# Absolute to Relative, & Back Again Baseline Safety Assessment Strategic Objective: Continual Improvement

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California Dig Safe Board

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## Contents

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- 1. Comparing States, or
  - Toward a replacement for "damages per thousand tickets," or
  - Economic analysis of damages, or
  - Why are we doing this?
- 2. Risk analysis in damage prevention (or was that safety?)
- 3. Model to predict success (& failure)



1. Comparing States

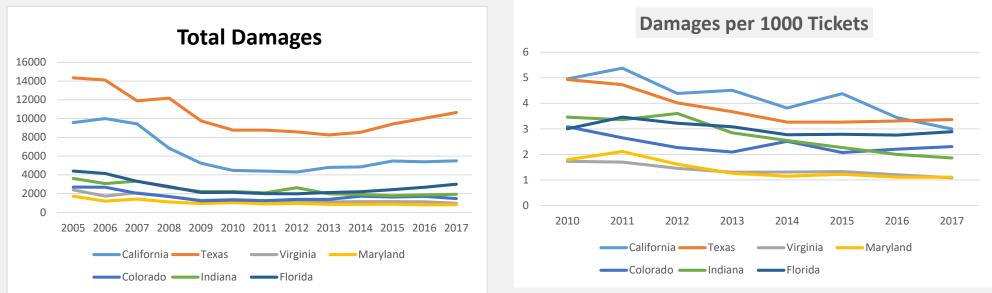
## "There are three kinds of lies: lies, damned lies, and statistics." -- Mark Twain (maybe)

# "Damages per 1000"

• The Good

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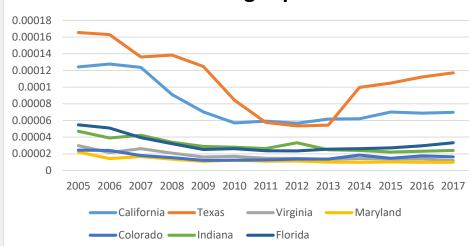
- Controls for changes in the economy (mostly)
- Allows comparison between states (kind of)
- The Bad
  - Measures two separate things: damages and notifications
    - If one-call center practices change, so does denominator



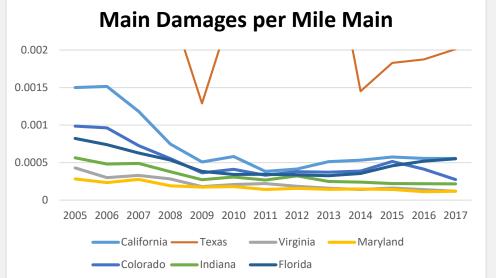
All data from Pipeline and Hazardous Materials Safety Administration (PHMSA) Pipeline Data Mart

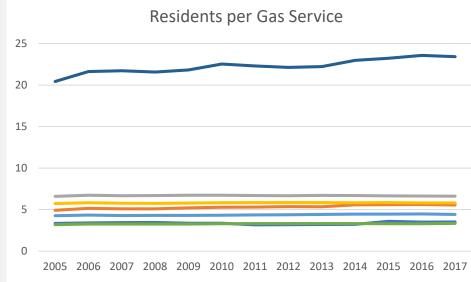


Damages per 10,000 Residents



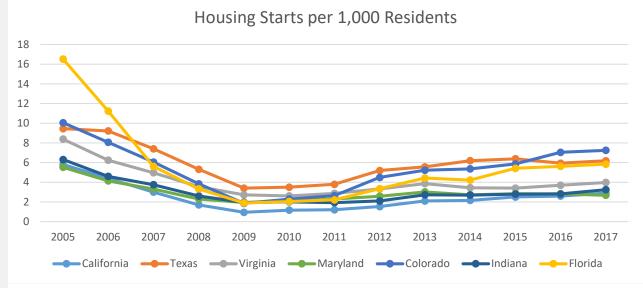
**Gas Service Damages per # Services** 



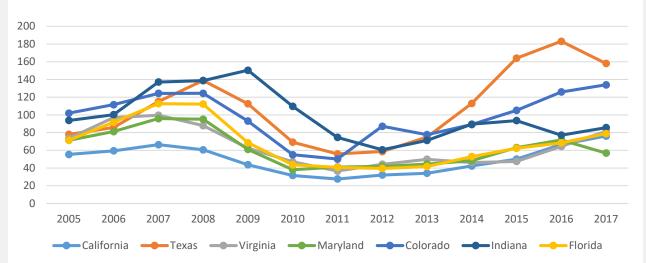


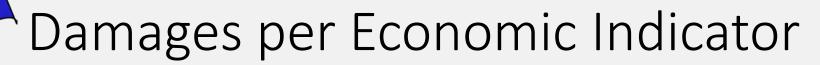
## Baseline Economic Data

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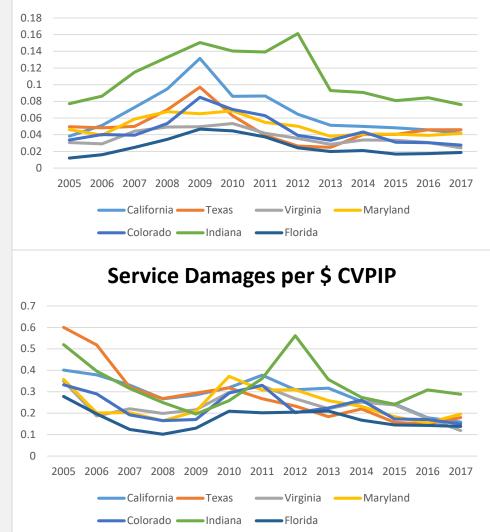
Construction Value put in Place per 100K Residents



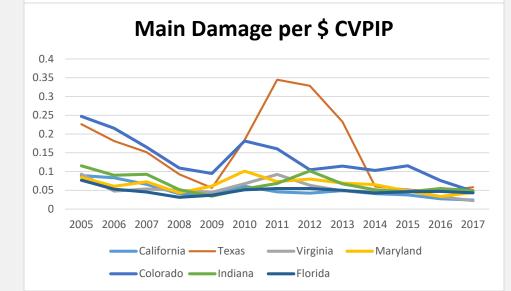


#### Service Damage per Housing Start

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Main Damages per Housing Start



## 2. Safety Risk Assessment

"All models are wrong, but some are useful." – George Box

- A decision-making process for comparing unlike negative outcomes
- May or may not be mathematical
- Depends on a calculated quantity, "risk", which is a construct and has no independent existence
- A tool

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# stakeholders  $\mathbf{R} = \sum \mathbf{F}_{requency} \mathbf{E}_{roosure} \mathbf{C}_{onsequence}$ 



# $\mathbf{R} = \mathbf{F} \mathbf{x} \mathbf{E} \mathbf{x} \mathbf{C}$ $\mathbf{P}_{robability}$

CA Damage Data (2005-2017)

## Probability

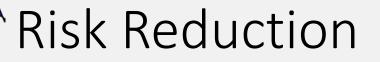
	Transmission	Main	Service	
Fatal	2	0	0	
Injury	0	0	1	
Reportable	44	44	9	
Non-Reportab	le 0	11643	68608	

## Consequence

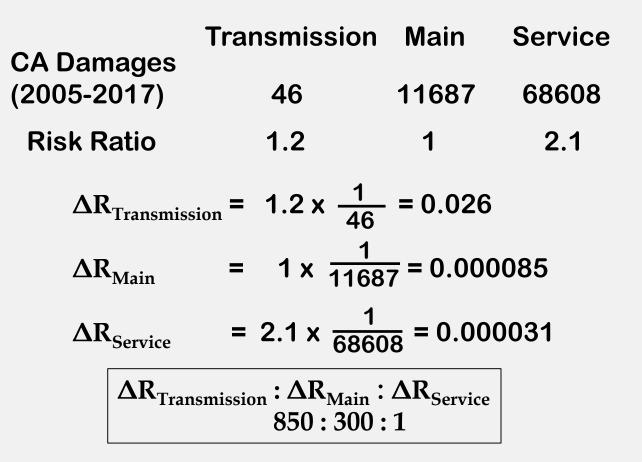
Fatal	<sub>10^</sub> 4	= 10,000
Injury	<sub>10^</sub> 3	= 1,000
Reportable	<sub>10^</sub> 2	= 100
Non-Reportable	<sub>10^</sub> 1	= 10

What did we do wrong?

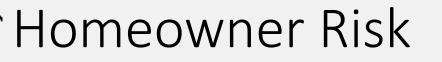
CALIFORNIA		CALIF	ORNIA D	ig Safe Boai	RD	
DIG SAFE	]	<b>R</b> =	F x	ExC	Saj	fety-only!
US Damage Data (	2005-2017)		P	robability		"Only!
	Prob	abilit	У		Consequ	Jence
٦	<b>Fransmission</b>	Main	Service			
Fatal	11	20	8		Fatal	10^4 = 10,000
Injury	29	36	42		Injury	$10^{3} = 1,000$
Total Risk (P x ( Risk Ratio	<u>C) 139000</u> 1.2	<u>116000</u> 1	<u>242000</u> 2.1	No Tic		Risk = 27%
					(et Safot	
Fatal (no ticket)		20	8		arety	Rici
Injury (no ticke	t) 29	36	42			"SK = 270/
Tot. No Tik Risk	× 77000	14000	43000			



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Reducing service damages by ~ 850 is risk-equivalent to reducing transmission damages by 1



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## Probability

Ti	ransmission	Main	Service	Total
All stakeholders All stakeholder (ratio)	46 ) 1	11687 250	68608 1500	1751
Homeowner (ratio)	1	250	5000	5251

 $\Delta R_{\text{Transmission}} : \Delta R_{\text{Main}} : \Delta R_{\text{Service}} = 850 : 300 : 1$ 

Assume homeowners contribute 25% of all damages

(math)

Homeowners contribute to 18% of the total risk!

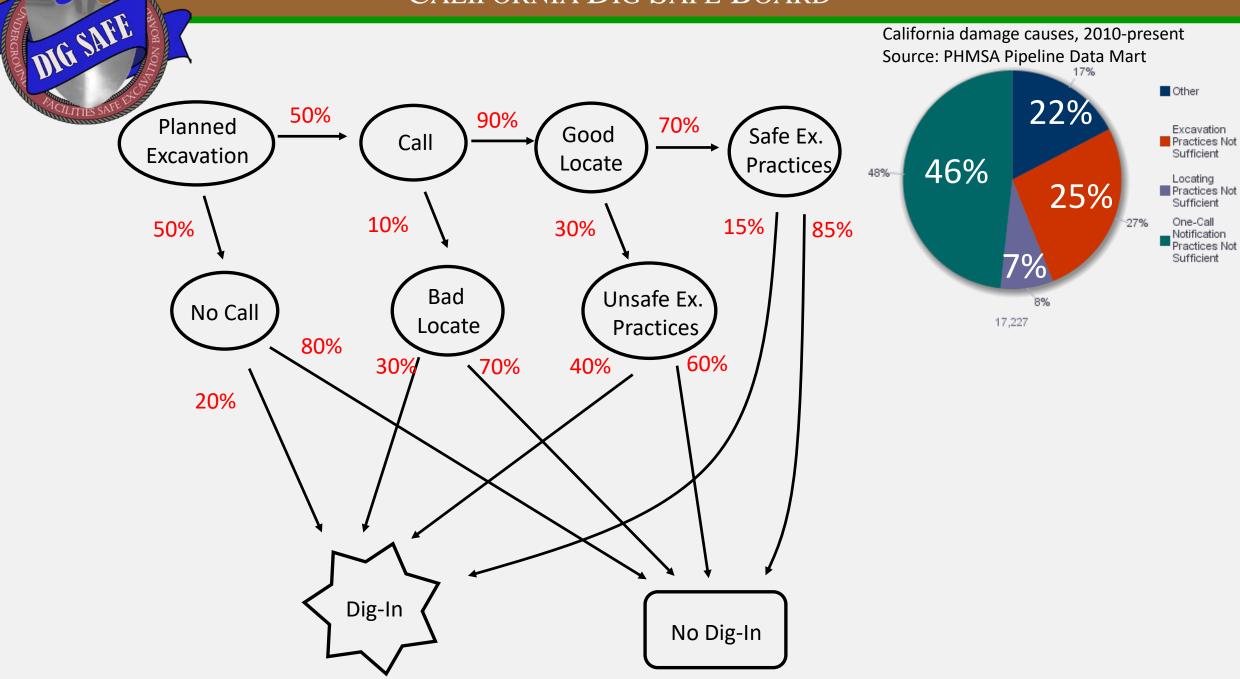
## 3. Models: Event Trees

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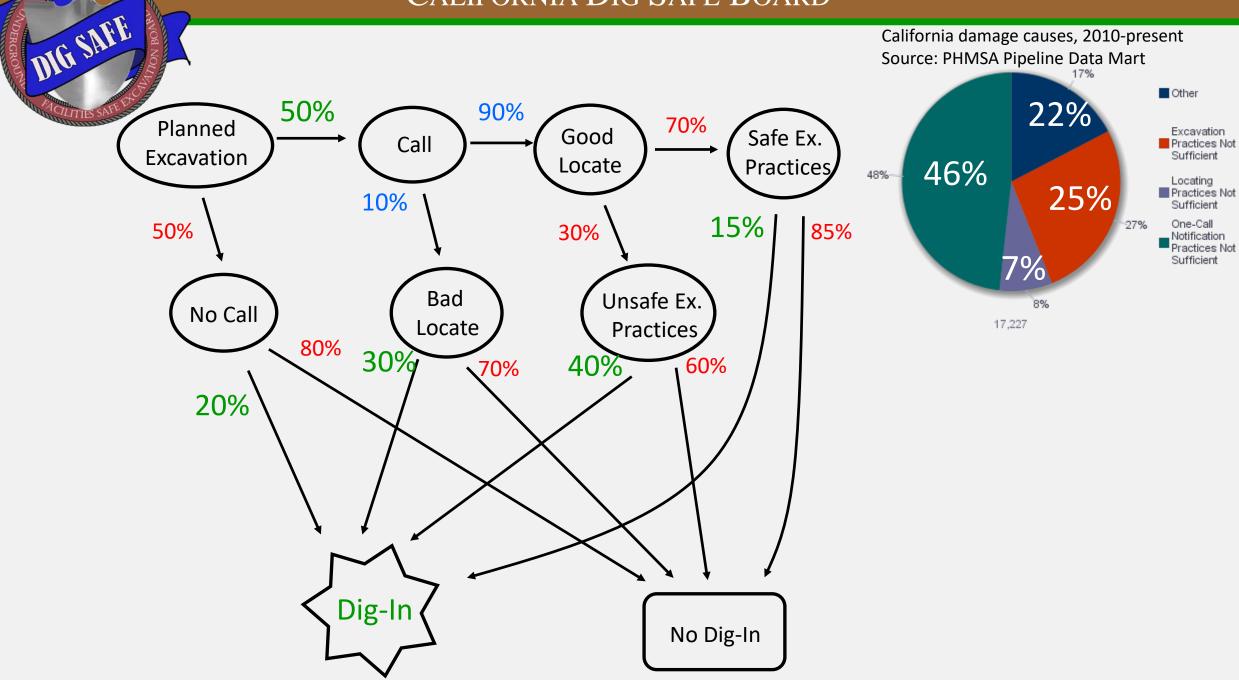
"There is no knowledge of external reality without the anticipation of future experience...there is no knowledge without interpretation."

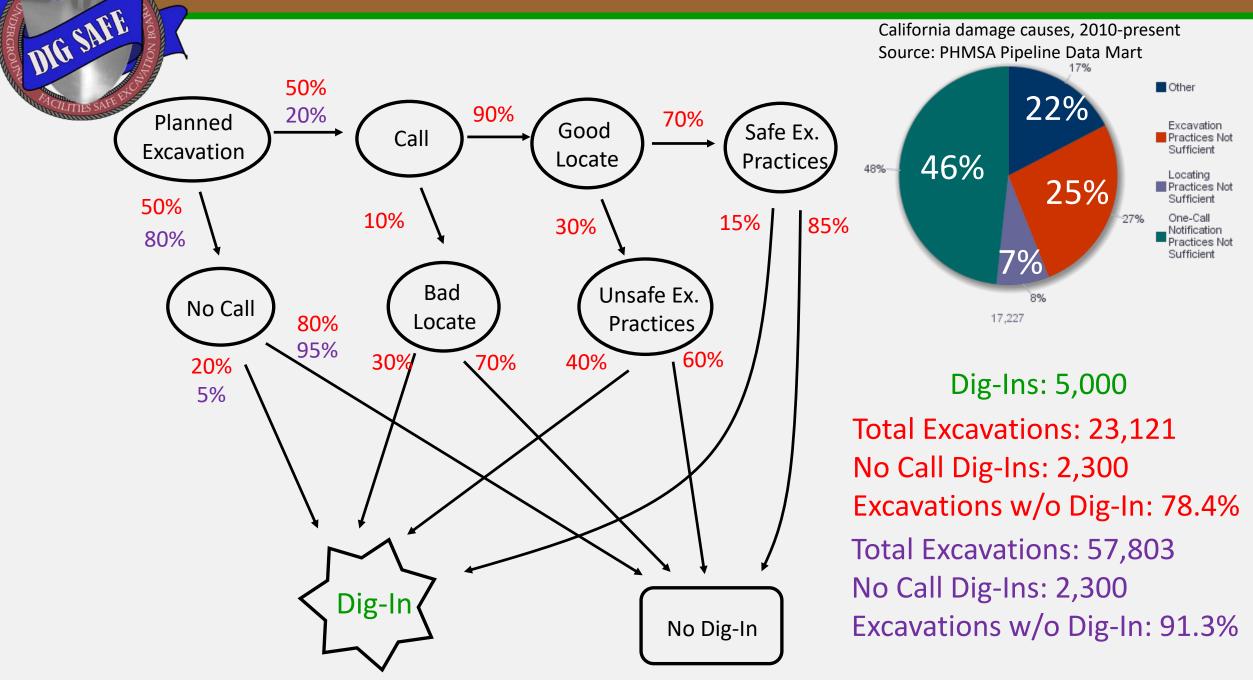
- Charles I. Lewis, Mind and the World-Order

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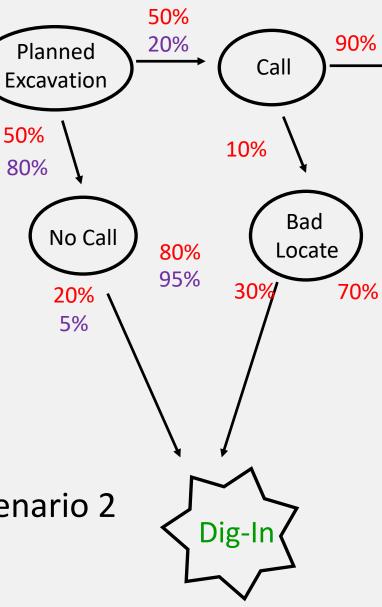


 $\mathbf{R} = \mathbf{F} \mathbf{x} \mathbf{E} \mathbf{x} \mathbf{C}$ 

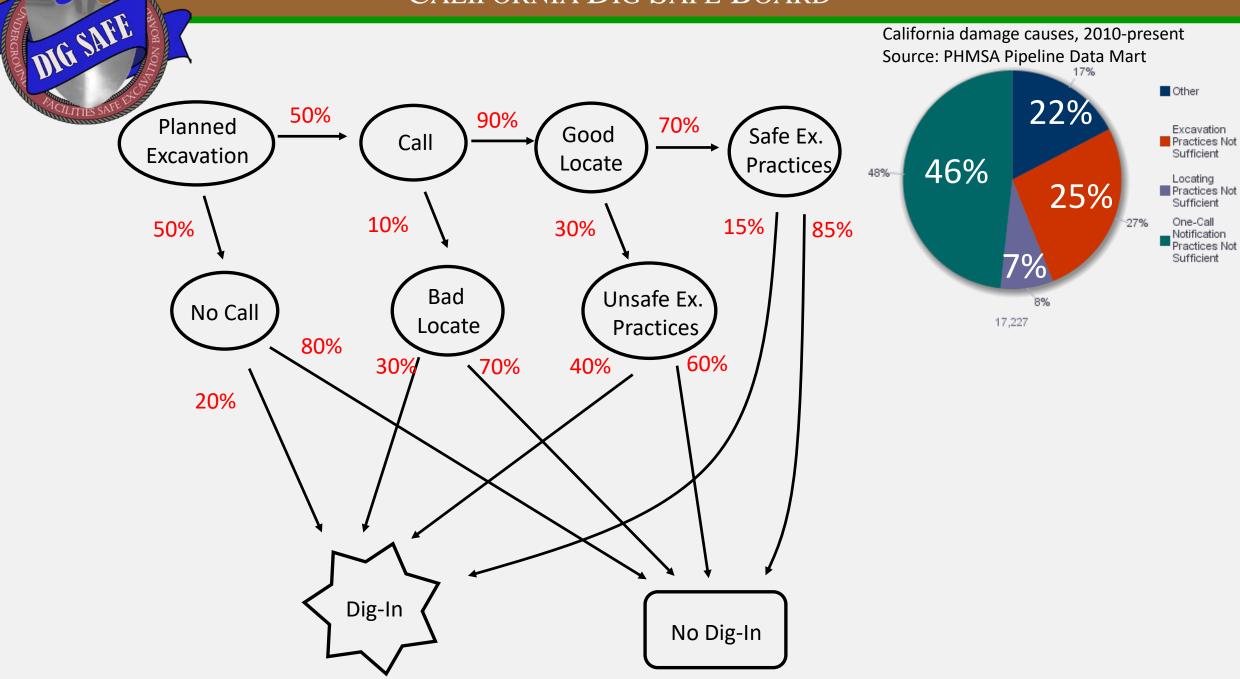
No Call Dig-In Probability  $P_1 = F_1 \times E_1 = 2,300/yr$  $P_2 = F_2 \times E_2 = 2,300/yr$ Number of No Call Excavations  $\mathbf{F}_1 = 23,121 \ge 50\% = 11,560$  $F_{2} = 57,803 \times 80\% = 46,242$ 

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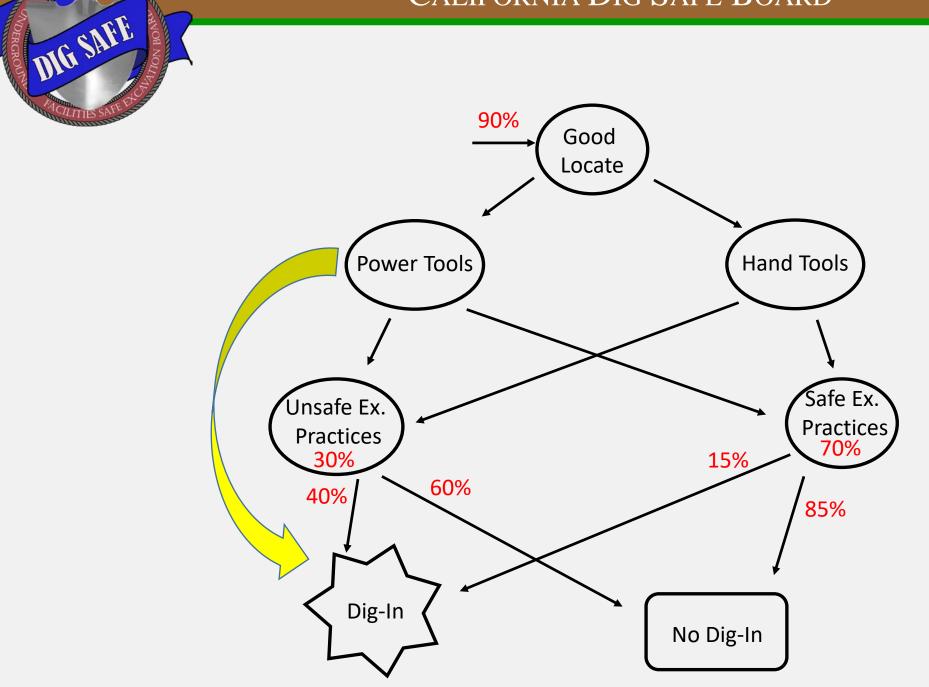
Same risk reduction requires 4x more interventions in Scenario 2



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## Conclusion

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- Macro-scale indicators can, with some effort, be used to compare state programs, but there is little value in doing so.
- We can't accurately predict the outcomes of our interventions by only looking at damage data. We need data about work done without damage.
- Event trees can be used to model the excavation landscape and predict intervention outcomes.
- Probabilities can be used in risk assessment, which can help determine priorities (but risk assessment may not be useful to compare risks across different infrastructure).