

Cryptographically secure multiparty evaluation of system reliability

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UTOPIÆ

Uncertainty
Treatment and
Optimisation in
Aerospace
Engineering



Durham
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Introduction

General Motivation

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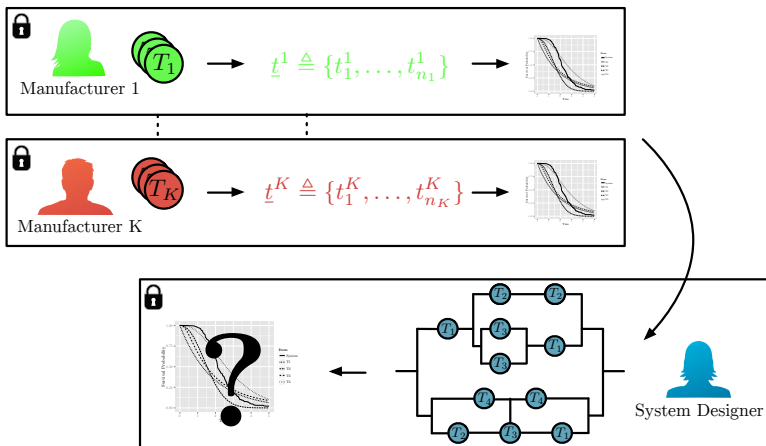
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- donation of sensitive/personal data (e.g. medical/genetic studies);
- complex models on constrained devices (e.g. smart watches)
- running confidential algorithms on confidential data (e.g. engineering reliability — topic of this talk)

Motivation in Reliability Theory

Inference on system/network reliability whilst *maintaining privacy requirements* of all parties.



Homomorphic Encryption

Encryption the solution?

Encryption can provide security guarantees ...

$$\text{Enc}(k_p, m) \rightleftharpoons c$$

Easy

Hard without k_s

$$\text{Dec}(k_s, c) = m$$

... but is typically 'brittle'.

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 \end{array}
 \quad \xrightarrow{\oplus}$$

Limitations of homomorphic encryption

- 1 Message space (what we can encrypt)
 - Commonly only easy to encrypt binary/integers/polynomials
- 2 Cipher text size (the result of encryption)
 - Present schemes all inflate the size of data substantially (e.g. 1MB \rightarrow 16.4GB)
- 3 Computational cost (computing without decrypting)
 - 1000's additions per sec
 - \approx 50 multiplications per sec
- 4 Division and comparison operations (equality/inequality checks)
 - Not possible in current schemes!
- 5 Depth of operations
 - After a certain depth of multiplications, need to 'refresh' cipher text: hugely time consuming, so avoid!

Encrypted Reliability Theory

System lifetimes

Let $C_t^k \in \{0, 1, \dots, M_k\}$ be random variable denoting number of components of type k surviving at time t . Then, survival function of system lifetime T_S is:

$$\begin{aligned}\mathbb{P}(T_S > t) &= \sum_{l_1=0}^{M_1} \cdots \sum_{l_K=0}^{M_K} \Phi(l_1, \dots, l_K) \mathbb{P}\left(\bigcap_{k=1}^K \{C_t^k = l_k\}\right) \\ &= \sum_{l_1=0}^{M_1} \cdots \sum_{l_K=0}^{M_K} \Phi(l_1, \dots, l_K) \prod_{k=1}^K \mathbb{P}(C_t^k = l_k)\end{aligned}$$

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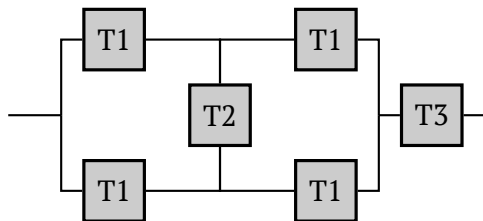
Note: this is a homogeneous polynomial of degree $K + 1$ in the survival signature and component survival probabilities \implies can evaluate encrypted.

Propagating uncertainty as a Bayesian

$$\begin{aligned}
 & P(T_{S^*} > t \mid \underline{y}_1, \dots, \underline{y}_K) \\
 &= \int \dots \int P(T_{S^*} > t \mid p_t^1, \dots, p_t^K) P(dp_t^1 \mid \underline{y}_1) \dots P(dp_t^K \mid \underline{y}_K) \\
 &= \int \dots \int \left[\sum_{l_1=0}^{M_1} \dots \sum_{l_K=0}^{M_K} \Phi(l_1, \dots, l_K) P \left(\bigcap_{k=1}^K \{C_t^k = l_k \mid p_t^k\} \right) \right] \\
 & \qquad \qquad \qquad \times P(dp_t^1 \mid \underline{y}_1) \dots P(dp_t^K \mid \underline{y}_K) \\
 &= \sum_{l_1=0}^{M_1} \dots \sum_{l_K=0}^{M_K} \Phi(l_1, \dots, l_K) \prod_{k=1}^K \int P(C_t^k = l_k \mid p_t^k) P(dp_t^k \mid \underline{y}_k)
 \end{aligned}$$

A homogeneous polynomial of degree $K + 1$ in the survival signature and posterior predictive component survival probabilities at each time point \implies can still evaluate encrypted.

Survival signature toy example

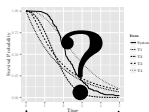
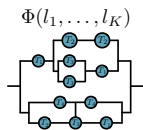


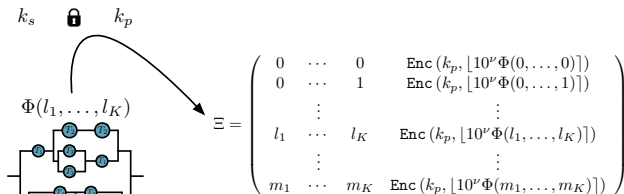
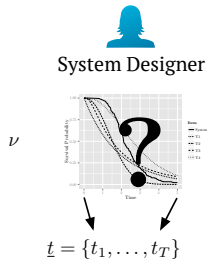
T1	T2	T3	Φ	T1	T2	T3	Φ
0	0	1	0	0	1	1	0
1	0	1	0	1	1	1	0
2	0	1	0.33	2	1	1	0.67
3	0	1	1	3	1	1	1
4	0	1	1	4	1	1	1

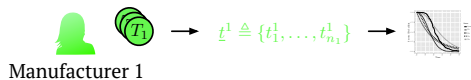
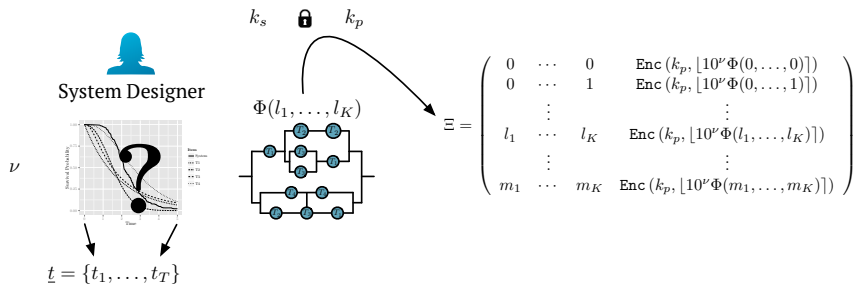
Table 1: Survival signature for a bridge system, omitting all rows with $T3 = 0$, since $\Phi = 0$ for these.

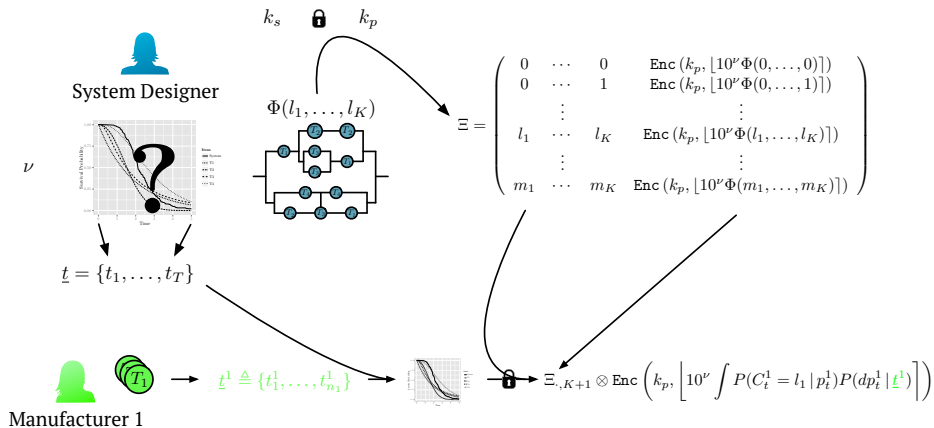


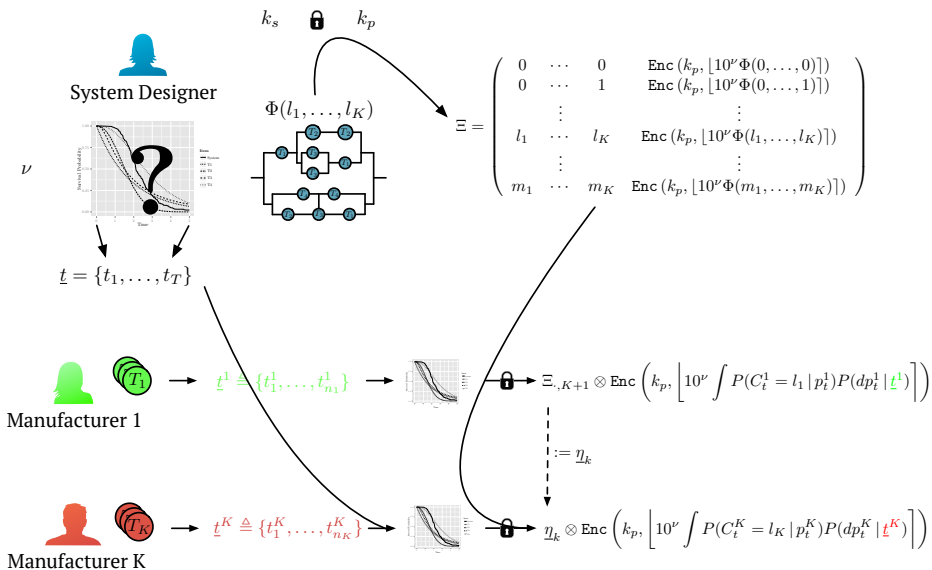
System Designer

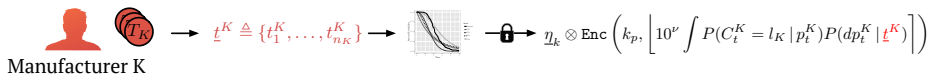
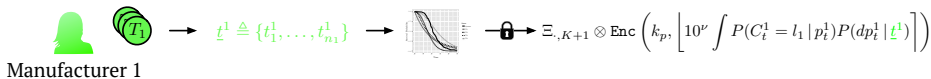
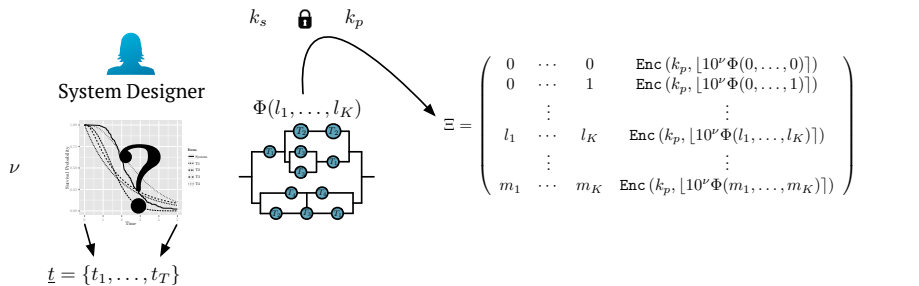
 ν 
 $\underline{t} = \{t_1, \dots, t_T\}$
 k_s  k_p


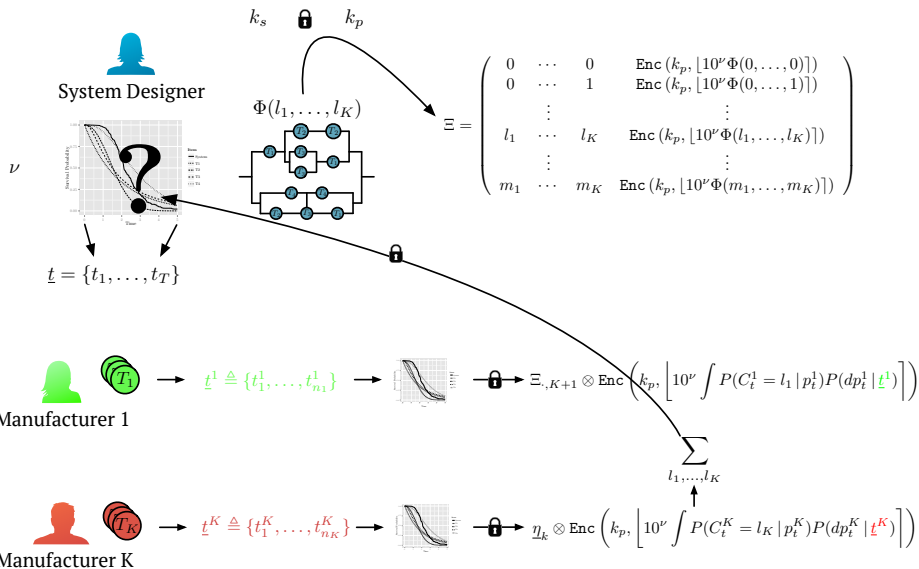












Example

Example system

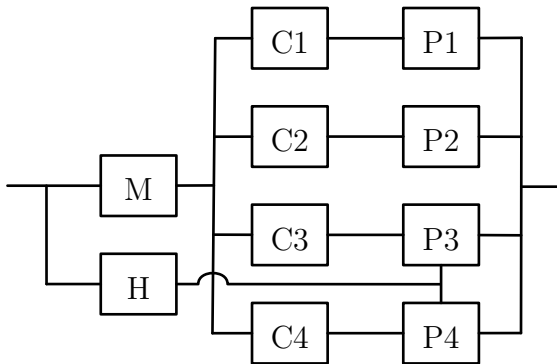


Figure 1: Simple automotive braking system. The master brake cylinder (M) engages all the four wheel brake cylinders (C1 – C4). These in turn each trigger a braking pad assembly (P1 – P4). The hand brake (H) goes directly to the rear brake pad assemblies P3 and P4; the vehicle brakes when at least one of the brake pad assemblies is engaged.

Experimental results

In order to examine the practicality of the problem, perform a full encrypted analysis using Amazon EC2 cloud computing service to mimic a global supply chain.

Role	Physical Server Location	Server Type
System designer	Dublin, Ireland	m4.10xlarge
Manufacturer C	Northern California, USA	m4.10xlarge
Manufacturer H	São Paulo, Brazil	c3.8xlarge
Manufacturer M	Sydney, Australia	r3.4xlarge
Manufacturer P	Tokyo, Japan	i2.8xlarge

Precision was set to $\nu = 5$ and system designer specifies an evenly spaced time grid of 100 points $t \in [0, 5]$.

Computational cost (I)

Role	Action	Timing / Size	
System designer Dublin, Ireland	Generation of (k_p, k_s)		0.3 secs
	Encryption of $\Xi^{(\Phi)}$	1 min	41.1 secs
	Saving $\Xi^{(\Phi)}$ to disk	2 min	41.3 secs
	Compressing $\Xi^{(\Phi)}$ on disk		48.0 secs
	Size of $\Xi^{(\Phi)}$ on disk		5.5GB

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	<i>Transfer $\Xi^{(\Phi)}$ to Manufacturer C</i>	11 min	37.5 secs
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<i>Transfer $\Xi^{(\Phi)}$ to Manufacturer C</i>		11 min	37.5 secs
Manufacturer C Northern California, USA	Decompress & load $\Xi^{(\Phi)}$ from disk	10 min	22.4 secs
	Update $\Xi^{(\Phi)}$	6 min	18.3 secs
	Saving & compressing $\Xi^{(\Phi)}$ to disk	2 min	9.8 secs

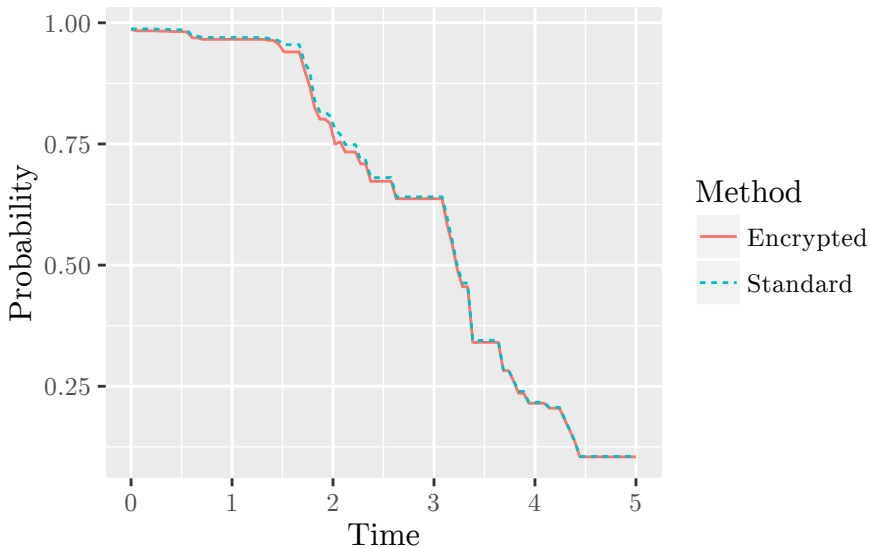
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	Update $\Xi^{(\Phi)}$	6 min	18.3 secs
	Saving & compressing $\Xi^{(\Phi)}$ to disk	2 min	9.8 secs
<i>Transfer $\Xi^{(\Phi)}$ to Manufacturer H</i>		11 min	24.4 secs
Manufacturer H São Paulo, Brazil	Decompress & load $\Xi^{(\Phi)}$ from disk	10 min	13.2 secs
	Update $\Xi^{(\Phi)}$	7 min	23.1 secs
	Saving & compressing $\Xi^{(\Phi)}$ to disk	4 min	45.2 secs
<i>Transfer $\Xi^{(\Phi)}$ to Manufacturer M</i>		20 min	16.5 secs
Manufacturer M Sydney, Australia	Decompress & load $\Xi^{(\Phi)}$ from disk	9 min	41.0 secs
	Update $\Xi^{(\Phi)}$	11 min	28.2 secs
	Saving & compressing $\Xi^{(\Phi)}$ to disk	2 min	54.2 secs

Computational cost (II)

Role	Action	Timing / Size	
	<i>Transfer $\Xi^{(\Phi)}$ to Manufacturer P</i>	6 min	40.7 secs
Manufacturer P Tokyo, Japan	Decompress & load $\Xi^{(\Phi)}$ from disk	9 min	57.1 secs
	Update $\Xi^{(\Phi)}$	7 min	13.5 secs
	Compute ξ		6.1 secs
	Saving & compressing ξ to disk		2.5 secs
	Size of ξ on disk		58.4MB
	<i>Transfer ξ to System Designer</i>		39.5 secs
System designer Dublin, Ireland	Decompress & load ξ from disk		5.9 secs
	Decryption of ξ		8.6 secs
Total:		2 hr	18 min 38.4 secs

Result



R package

HomomorphicEncryption R package (Aslett 2014)

```
library("HomomorphicEncryption")
p <- parsHelp("FandV", lambda=128, L=5)
k <- keygen(p)
c1 <- enc(k$pk, 2); c2 <- enc(k$pk, 3)
cres <- c1 + c2 * c1
dec(k$sk, cres)
```

```
[1] 8
```

```
cmat <- enc(k$pk, matrix(1:9, nrow=3))
cmat2 <- cmat %*% cmat
dec(k$sk, cmat2)
```

```
      [,1] [,2] [,3]
[1,]   30   66  102
[2,]   36   81  126
[3,]   42   96  150
```

References

Aslett, L. J. M. (2014). HomomorphicEncryption: Fully homomorphic encryption. *R package version 0.1*.

Aslett, L. J. M., Coolen, F. P. A., & Wilson, S. P. (2015). Bayesian inference for reliability of systems and networks using the survival signature. *Risk Analysis*, 35/9: 1640–51. DOI: 10.1111/risa.12228

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