Google

Test Selection Safety Evaluation Framework

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Goal: to make a generic framework for evaluating test scheduling algorithms at scale from the historical record.

Project Overview

- Implementation:
 - 1. Determine safety information for historical changelists
 - 2. Evaluate the safety of test selection algorithms
 - 3. Implement optimistic, pessimistic and random test selection algorithms

Project Overview

Used over 2 datasets:

Small Dataset	Large Dataset
2 days of CL data (6-8 Dec 2017)	1 month of CL data (October 2017)
11k changelists	900k changelists
1k total targets	4m total targets
430k times targets were affected	16b times targets were affected

Determining safety

- Safety = would skipping this test target miss a transition?
- Transition = a change in target results, either from failing->passing or passing->failing

Time	
	_
Time -	

Changelist	CL1	CL2
Target Result	Р	Р
Safety	-	Safe
Transition	-	P->P

^{* =} affected

	_
Timo	
111116	

Changelist	CL1	CL2
Target Result	F	F
Safety	-	Safe
Transition	-	F->F

^{* =} affected

Time	\longrightarrow

Changelist	CL1	CL2	CL3
Target Result	Р	*	Р
Safety	-	Safe	Safe
Transition	-	P->P	P->P

^{* =} affected

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Time	
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Changelist	CL1	CL2	CL3
Target Result	F	*	F
Safety	-	Safe	Safe
Transition	-	F->F	F->F

^{* =} affected

Time	
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Changelist	CL1	CL2
Target Result	Р	F
Safety	-	Unsafe
Transition	-	P->F

^{* =} affected

Timo	
1111116	

Changelist	CL1	CL2
Target Result	F	Р
Safety	-	Unsafe
Transition	-	F->P

^{* =} affected

Maybe Unsafe Targets skipping this target might miss a transition

Time	
rime	

Changelist	CL1	CL2	CL3
Target Result	Р	*	F
Safety	-	Maybe unsafe	Maybe unsafe
Transition	-	P->F	P->F

^{* =} affected

Maybe Unsafe Targets skipping this target might miss a transition

Time	
Time	

Changelist	CL1	CL2	CL3
Target Result	F	*	Р
Safety	-	Maybe unsafe	Maybe unsafe
Transition	-	F->P	F->P

^{* =} affected

Can we skip targets safely?

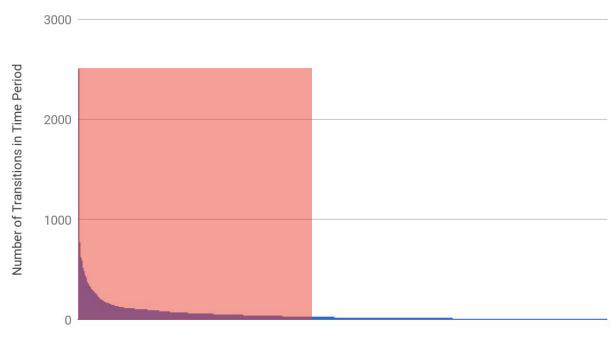
- This information is used to determine whether skipping a target is safe
- All non-definitive pass or fail results treated as affected
- We calculate the safety of skipping tests at rates from 0-100% for an algorithm

Input data

- Input taken from Spanner backup's Result and Affected tables
- Used 3 methods to eliminate flakes from the data
 - Only take pass and fail results
 - Removing target results identified as flaky by Kellogs
 - Removing targets with over X transitions in the time period

Removing high transition count targets

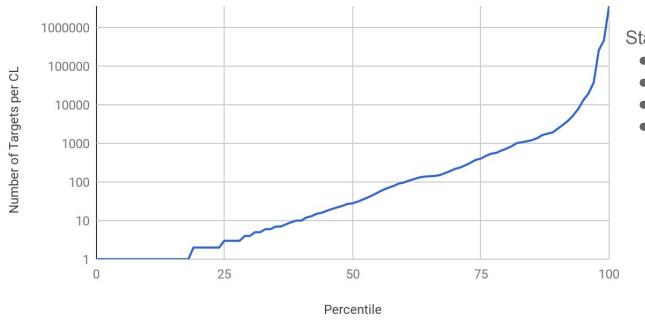
Target Transition Count Histogram (for >10 transition targets)



Remove targets with > 30 transitions (~3k targets)

Targets per CL Distribution

Affected Targets per CL distribution

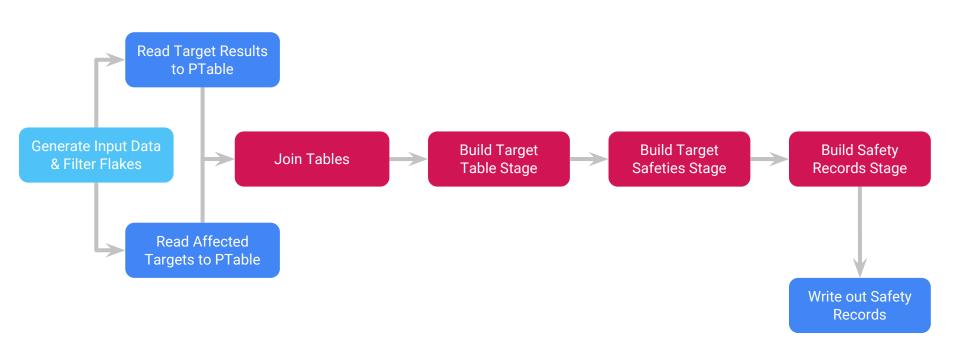


Stats:

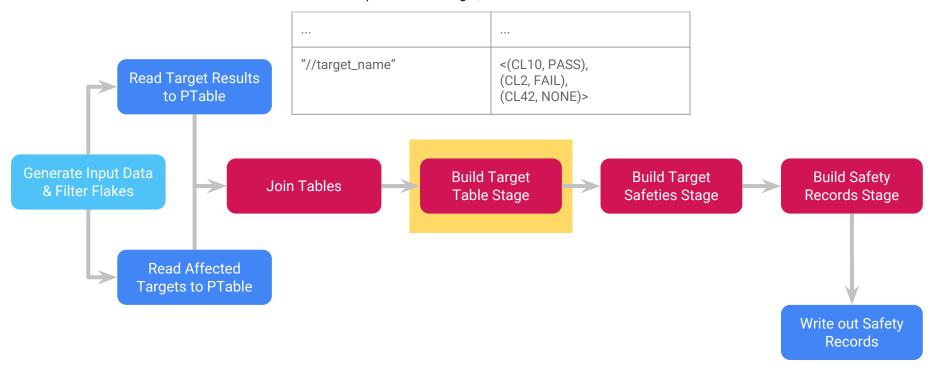
- Median 38 tests!
- 90th percentile 2,604
- 95th percentile 4,702
- 99th percentile 55,730

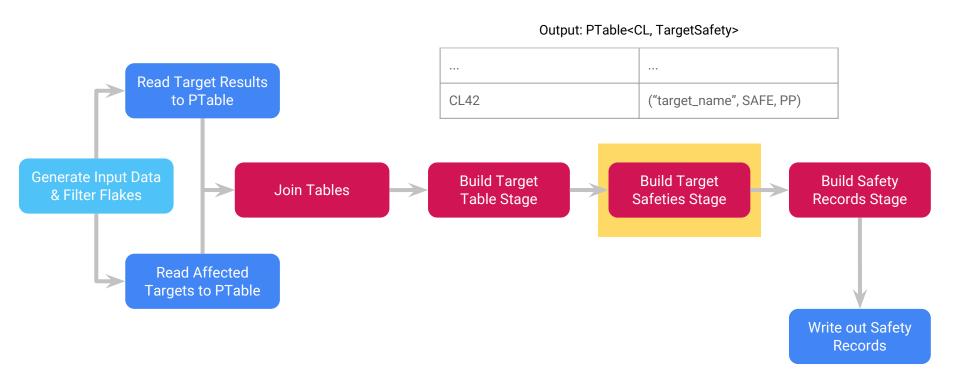
Implementation: Safety Data Builder

This package creates safety data given the historical changelist data as input.

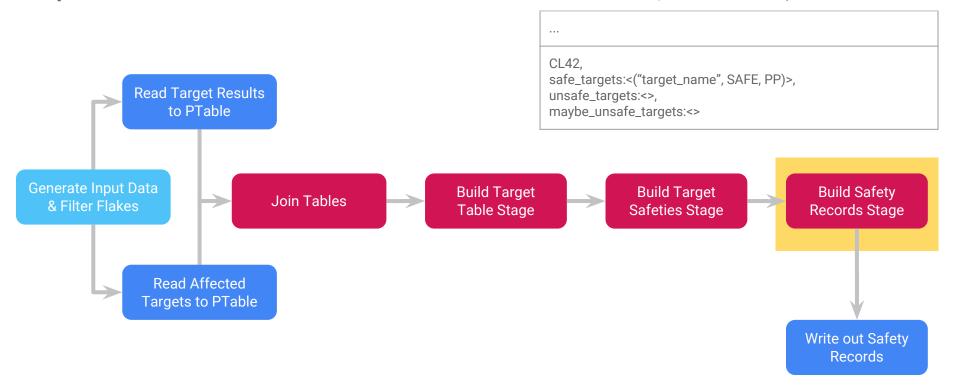


Output: PTable<Target, Collection<ClAndResult>>





Output PCollection<SafetyRecord>



Safety Data Results

	Small Data Set	Large Data Set
Total CLs	10170	891,621
CLs with only safe targets	96.4% (9801)	90.2% (804,160)
CLs with maybe unsafe	3.4% (346)	8.3% (73,897)
CLs with unsafe	0.2% (25)	1.5% (13,564)
Total target affecteds	428,938	15,931,019,923
Safe target affecteds	99.9% (428,547)	99.98% (15,927,853,638)
Maybe unsafe target affecteds	0.09% (365)	0.019% (3,054,667)
Unsafe target affecteds	0.01% (26)	0.0001% (111,618)

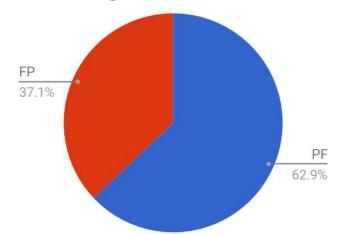


Culprit finding works!

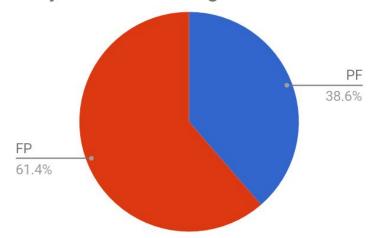


Culprit finding works!

Unsafe Target Transitions



Maybe Unsafe Target Transitions



Implementation: Algorithm Evaluator

This package evaluates the safety of using an algorithm to select tests to skip for a changelist.

Evaluator Implementation

- For every changelist in the safety data, it will call an algorithm with skip rates from 0 to 100%
- Using the targets returned by the algorithm, determines if that selection was safe or not
 - Safe = no unsafe or maybe unsafe tests were skipped
 - Maybe unsafe = maybe unsafe tests were skipped but no unsafe tests
 - Unsafe = unsafe tests were skipped

Algorithms

 Algorithms are implementations of the interface TestSelectionAlgorithm which contains the method

```
ImmutableSet<Target> skipTargets(long cl, Iterable<Target> targets, int
numToSkip)
```

Changelist's Affected Targets

1 Safe target

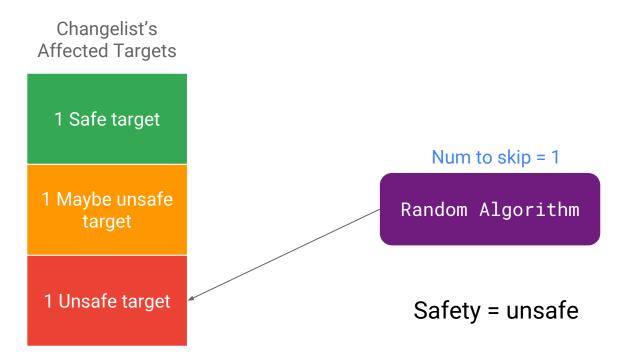
1 Maybe unsafe target

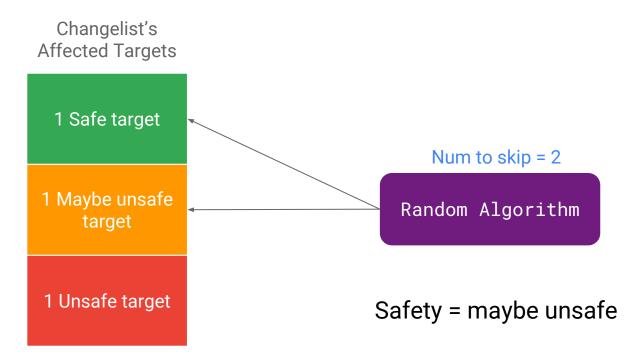
1 Unsafe target

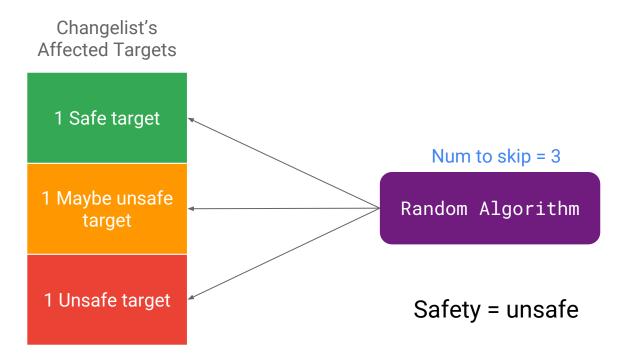
Num to skip = 0

Random Algorithm

Safety = safe







Changelist's Affected Targets

1 Safe target

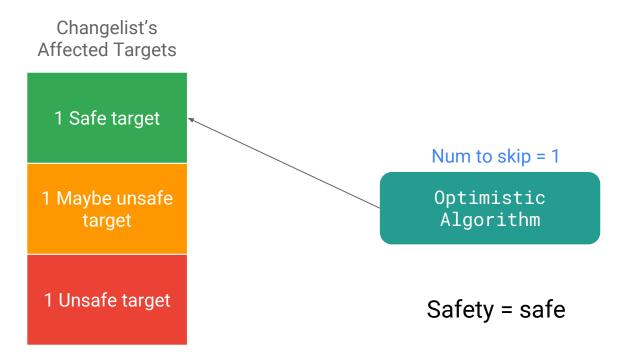
1 Maybe unsafe target

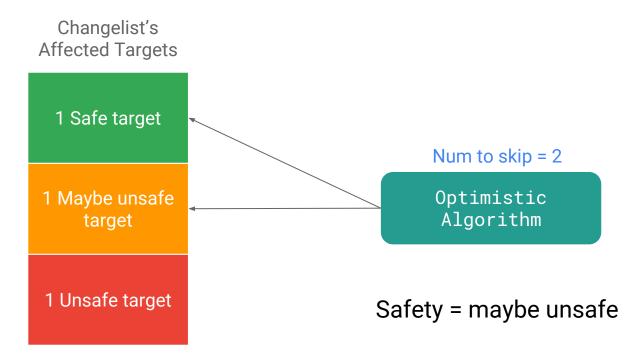
1 Unsafe target

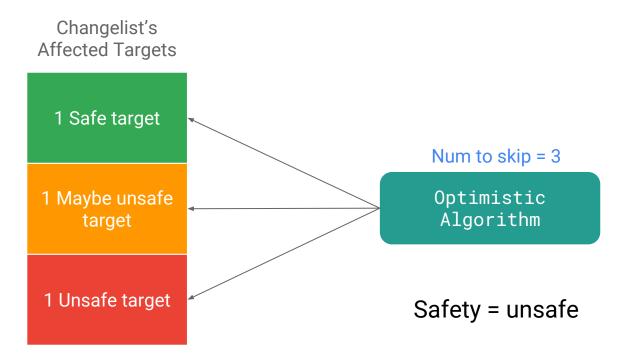
Num to skip = 0

Optimistic Algorithm

Safety = safe







Changelist's Affected Targets

1 Safe target

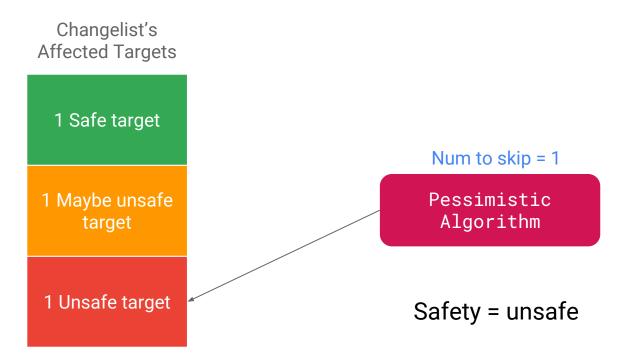
1 Maybe unsafe target

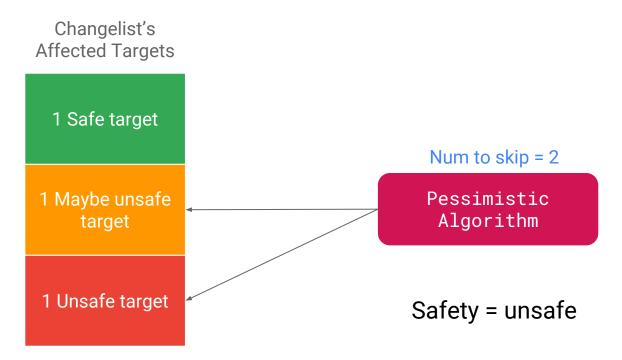
1 Unsafe target

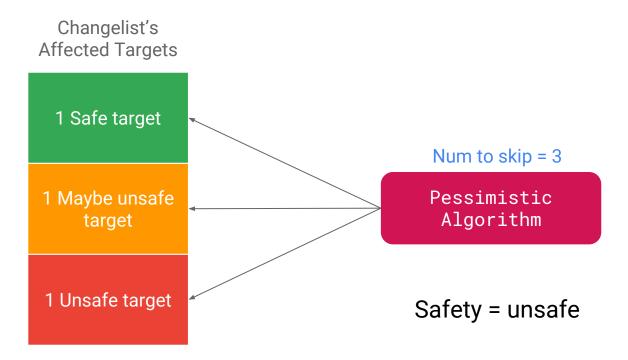
Num to skip = 0

Pessimistic Algorithm

Safety = safe





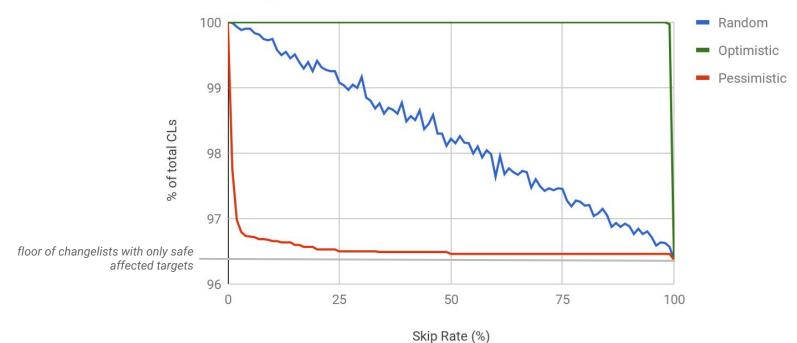


Pipeline performance

- Safety data builder ran in 35 mins
- Algorithm evaluator
 - o Optimistic ran in 2h 40m
 - o Pessimistic ran in 3h 5m
 - o Random ran in 4h 40m

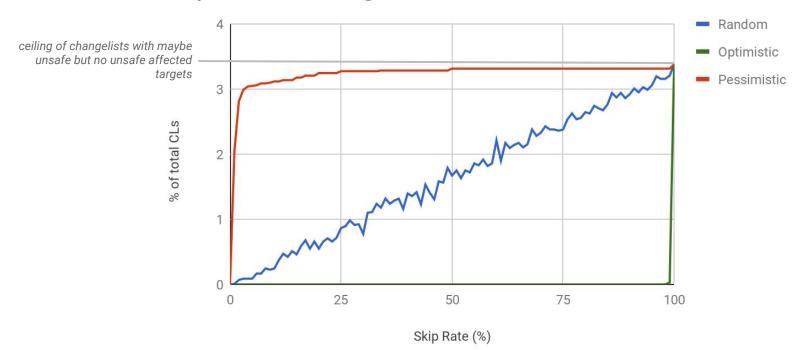
Small dataset results

Safe Changelists



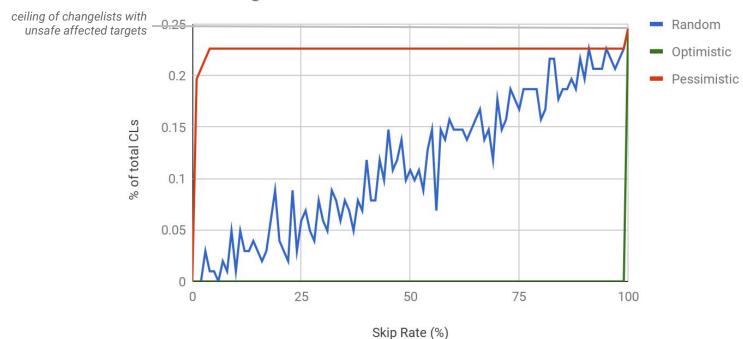
Small dataset results

Maybe Unsafe Changelists



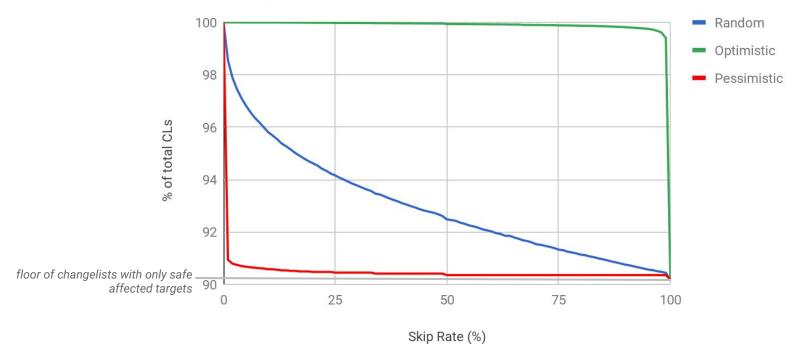
Small dataset results

Unsafe Changelists



Large dataset results

Safe Changelists



Why is random a curve?

- Previously we had predicted a straight line for random
- Small data set has a straight line

Probability Distribution

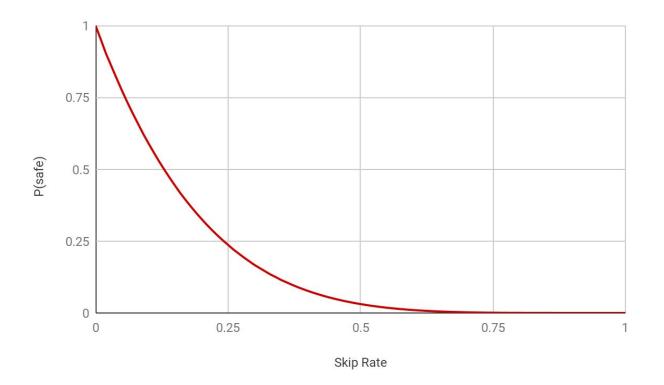
$$\binom{n}{k}$$
 = number of ways to select k items from n total items

$$P(\text{select k only safe targets}) = \frac{\text{number of ways to select k safe targets}}{\text{number of ways to select any k targets}}$$

$$=rac{inom{n}{Np}}{inom{N}{Np}}$$

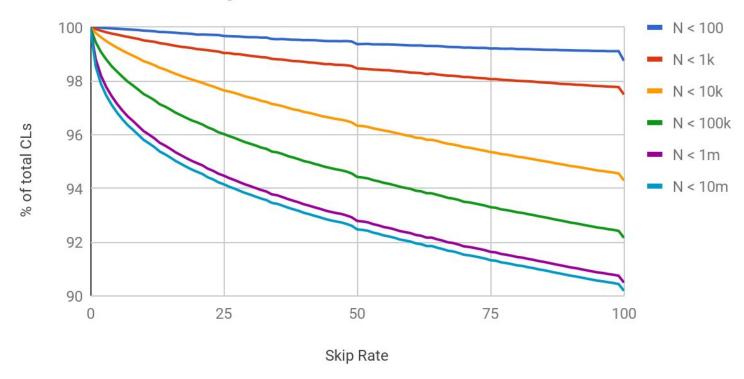
 $= \frac{\binom{n}{Np}}{\binom{N}{Nm}}$ Where n = number of safe affected targets N = total number of affected targets p = % of targets being selected

Probability Distribution where N = 1000, n = 995



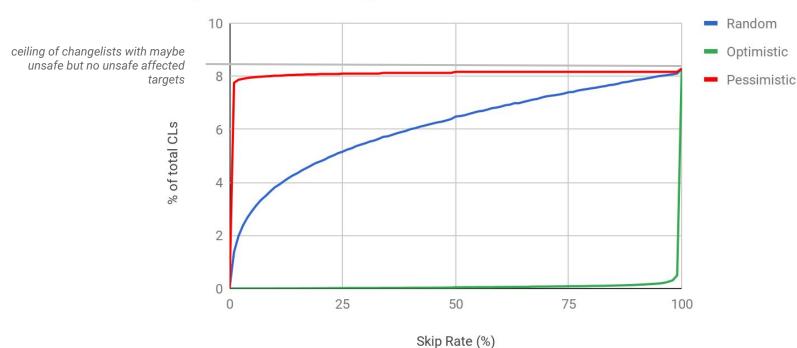
Random Algorithm Safe Changelists

N = number of affected targets for a CL



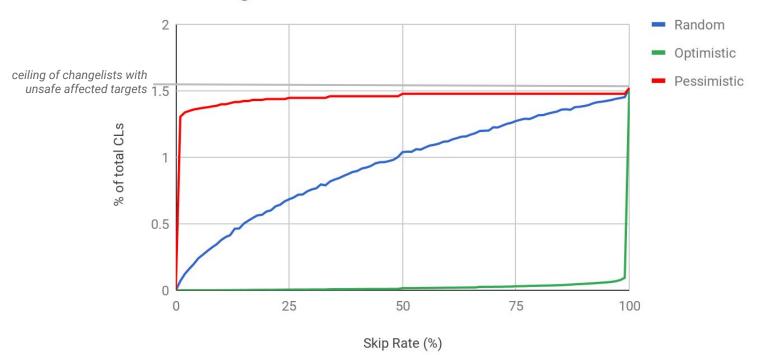
Large dataset results

Maybe Unsafe Changelists



Large dataset results

Unsafe Changelists



Conclusions

- The project was completed!
- We now have an offline method to evaluate test scheduling algorithms and a baseline for future comparison

Continuing the project

- Better flake exclusion
 - Filter using ratio transitions:results
 - Find the point where Kellogs doesn't identify the target as flaky
- Rerunning Elbaum experiments
 - An algorithm which prioritizes targets based on the number of transitions in some previous window of time
- Evaluating Efficacy machine learning model

Questions?

Creating safeties

```
for all targets {
     for (result in sorted target results) {
           if (result = affected) {
                add result to pending results
                continue;
           if (previous result = result) {
                mark this result and all pending results as safe
           } else (if no pending results) {
                mark this result as unsafe
           } else {
                mark this result and all pending results as maybe safe
           previous result = result;
```

Evaluating algorithms

```
for (changelist) {
     retrieve affected targets for changelist
     for (skip rate = 0..100%) {
           skipped targets = algorithm.skipTargets(affected targets, skip rate * num affected
           targets);
           if (skipped targets contain sunsafe targets) {
                mark this test selection as unsafe
           } else if (skipped targets contains maybe unsafe targets) {
                mark this test selection as maybe unsafe
           } else {
                mark this test selection as safe
```

safety_record.proto

```
// Represents the safety information for all affected targets at a CL.
message SafetyRecord {
  optional int64 changelist = 1;
  repeated TargetSafety safe_targets = 2;
  repeated TargetSafety unsafe_targets = 3;
  repeated TargetSafety maybe_unsafe_targets = 4;
// Represents the safety of skipping a test target.
message TargetSafety {
  optional string target_name = 1;
  optional Safety safety = 2;
  optional Transition transition = 3;
```

safety_result.proto

```
// Represents the safety information for using a test selection algorithm on a
// CL's test targets with a given target skip rate.
message SafetyResult {
  optional int64 changelist = 1;
  optional string algorithm_name = 2;
  optional int32 skip_rate = 3;

  optional tko.testselectionevaluation.TargetSafety.Safety safety = 4;

  repeated string unsafe_skipped = 5;
  repeated string maybe_unsafe_skipped = 6;

  optional int32 total_skipped = 7;
}
```

Optimistic Algorithm Implementation

```
// Constructor takes a SafetyRecord as input
private final SafetyRecord record;
skipTargets(changelist, targetList, numToSkip) {
     remainingToSkip = numToSkip;
     skip targets from record.safe_targets with limit remainingToSkip
     remainingToSkip -= safe skipped targets
     skip targets from record.maybe_safe_targets with limit remainingToSkip
     remainingToSkip -= maybe unsafe skipped targets
     skip targets from record.unsafe_targets with limit remainingToSkip
     remainingToSkip -= unsafe skipped targets
     return skipped targets
```