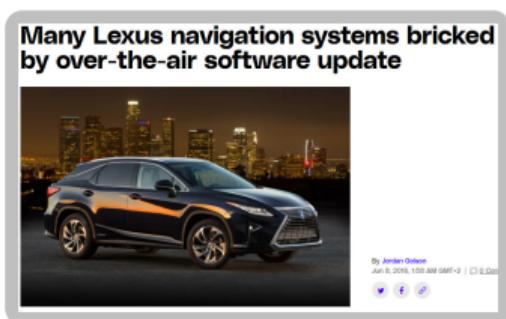
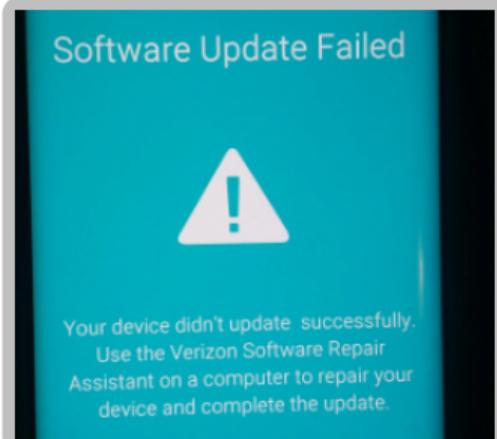




Covering T-Wise Interactions of Deployed Configurations

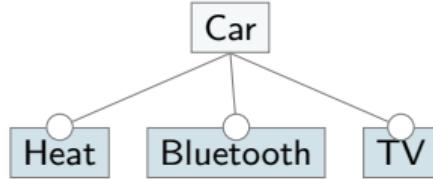
Rahel Sundermann, Sabrina Boehm, Sebastian Krieter, Malte Lochau, Thomas Thüm | VaMoS 2025 | February 05

Motivation Software Updates



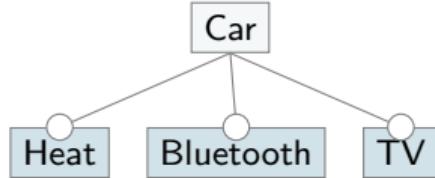
Motivation Testing a Product

Product Line	#Features	#Configurations
Automotive01	2,513	$\approx 5.4 \times 10^{217}$



Motivation Testing a Product

Product Line	#Features	#Configurations	Pair-wise Sample Size
Automotive01	2,513	$\approx 5.4 \times 10^{217}$	1,167

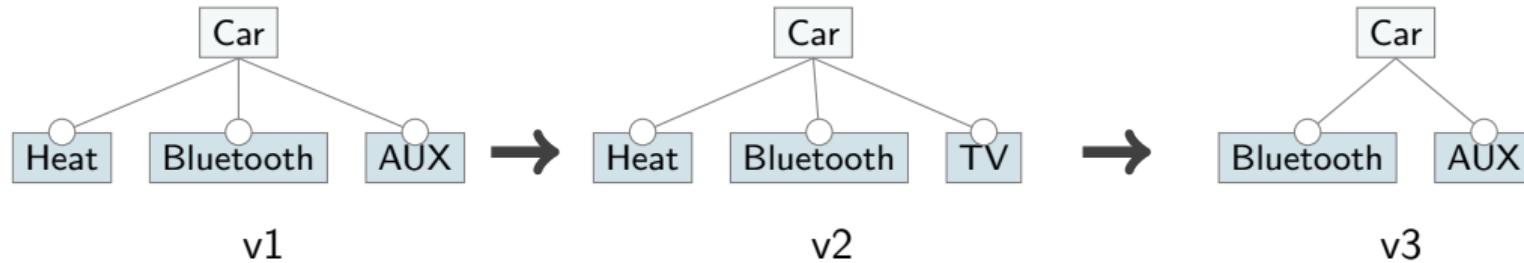


*Bluetooth \wedge Heat
Bluetooth \wedge \neg Heat
 \neg Bluetooth \wedge Heat
 \neg Bluetooth \wedge \neg Heat
Bluetooth \wedge TV*

...

Motivation Testing a Product

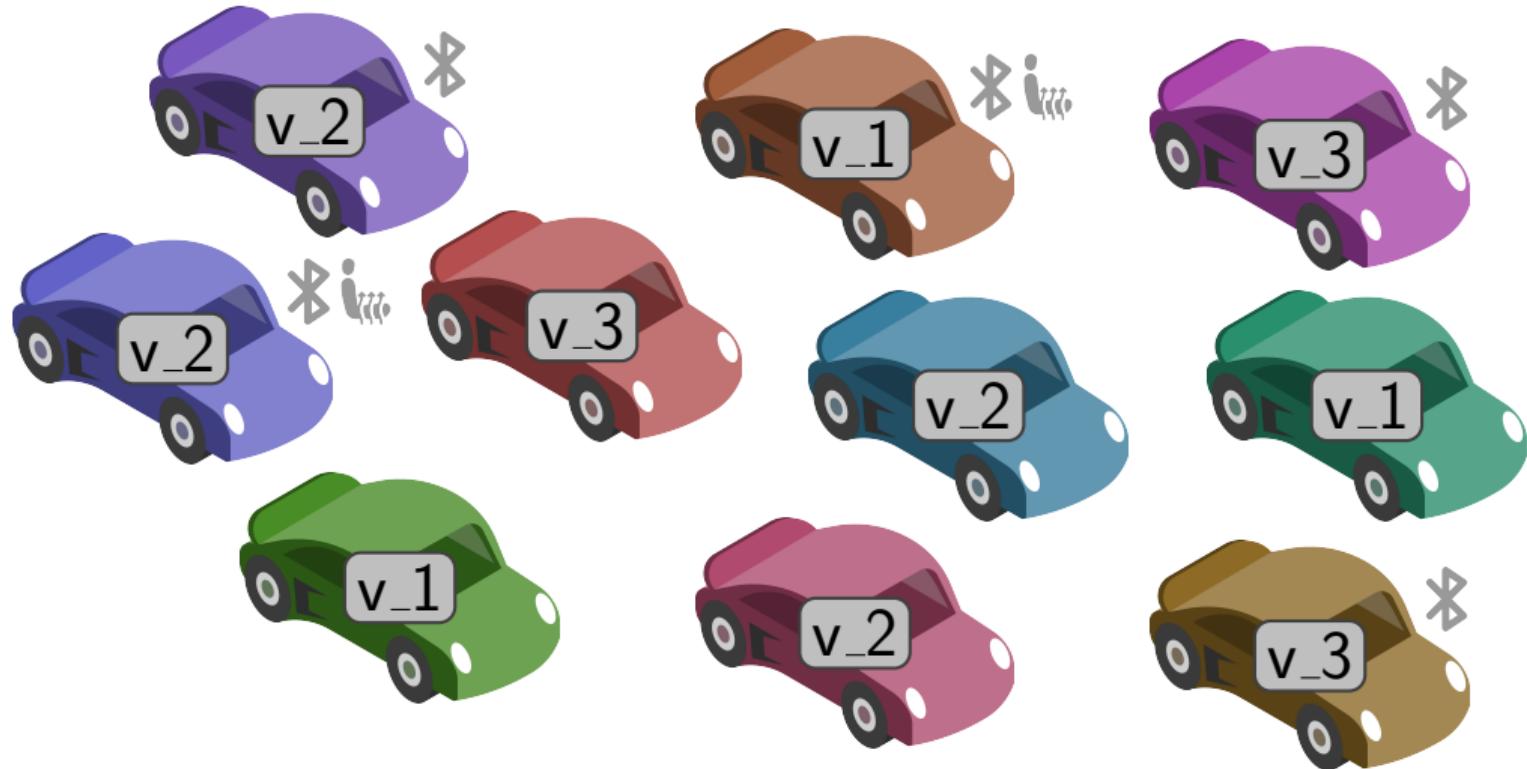
Product Line	#Features	#Configurations	Pair-wise Sample Size
Automotive01_v1	2,513	$\approx 5.4 \times 10^{217}$	1,167
Automotive01_v2	2,520	$\approx 5.4 \times 10^{217}$	1,185
Automotive01_v3	2,533	$\approx 5.4 \times 10^{217}$	1,172



Motivation Testing Multiple Versions



Idea



Idea

$Bluetooth \wedge Heat$
 $Bluetooth \wedge \neg Heat$
 $\neg Bluetooth \wedge Heat$
 $\neg Bluetooth \wedge \neg Heat$



T-Wise Field Coverage

How to measure coverage of **multiple model versions** and only for **deployed configurations**?

T-Wise Field Coverage

How to measure coverage of **multiple model versions** and only for **deployed configurations**?
⇒ Analogous to t -wise interaction sample: ratio of covered interaction to total interactions

T-Wise Field Coverage

How to measure coverage of **multiple model versions** and only for **deployed configurations**?
⇒ Analogous to t -wise interaction sample: ratio of covered interaction to total interactions

Field coverage for any $\mathcal{S} \subset \mathcal{S}_{Field}$

$$\text{coverage}^t(\mathcal{S}, \mathcal{S}_{Field}) = \frac{|\{i \in I^t : \exists c \in \mathcal{S}\}|}{|\{i \in I^t : \exists c \in \mathcal{S}_{Field}\}|}$$

features over all versions: $L = \bigcup_{c \in \mathcal{S}_{Field}} c$

t -wise interactions: $I^t = \{i \subset L : |i| = t\}$

Greedy Algorithm for T-Wise Field Sampling

Input: Set of field configurations $\mathcal{S}_{Field} = \{c_1, \dots, c_n\}$

Output: Subset of \mathcal{S}_{Field} with 100% field coverage

Greedy Algorithm for T-Wise Field Sampling

Input: Set of field configurations $\mathcal{S}_{Field} = \{c_1, \dots, c_n\}$

Output: Subset of \mathcal{S}_{Field} with 100% field coverage

1. $\mathcal{S}' = \emptyset$

1. Start with empty sample \mathcal{S}'

Greedy Algorithm for T-Wise Field Sampling

Input: Set of field configurations $\mathcal{S}_{Field} = \{c_1, \dots, c_n\}$

Output: Subset of \mathcal{S}_{Field} with 100% field coverage

1. Start with empty sample \mathcal{S}'
2. Count occurrences of each t-wise interaction

1. $\mathcal{S}' = \emptyset$
2. $count(i) = |\{c \in \mathcal{S}_{Field} : i \subseteq c\}|$

Greedy Algorithm for T-Wise Field Sampling

Input: Set of field configurations $\mathcal{S}_{Field} = \{c_1, \dots, c_n\}$

Output: Subset of \mathcal{S}_{Field} with 100% field coverage

1. Start with empty sample \mathcal{S}'
2. Count occurrences of each t-wise interaction
3. Add configurations with unique interactions to \mathcal{S}'

1. $\mathcal{S}' = \emptyset$
2. $count(i) = |\{c \in \mathcal{S}_{Field} : i \subseteq c\}|$
3. $\mathcal{U} = \{c \in \mathcal{S}_{Field} : \exists i \subseteq c : count(i) = 1\}$
 $\mathcal{S}' = \mathcal{S}' \cup \mathcal{U}, \mathcal{S}_{Field} = \mathcal{S}_{Field} \setminus \mathcal{U}$

Greedy Algorithm for T-Wise Field Sampling

Input: Set of field configurations $\mathcal{S}_{Field} = \{c_1, \dots, c_n\}$

Output: Subset of \mathcal{S}_{Field} with 100% field coverage

1. Start with empty sample \mathcal{S}'
2. Count occurrences of each t-wise interaction
3. Add configurations with unique interactions to \mathcal{S}'
 1. Update score for each configuration in \mathcal{S}_{Field}

1. $\mathcal{S}' = \emptyset$
2. $count(i) = |\{c \in \mathcal{S}_{Field} : i \subseteq c\}|$
3. $\mathcal{U} = \{c \in \mathcal{S}_{Field} : \exists i \subseteq c : count(i) = 1\}$
 $\mathcal{S}' = \mathcal{S}' \cup \mathcal{U}, \mathcal{S}_{Field} = \mathcal{S}_{Field} \setminus \mathcal{U}$
 1. $score(c) = \sum_{i \in I_t} \frac{1}{count(i)}$

Greedy Algorithm for T-Wise Field Sampling

Input: Set of field configurations $\mathcal{S}_{Field} = \{c_1, \dots, c_n\}$

Output: Subset of \mathcal{S}_{Field} with 100% field coverage

1. Start with empty sample \mathcal{S}'
2. Count occurrences of each t-wise interaction
3. Add configurations with unique interactions to \mathcal{S}'
 1. Update score for each configuration in \mathcal{S}_{Field}
 2. Add configuration with highest score to \mathcal{S}'

1. $\mathcal{S}' = \emptyset$
2. $count(i) = |\{c \in \mathcal{S}_{Field} : i \subseteq c\}|$
3. $\mathcal{U} = \{c \in \mathcal{S}_{Field} : \exists i \subseteq c : count(i) = 1\}$
 $\mathcal{S}' = \mathcal{S}' \cup \mathcal{U}, \mathcal{S}_{Field} = \mathcal{S}_{Field} \setminus \mathcal{U}$
 1. $score(c) = \sum_{i \in I_t} \frac{1}{count(i)}$
 2. $c_{max} = \arg \max_{c \in \mathcal{S}_{Field}} score(c)$
 $\mathcal{S}' = \mathcal{S}' \cup \{c_{max}\},$
 $\mathcal{S}_{Field} = \mathcal{S}_{Field} \setminus \{c_{max}\}$

Greedy Algorithm for T-Wise Field Sampling

Input: Set of field configurations $\mathcal{S}_{Field} = \{c_1, \dots, c_n\}$

Output: Subset of \mathcal{S}_{Field} with 100% field coverage

1. Start with empty sample \mathcal{S}'
2. Count occurrences of each t-wise interaction
3. Add configurations with unique interactions to \mathcal{S}'
 1. Update score for each configuration in \mathcal{S}_{Field}
 2. Add configuration with highest score to \mathcal{S}'
4. Repeat until all interactions are contained in \mathcal{S}'

1. $\mathcal{S}' = \emptyset$
2. $count(i) = |\{c \in \mathcal{S}_{Field} : i \subseteq c\}|$
3. $\mathcal{U} = \{c \in \mathcal{S}_{Field} : \exists i \subseteq c : count(i) = 1\}$
 $\mathcal{S}' = \mathcal{S}' \cup \mathcal{U}, \mathcal{S}_{Field} = \mathcal{S}_{Field} \setminus \mathcal{U}$
 1. $score(c) = \sum_{i \in I_t} \frac{1}{count(i)}$
 2. $c_{max} = \arg \max_{c \in \mathcal{S}_{Field}} score(c)$
 $\mathcal{S}' = \mathcal{S}' \cup \{c_{max}\},$
 $\mathcal{S}_{Field} = \mathcal{S}_{Field} \setminus \{c_{max}\}$

T-Wise Field Sampling: Example



T-Wise Field Sampling: Example



$Bluetooth \wedge \neg Heat : 2$ $\neg Bluetooth \wedge \neg Heat : 5$

$Bluetooth \wedge Heat : 1$ $\neg Bluetooth \wedge Heat : 0$

T-Wise Field Sampling: Example



$Bluetooth \wedge \neg Heat : 2$ $\neg Bluetooth \wedge \neg Heat : 5$

$Bluetooth \wedge Heat : 0$ $\neg Bluetooth \wedge Heat : 0$

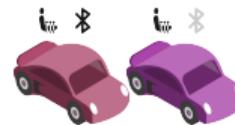
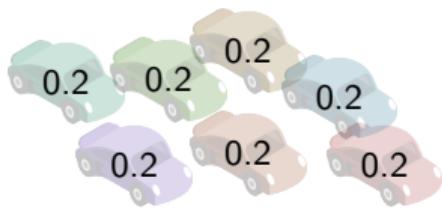
T-Wise Field Sampling: Example



$Bluetooth \wedge \neg Heat : 2$ $\neg Bluetooth \wedge \neg Heat : 5$

$Bluetooth \wedge Heat : 0$ $\neg Bluetooth \wedge Heat : 0$

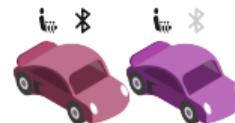
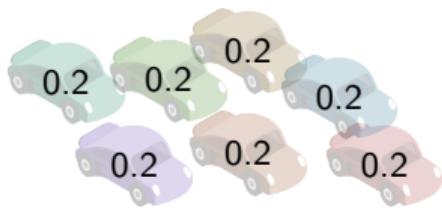
T-Wise Field Sampling: Example



$Bluetooth \wedge \neg Heat : 2$ $\neg Bluetooth \wedge \neg Heat : 5$

$Bluetooth \wedge Heat : 0$ $\neg Bluetooth \wedge Heat : 0$

T-Wise Field Sampling: Example



$Bluetooth \wedge \neg Heat : 0$ $\neg Bluetooth \wedge \neg Heat : 5$

$Bluetooth \wedge Heat : 0$ $\neg Bluetooth \wedge Heat : 0$

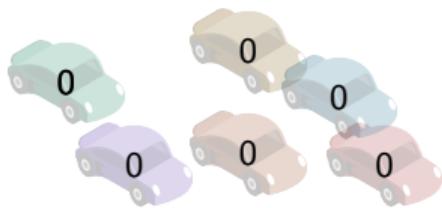
T-Wise Field Sampling: Example



$Bluetooth \wedge \neg Heat : 0$ $\neg Bluetooth \wedge \neg Heat : 5$

$Bluetooth \wedge Heat : 0$ $\neg Bluetooth \wedge Heat : 0$

T-Wise Field Sampling: Example



$Bluetooth \wedge \neg Heat : 0$ $\neg Bluetooth \wedge \neg Heat : 0$

$Bluetooth \wedge Heat : 0$ $\neg Bluetooth \wedge Heat : 0$

Evaluation Setup

Evaluation Setup

System	#Versions	#Features	#Field Config.
Soletta	173	457	11,950
Toybox	62	334	1,039
Busybox	454	1,050	31,279
Fiasco	33	258	10,872
Uclibc	177	272	110,524

Evaluation Setup

System	#Versions	#Features	#Field Config.
Soletta	173	457	11,950
Toybox	62	334	1,039
Busybox	454	1,050	31,279
Fiasco	33	258	10,872
Uclibc	177	272	110,524
ERP	1	1,920	61
Agrib	1	2,238	5,749

Evaluation Setup

System	#Versions	#Features	#Field Config.
Soletta	173	457	11,950
Toybox	62	334	1,039
Busybox	454	1,050	31,279
Fiasco	33	258	10,872
Uclibc	177	272	110,524
ERP	1	1,920	61
Agrib	1	2,238	5,749
Automotive08	unknown	221	233,309
Automotive09	unknown	279	247,780
Automotive10	unknown	279	45,387
Automotive11	unknown	218	27,808
Automotive12	unknown	147	129

Evaluation Setup

System	#Versions	#Features	#Field Config.
Soletta	173	457	11,950
Toybox	62	334	1,039
Busybox	454	1,050	31,279
Fiasco	33	258	10,872
Uclibc	177	272	110,524
ERP	1	1,920	61
Agrib	1	2,238	5,749
Automotive08	unknown	221	233,309
Automotive09	unknown	279	247,780
Automotive10	unknown	279	45,387
Automotive11	unknown	218	27,808
Automotive12	unknown	147	129

Algorithms

- YASA

Evaluation Setup

System	#Versions	#Features	#Field Config.
Soletta	173	457	11,950
Toybox	62	334	1,039
Busybox	454	1,050	31,279
Fiasco	33	258	10,872
Uclibc	177	272	110,524
ERP	1	1,920	61
Agrib	1	2,238	5,749
Automotive08	unknown	221	233,309
Automotive09	unknown	279	247,780
Automotive10	unknown	279	45,387
Automotive11	unknown	218	27,808
Automotive12	unknown	147	129

Algorithms

- YASA
- Field Sampling
 - Random
 - Set-Cover
- Scoring
$$score(i) = count(i)$$
$$score(i) = \frac{1}{count(i)}$$

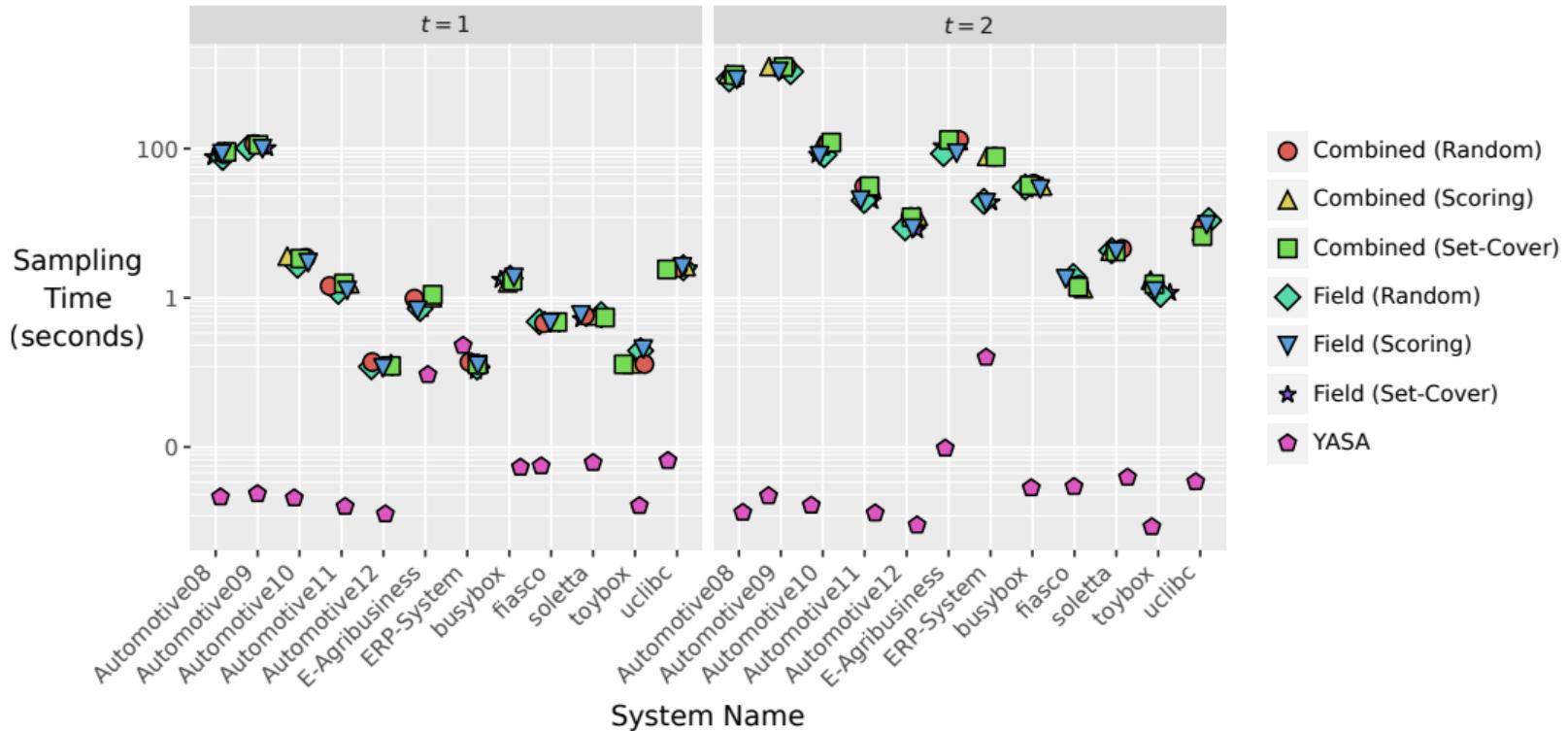
Evaluation Setup

System	#Versions	#Features	#Field Config.
Soletta	173	457	11,950
Toybox	62	334	1,039
Busybox	454	1,050	31,279
Fiasco	33	258	10,872
Uclibc	177	272	110,524
ERP	1	1,920	61
Agrib	1	2,238	5,749
Automotive08	unknown	221	233,309
Automotive09	unknown	279	247,780
Automotive10	unknown	279	45,387
Automotive11	unknown	218	27,808
Automotive12	unknown	147	129

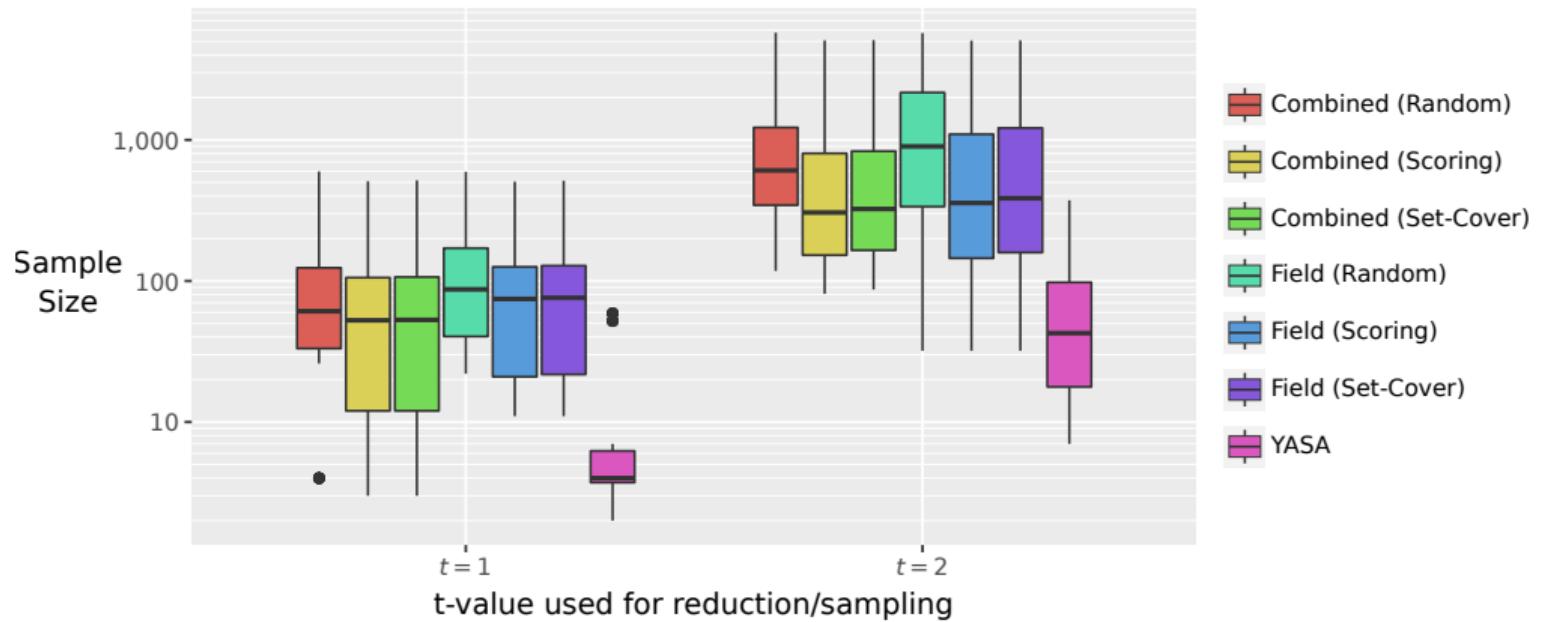
Algorithms

- YASA
- Field Sampling
 - Random
 - Set-Cover
- Scoring
 - $score(i) = count(i)$
 - $score(i) = \frac{1}{count(i)}$
- Combined
 - Random
 - Set-Cover
 - Scoring

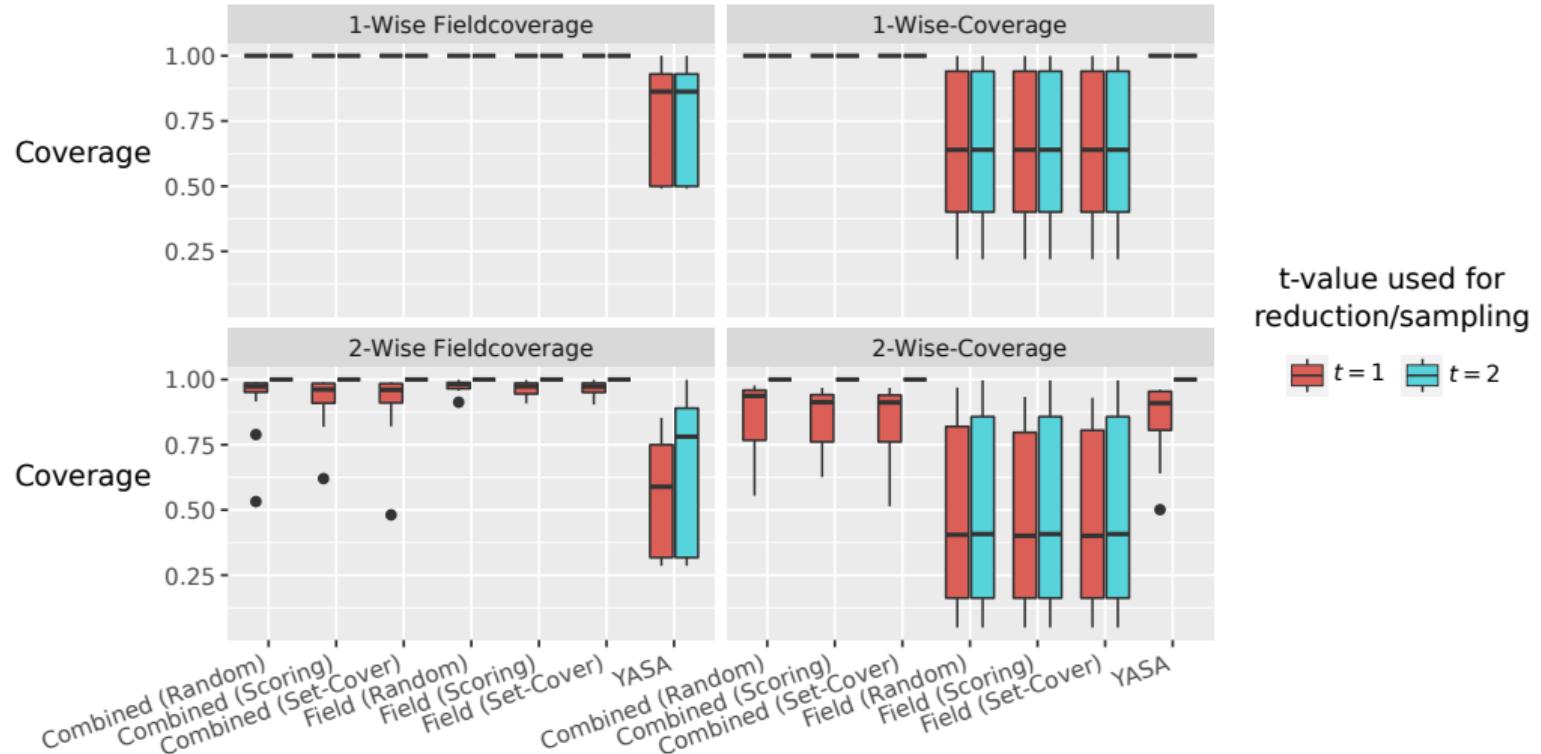
Evaluation Results Sampling Time



Evaluation Results Sample Size



Evaluation Results Coverage



Future Work

- Investigate alternative algorithms and scoring functions
- Take planned configurations into account

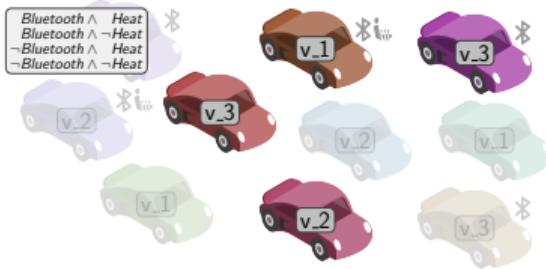
Conclusion

Motivation Testing Multiple Versions



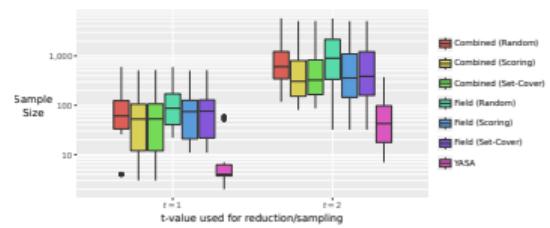
Sundermann et al. – 1. Motivation

Idea



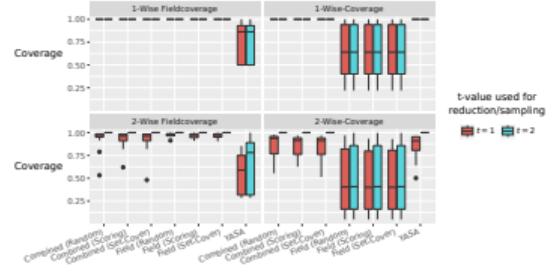
Sundermann et al. – 2. Idea

Evaluation Results Sample Size



Sundermann et al. – 4. Evaluation

Evaluation Results Coverage



Sundermann et al. – 4. Evaluation