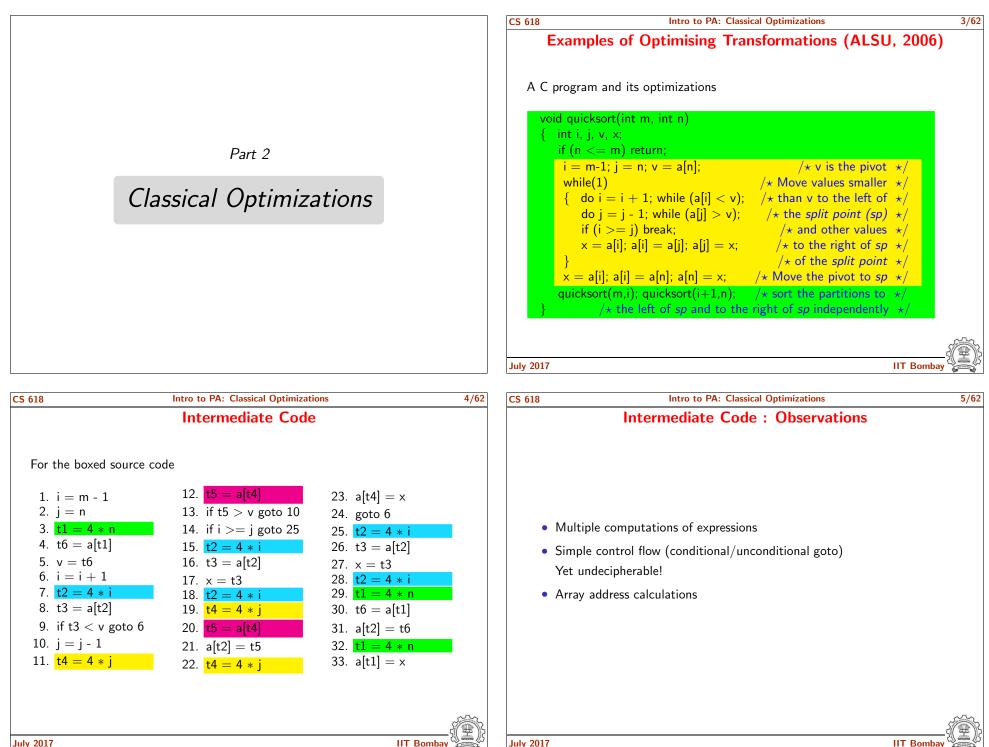
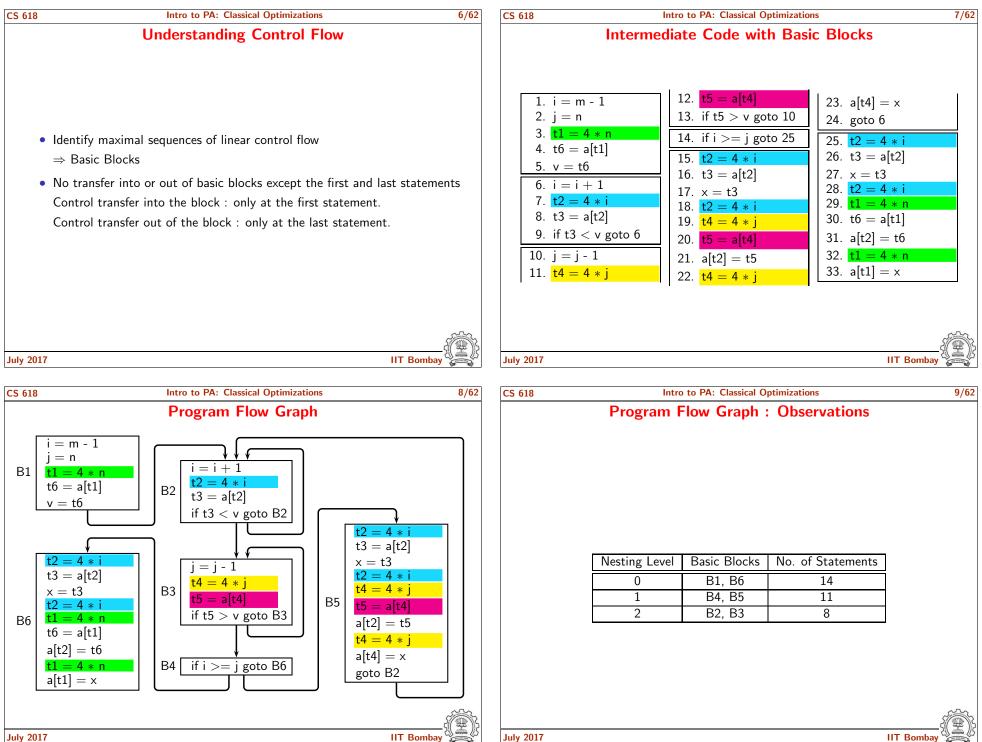
Introduction to Program Analysis	
Uday Khedker (www.cse.iitb.ac.in/~uday) Department of Computer Science and Engineering, Indian Institute of Technology, Bombay	Part 1 About These Slides

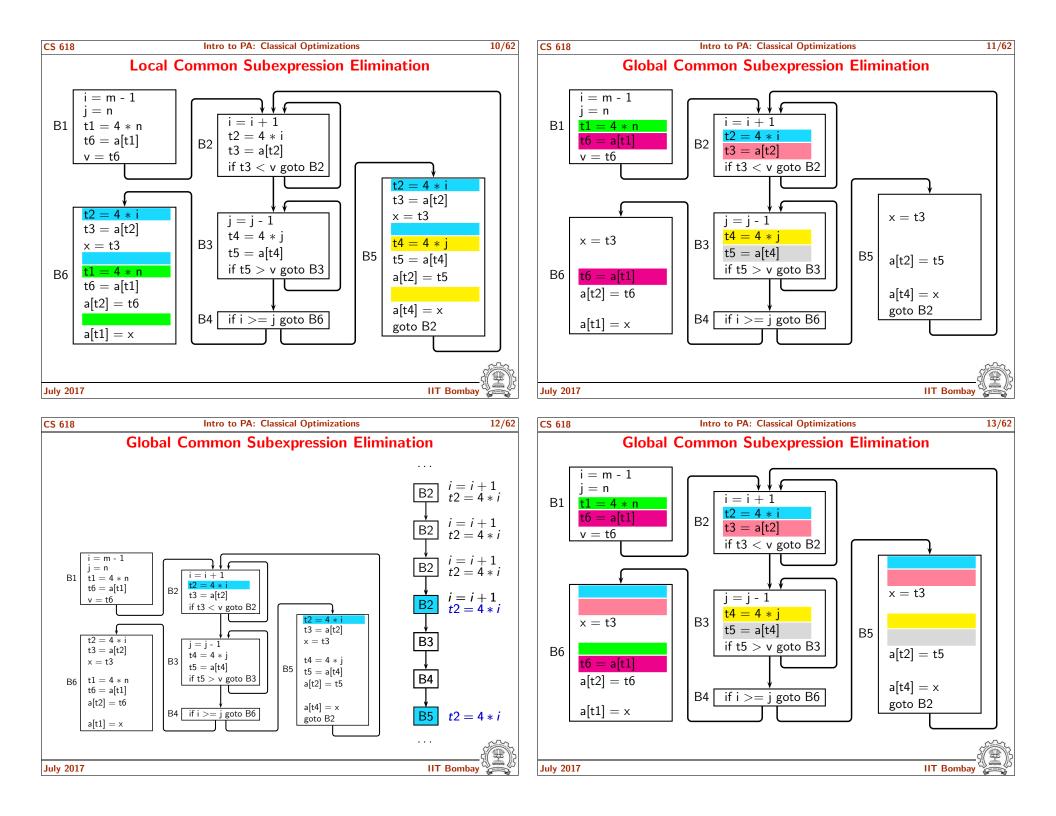
CS 618	Intro to PA: About These Slides	1/62	CS 618	Intro to PA: Outline	2/62
	Copyright			Motivating the Need of Program Analysis	
	onstitute the lecture notes for CS618 Program Anal nd have been made available as teaching material a	-			
Data Flov Group). 2 (Indian ec	edker, Amitabha Sanyal, and Bageshri Karkare. <i>w Analysis: Theory and Practice</i> . CRC Press (Taylo 2009. dition published by Ane Books in 2013) e above book, some slides are based on the materia			 Come representative examples Classical optimizations performed by compilers Optimizing heap memory usage Course details, schedule, assessment policies etc. 	
following books			• F	Program Model	
	o, M. Lam, R. Sethi, and J. D. Ullman. <i>Compilers:</i> es, and Tools. Addison-Wesley. 2006.	Principles,	• 9	Soundness and Precision	
	cht. <i>Flow Analysis of Computer Programs</i> . Elsevier Iland Inc. 1977.				
These slides ar	re being made available under GNU FDL v1.2 or lat	ter purely for			
academic or re	esearch use.	8			
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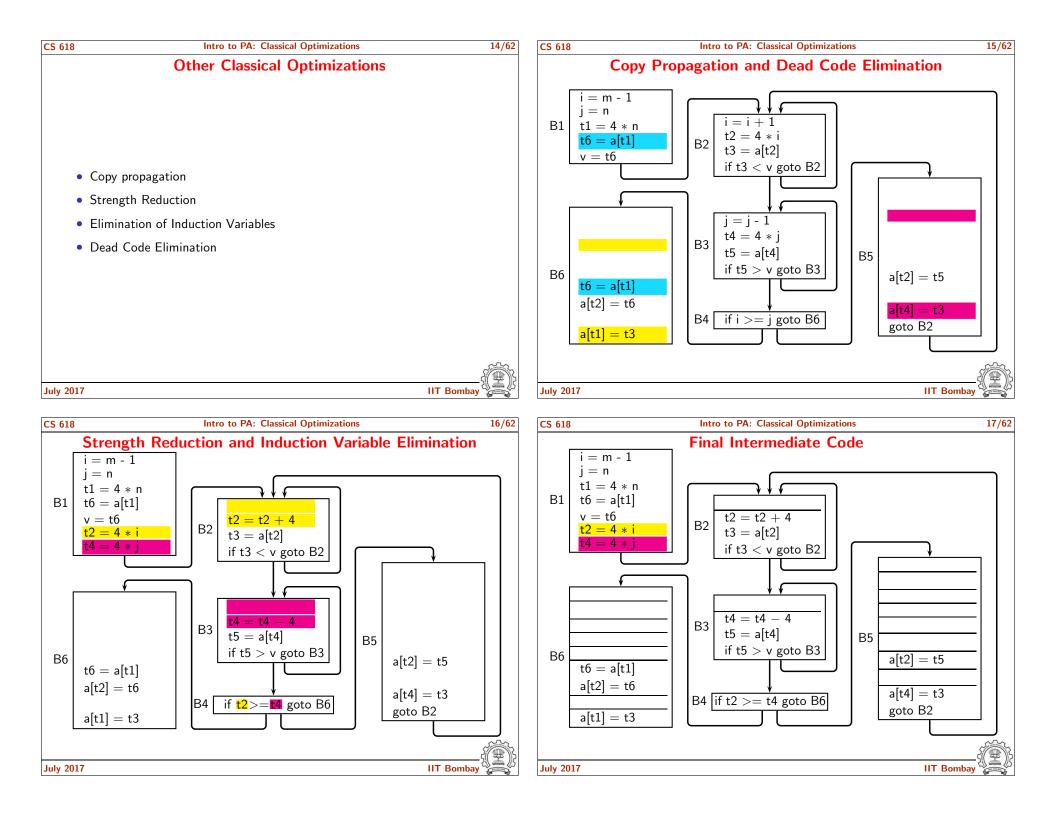


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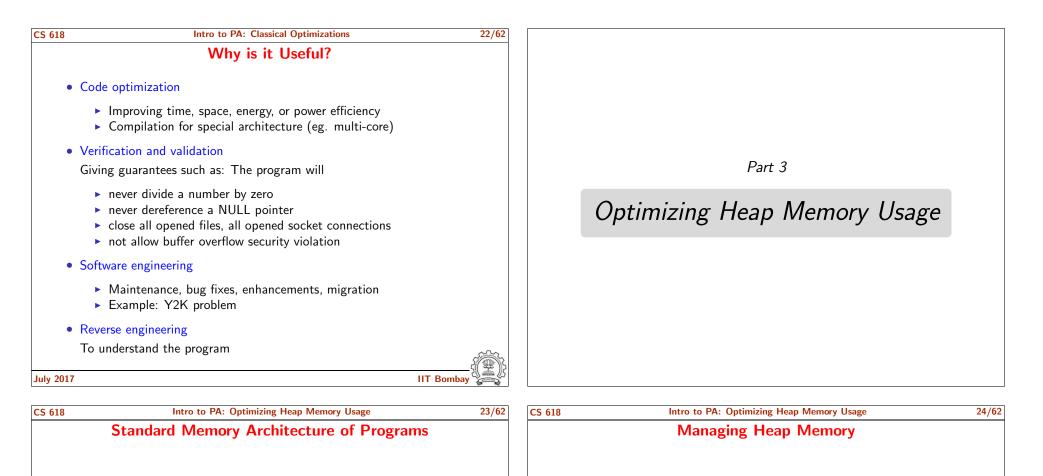


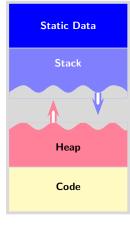
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	Intro to PA: Classical Optimization	ons	18/62 CS 618	Intro to PA: Classical Optimizations	19/
Οι	otimized Program Flow	Graph		Observations	
f we assume that a lo omputations saved at	0 14 1 11 2 8 op is executed 10 times, then	imized 10 4 6 the number of	• •	Optimizations are transformations based on some information. Systematic analysis required for deriving the information. We have looked at data flow optimizations. Many control flow optimizations can also be performed.	
2017		C IIT Bombay	July 2017	IIT B	ombay
Categories of	Intro to PA: Classical Optimization		20/62 CS 618	Intro to PA: Classical Optimizations What is Program Analysis?	21
	Optimizing Transformat				21,
Categories of Code Motion Redundancy Elimir	Optimizing Transformation Machine Independent	tions and Analyses	Disco	What is Program Analysis?	21,
Categories of Code Motion Redundancy Elimin Control flow Optim	Optimizing Transformation nation Machine Independent ization Machine Dependent cions Machine Dependent uling Machine Dependent	tions and Analyses Flow Analysis (Data + Control) Dependence Analysis	Disco	What is Program Analysis? vering information about a given program Representing the dynamic behaviour of the program	21,
Categories of Code Motion Redundancy Elimir Control flow Optim Loop Transformat	Optimizing Transformation nation Machine Independent ization Machine Dependent uling ion ation Machine Dependent	Flow Analysis (Data + Control) Dependence Analysis (Data + Control) Several Independent	Disco	What is Program Analysis? vering information about a given program Representing the dynamic behaviour of the program Most often obtained without executing the program Static analysis Vs. Dynamic Analysis Example of loop tiling for parallelization	21





Heap allocation provides the flexibility of

• *Variable Sizes.* Data structures can grow or shrink as desired at runtime.

(Not bound to the declarations in program.)

• Variable Lifetimes. Data structures can be created and destroyed as desired at runtime.

(Not bound to the activations of procedures.)

Decision 1: When to Allocate?

- Explicit. Specified in the programs. (eg. Imperative/OO languages)
- Implicit. Decided by the language processors. (eg. Declarative Languages)

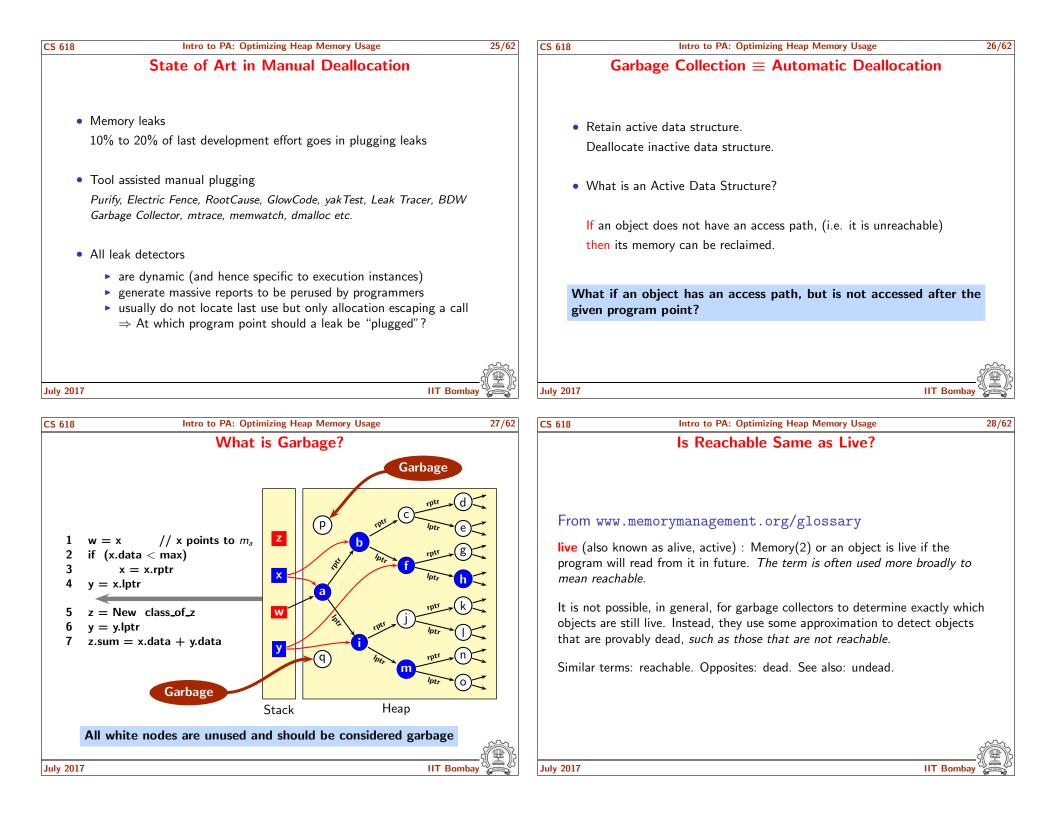
Decision 2: When to Deallocate?

- Explicit. Manual Memory Management (eg. C/C++)
- Implicit. Automatic Memory Management aka Garbage Collection (eg. Java/Declarative languages)

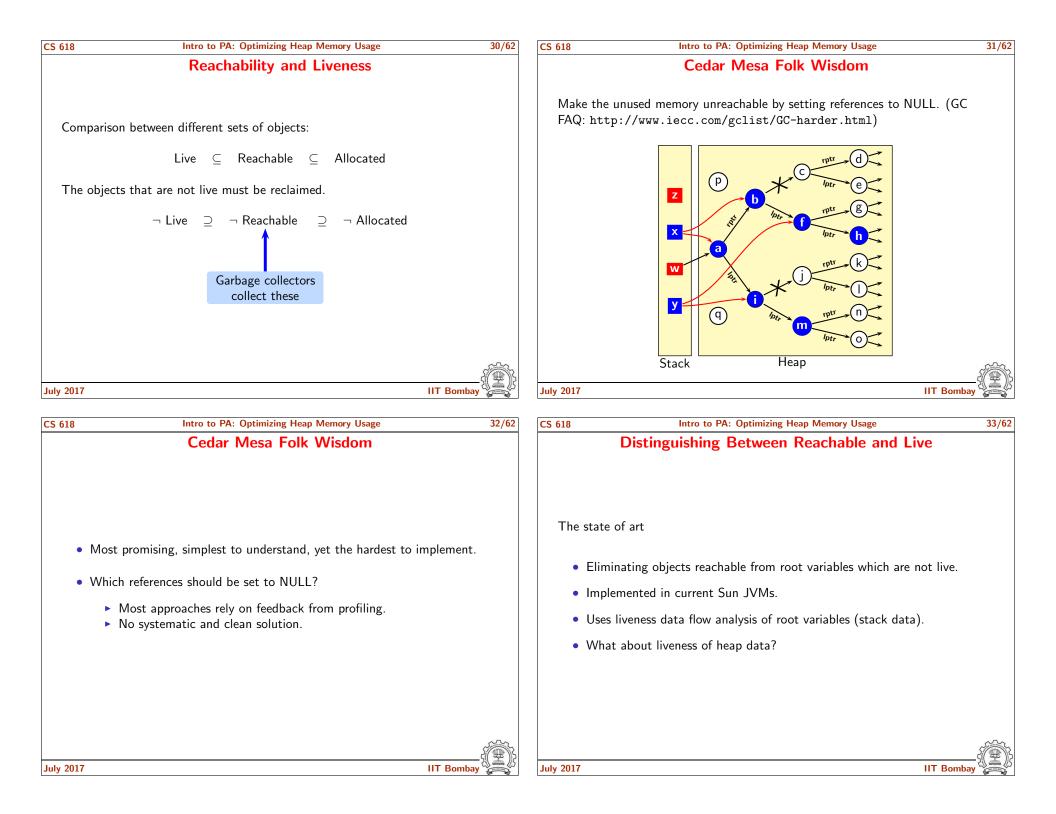
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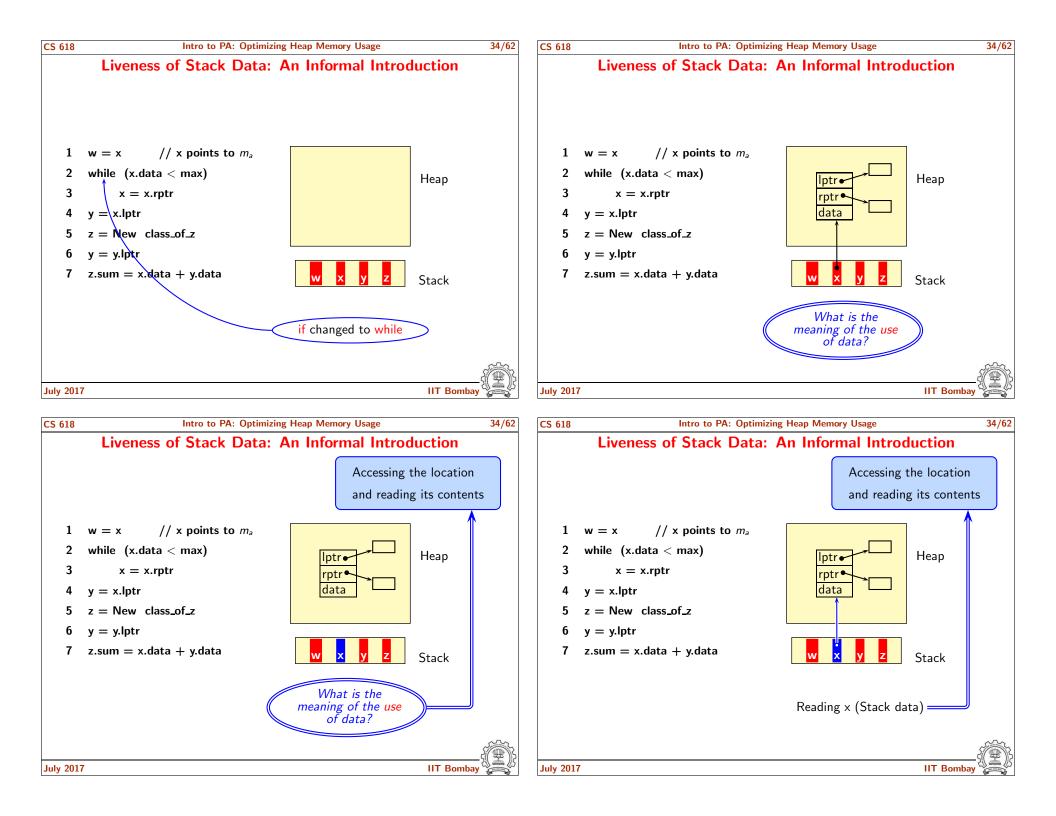


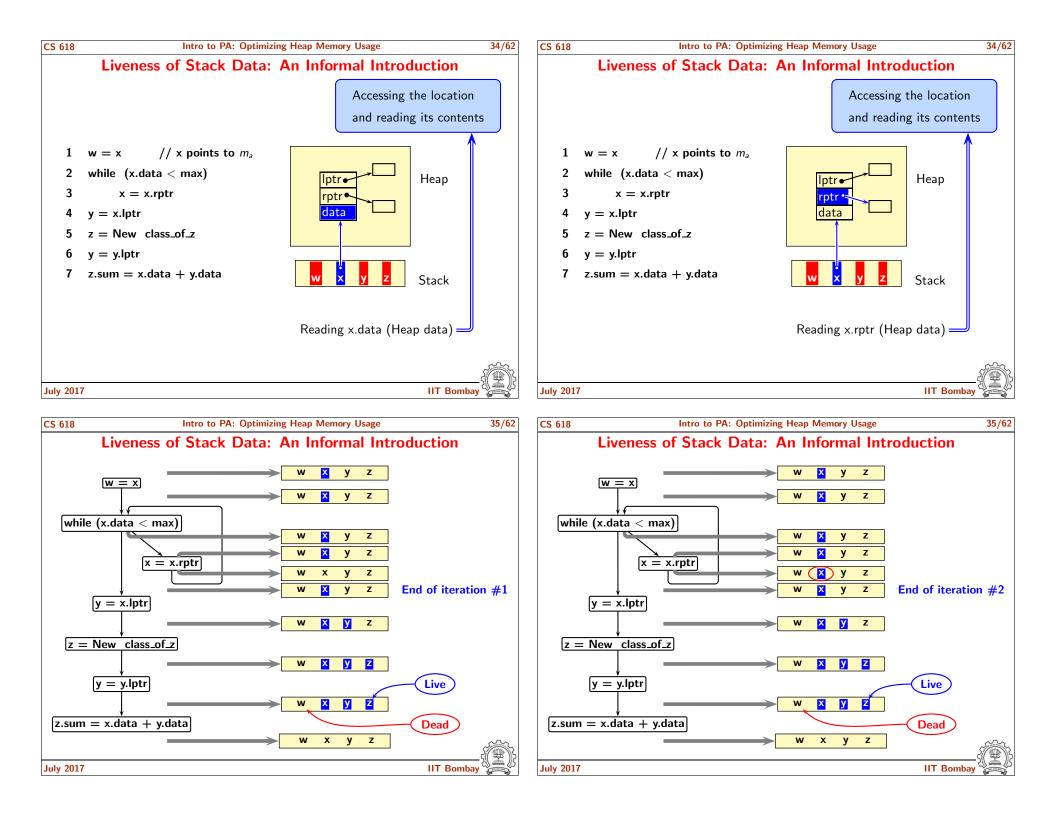


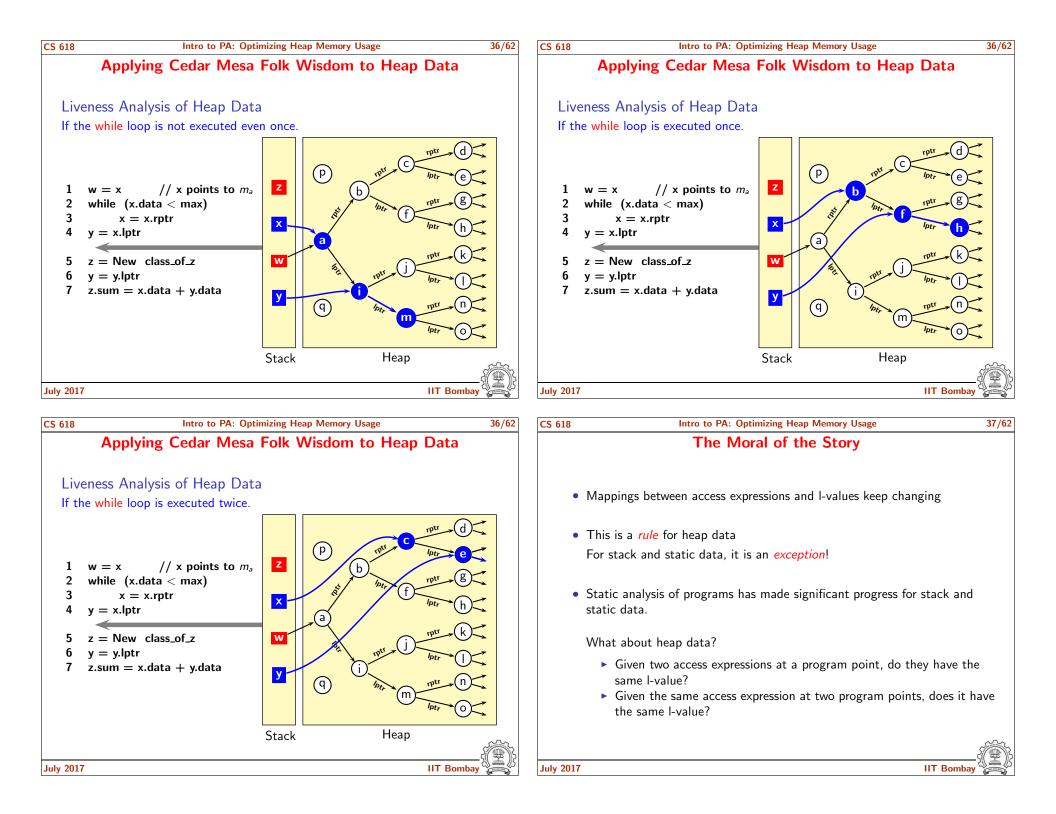


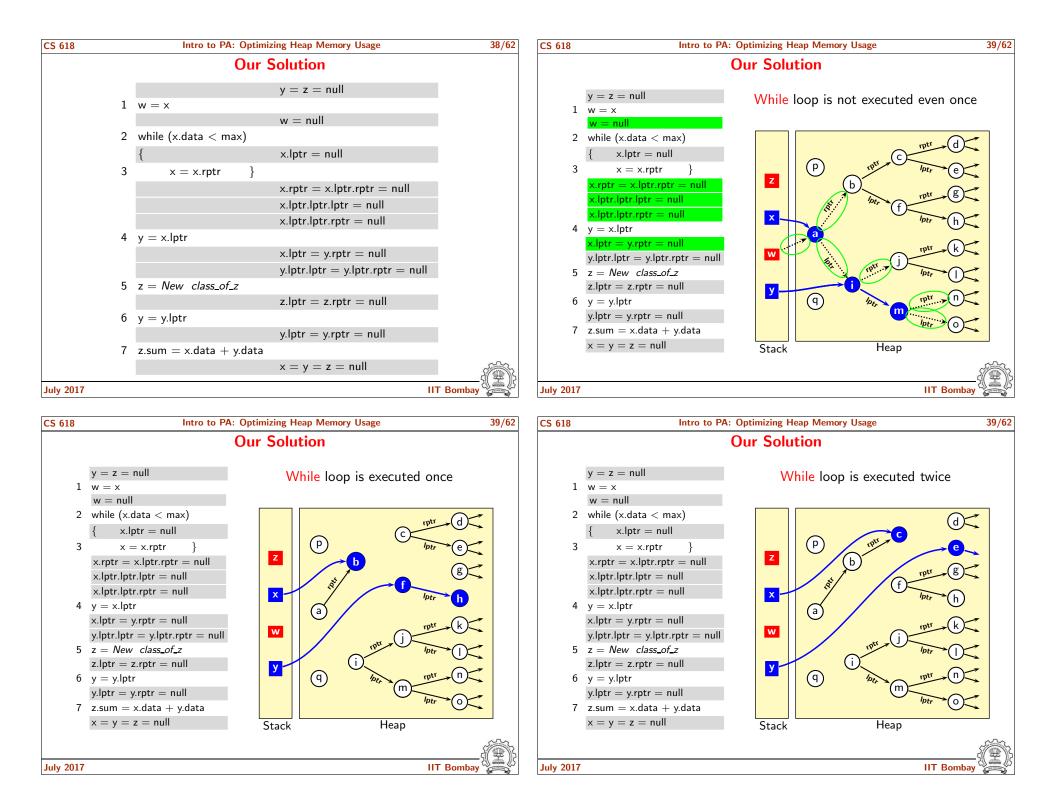
CS 618 Intro to PA: Optimizing Heap Memory Usage	29/62	CS 618	Intro to PA: Optimizing Heap Memory Usage	30/62
Is Reachable Same as Live?			Reachability and Liveness	
		Compari	son between different sets of objects:	
• Not really. Most of us know that.			Live ? Reachable ? Allocated	
Even with the state of art of garbage collection, 24% memory remains unclaimed	to 76% unused	The obje	ects that are not live must be reclaimed.	
• The state of art compilers, virtual machines, garbage distinguish between the two	collectors cannot			
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CS 618 Intro to PA: Optimizing Heap Memory Usage	30/62	CS 618	Intro to PA: Optimizing Heap Memory Usage	30/62
Reachability and Liveness			Reachability and Liveness	
Comparison between different sets of objects:		Compari	son between different sets of objects:	
$Live \subseteq Reachable \subseteq Allocated$	I		$Live \ \subseteq \ Reachable \ \subseteq \ Allocated$	
The objects that are not live must be reclaimed.		The obje	ects that are not live must be reclaimed.	
			\neg Live ? \neg Reachable ? \neg Allocated	
				<u>,</u>
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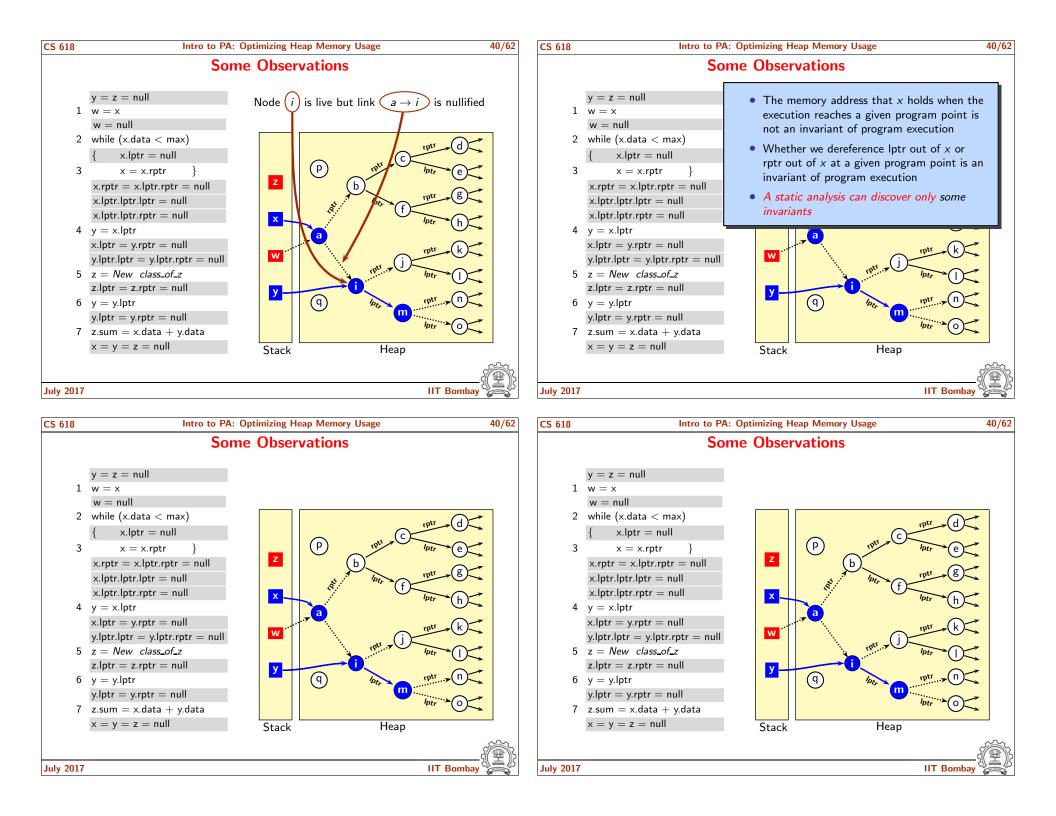


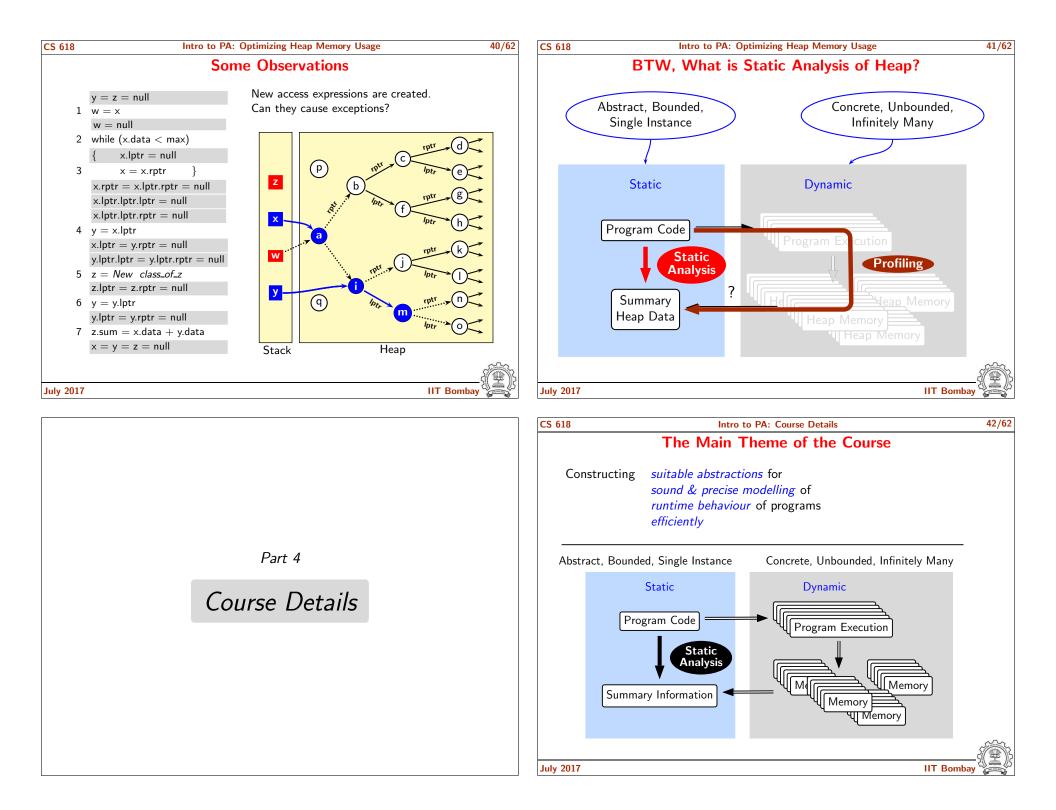




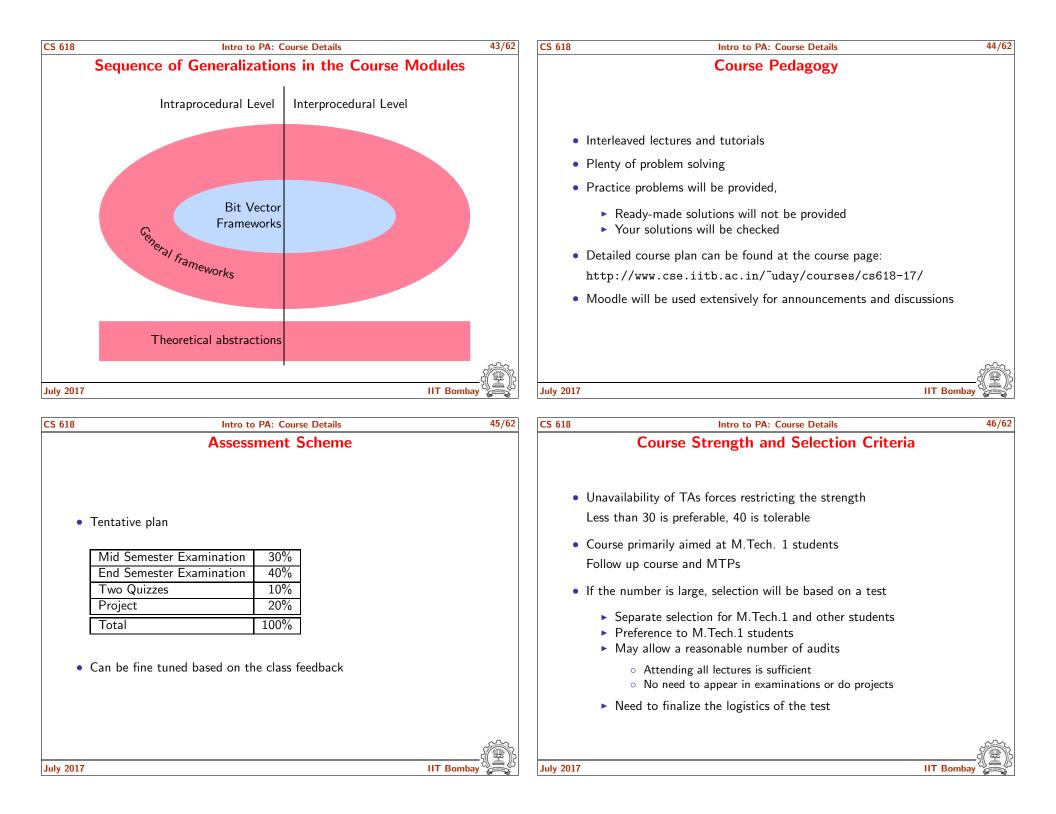


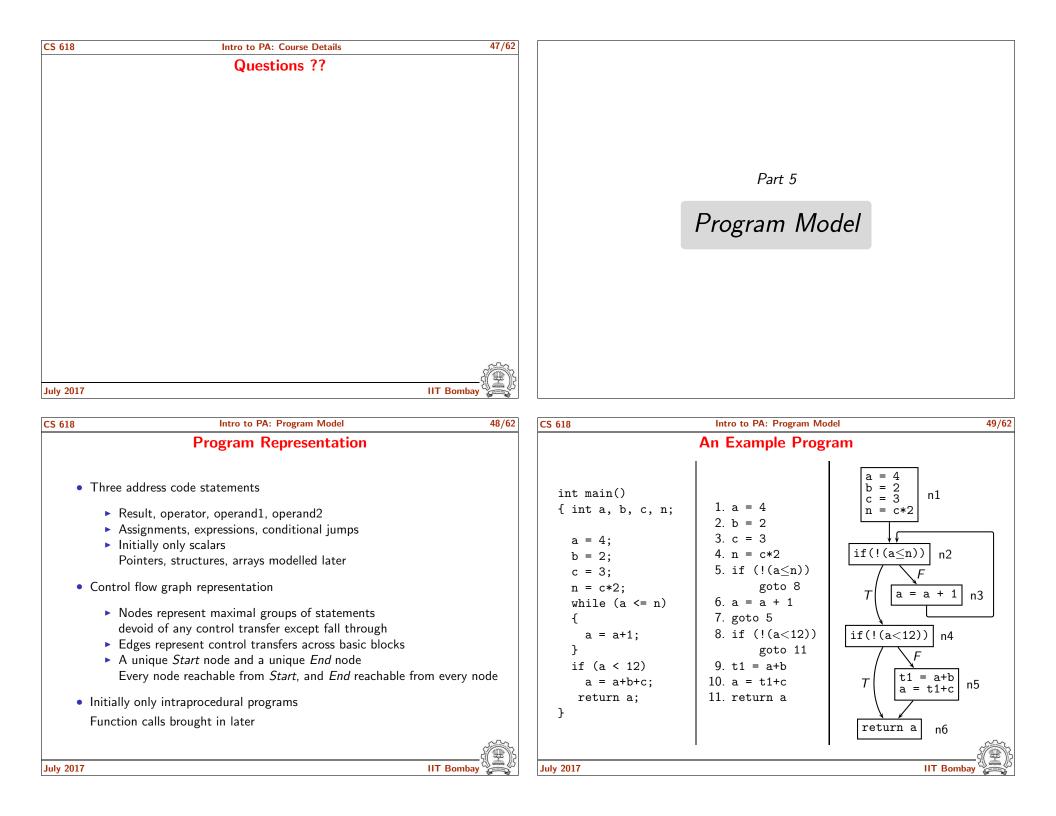


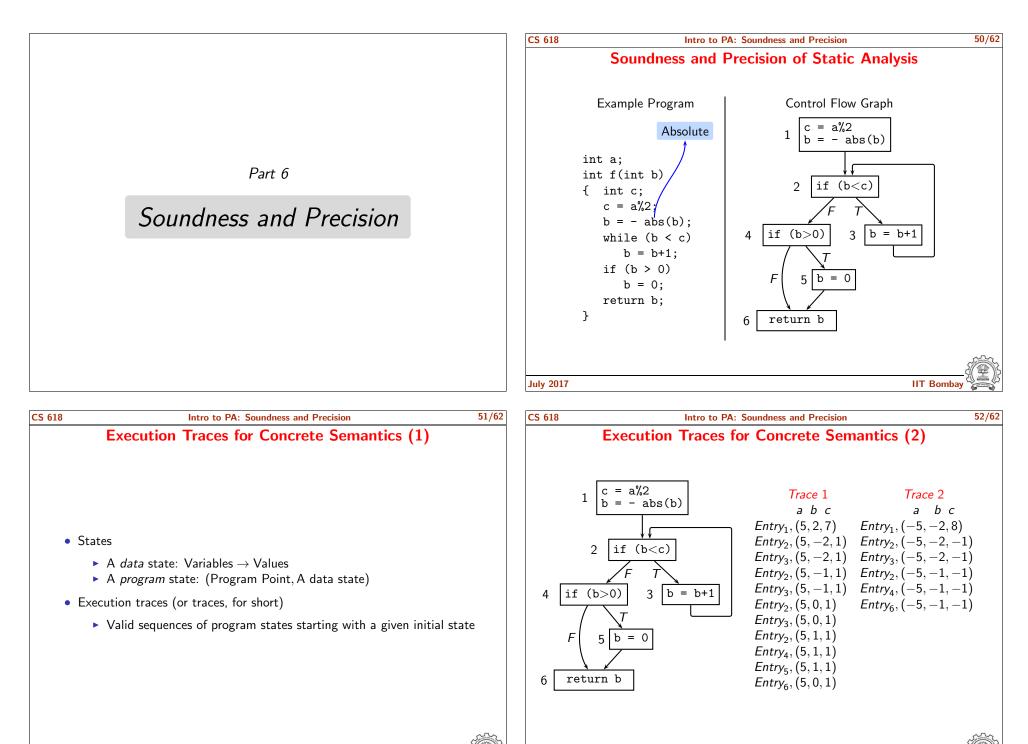




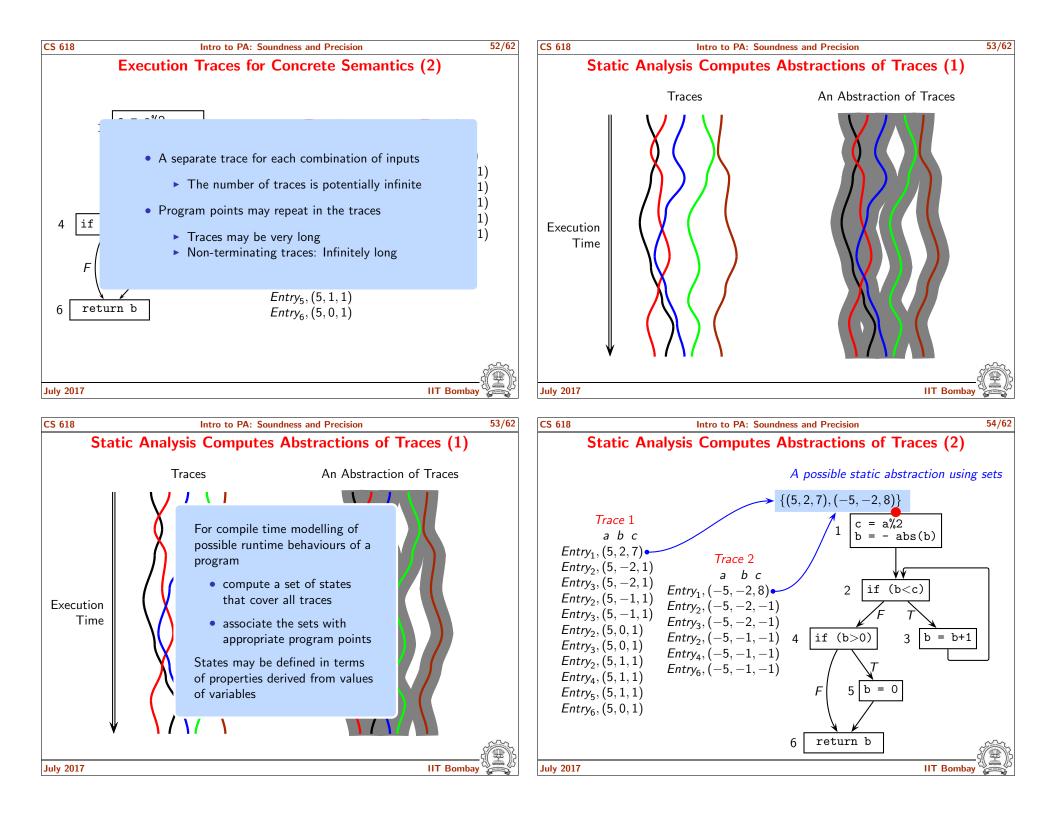


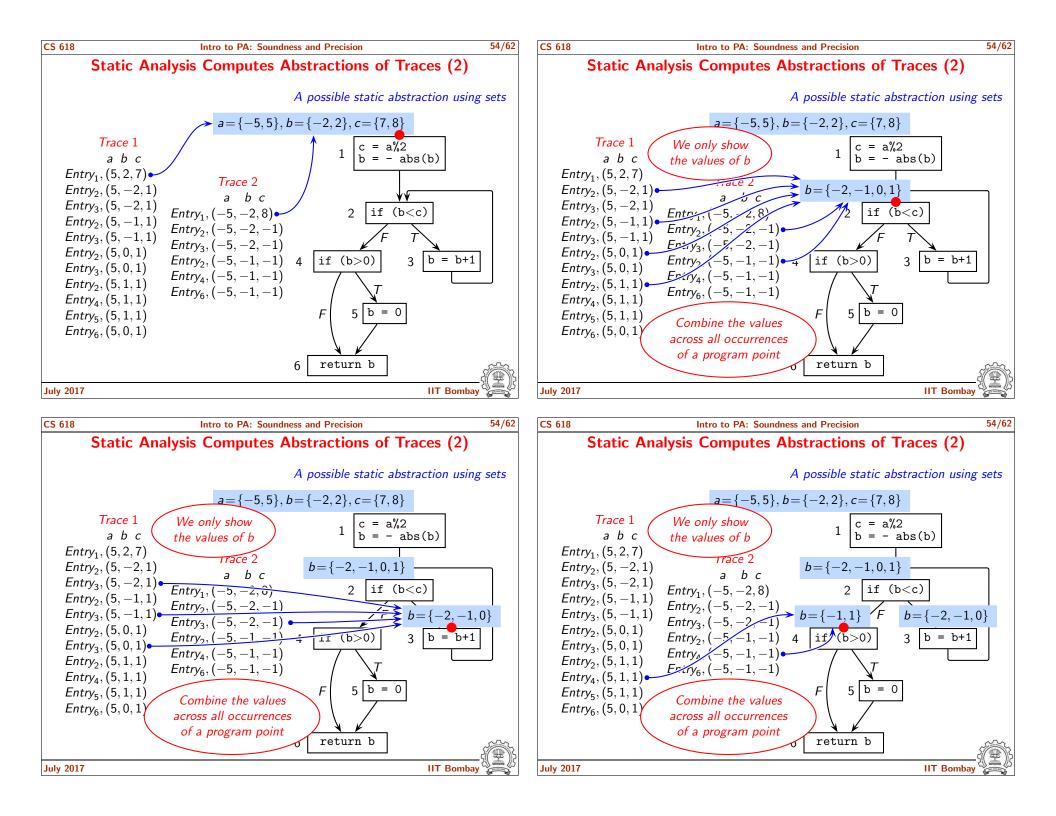


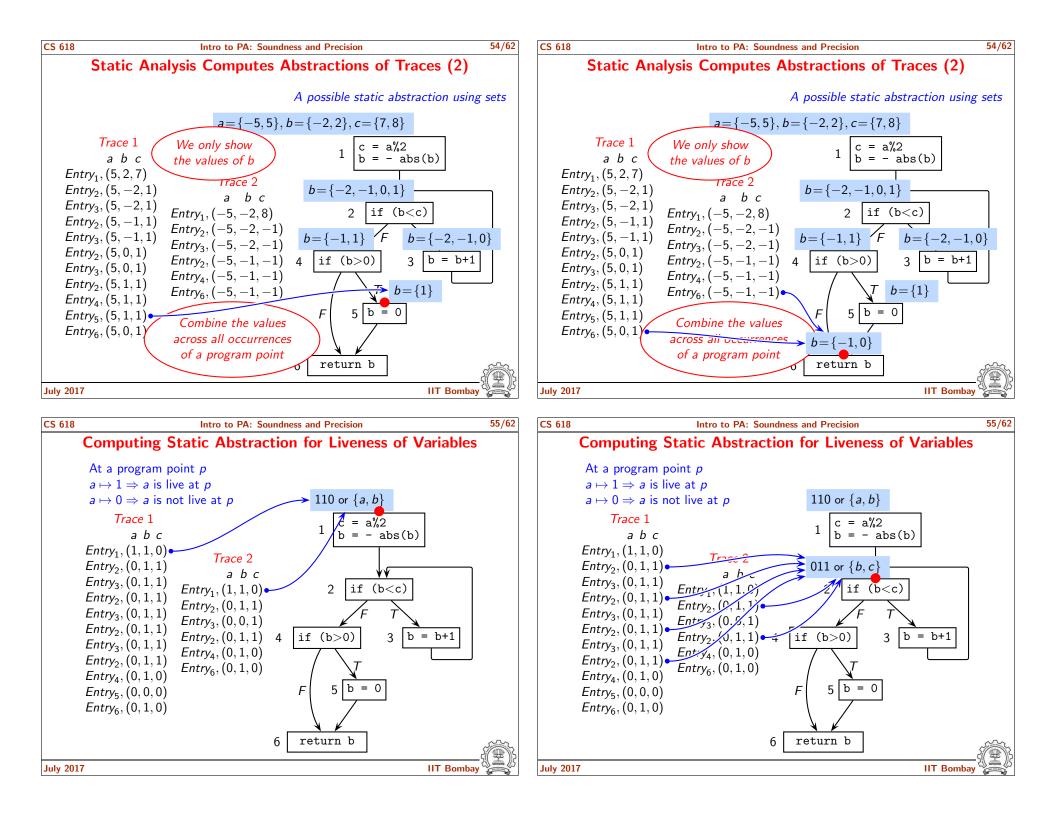


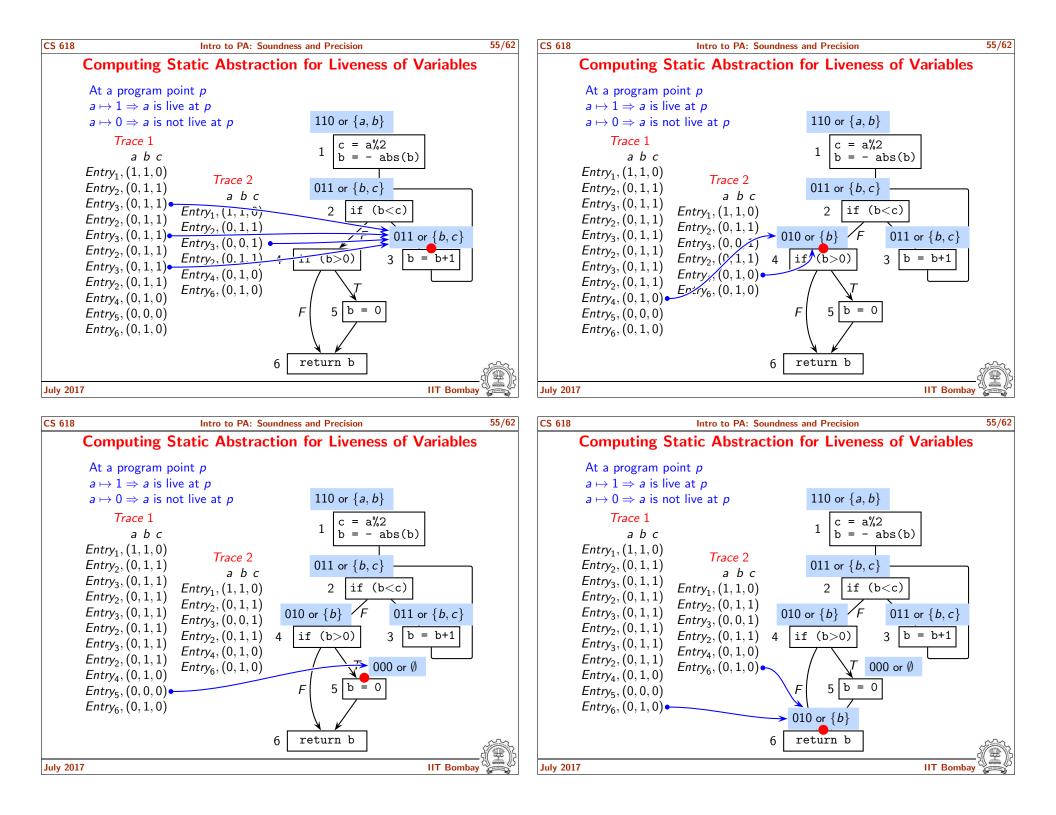


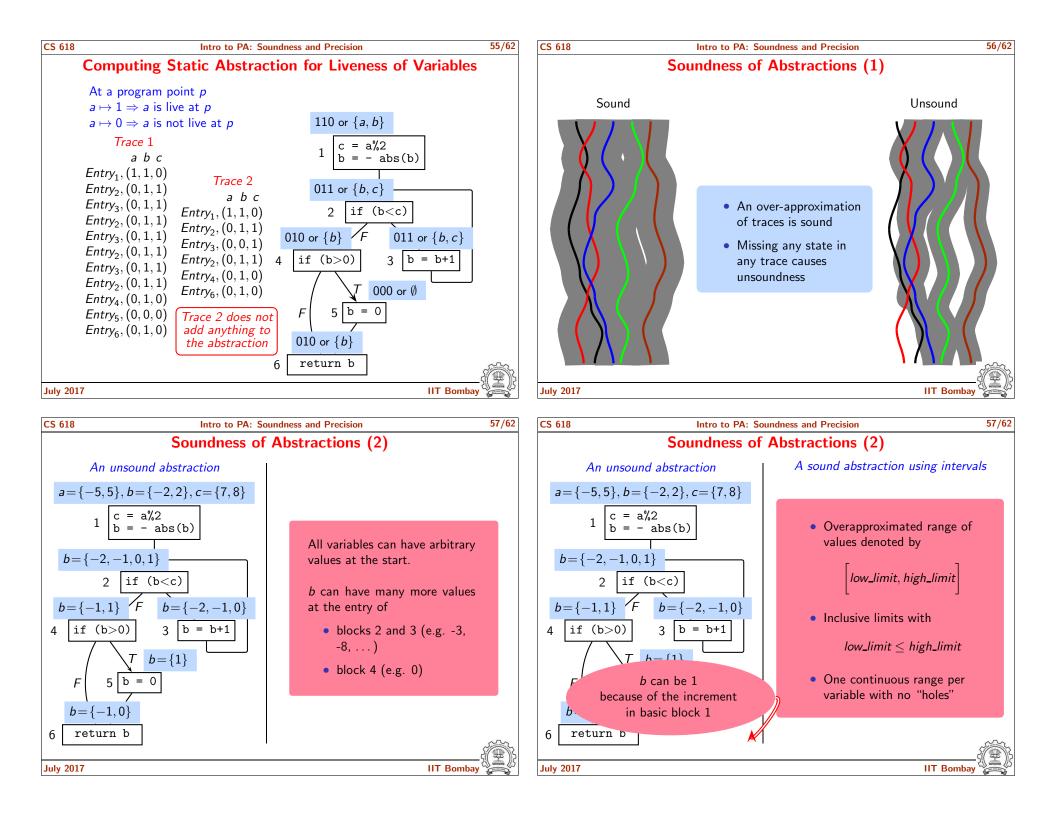
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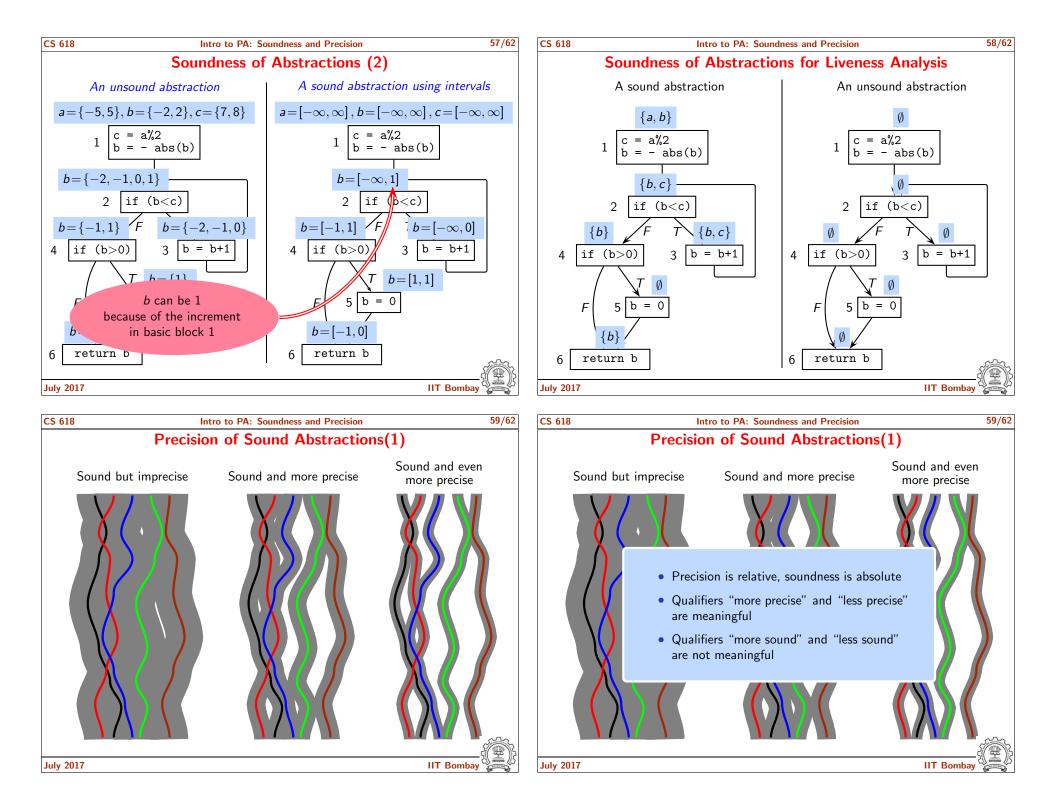


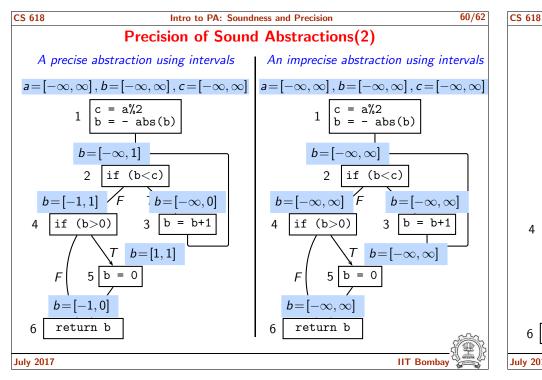


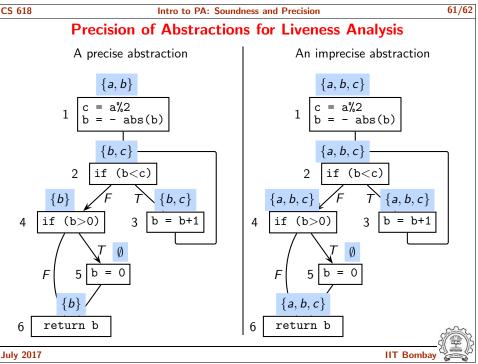












CS 618	Intro to PA: Soundness and Precision	62/62
	Limitations of Static Analysis	
•	In general, the computation of <i>exact</i> static abstraction is <i>undecidable</i>	
	 Possible reasons Values of variables not known Branch outcomes not known Infinitely many paths in the presence of loops or recursion Infinitely many values 	
	 We have to settle for some imprecision How are data states compared to distinguish between a sound and unsound (or a precise or an imprecise result)? 	
	We have introduced the concepts intuitivelyWill define them formally in a later module	
•	Goodness of a static analysis lies in minimizing imprecision without compromising on soundness	
	Additional expectations: Efficiency and scalability	

