
Fine grained control over confidentiality



We can summarize provenance chains to save space, make audits fast



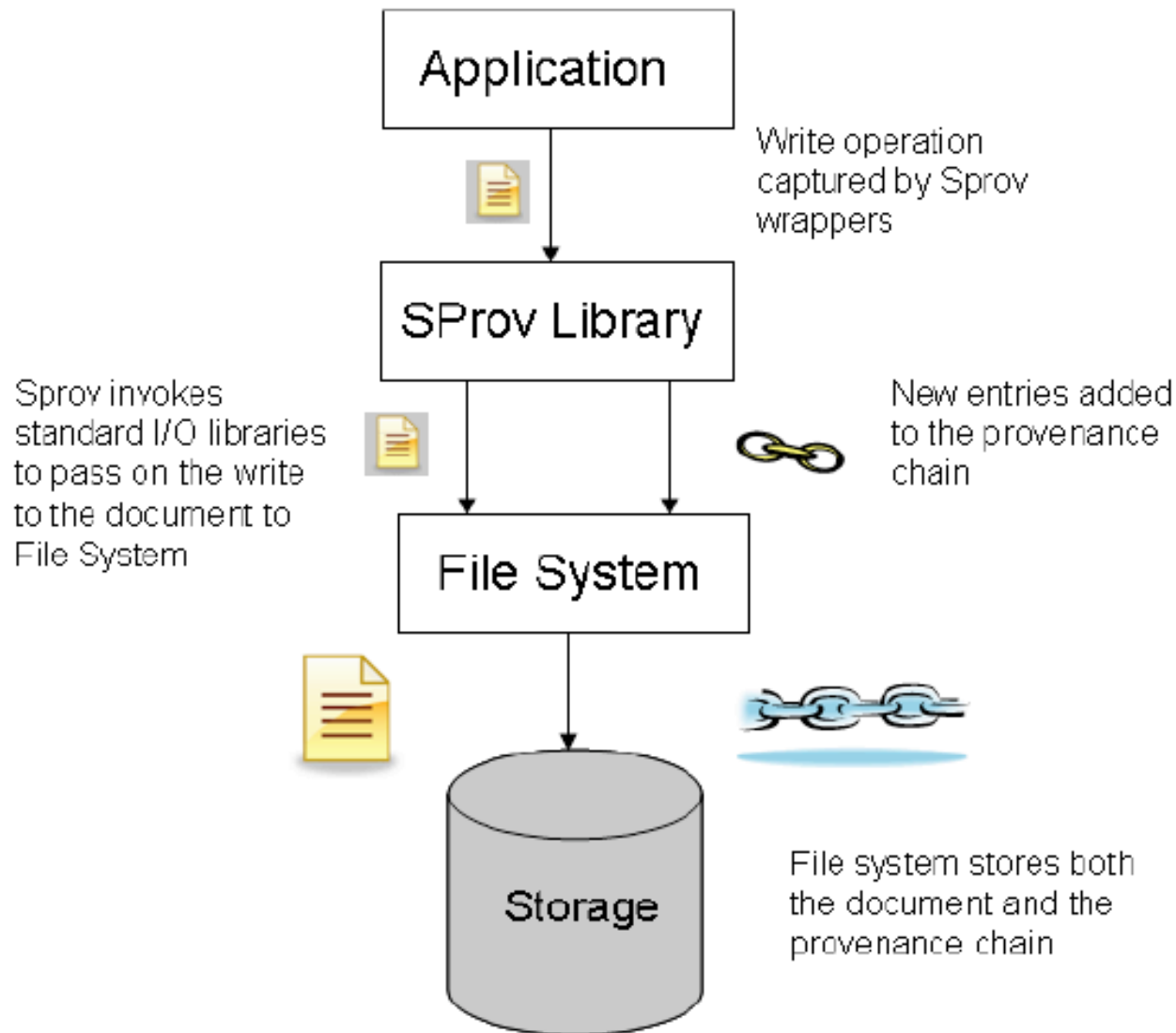
Each entry has **1** checksum, calculated from **1** previous checksum



Each entry has **n** checksums, each of them calculated from **1** previous checksum

We can systematically remove entries from the chain while still being able to prove integrity of chain

Our Sprov application-level library requires almost no application changes



- Sprov provides the file system APIs from `stdio.h`
- To add secure provenance, simply **relink** applications with Sprov library instead of `stdio.h`

Experimental settings

Crypto settings

- 1024 bit DSA signatures
- 128 bit AES encryption
- SHA-1 for hashes

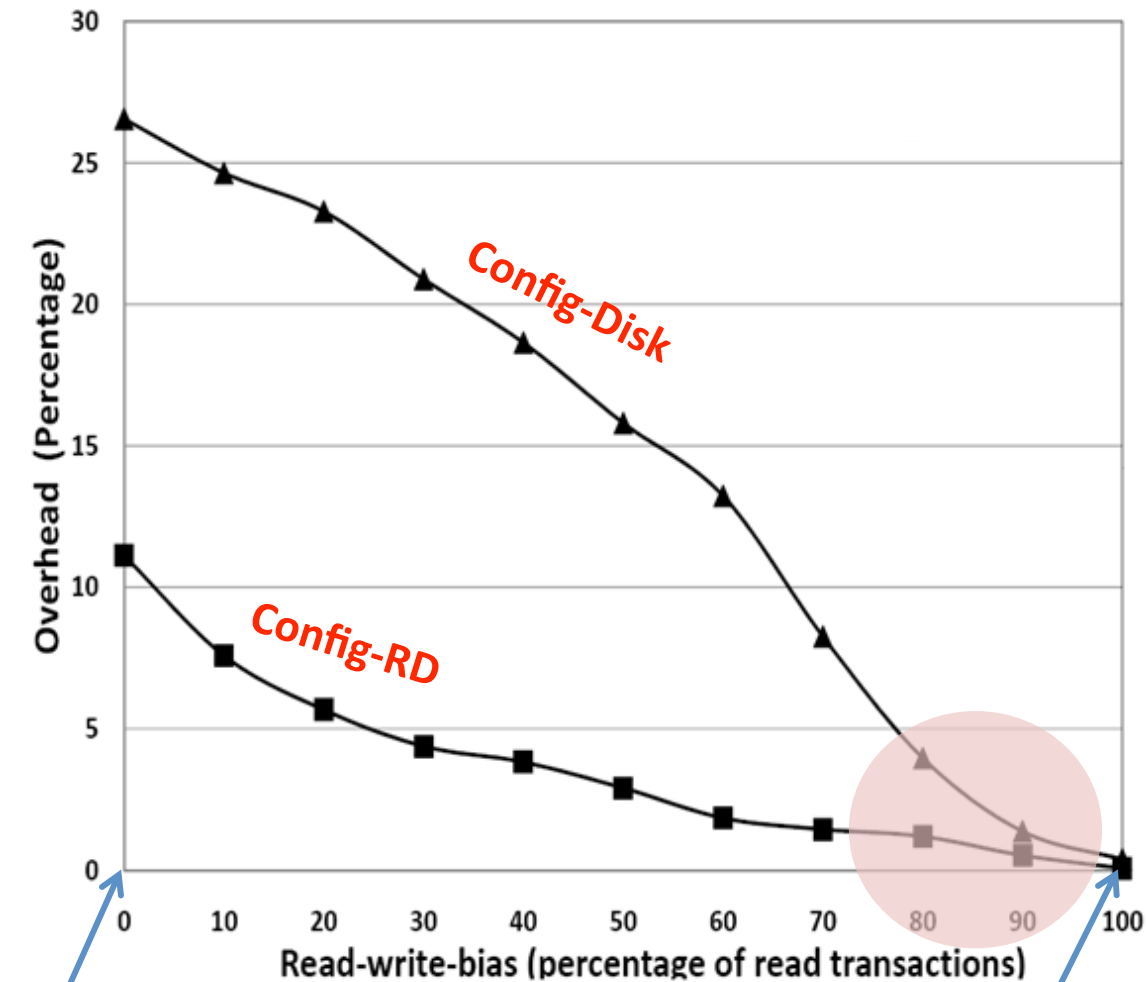
Experiment platform

- Linux 2.6.11 with ext3
- Pentium 3.4 GHz, 2GB RAM,
- Disks: Seagate Barracuda 7200 rpm, WD Caviar SE16 7200 rpm

Modes

- **Config-Disk** : Provenance chains stored on Disk
- **Config-RD**: Provenance chains stored on RAM Disk buffer, and periodically saved to disk

Postmark small file benchmark: Overhead < 5% for realistic workloads



100% writes,
0% reads

0% writes,
100% reads

- **20,000** small files (8KB-64KB) subjected to 100% to 0% write load with the Postmark benchmark
- At 100% write load, execution time overhead of using secure provenance over the no-provenance case is approx. 27% (12% with RD)
- At 50% write load, overheads go down to 16% (3% with RD)
- Overheads are less than **5%** with 20% or less write load

Hybrid workloads: Simulating real file systems

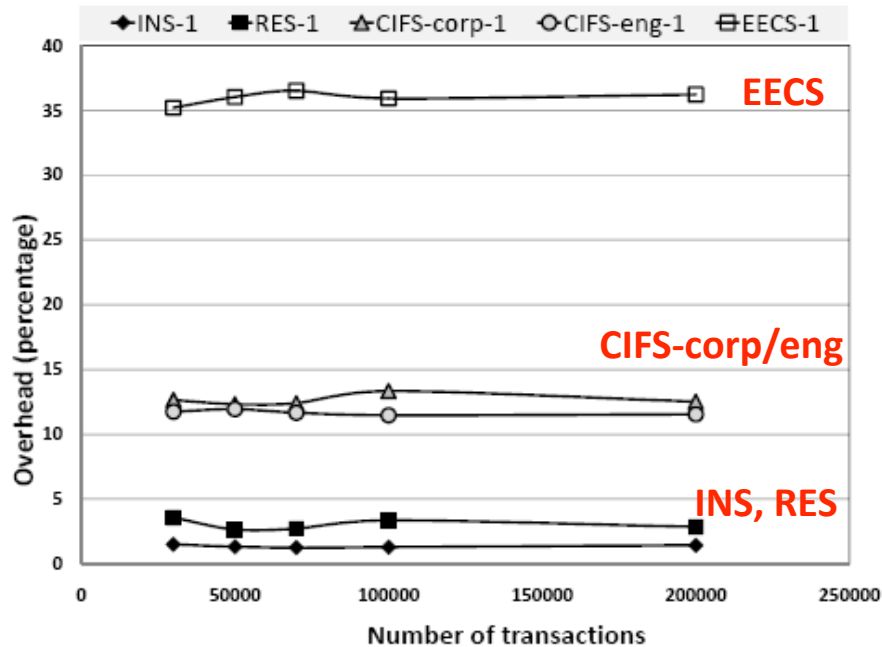
File system distribution:

- File size distribution in real file systems follows the [log normal distribution](#) [Bolosky and Douceur 99]
- Median file size = 4KB , mean file size = 80KB
- We created a file system with 20,000 files, using the lognormal parameters $\mu = 8.46$, $\sigma = 2.4$
- In addition, we included a few large (1GB+) files

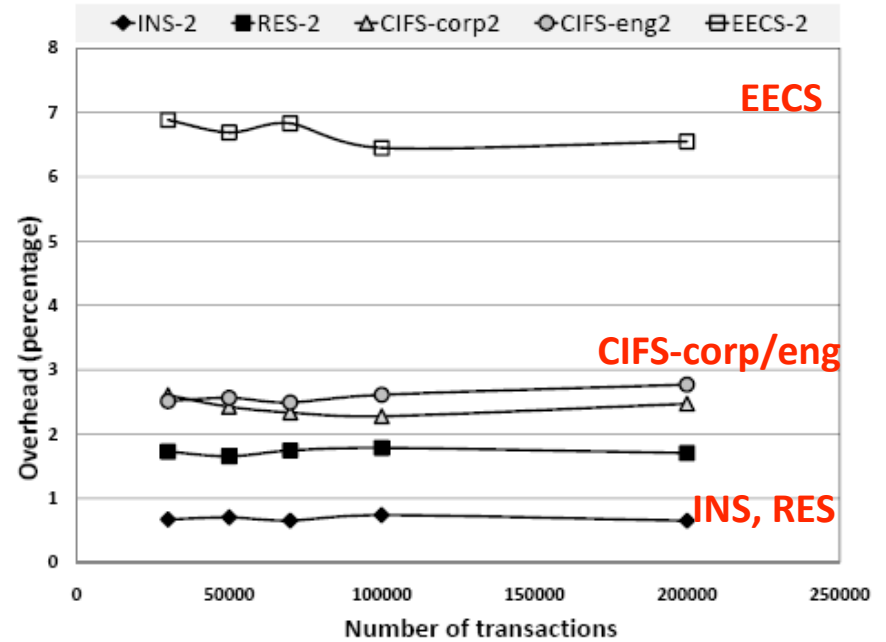
Workload

- **INS**: Instructional lab (**1.1%** writes) [Roselli 00]
- **RES**: A research lab (**2.9%** writes) [Roselli 00]
- **CIFS-Corp**: (**15%** writes) [Leung 08]
- **CIFS-Eng**: (**17%** writes) [Leung 08]
- **EECS**: (**82%** writes) [Ellard 03]

Typical real life workloads: 1 - 13% overhead



Config-Disk



Config-RD

- **INS** and **RES** are read-intensive (80%+ reads), so overheads are very low in both cases.
- **CIFS-corp** and **CIFS-eng** have 2:1 ratio of reads and writes, overheads are still low (range from 12% to 2.5%)
- **EECS** has very high write load (82%+), so the overhead is higher, but still less than 35% for Config-Disk, and less than 7% for Config-RD

Summary: Secure provenance possible at **low cost**

Yes, We CAN achieve secure provenance with integrity and confidentiality assurances with reasonable overheads

- For most real-life workloads, overheads are between 1% and 15% only



More info at <http://tinyurl.com/secprov>