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# Continuity and Robustness of Programs

Seminar: Robustness of Hardware and Software Systems

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Saarbrücken, February 21, 2014

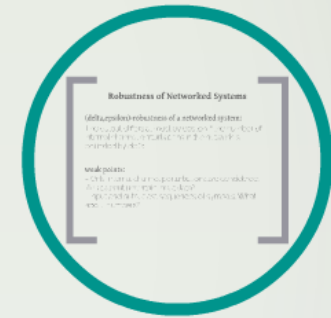
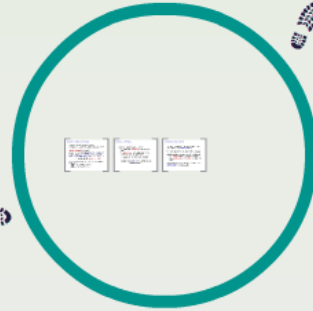


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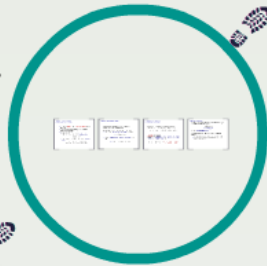
# ROBUSTNESS

Lipschitz continuity

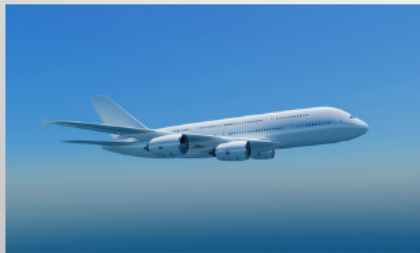


networked systems

continuity



(delta,epsilon)-robustness



safety-critical systems



*safety-critical systems*

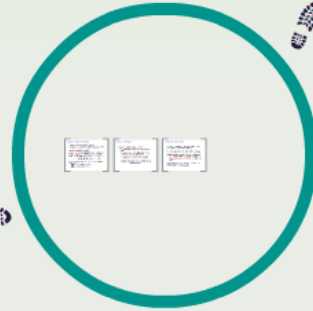


caches

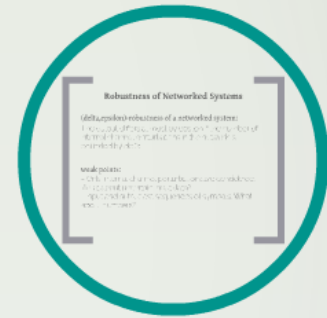


# ROBUSTNESS

Lipschitz  
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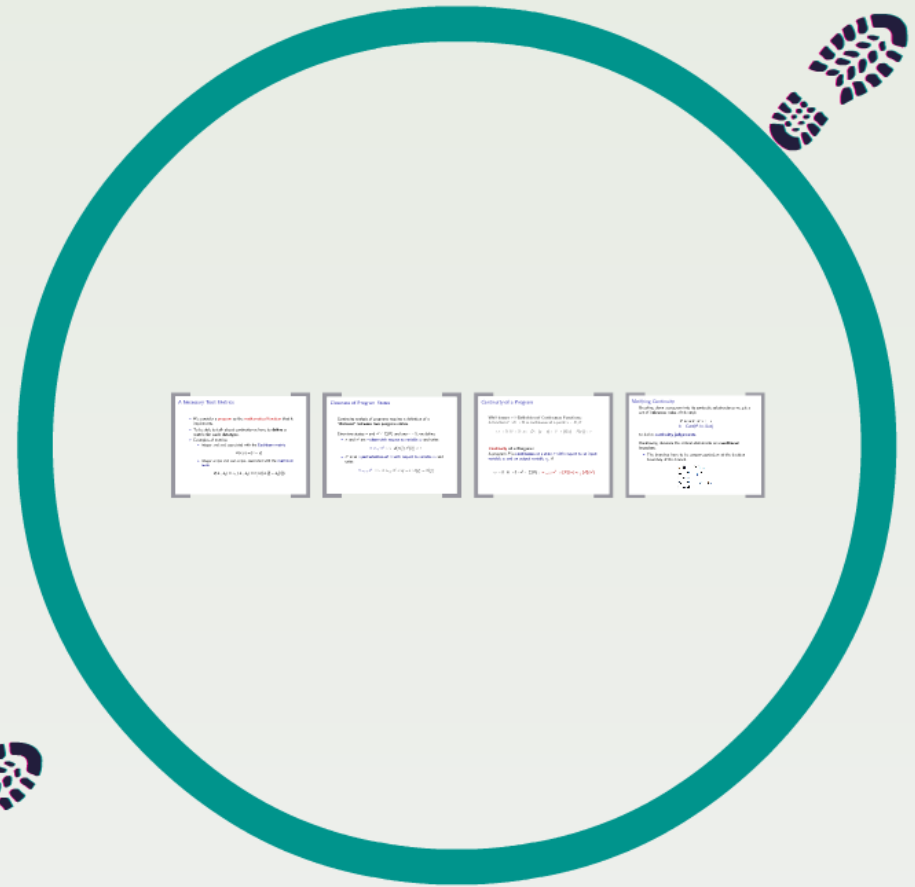
( $\delta, \epsilon$ )-  
robustness



safety-critical systems



# continuity



## A Necessary Tool: Metrics

- ▶ We consider a **program** as the **mathematical function** that it implements.
- ▶ To be able to talk about continuity we have to **define a metric for each datatype**.
- ▶ Examples of metrics:
  - ▶ *integer* and *real*, associated with the **Euclidean metric**

$$d(x, y) = |x - y|$$

- ▶ *integer arrays* and *real arrays*, associated with the **maximum norm**

$$d(A_1, A_2) = L_\infty(A_1, A_2) = \max_i (|A_1[i] - A_2[i]|)$$

## Closeness of Program States

Continuity analysis of programs requires a definition of a “**distance**” between two program states.

Given two states  $\sigma$  and  $\sigma' \in \Sigma(P)$  and any  $\epsilon > 0$ , we define:

- ▶  $\sigma$  and  $\sigma'$  are  $\epsilon$ -**close** with respect to variable  $x_i$  and write

$$\sigma \approx_{\epsilon,i} \sigma' :\Leftrightarrow d(\sigma(i), \sigma'(i)) < \epsilon$$

- ▶  $\sigma'$  is an  $\epsilon$ -**perturbation of**  $\sigma$  with respect to variable  $x_i$  and write

$$\sigma \equiv_{\epsilon,i} \sigma' :\Leftrightarrow \sigma \approx_{\epsilon,i} \sigma' \wedge \forall j \neq i : \sigma(j) = \sigma'(j)$$

# Continuity of a Program

## Well-known $\epsilon$ - $\delta$ -Definition of Continuous Functions:

A function  $f : D \rightarrow \mathbb{R}$  is continuous at a point  $x \in D$ , if

$$\forall \epsilon > 0 \exists \delta > 0 \forall y \in D : |x - y| < \delta \Rightarrow |f(x) - f(y)| < \epsilon$$

## Continuity of a Program:

A program  $P$  is **continuous** at a state  $\sigma$  with respect to an input variable  $x_i$  and an output variable  $x_j$ , if

$$\forall \epsilon > 0 \exists \delta > 0 \forall \sigma' \in \Sigma(P) : \sigma \equiv_{\delta,i} \sigma' \Rightarrow \llbracket P \rrbracket(\sigma) \approx_{\epsilon,j} \llbracket P \rrbracket(\sigma')$$



## Verifying Continuity

Breaking down a program into its syntactic substructures we get a set of **inference rules** of the style

$$\frac{P \text{ is SKIP or } x := e}{b \vdash \text{Cont}(P, \text{In}, \text{Out})}$$

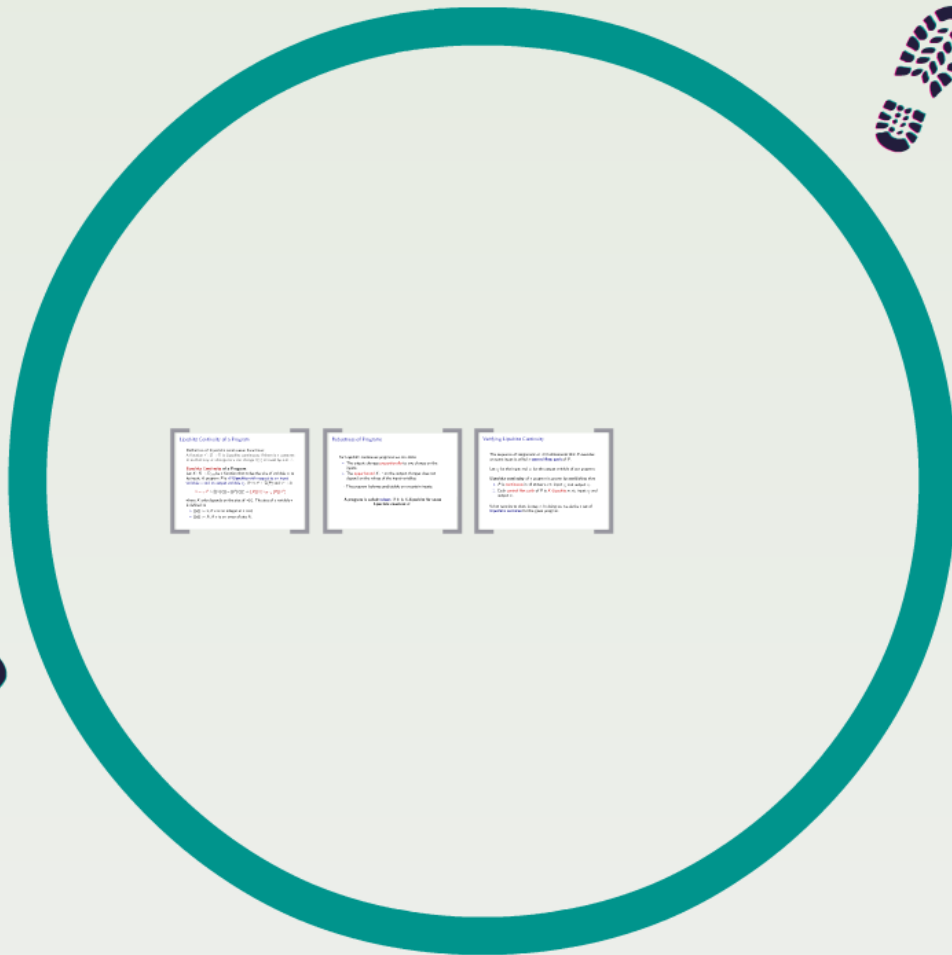
to derive **continuity judgements**.

Disallowing divisions the critical statements are **conditional branches**.

- ▶ The branches have to be *output-equivalent* at the decision boundary of the branch.

```
1: if  $x > 2$  then  
2:    $y := \frac{1}{2} \cdot x$   
3: else  
4:    $y := -5x + 11$   
5: end if
```

# Lipschitz continuity



**Definition of Lipschitz Continuity**  
A function  $f: X \rightarrow Y$  is said to be Lipschitz continuous if there exists a constant  $L \geq 0$  such that for all  $x, y \in X$ ,  
$$\|f(x) - f(y)\| \leq L \|x - y\|$$

**Properties of Lipschitz**  
The Lipschitz constant  $L$  is the smallest constant such that the inequality holds. If  $f$  is Lipschitz continuous with constant  $L$ , then  $f$  is also continuous.

**Example of Lipschitz Continuity**  
The function  $f(x) = \sin(x)$  is Lipschitz continuous with constant  $L = 1$ . This is because the derivative of  $\sin(x)$  is  $\cos(x)$ , which has a maximum absolute value of 1.

**Definition of Lipschitz Continuity**  
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**Conclusion for the Approximation**  
The Lipschitz constant  $L$  is the smallest constant such that the inequality holds. If  $f$  is Lipschitz continuous with constant  $L$ , then  $f$  is also continuous.

# Lipschitz Continuity of a Program

## Definition of Lipschitz continuous Functions:

A function  $f : D \rightarrow \mathbb{R}$  is Lipschitz continuous, if there is a constant  $K$  so that any  $\pm\epsilon$ -change to  $x$  can change  $f(x)$  at most by  $\pm K \cdot \epsilon$ .

## Lipschitz Continuity of a Program:

Let  $K : \mathbb{N} \rightarrow \mathbb{R}_{\geq 0}$  be a function that takes the size of variable  $x_i$  as its input. A program  $P$  is  **$K$ -Lipschitz** with respect to an input variable  $x_i$  and an output variable  $x_j$ , if  $\forall \sigma, \sigma' \in \Sigma(P)$  and  $\forall \epsilon > 0$

$$\sigma \equiv_{\epsilon, i} \sigma' \wedge (\|\sigma(i)\| = \|\sigma'(i)\|) \Rightarrow \llbracket P \rrbracket(\sigma) \approx_{K \cdot \epsilon, j} \llbracket P \rrbracket(\sigma')$$

where  $K$  only depends on the size of  $\sigma(i)$ . The size of a variable  $v$  is defined as

- ▶  $\|v\| := 1$ , if  $v$  is an integer or a real,
- ▶  $\|v\| := N$ , if  $v$  is an array of size  $N$ .

# Robustness of Programs

For Lipschitz continuous programs we can state:

- ▶ The output changes **proportionally** to any change on the inputs.
- ▶ The **upper bound**  $K \cdot \epsilon$  on the output changes does not depend on the values of the input variables.

→ The program behaves predictably on uncertain inputs.

**A program is called **robust**, if it is  $K$ -Lipschitz for some Lipschitz constant  $K$ .**

## Verifying Lipschitz Continuity

The sequence of assignment or SKIP-statements that  $P$  executes on some input is called a **control flow path** of  $P$ .

Let  $x_j$  be the input and  $x_i$  be the output variable of our program.

**Lipschitz continuity** of a program is proven by establishing that

1.  $P$  is **continuous** in all states w.r.t. input  $x_j$  and output  $x_i$ .
2. Each **control flow path** of  $P$  is  **$K$ -Lipschitz** w.r.t. input  $x_j$  and output  $x_i$ .

What remains to show is step 2. In doing so, we derive a set of **Lipschitz matrices** for the given program.

## Conclusion for this Approach

- ▶ We asked for a theory about **robustness** of programs to **uncertainty**.
- ▶ **Lipschitz continuity** is an adequate answer to this question. It is a **strong property**.
- ▶ **Weak points:**
  - ▶ Is it satisfactory to live without **divisions**?
  - ▶ The degree of **automation** remains unclear.
  - ▶ No reasonable handling of **discrete input data**.
  - ▶ Not applicable to **reactive or concurrent systems**.

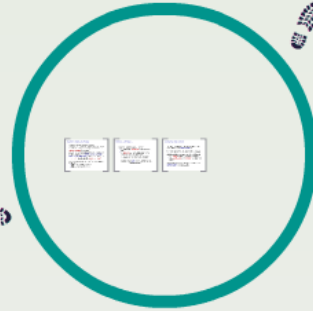


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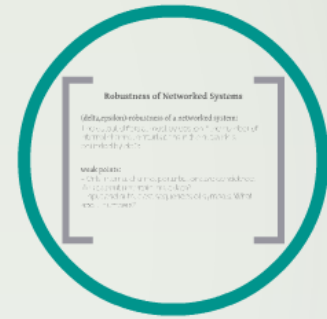
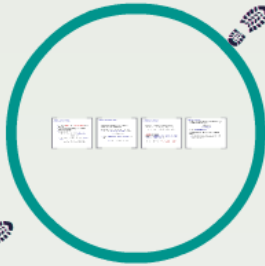


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Lipschitz  
continuity



continuity



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systems



(delta,epsilon)-  
robustness



safety-critical systems





### Symbolic Robustness Analysis

(delta,epsilon)-robustness w.r.t. variable x:

The outputs differ at most by epsilon if the input variable x is perturbed at most by delta (and all other variables remain unchanged).

weak points:

- How to choose the parameter delta?
- No direct adaption to closed loop systems.
- Floating-point numbers and non-linear arithmetic cannot be handled.

# (delta,epsilon)- robustness



# Symbolic Robustness Analysis

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The outputs differ at most by epsilon if the input variable  $x$  is perturbed at most by delta (and all other variables remain unchanged).

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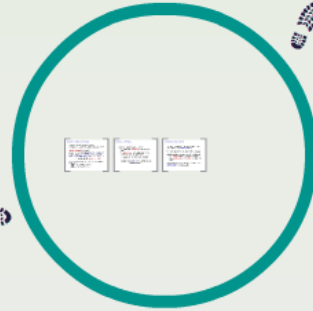


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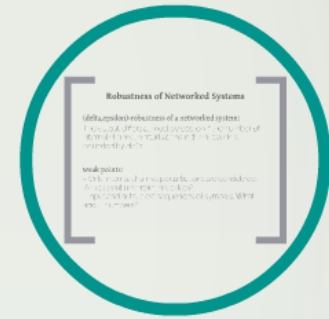
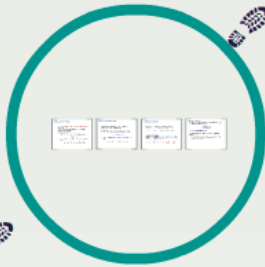


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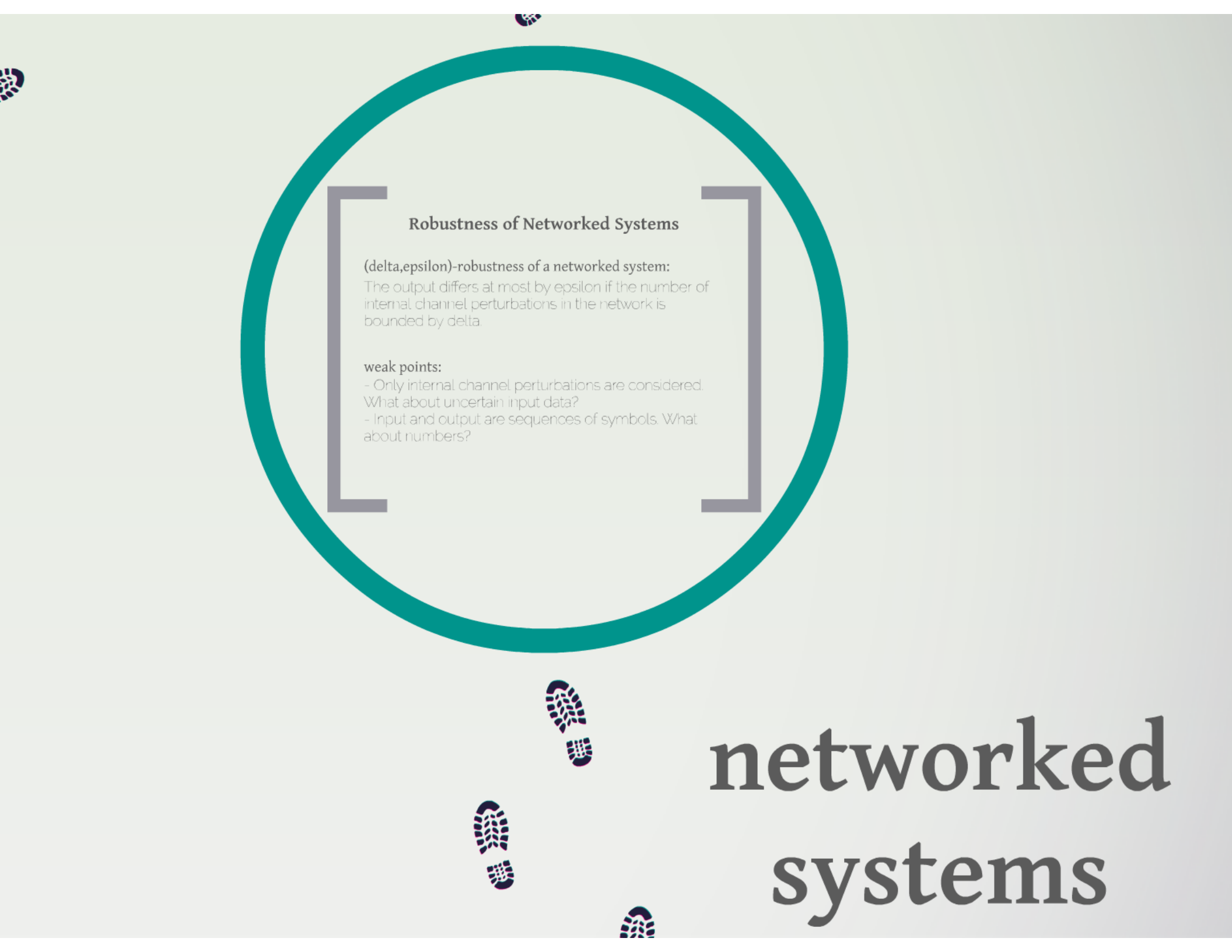
## Robustness of Networked Systems

### ( $\delta, \epsilon$ )-robustness of a networked system:

The output differs at most by  $\epsilon$  if the number of internal channel perturbations in the network is bounded by  $\delta$ .

### weak points:

- Only internal channel perturbations are considered. What about uncertain input data?
- Input and output are sequences of symbols. What about numbers?



networked  
systems

# Robustness of Networked Systems

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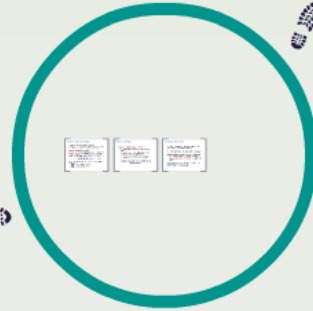


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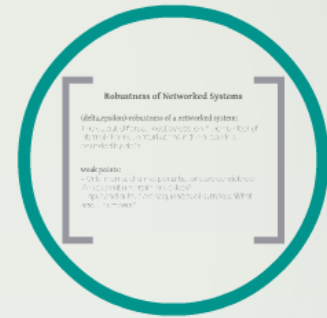
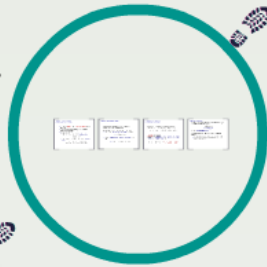


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continuity



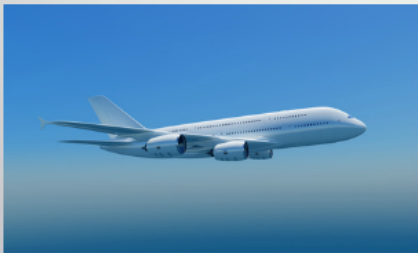
continuity



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(delta,epsilon)-  
robustness



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# caches

## Cache replacement policies

**Replacement policies ask:**  
Which memory block should we replace upon a cache miss?

**Sensitivity of a policy:**  
To what extent does the execution history influence the number of cache hits and cache misses?

**(r,c)-robustness of a policy:**  
Let  $A(s)$  be the number of cache misses on input sequence  $s$ . Where  $r = \delta(s) \ll \delta$  for a fixed  $\delta$ , it holds that  
 $A(s) \leq r \cdot A(s) + c$

## competitive analysis

Analyze the performance of an online algorithm compared to the optimal offline algorithm.

**(r,c)-competitiveness of a policy:**  
For all input sequences  $s$  it holds that:  
 $A(s) \leq r \cdot \text{OPT}(s) + c$   
where OPT is the optimal offline algorithm.



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## Replacement policies ask:

Which memory block should we replace upon a cache miss?

## Sensitivity of a policy:

To what extent does the execution history influence the number of cache hits and cache misses?

## (r,c)-robustness of a policy:

Let  $A(s)$  be the number of cache misses on input sequence  $s$ . Whenever  $\text{dist}(s_1, s_2) \leq \delta$  for a fixed  $\delta$ , it holds that

$$A(s_1) \leq r * A(s_2) + c$$

# competitive analysis

Analyze the performance of an online algorithm compared to the optimal offline algorithm.

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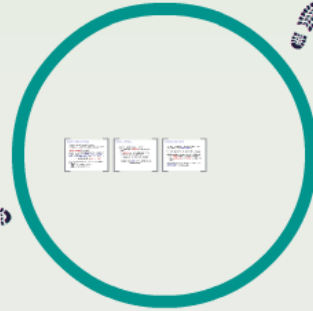


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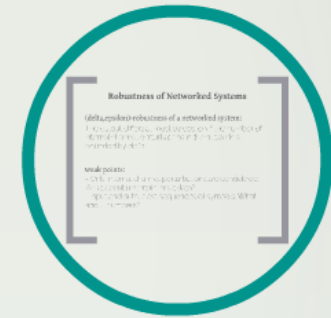
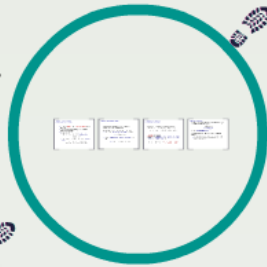


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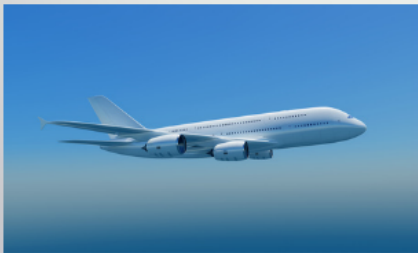
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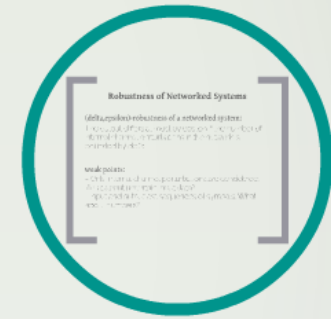
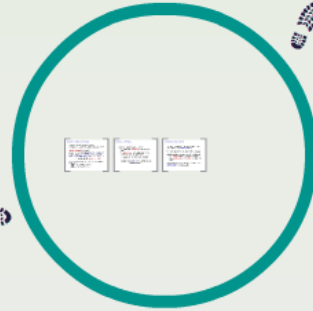


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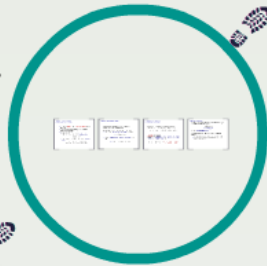
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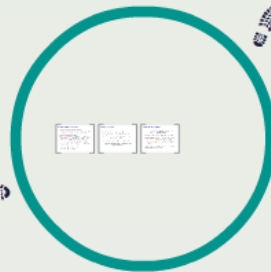


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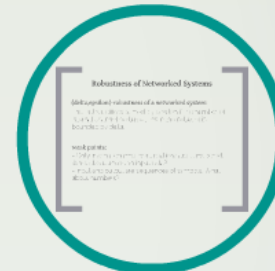
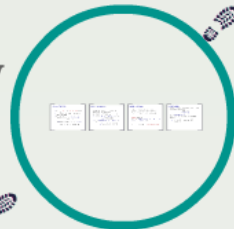


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Thanks for your attention!

## Literature

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## Image Sources

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