

# Inner Approximation of the Range of Vector-valued Functions, Episode 2

SWIM 2011

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# Aim

## Static Analysis

For a given program

**Input** Sets defining the possible values for the variables of this program;

**Output** Results guaranteed to be reached with input variables.

## Robust Control

For a given model

**Input** Initial state and control input (with uncertainty)

**Output** Intermediate and final states

## In both cases

Find range of vector-valued function  $f : \mathbb{R}^m \rightarrow \mathbb{R}^n$

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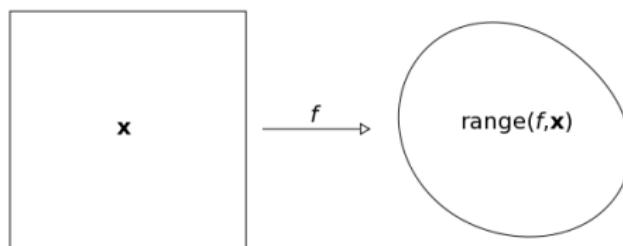
## In both cases

Find range of vector-valued function  $f : \mathbb{R}^m \rightarrow \mathbb{R}^n$

# The range

In general

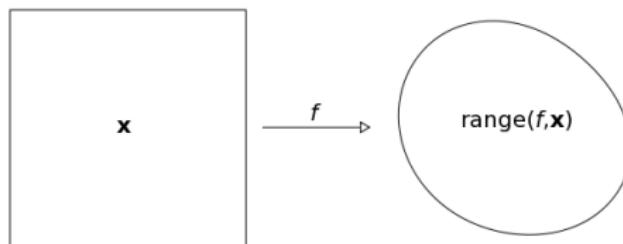
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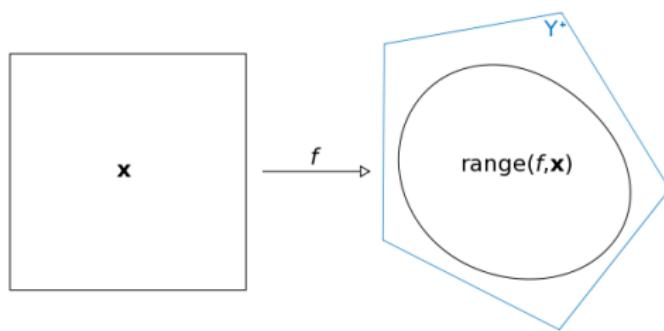
Issue

Computing the range of a function is intractable in general

# The range

In general

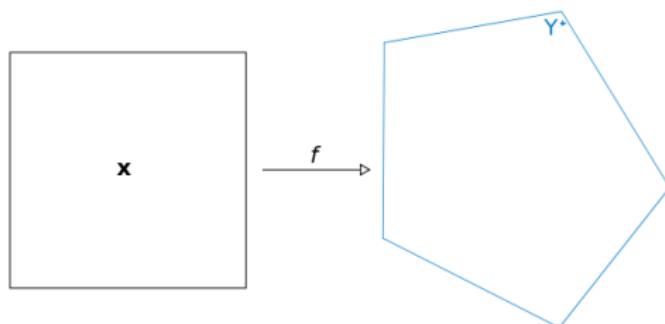
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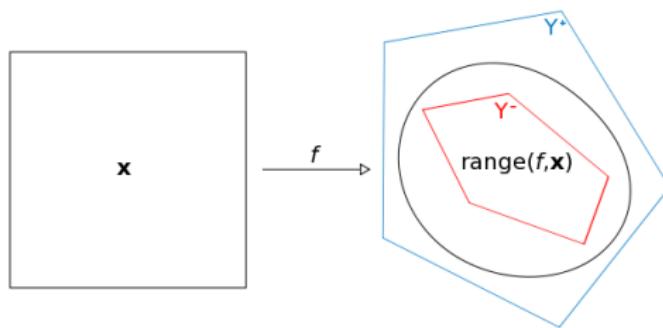
Issue

How pessimistic is the over estimation?

# The range

Here

Find  $\mathbb{Y}^- \subseteq \text{range}(f, \mathbf{x}) \subseteq \mathbb{Y}^+$



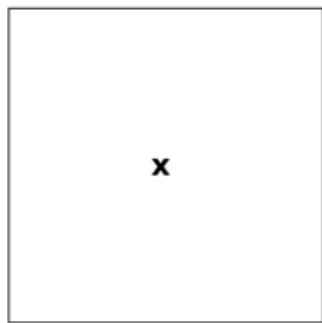
# State of the art

## Inner-approximation

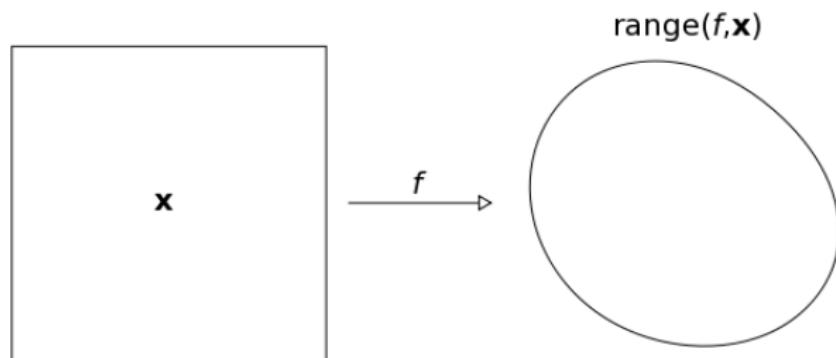
(Kieffer, Jaulin & Walter 1998)  $f$  must be *globally* invertible

(Goldsztein & Jaulin 2010)  $f$  must be *locally* invertible (dimension of domain and codomain must be equal)

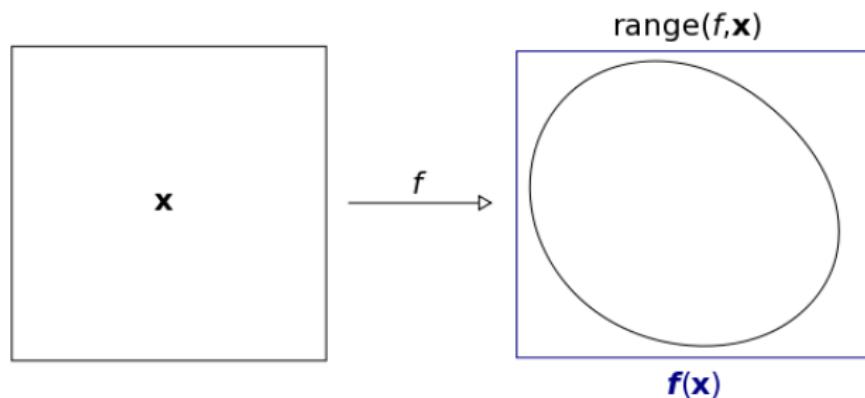
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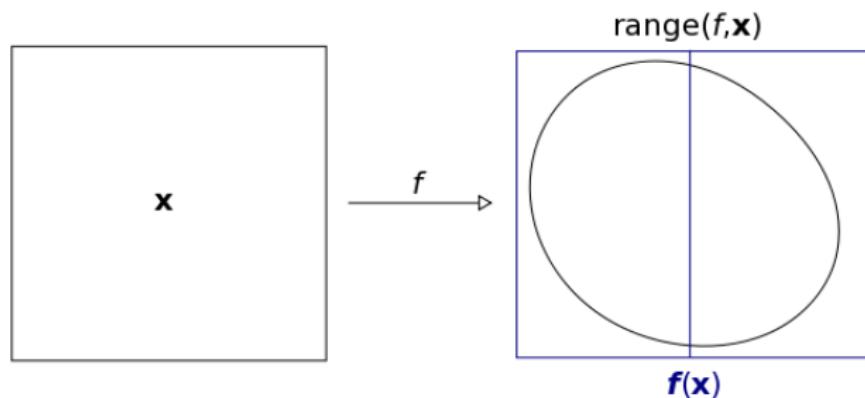
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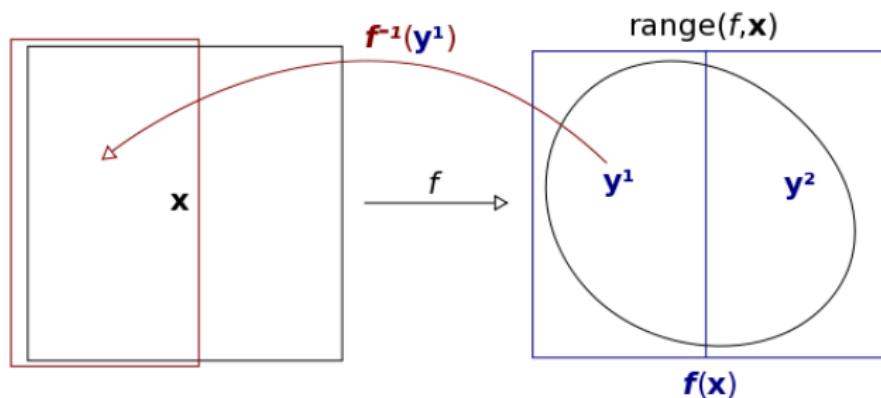
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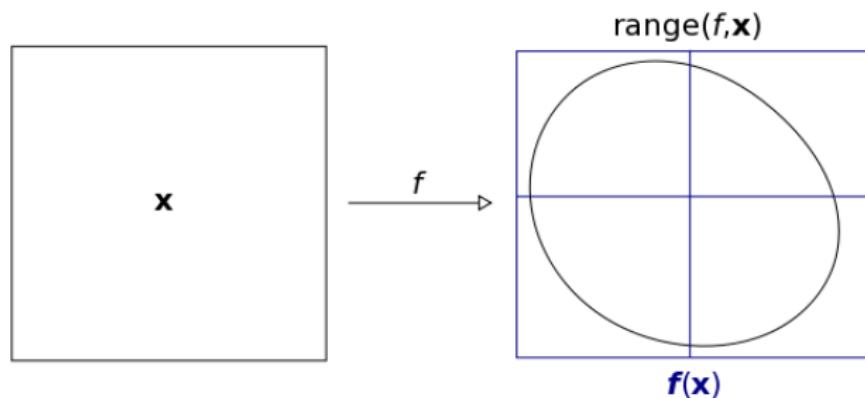
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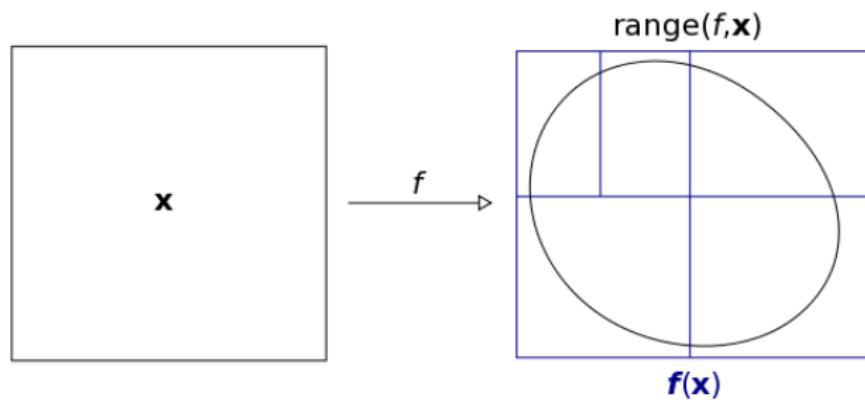
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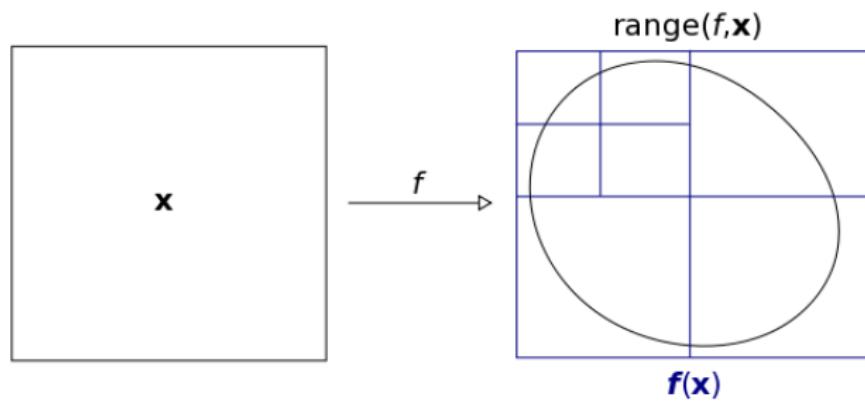
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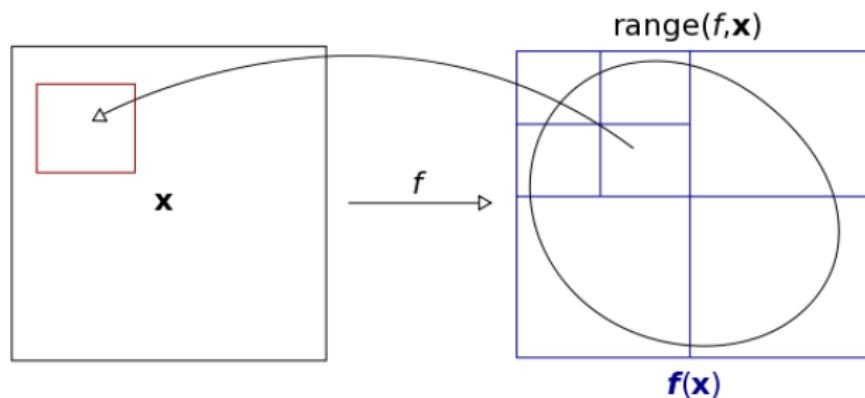
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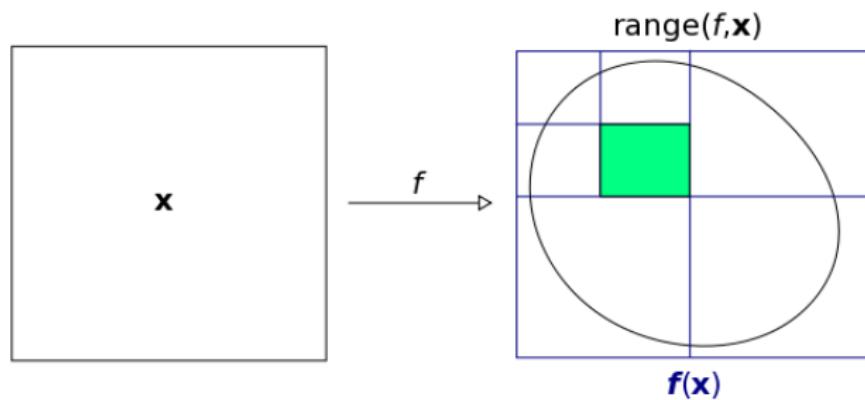
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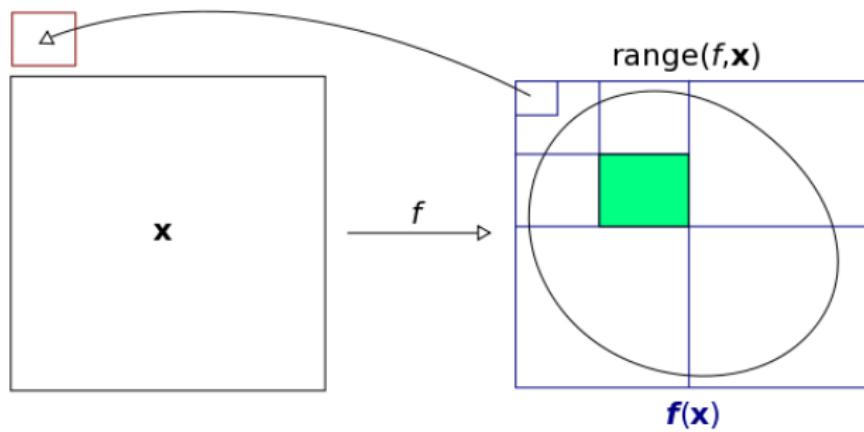
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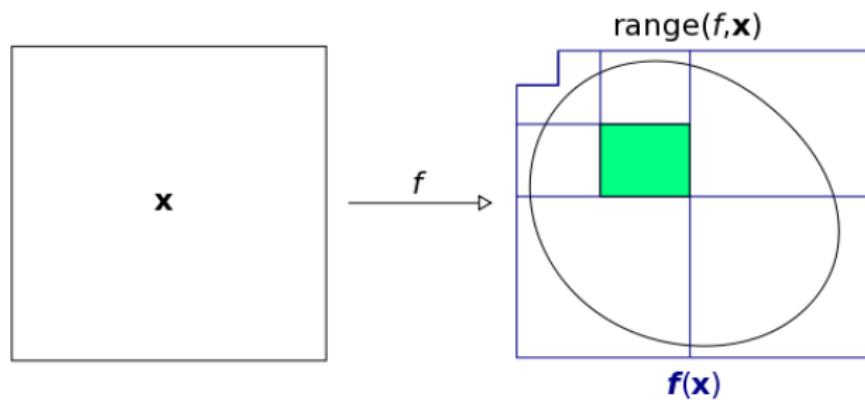
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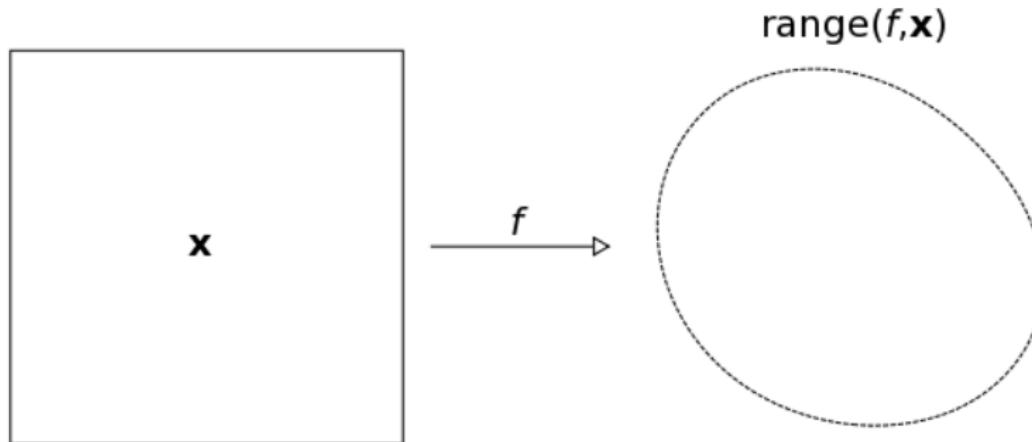
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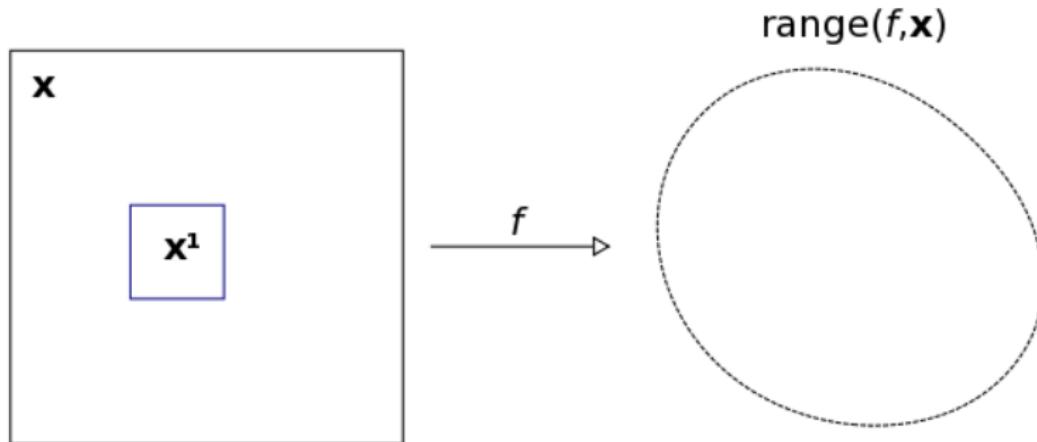
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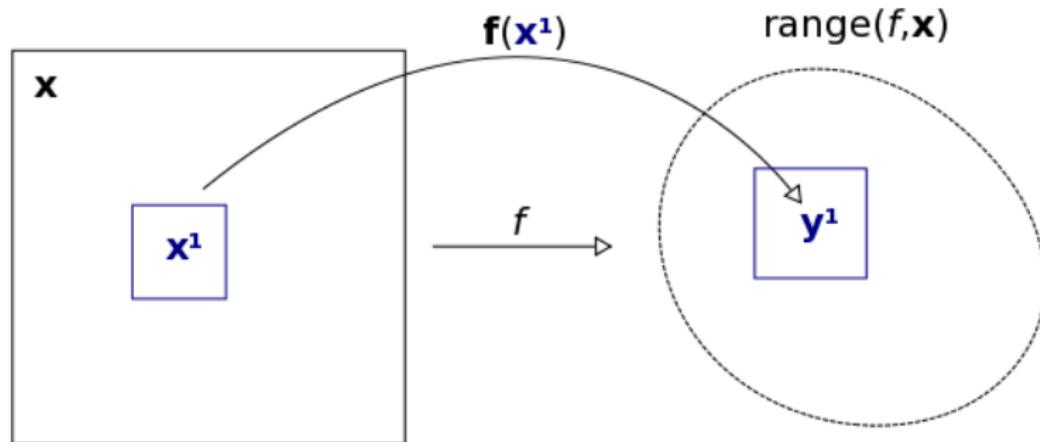
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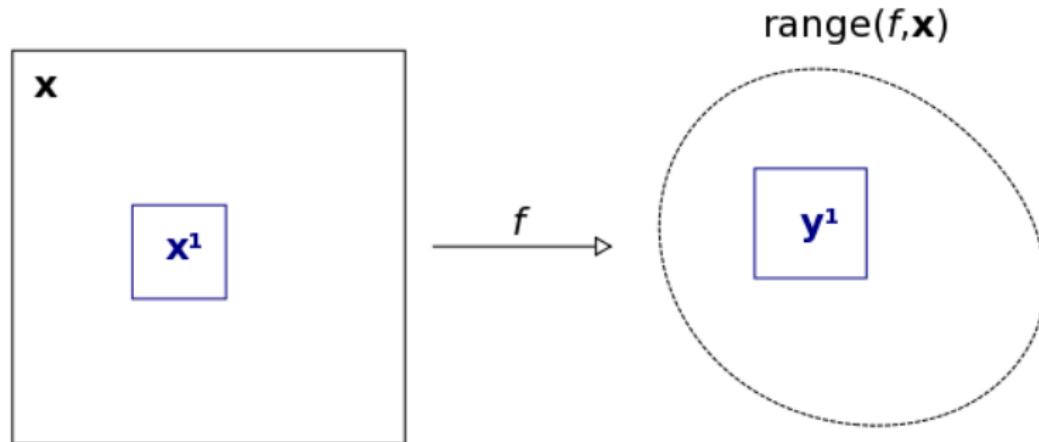
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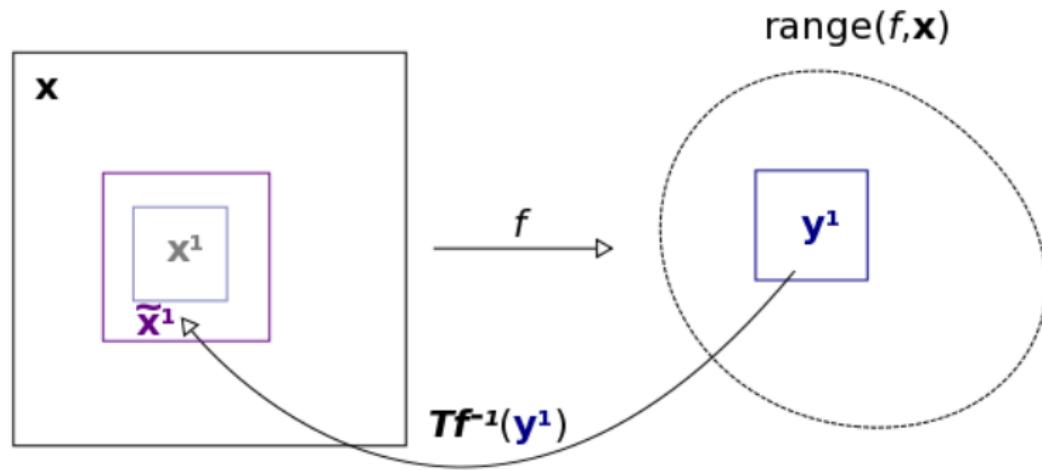
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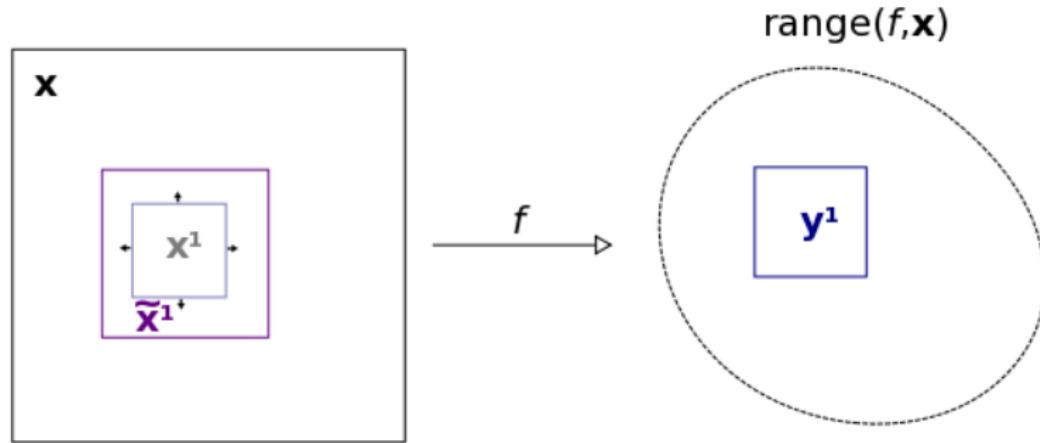
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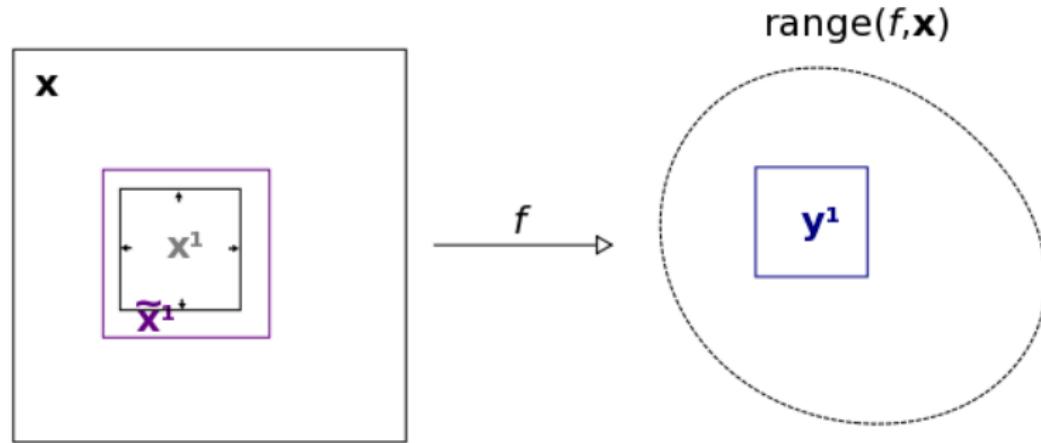
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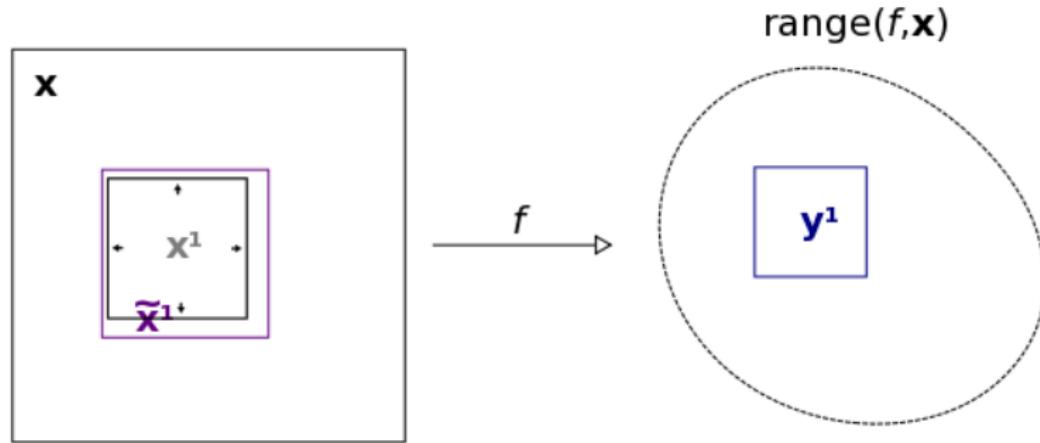
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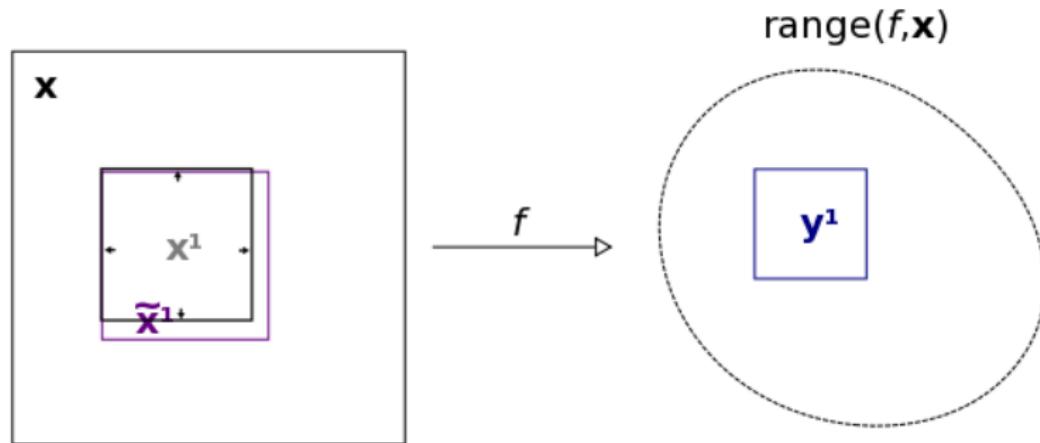
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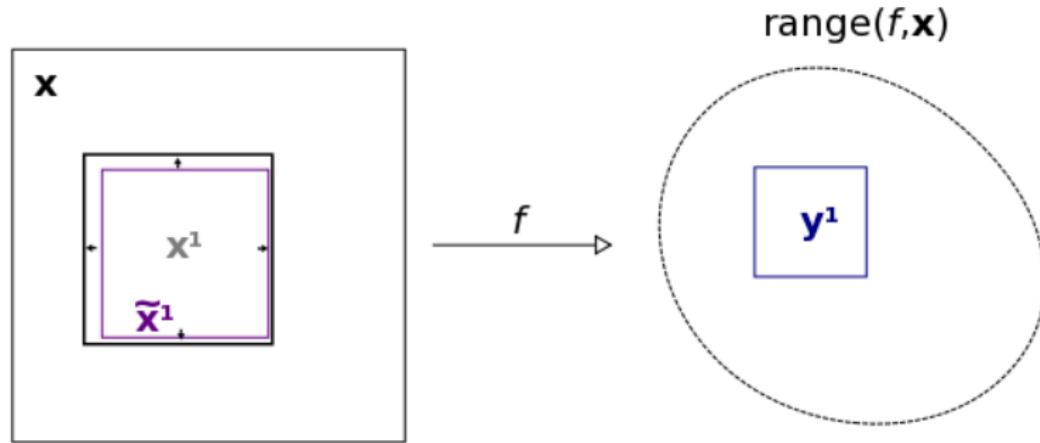
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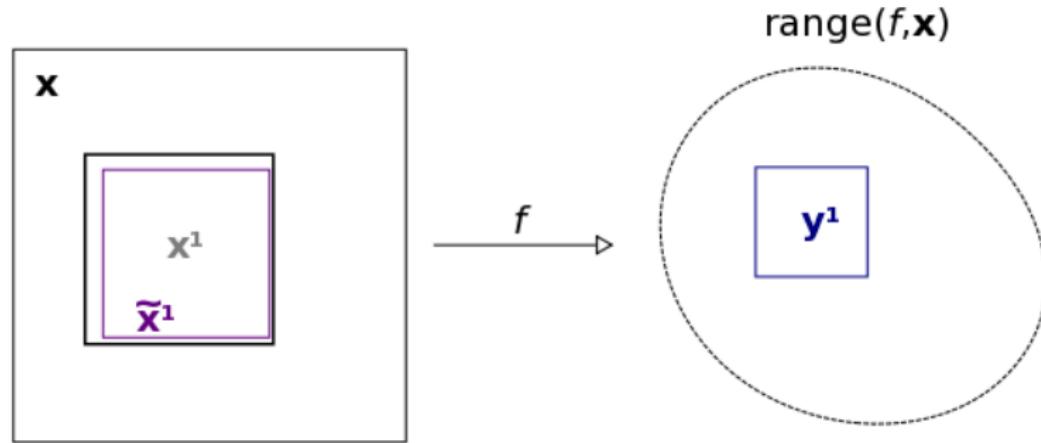
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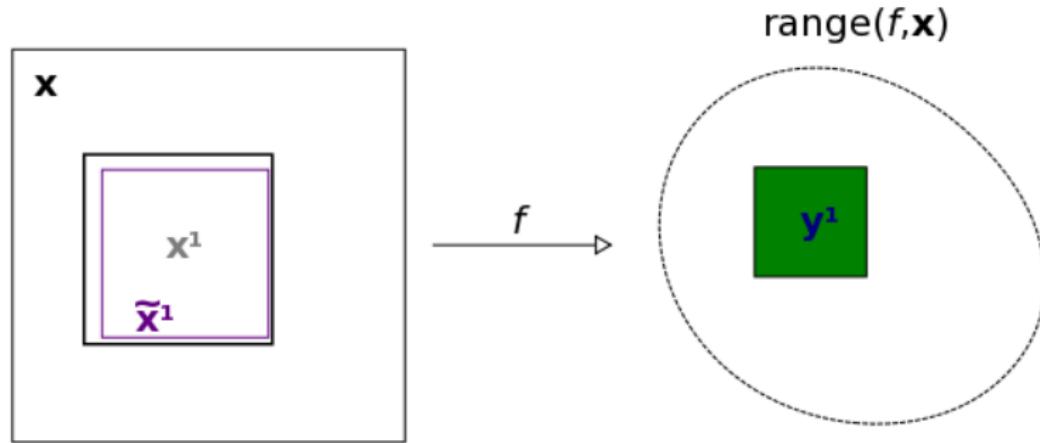
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# Outline

- 1 Method
- 2 Extension
- 3 Experiments
- 4 Conclusion

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# Idea

Starting Point : Order 1 Taylor Expansion

$$f(x) = f(\tilde{x}) + J(x - \tilde{x})$$

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Let consider  $y = f(\tilde{x}) + J(x - \tilde{x})$

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$$y - f(\tilde{x}) = J(x - \tilde{x})$$

$$y - f(\tilde{x}) = (\text{Diag } J + \text{OffDiag } J)(x - \tilde{x})$$

## Split

$$(\text{Diag } J)_{ij} = \begin{cases} (J)_{ij} & \text{if } i = j \\ 0 & \text{otherwise} \end{cases} \quad (\text{OffDiag } J)_{ij} = \begin{cases} (J)_{ij} & \text{if } i \neq j \\ 0 & \text{otherwise} \end{cases}$$

## Limitation

This split implies that dimensions of domain and codomain have to be equal.

# Idea

Let consider  $y = f(\tilde{x}) + J(x - \tilde{x})$

$$y - f(\tilde{x}) = J(x - \tilde{x})$$

$$y - f(\tilde{x}) = (\text{Diag } J + \text{OffDiag } J)(x - \tilde{x})$$

$$x = \tilde{x} + \text{Diag}^{-1} J(y - f(\tilde{x}) - \text{OffDiag } J(x - \tilde{x}))$$

## Constraint

$J_{ii}$  must not be equal to 0

# Idea

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## Criteria

For a box  $\mathbf{y} \in \mathbb{R}^n$ :

$$\begin{aligned} \tilde{x} + \text{Diag}^{-1} \mathbf{J}(\mathbf{y} - f(\tilde{x}) - \text{OffDiag } \mathbf{J}(x - \tilde{x})) &\subseteq \text{int } \mathbf{x} \\ &\Downarrow \\ \mathbf{y} &\subseteq \text{range}(f, \mathbf{x}) \end{aligned}$$

# Extension: $m \neq n$

## Criteria

For a box  $\mathbf{y} \in \mathbb{R}^n$ :

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## Limitation

This criteria is only valid when dimensions of domain  $m$  and codomain  $n$  of  $f$  are equal

## Idea

Transformation of  $f : \mathbb{R}^m \rightarrow \mathbb{R}^n$  to a function  $g : \mathbb{R}^\beta \rightarrow \mathbb{R}^\beta$ , with  $\beta = \max(m, n)$  and with similar range properties.

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2 Extension

3 Experiments

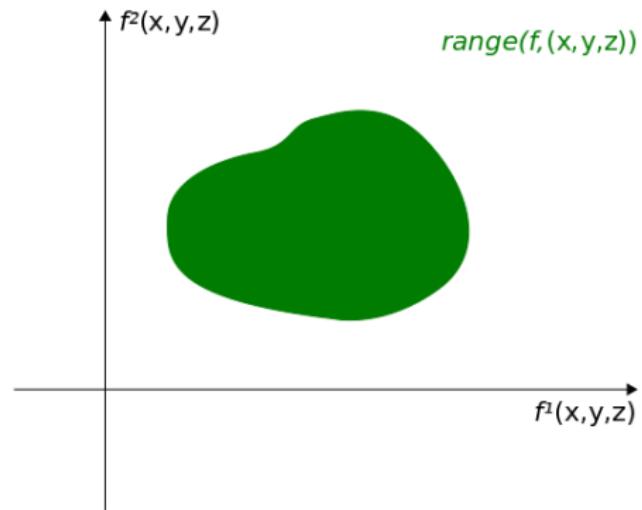
4 Conclusion

# Examples: $m > n$

## Example 1

$$f : \mathbb{R}^3 \rightarrow \mathbb{R}^2$$

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} \mapsto \begin{pmatrix} f^1(x, y, z) \\ f^2(x, y, z) \end{pmatrix}$$

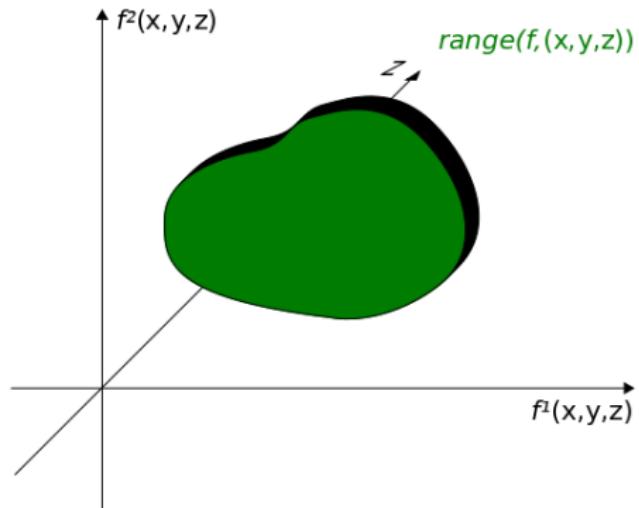


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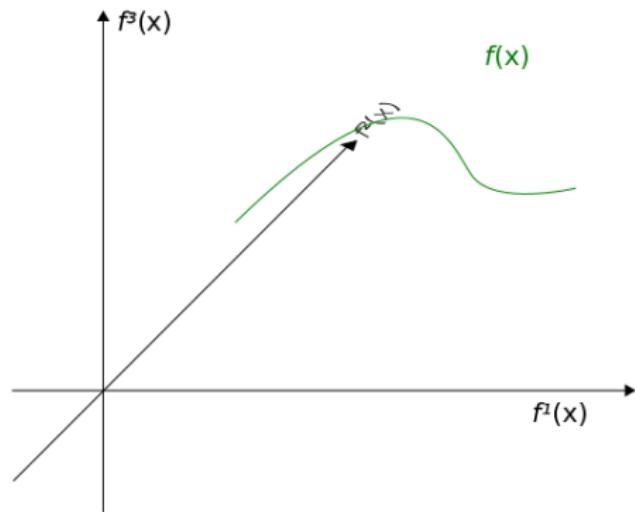


# Examples: $m < n$

## Example 1

$$f : \mathbb{R} \rightarrow \mathbb{R}^3$$

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## Construction

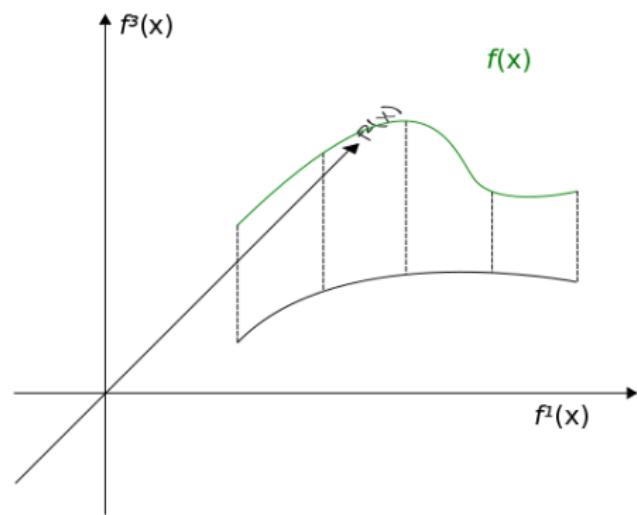
$f$  is of constant rank 2 in a given box  $x$

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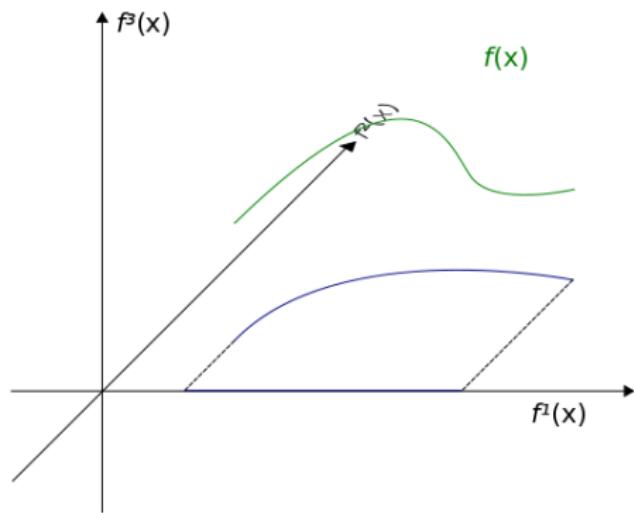
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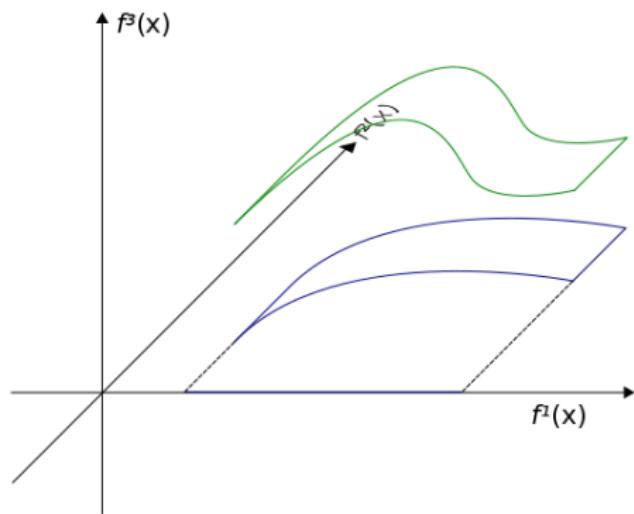
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$$g : \mathbb{R}^2 \rightarrow \mathbb{R}^3$$

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## Construction

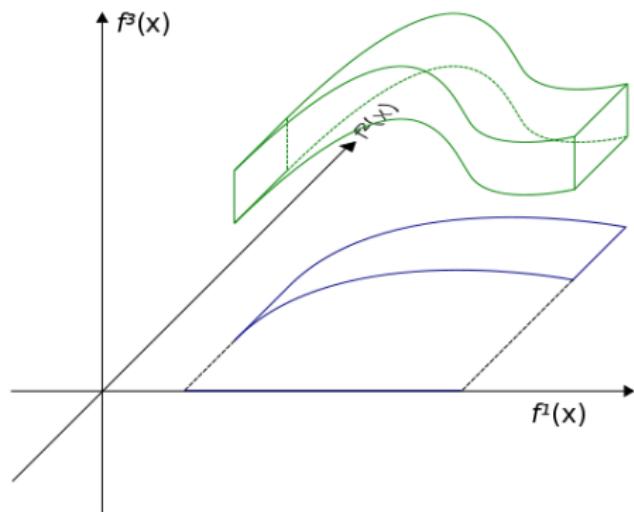
We add  $y$  to the domain of  $f$

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## Construction

We add  $z$  to the domain of  $f$

# Extension

## Constant Rank Theorem

This method is supported by functions verifying the *constant rank theorem*.

For a function  $f : \mathbb{R}^m \rightarrow \mathbb{R}^n$  of constant rank  $k$ :

$k = n$  we obtain the inner-approximation of the range of  $f$

Otherwise we obtain the inner-approximation of the projection of  $k$  components of  $f$

## Function $g$

For  $\alpha = \min(m, n)$  and  $\beta = \max(m, n)$ :

Let  $g : \mathbb{R}^k \times \mathbb{R}^{\alpha-k} \times \mathbb{R}^{\beta-\alpha} \rightarrow \mathbb{R}^k \times \mathbb{R}^{\alpha-k} \times \mathbb{R}^{\beta-\alpha}$

$$(u, v_1, v_2) \mapsto \begin{pmatrix} f_{[1..k]}(u, v_1, v_2) \\ f_{[k+1..\alpha]}(u, v_1, v_2) + v_1 \\ f_{[\alpha+1..\beta]}(u, v_1, v_2) + v_2 \end{pmatrix}$$

with  $f_{[\alpha+1..\beta]}(u, v_1, v_2) = 0_{\beta-\alpha}$  if  $\alpha = n$

Jacobian of  $g$  :  $J^g$

$$J^g = \begin{pmatrix} J^{f,[1..k],u} & J^{f,[1..k],v_1} & J^{f,[1..k],v_2} \\ J^{f,[k+1..\alpha],u} & | & 0 \\ J^{f,[\alpha+1..\beta],u} & 0 & | \end{pmatrix}$$

with  $J^{f,[a..b],t} = \left(\frac{\partial f_i}{\partial t}\right)_{i \in [a..b]}$

# Extension: criteria

## Criteria (Goldsztein & Jaulin 2010)

For a box  $\mathbf{y} \in \mathbb{R}^n$ :

$$\begin{aligned} \tilde{\mathbf{x}} + \text{Diag}^{-1} \mathbf{J}(\mathbf{y} - f(\tilde{\mathbf{x}}) - \text{OffDiag } \mathbf{J}(\mathbf{x} - \tilde{\mathbf{x}})) &\subseteq \text{int } \mathbf{x} \\ \Downarrow \\ \mathbf{y} &\subseteq \text{range}(f, \mathbf{x}) \end{aligned}$$

## Criteria: (Extension)

For a box  $\mathbf{y} = (\mathbf{y}_1, \mathbf{y}_2, \mathbf{y}_3) \in \mathbb{R}^k \times \mathbb{R}^{\alpha-k} \times \mathbb{R}^{\beta-\alpha}$ :

$$\begin{aligned} \tilde{\mathbf{u}} + \text{Diag}_{J_{f,[1..k]}, u}^{-1} (\mathbf{y}_1 - f_{[1..k]}(\tilde{\mathbf{u}}, \tilde{\mathbf{v}}_1, \tilde{\mathbf{v}}_2) - \text{OffDiag}_{J_{f,[1..k]}, u} (\mathbf{u} - \tilde{\mathbf{u}}) - \\ J_{f,[1..k], v_1} (\mathbf{v}_1 - \tilde{\mathbf{v}}_1) - J_{f,[1..k], v_2} (\mathbf{v}_2 - \tilde{\mathbf{v}}_2)) &\subseteq \text{int } \mathbf{u} \\ \Downarrow \\ \mathbf{y}_1 &\subseteq \text{Im}(f_1, \dots, f_k) \end{aligned}$$

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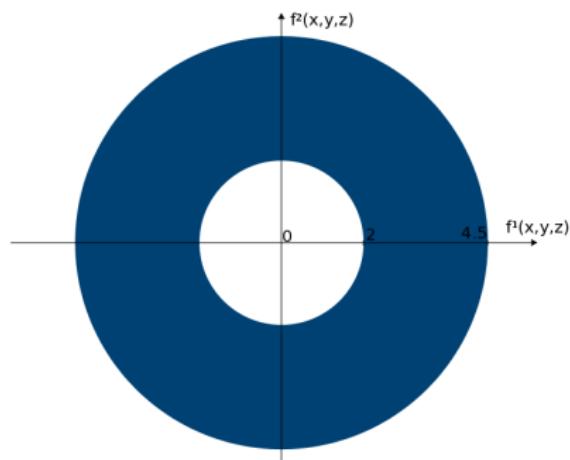
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## Implementation

- Computation in C++ using Profil/bias library for interval computations;
- Slight changes in algorithms provided in (Goldsztein & Jaulin 2010)

## Experiments

- Experiments have been runned on a MacBook Pro 13" Core i5 2,3 GHz

$m > n$ 

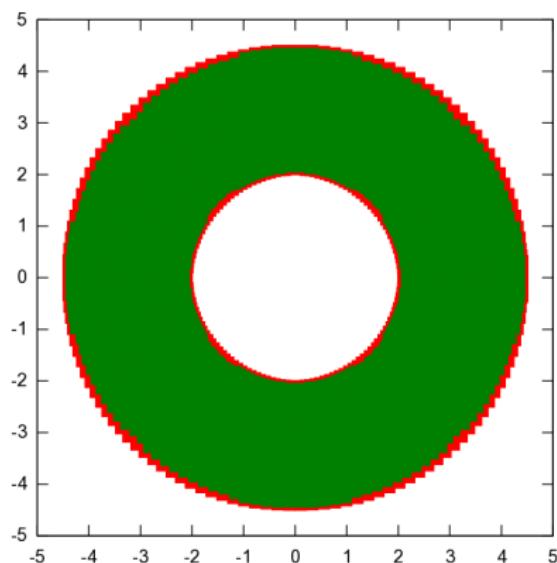
## Function definition

$$f : \mathbb{R}^3 \rightarrow \mathbb{R}^2$$

$$(x, y, z) \mapsto \begin{cases} (x + r \cos(z)) \cos(y) \\ (x + r \sin(z)) \sin(y) \end{cases}$$

$$\mathbf{x} = [3, 4] ; \quad \mathbf{y} = [0, 2\pi] ; \quad \mathbf{z} = [0, 2\pi]$$

$m > n$

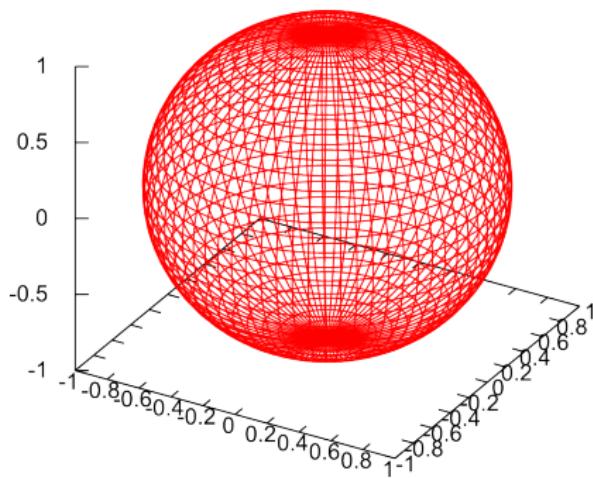


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$$(x, y, z) \mapsto \begin{cases} (x + r \cos(z)) \cos(y) \\ (x + r \sin(z)) \sin(y) \end{cases}$$

$$\mathbf{x} = [3, 4]; \quad \mathbf{y} = [0, 2\pi]; \quad \mathbf{z} = [0, 2\pi]$$

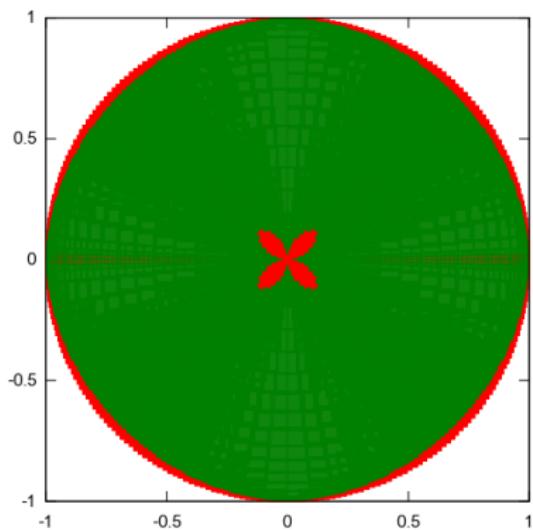
$m < n$ 

## Function definition

$$f : \mathbb{R}^2 \rightarrow \mathbb{R}^3$$

$$(u, v) \mapsto \begin{cases} f_1(u, v) = \cos(u) \cos(v) \\ f_2(u, v) = \sin(u) \cos(v) \\ f_3(u, v) = \sin(v) \end{cases}$$

$$\mathbf{u} = [0, 2\pi]; \quad \mathbf{v} = [-\pi, \pi]$$

$m < n$ 

### Function definition

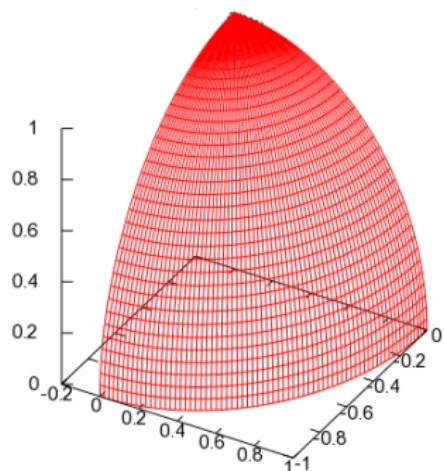
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$$\varepsilon = 0.04$$

$$\mathbf{u} = \left[ \frac{3\pi}{2} + \varepsilon, 2\pi - \varepsilon \right]; \quad \mathbf{v} = \left[ \varepsilon, \frac{\pi}{2} - \varepsilon \right], \quad \varepsilon > 0$$

$m < n$



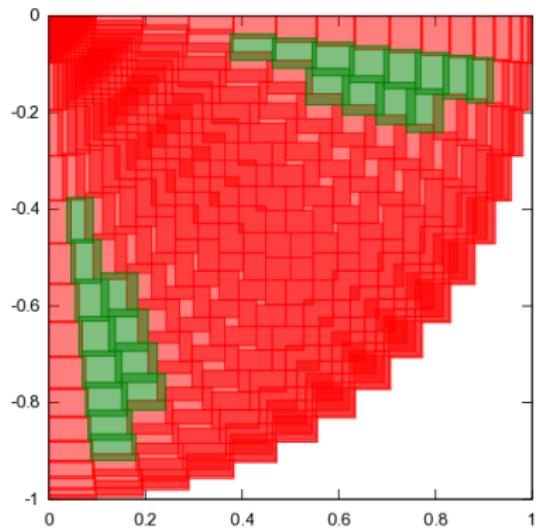
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$m < n$



$$\varepsilon = 0.1$$

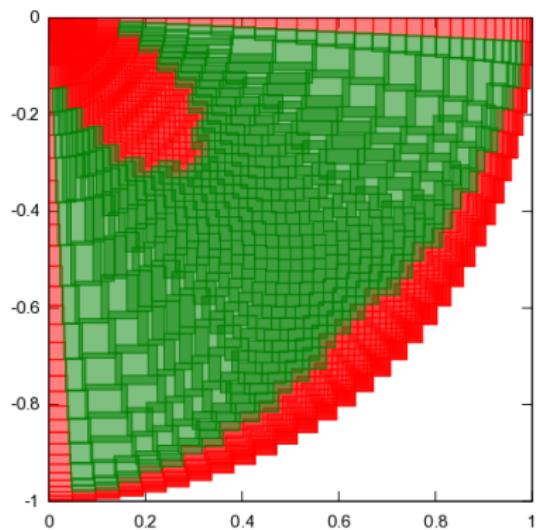
$$\mathbf{u} = \left[ \frac{3\pi}{2} + \varepsilon, 2\pi - \varepsilon \right]; \quad \mathbf{v} = \left[ \varepsilon, \frac{\pi}{2} - \varepsilon \right], \quad \varepsilon > 0$$

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### Function definition

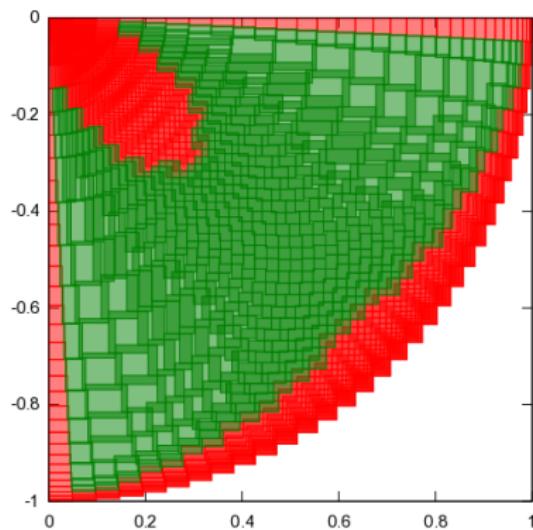
$$f : \mathbb{R}^2 \rightarrow \mathbb{R}^3$$

$$(u, v) \mapsto \begin{cases} f_1(u, v) = \cos(u) \cos(v) \\ f_2(u, v) = \sin(u) \cos(v) \\ f_3(u, v) = \sin(v) \end{cases}$$

$$\varepsilon = 0.08$$

$$\mathbf{u} = \left[ \frac{3\pi}{2} + \varepsilon, 2\pi - \varepsilon \right]; \quad \mathbf{v} = \left[ \varepsilon, \frac{\pi}{2} - \varepsilon \right], \quad \varepsilon > 0$$

$m < n$



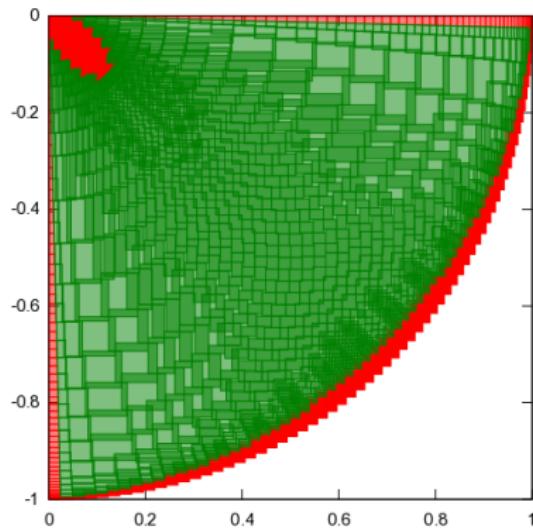
$$\varepsilon = 0.06$$

$$\mathbf{u} = \left[ \frac{3\pi}{2} + \varepsilon, 2\pi - \varepsilon \right]; \quad \mathbf{v} = \left[ \varepsilon, \frac{\pi}{2} - \varepsilon \right], \quad \varepsilon > 0$$

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$m < n$ 

$$\varepsilon = 0.04$$

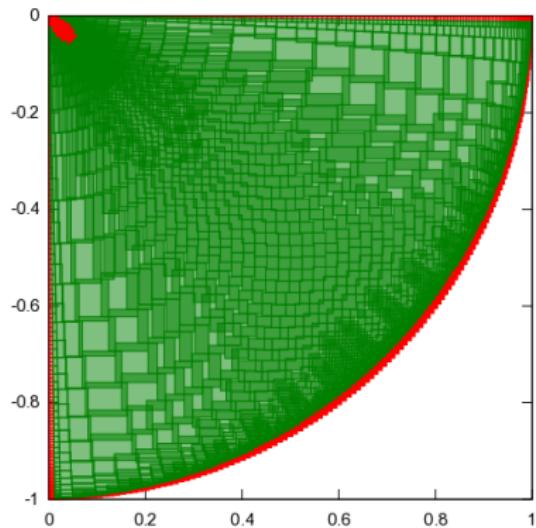
$$\mathbf{u} = \left[ \frac{3\pi}{2} + \varepsilon, 2\pi - \varepsilon \right]; \quad \mathbf{v} = \left[ \varepsilon, \frac{\pi}{2} - \varepsilon \right], \quad \varepsilon > 0$$

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### Function definition

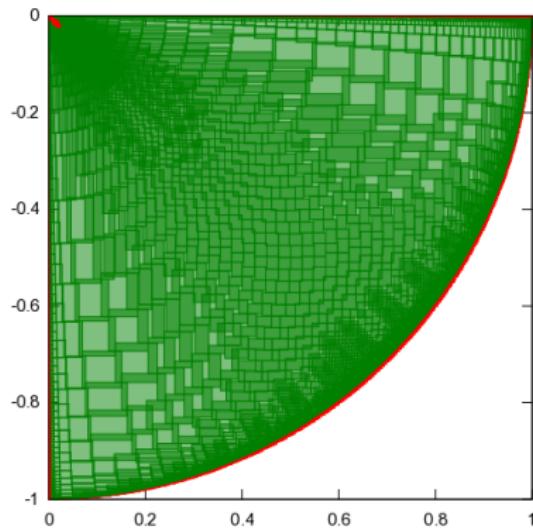
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$$\varepsilon = 0.02$$

$$\mathbf{u} = \left[ \frac{3\pi}{2} + \varepsilon, 2\pi - \varepsilon \right]; \quad \mathbf{v} = \left[ \varepsilon, \frac{\pi}{2} - \varepsilon \right], \quad \varepsilon > 0$$

$m < n$



### Function definition

$$f : \mathbb{R}^2 \rightarrow \mathbb{R}^3$$

$$(u, v) \mapsto \begin{cases} f_1(u, v) = \cos(u) \cos(v) \\ f_2(u, v) = \sin(u) \cos(v) \\ f_3(u, v) = \sin(v) \end{cases}$$

$$\varepsilon = 0.01$$

$$\mathbf{u} = \left[ \frac{3\pi}{2} + \varepsilon, 2\pi - \varepsilon \right]; \quad \mathbf{v} = \left[ \varepsilon, \frac{\pi}{2} - \varepsilon \right], \quad \varepsilon > 0$$

# Outline

- ① Method
- ② Extension
- ③ Experiments
- ④ Conclusion

# Conclusion

## Extension

This extension is able to

- Compute inner-approximation when  $k = n$ ;
- Compute  $k$  components of  $f$  otherwise.

## Future Works

- Deal with redundant computations when  $m > n$ ;
- Find submatrix of maximal constant rank (hints);
- Extension to other abstract domains (e.g. affine forms).

# Conclusion

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# References

-  Goldsztejn, A. & Jaulin, L. (2010), 'Inner approximation of the range of vector-valued functions', *Reliable Computing* **1**, 23.
-  Kieffer, M., Jaulin, L. & Walter, E. (1998), Guaranteed recursive nonlinear state estimation using interval analysis, in 'Decision and Control, 1998. Proceedings of the 37th IEEE Conference on', Vol. 4, IEEE, pp. 3966–3971.