

# Certified curve tracking using interval arithmetic

Joint work with Michael (Burr + Byrd)

Kisun Lee (Clemson University) - [kisunl@clemson.edu](mailto:kisunl@clemson.edu)  
SIAM Conference on Applied Algebraic Geometry (AG25)  
Real Solutions in Numerical Algebraic Geometry

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## 2025 SIAM Conference on Applied Algebraic Geometry (AG25)

Part of [MS51 Real Solutions in Numerical Algebraic Geometry - Part II of III](#)

### Implementing Real Polyhedral Homotopy

**Abstract.** In this talk, we revisit the Julia implementation of real polyhedral homotopy. The implementation is based on the real polyhedral homotopy algorithm established by Ergür and de Wolff (2023). This algorithm constructs a homotopy with solution paths that do not cross the real discriminant locus obtained by Viro's patchworking to track the optimal number of real solutions. Based on joint work with Lindberg and Rodriguez (2024), we pose open questions and future outlooks for finding real solutions to polynomial systems.

### Authors

- *Kisun Lee, Clemson University, U.S., kisunl@clemson.edu*
- *Julia Lindberg, University of Texas at Austin, U.S., julia.lindberg@math.utexas.edu*
- *Jose Rodriguez, University of Wisconsin-Madison, U.S., jose@math.wisc.edu*

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Part of [MS64 Real Solutions in Numerical Algebraic Geometry - Part III of III](#)

### Certified Curve Tracking

**Abstract.** In this talk, we consider the problem of computing a certified approximation to a regular curve in  $\mathbb{R}^n$ . Specifically, given a regular curve and an approximate point on it, our goal is to construct a tubular neighborhood containing the curve. Our approach uses certified homotopy tracking methods and interval arithmetic, extending the work of Duff-Lee and Guillemot-Lairez. As a key application, we provide a certified algorithm for approximating the image of a generic projection of a curve in  $\mathbb{R}^n$  to the plane.

### Authors

- Michael Byrd, Clemson University, U.S., [mbyrd6@clemson.edu](mailto:mbyrd6@clemson.edu)

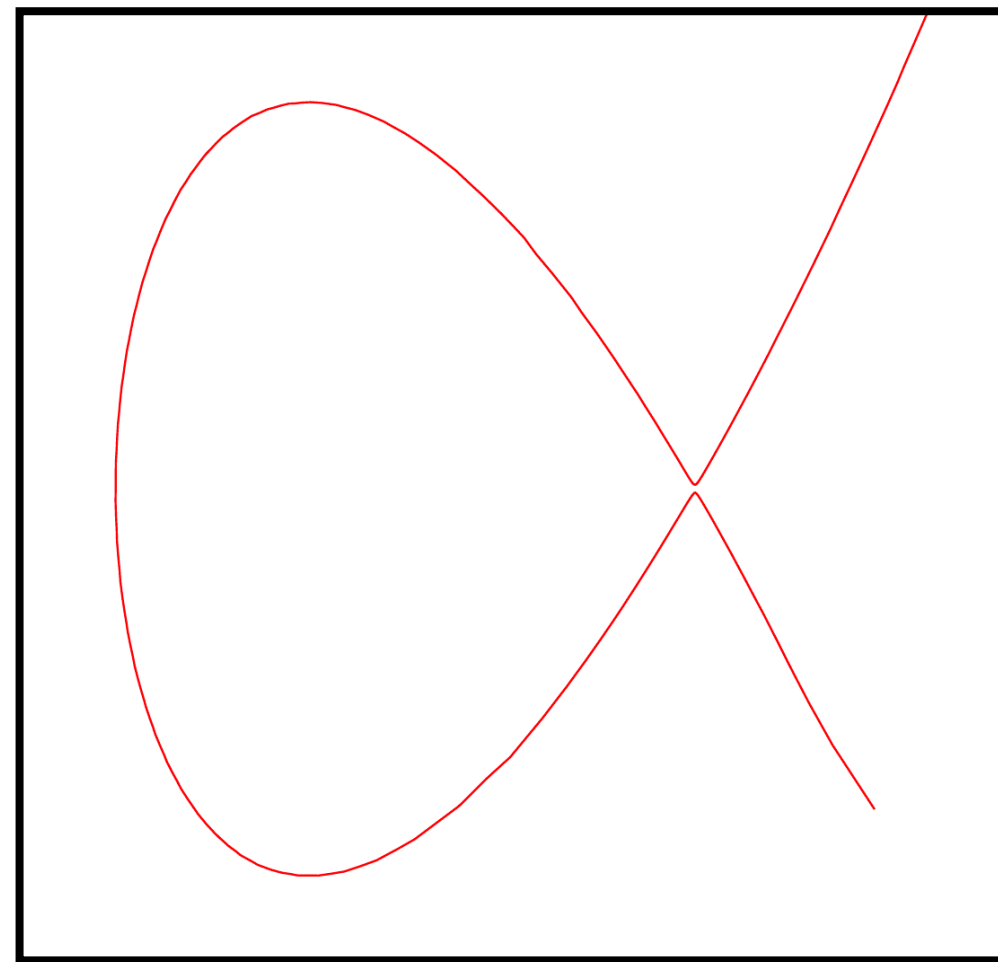
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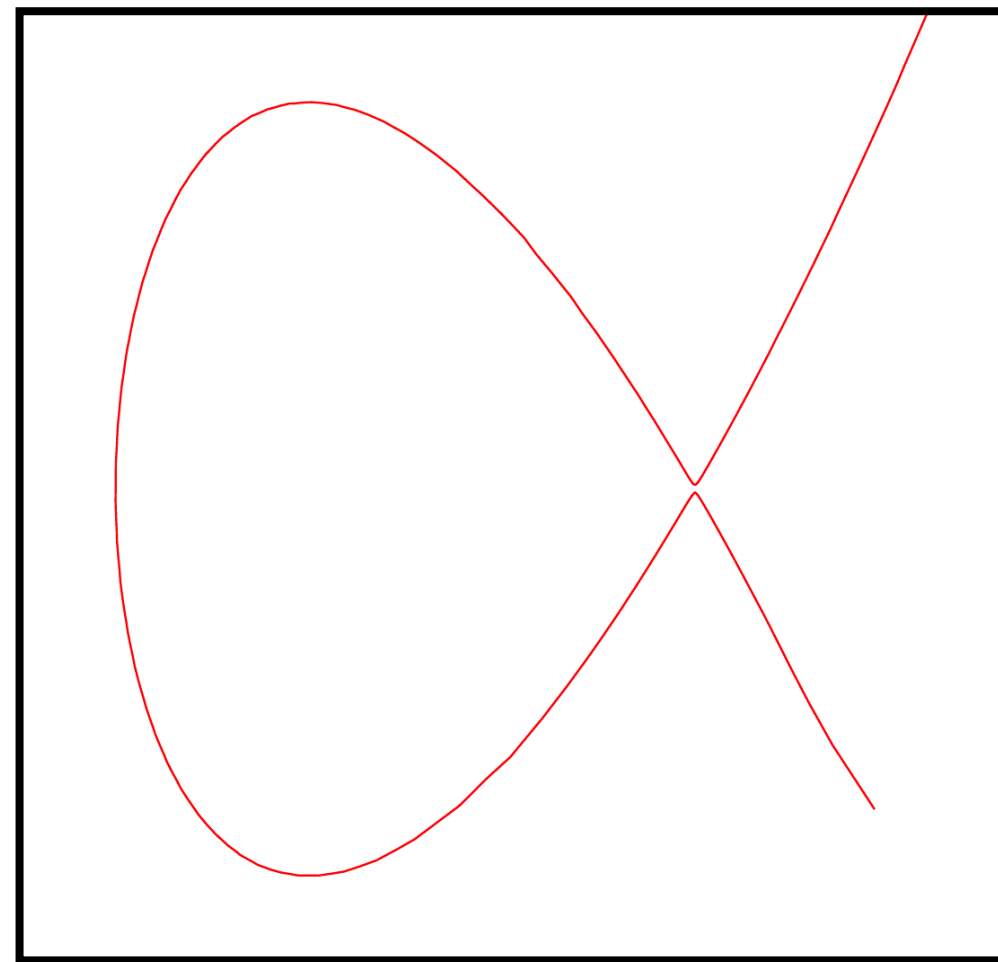
# Certified curve tracking

**Given:** a regular curve  $C \in \mathbb{R}[X_1, \dots, X_n]^{n-1}$ ,  
an (approximate) point  $x$  on  $C$ , and a compact region  $R$   
that contains  $x$ .



$$C = \{x^3 - 2.9995x - y^2 + 2\}$$

# Certified curve tracking

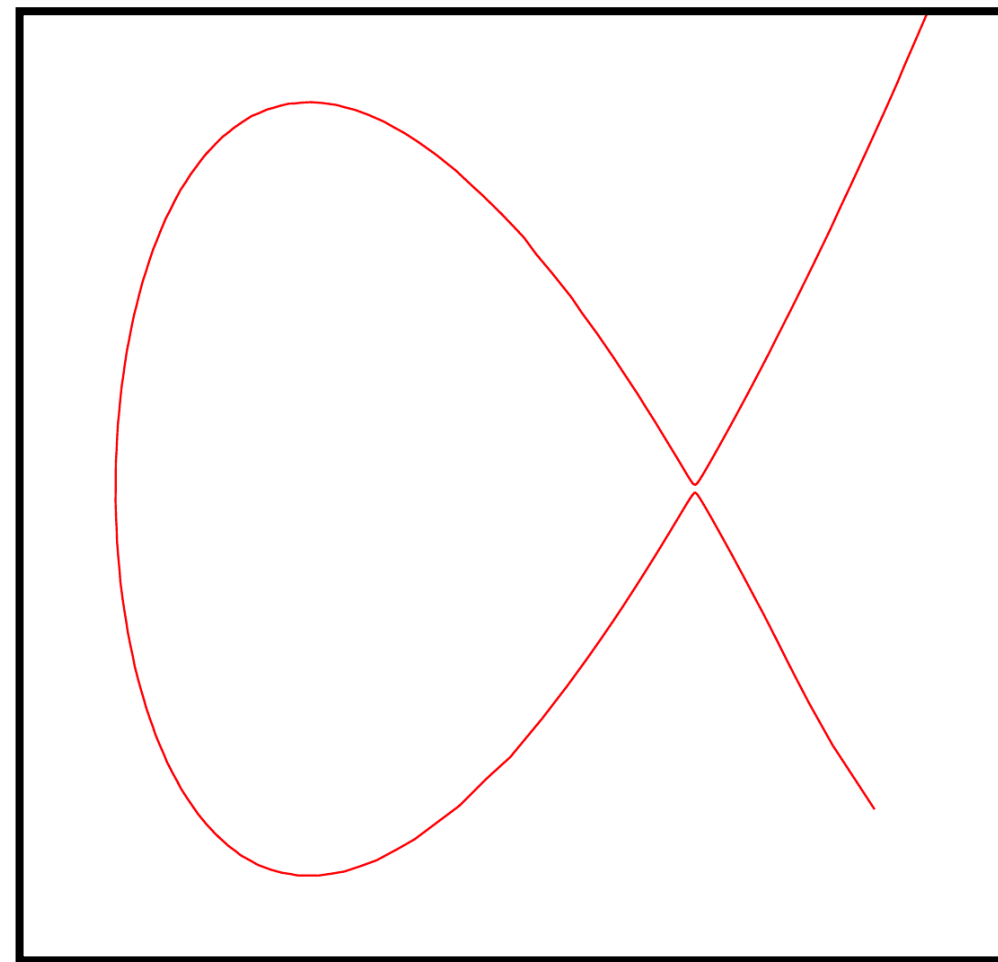


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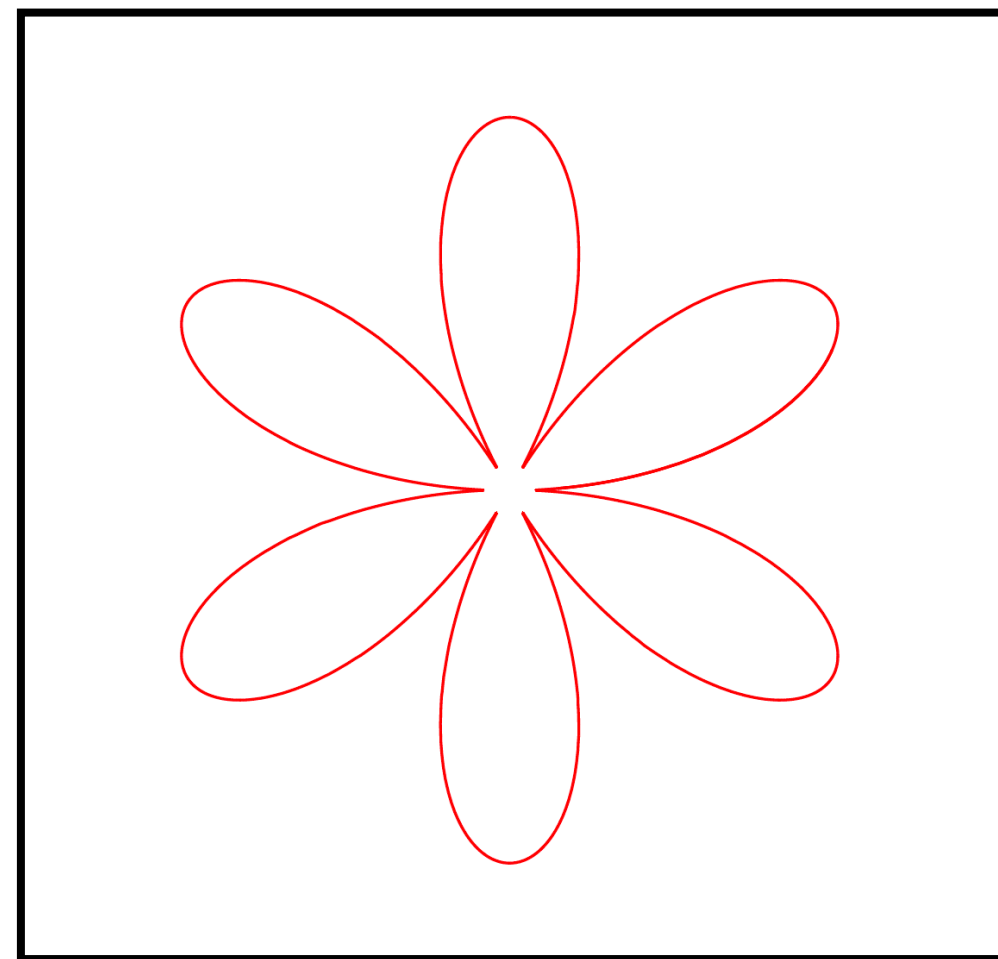
Starting at  $x$ , **track** along the curve  $C$  in a **certified**  
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Possible stopping criteria:

- Track until the computed curve hits the boundary of  $R$ ,



# Certified curve tracking



$$C = \{x^8 - (1 - \epsilon)x^6 + 4x^6y^2 - (3 + 15\epsilon)x^4y^2 + 6x^4y^4 \\ - (3 - 15\epsilon)x^2y^4 + 4x^2y^6 - (1 + \epsilon)y^6 + y^8\}$$

where  $\epsilon = .99$

**Given:** a regular curve  $C \in \mathbb{R}[X_1, \dots, X_n]^{n-1}$ ,  
an (approximate) point  $x$  on  $C$ , and a compact region  $R$   
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Starting at  $x$ , **track** along the curve  $C$  in a **certified**  
manner until a stopping criterion is satisfied.

Possible stopping criteria:

- Track until the computed curve hits the boundary of  $R$ ,  
or
- conclude that the curve is closed.



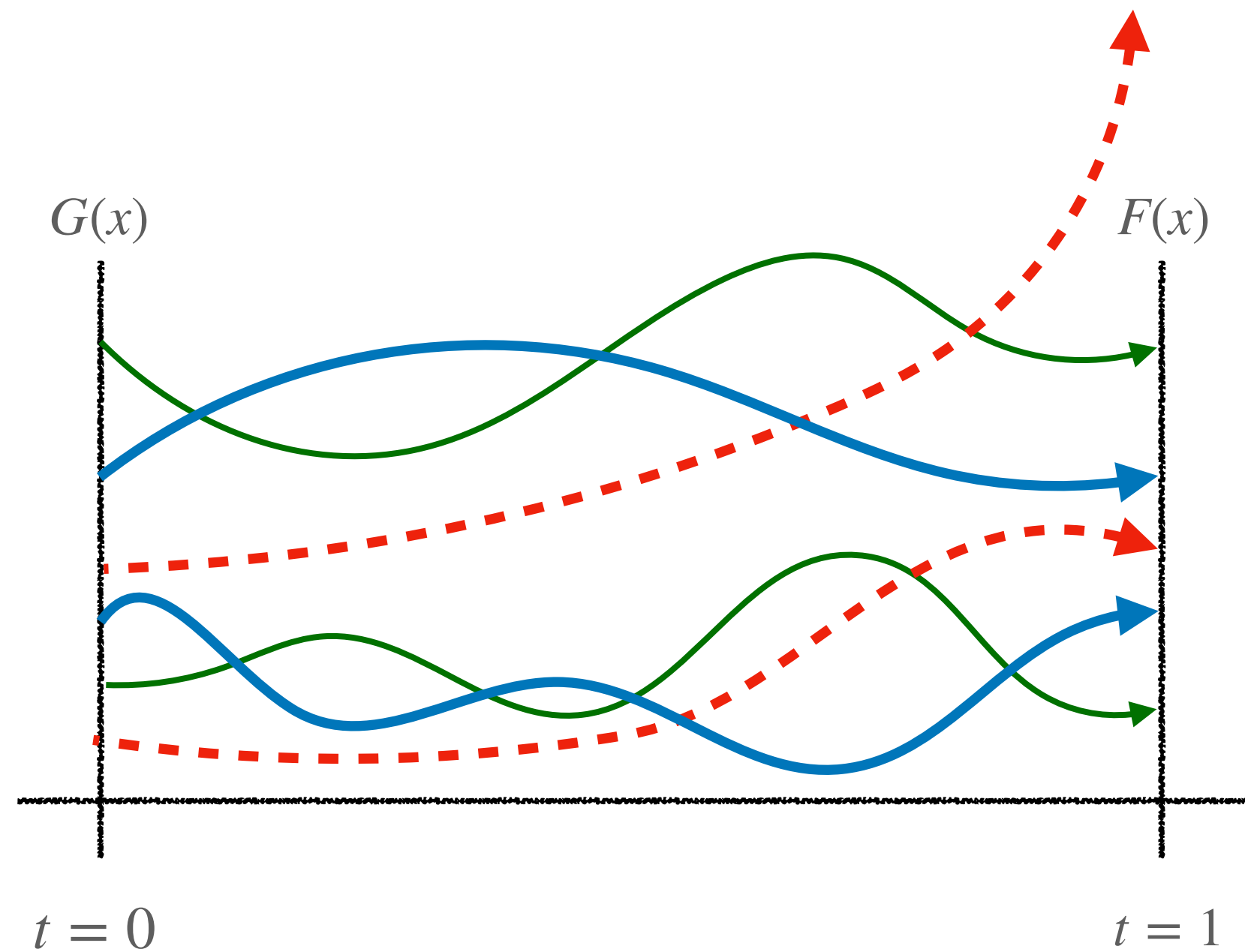
# Motivations

## Certified homotopy tracking

**Homotopy tracking** finds solutions to a polynomial system by tracking homotopy

$$H(X; t) = (1 - t)g(X) + tf(X), \quad t \in [0, 1].$$

Solve  $f$  (**target system**) by constructing a homotopy with  $g$  (**start system**) whose solutions are known.



# Motivations

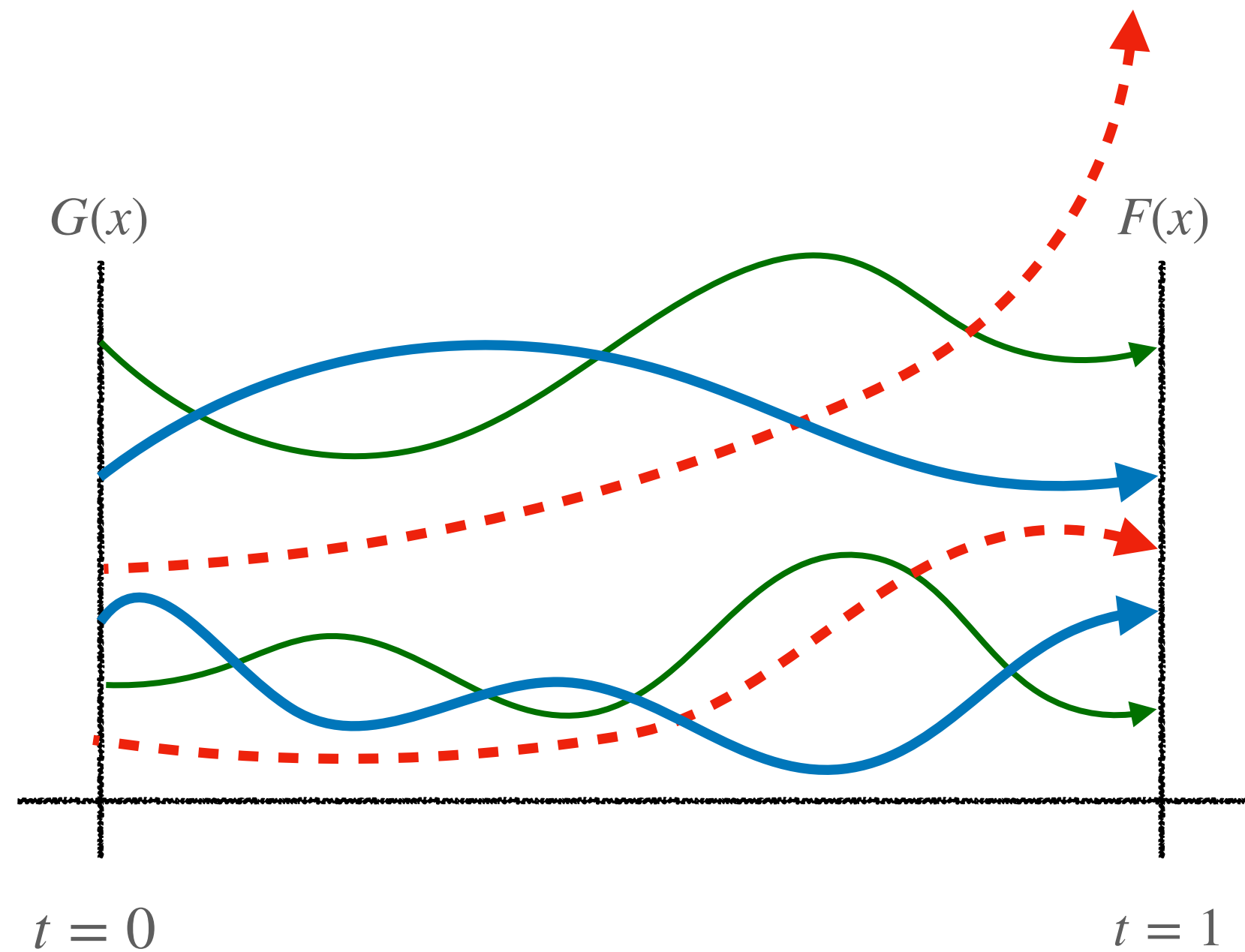
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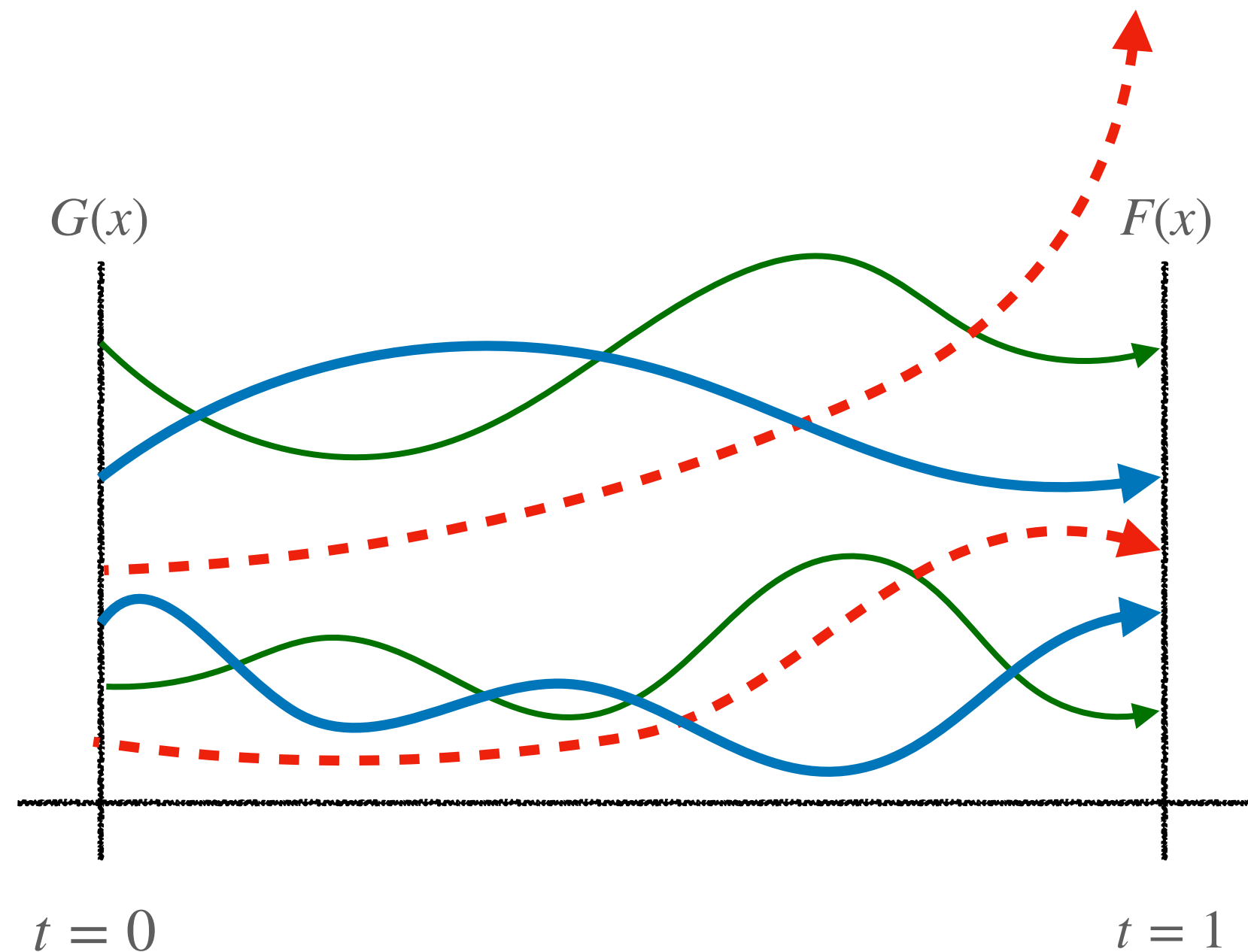
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A homotopy  $H(x; t)$  has one more variable than equations ( $H(X; t) := \{h_1(X; t), \dots, h_n(X; t)\} \subset \mathbb{C}[X_1, \dots, X_n, t]$ ).



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Likewise, a curve  $C$  has one more variable than equations ( $C := \{c_1, \dots, c_{n-1}\} \subset \mathbb{R}[X_1, \dots, X_n]$ ).

$G(x)$

$F(x)$

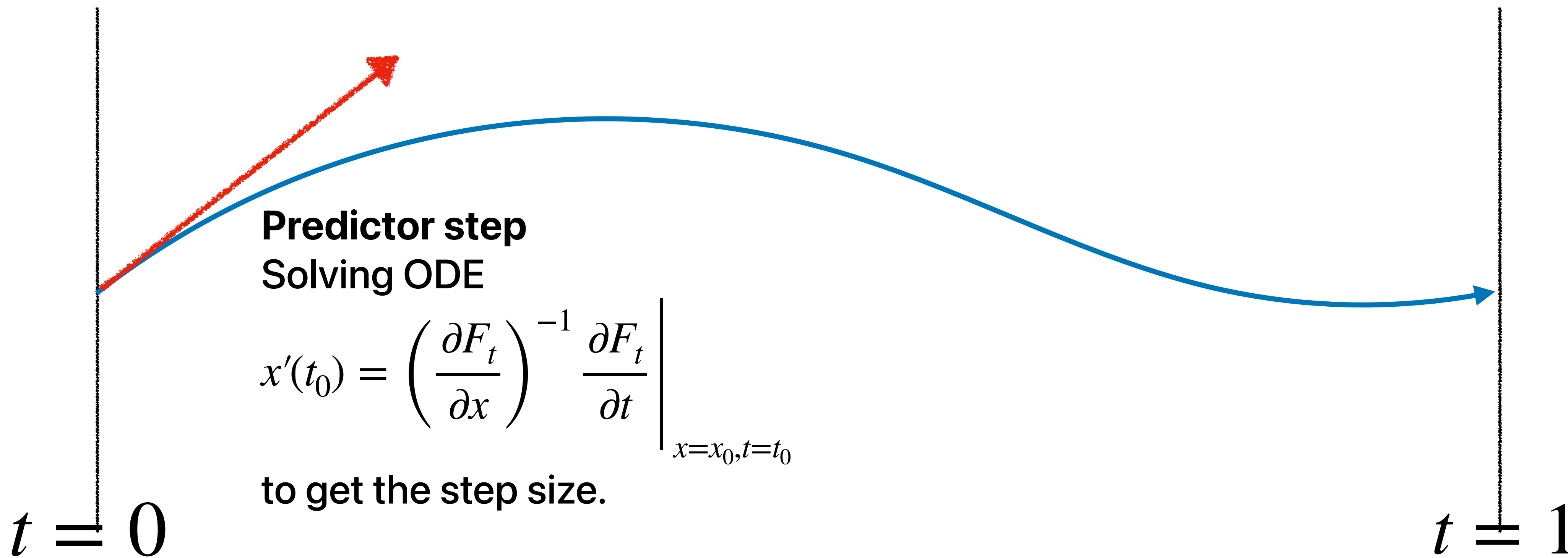
$t = 0$

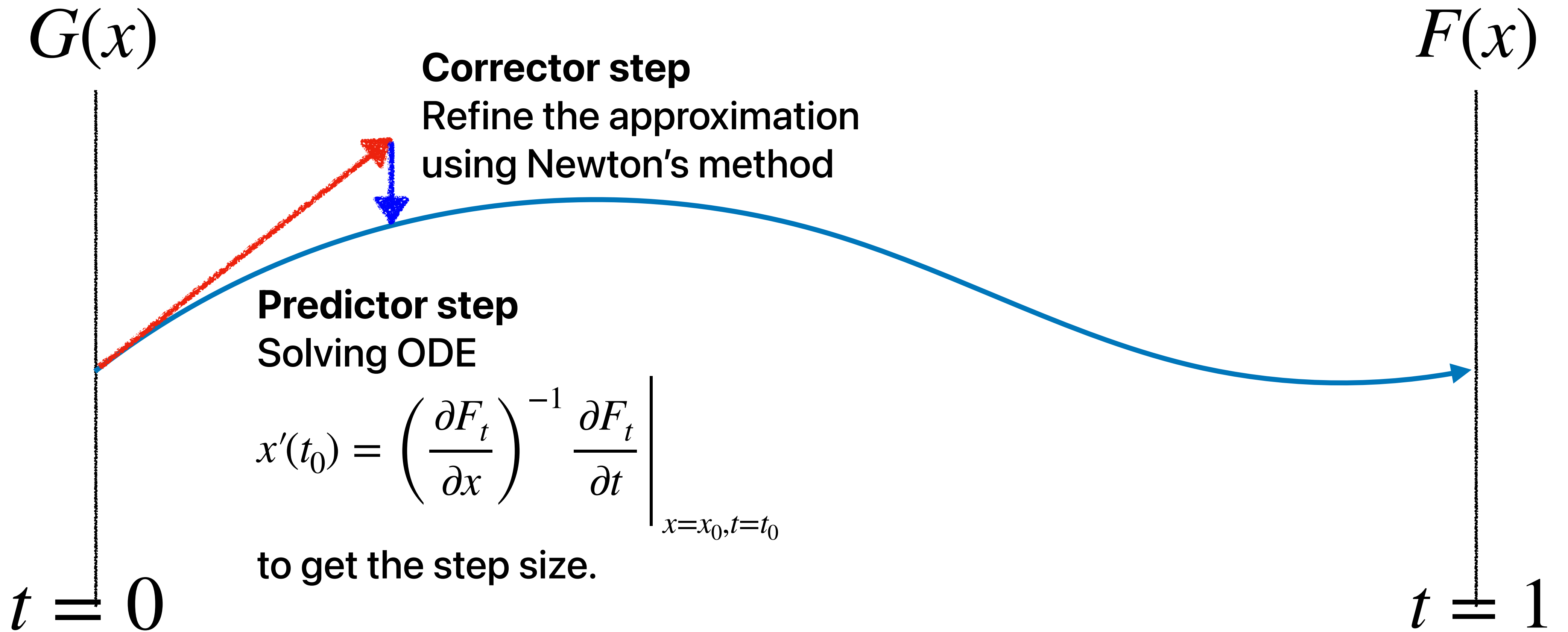
$t = 1$

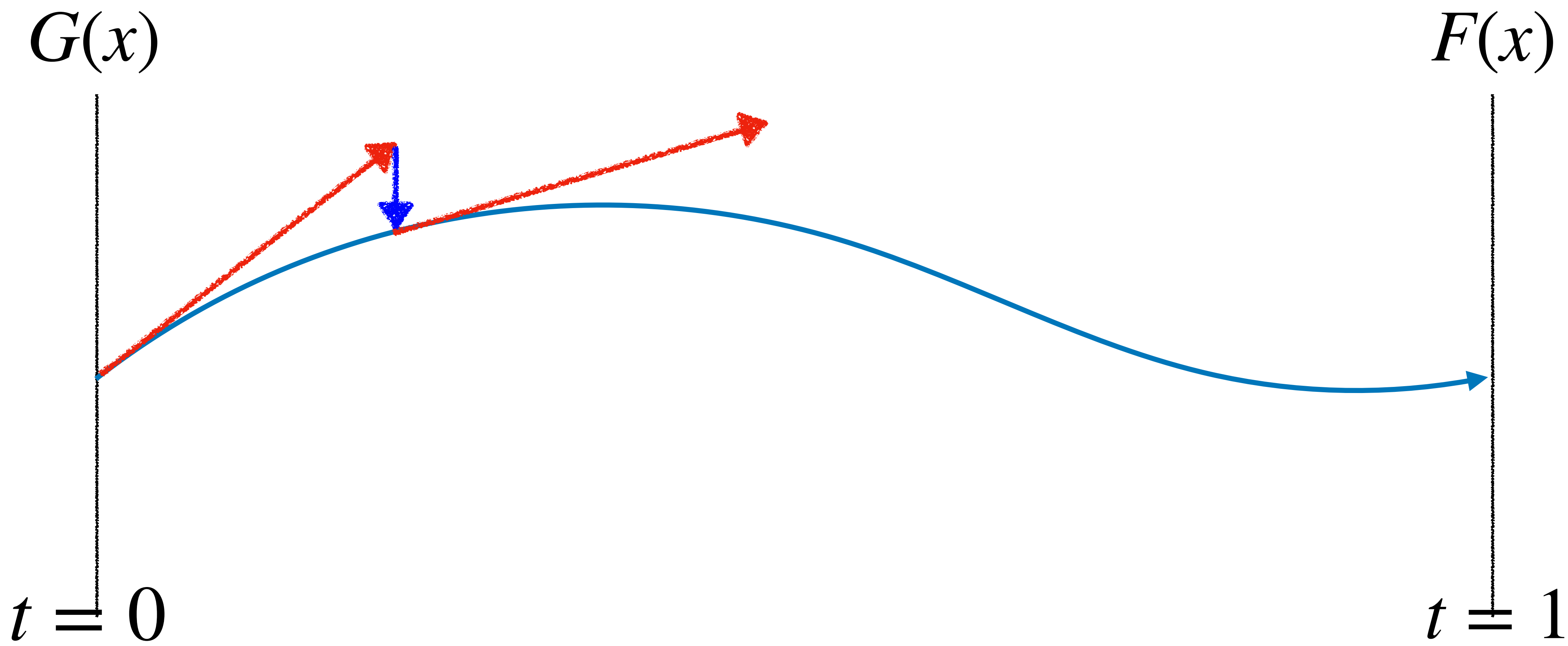


$G(x)$

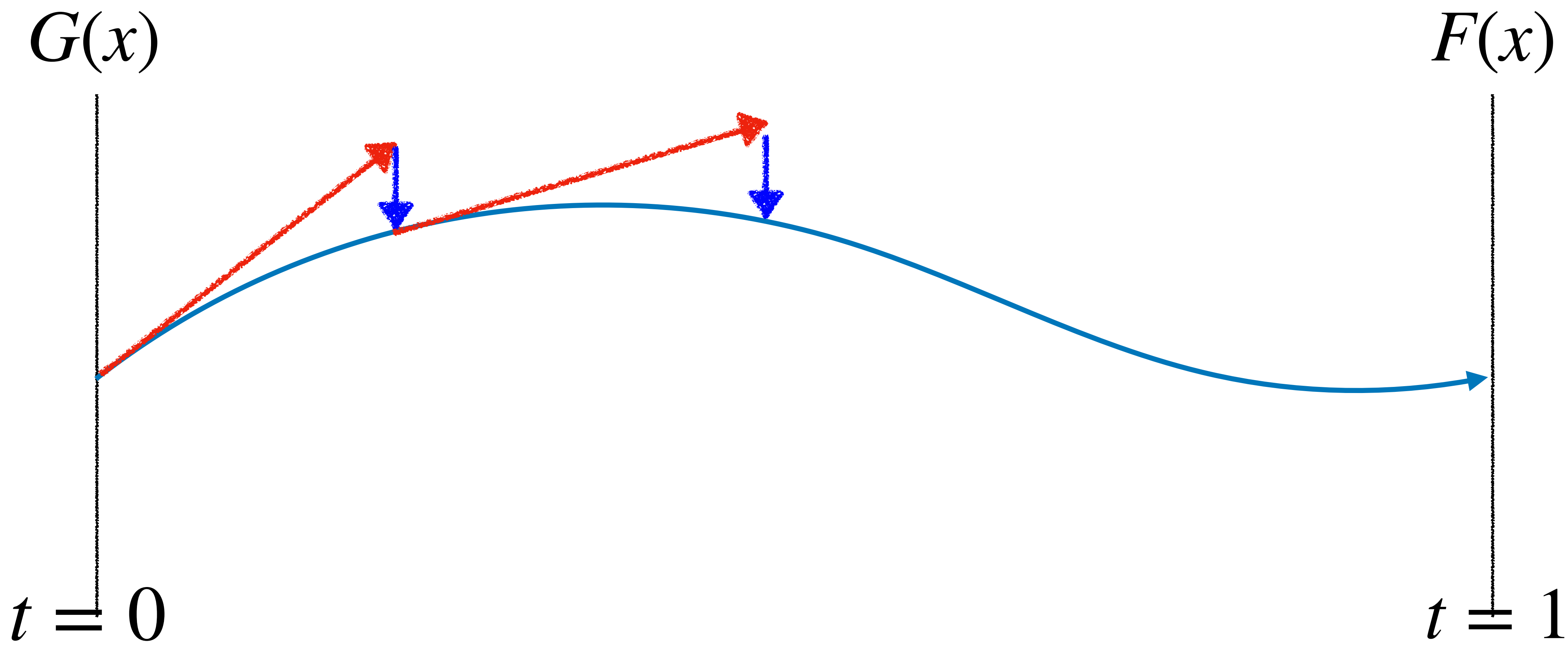
$F(x)$

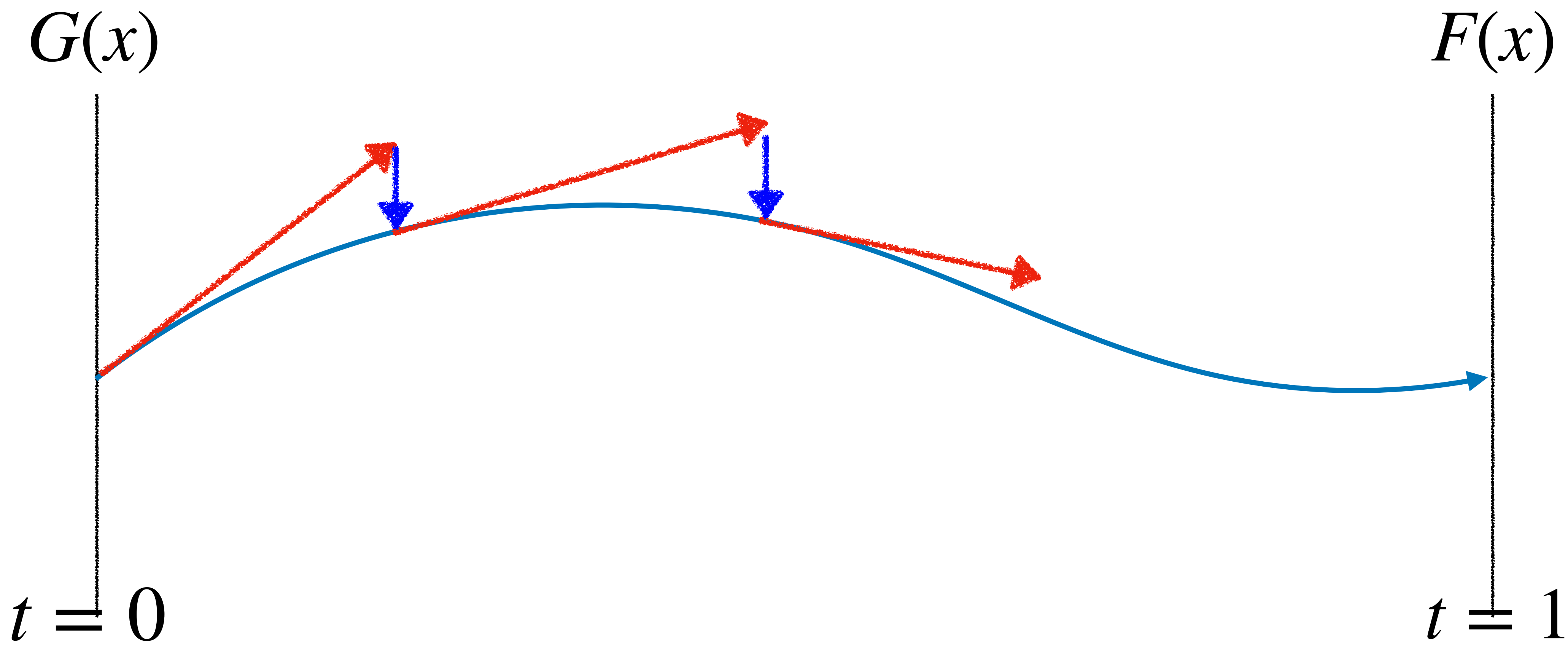










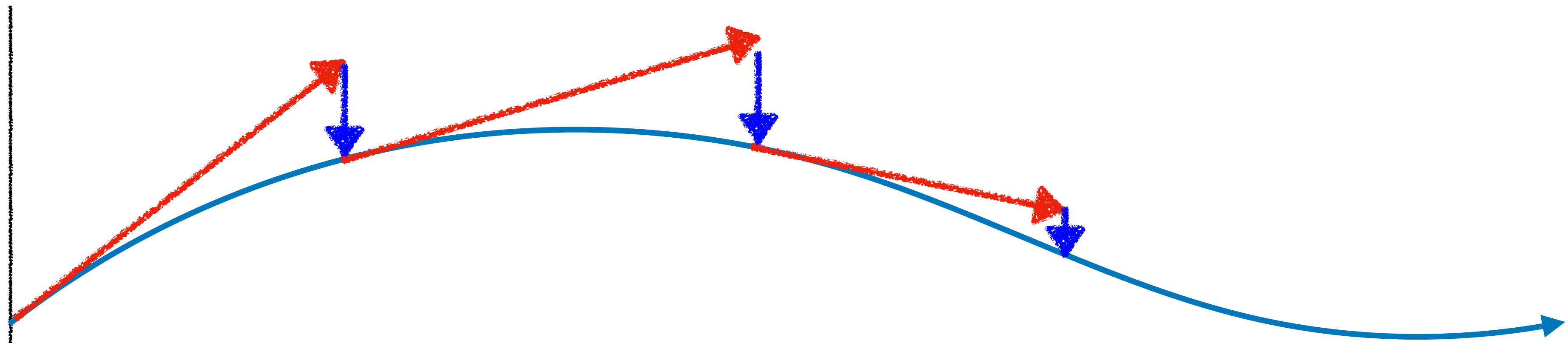


$G(x)$

$F(x)$

$t = 0$

$t = 1$

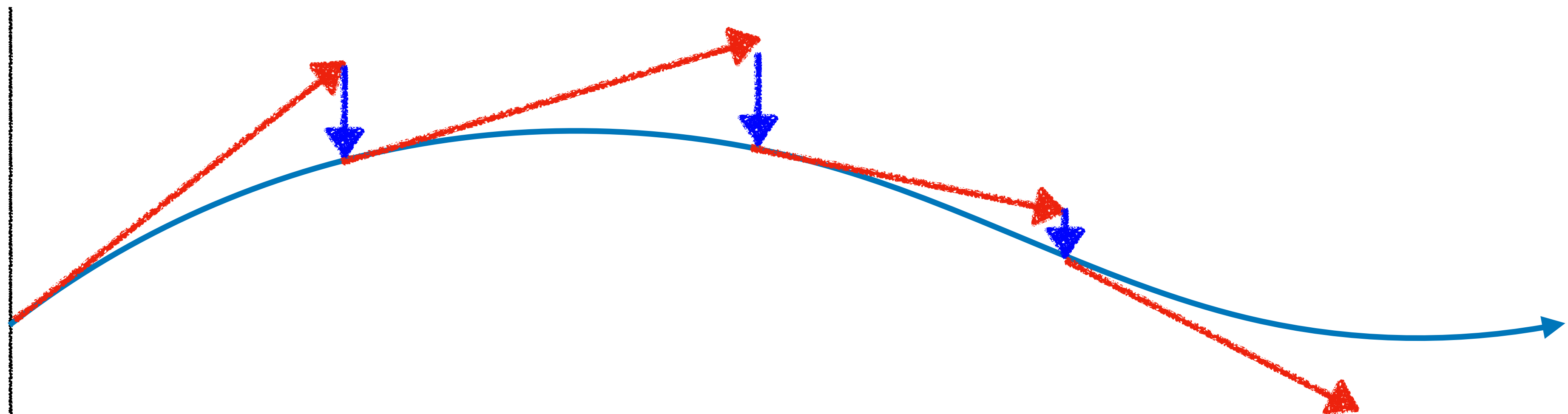


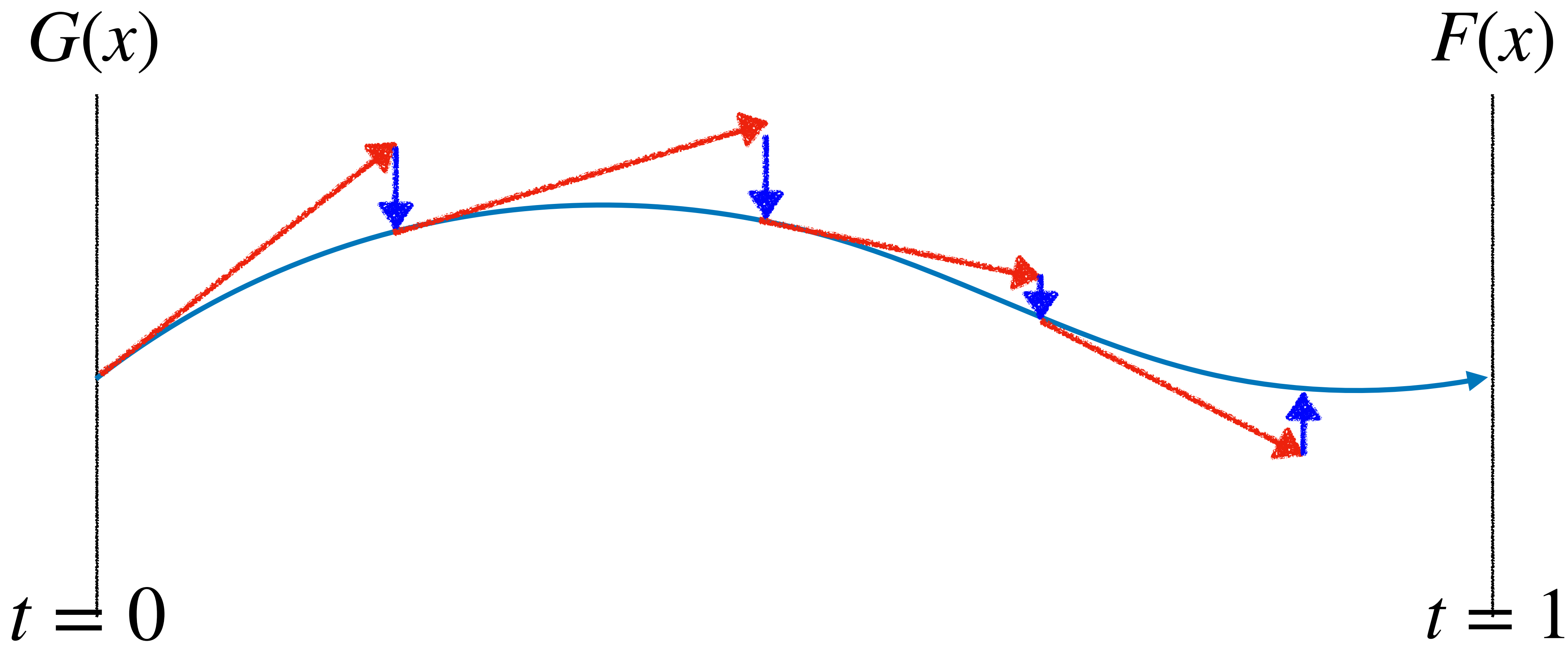
$G(x)$

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$t = 1$





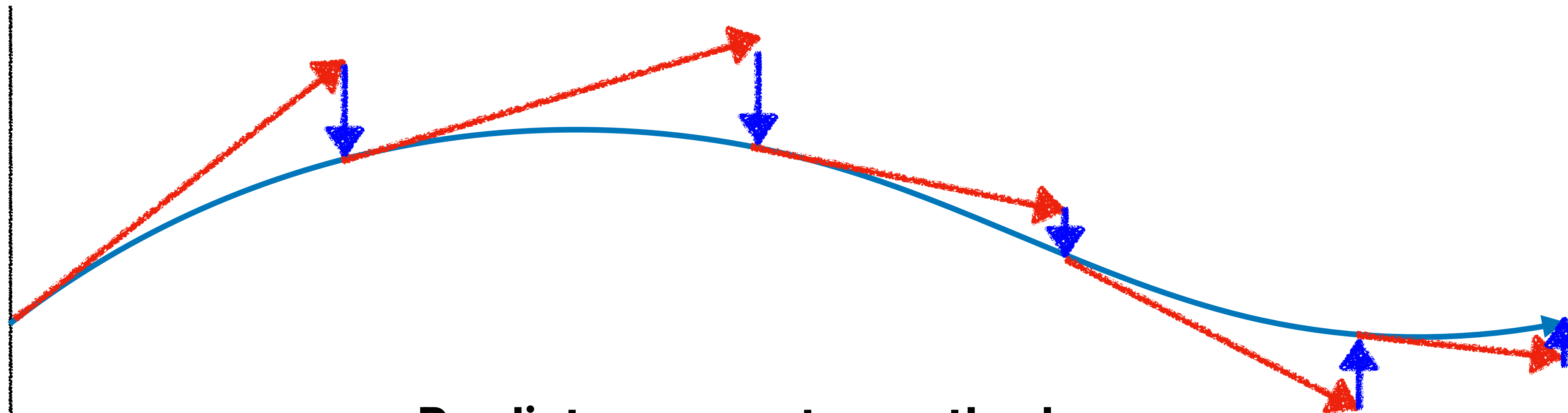
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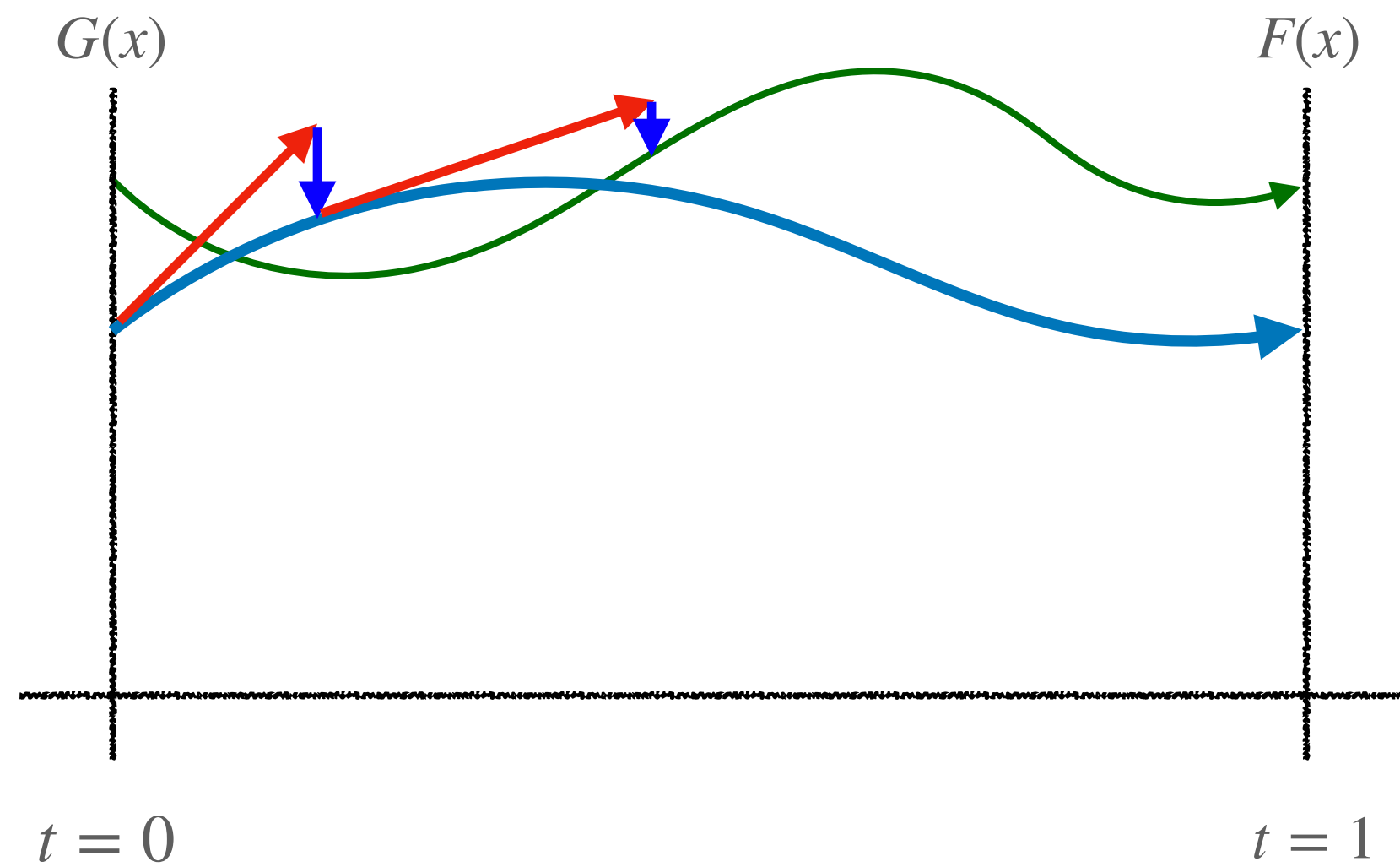
**Predictor-corrector method**



# Motivations

## Certified homotopy tracking

A path-jumping may occur when there is a nearby path.



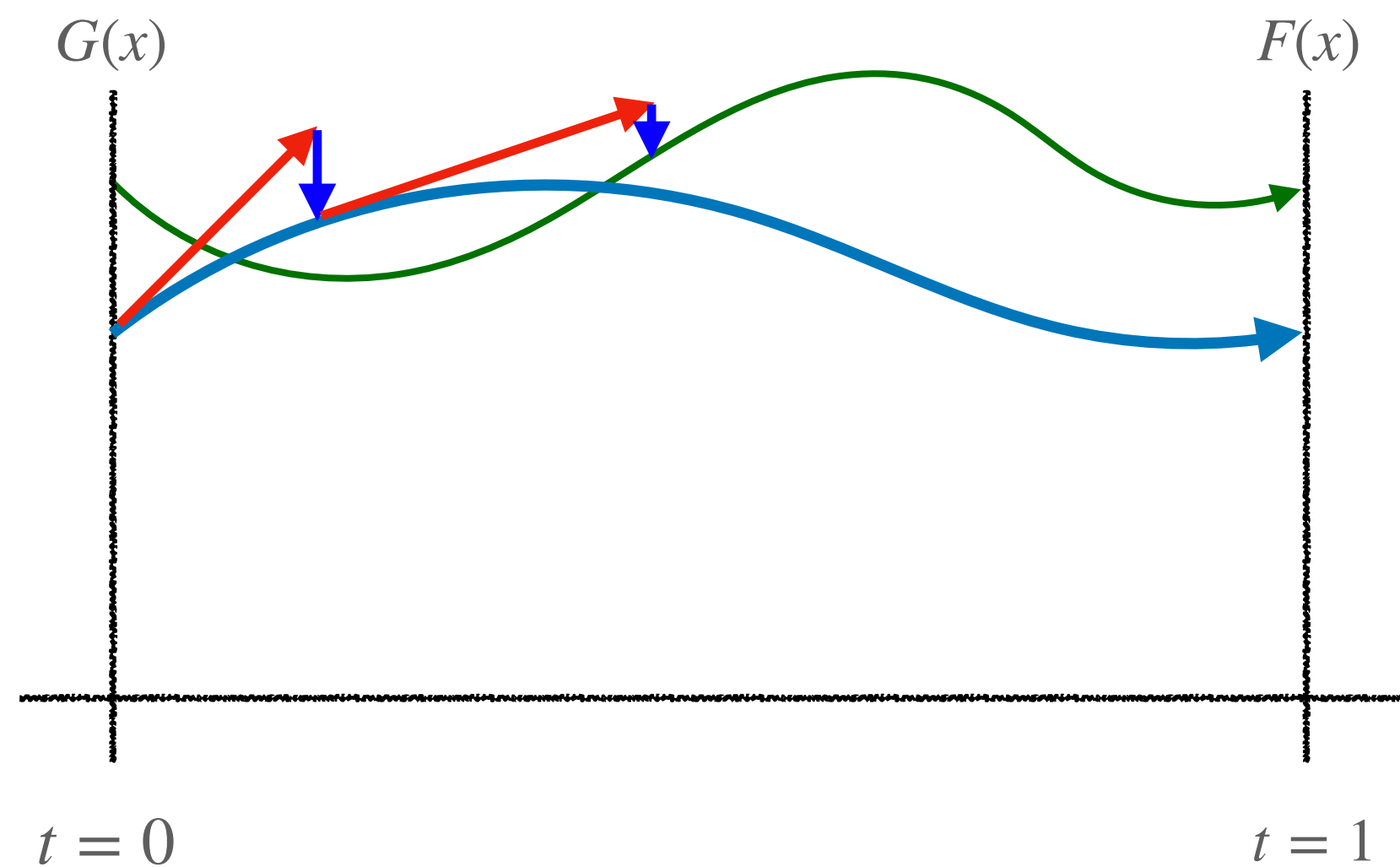


# Motivations

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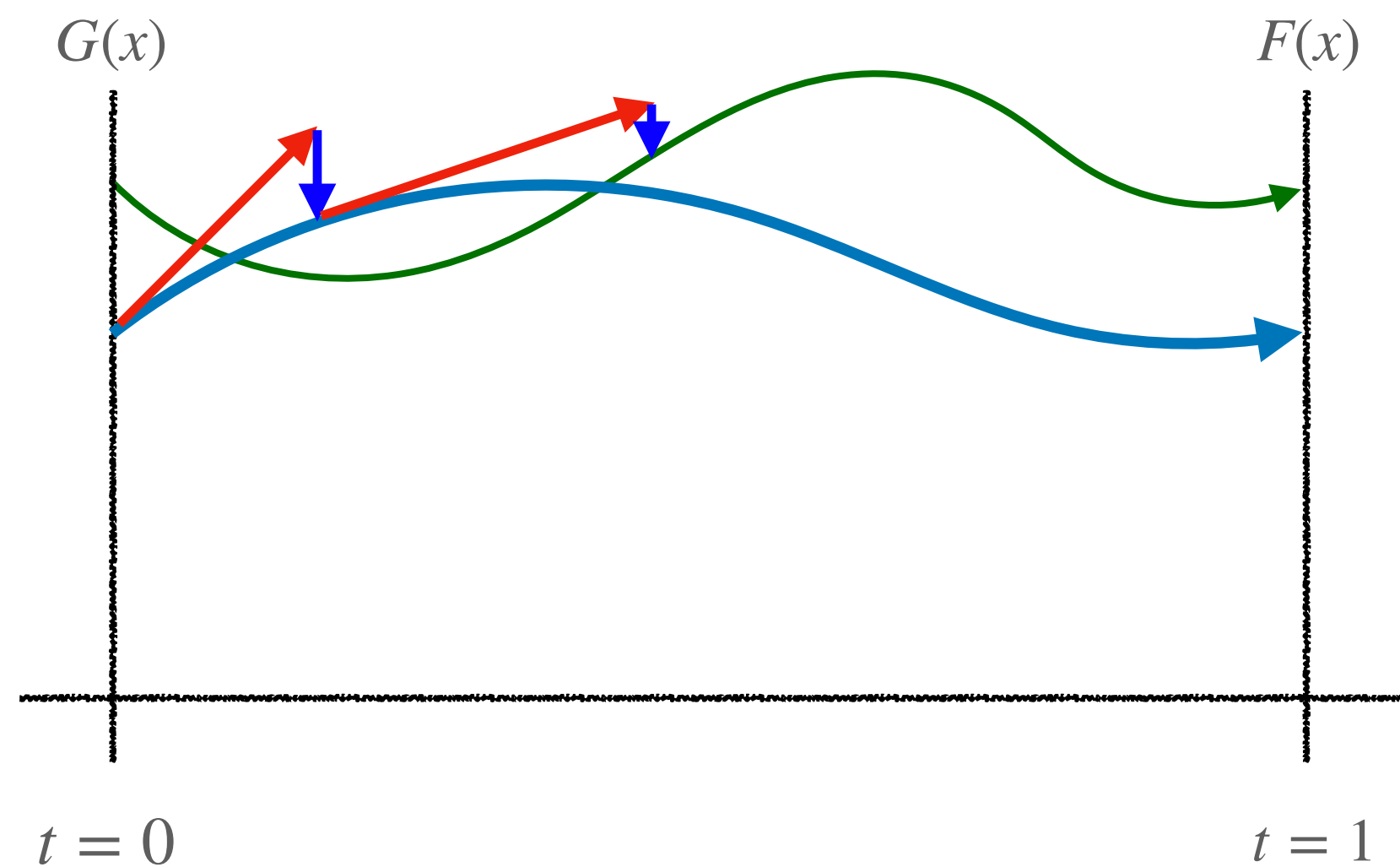
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**Certified homotopy tracking** guarantees tracking the solution path from start to finish **without jumping**.



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(Beltran-Leykin 2012 2013,  
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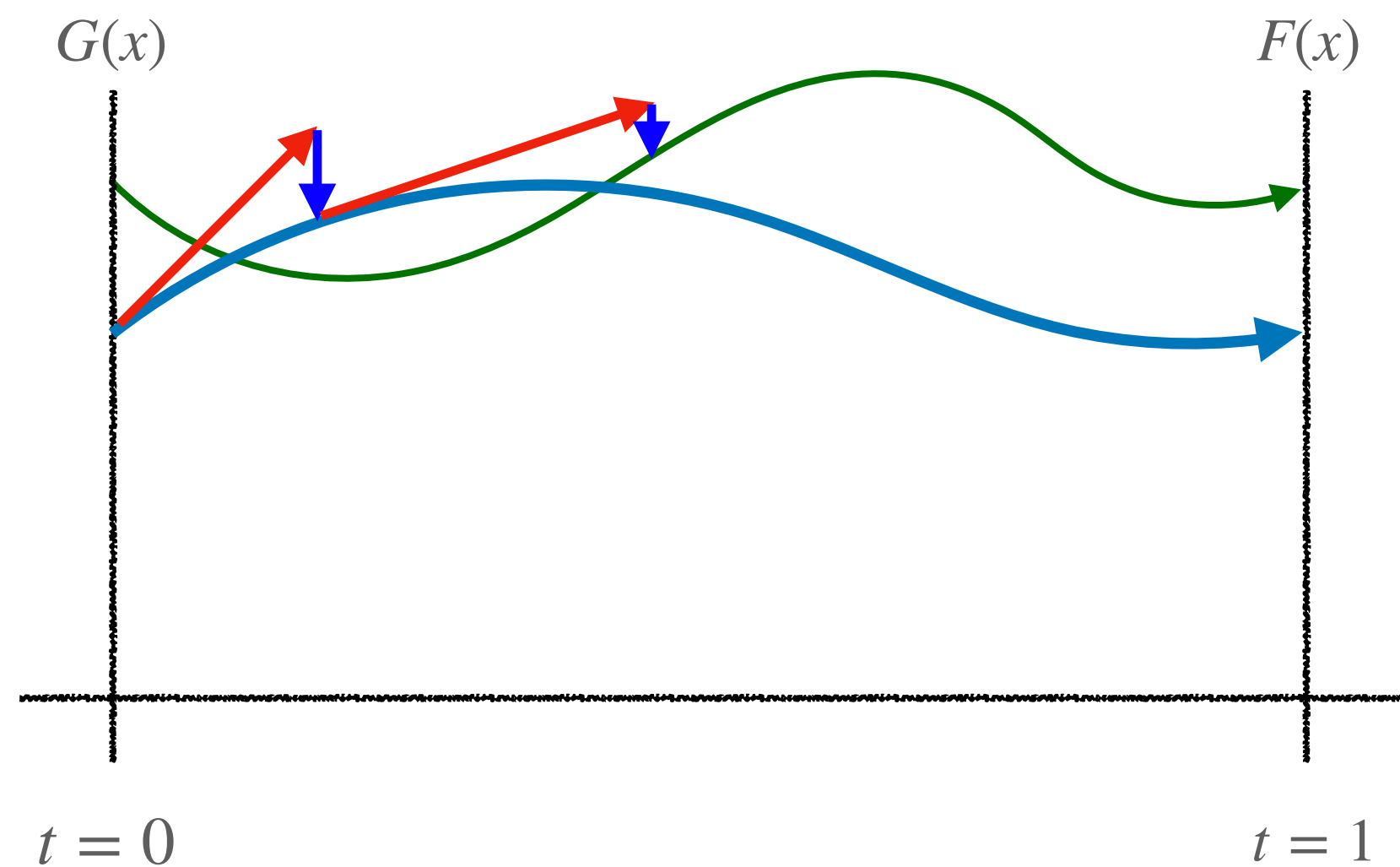
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# Motivations

Computing the geometry  
of curves

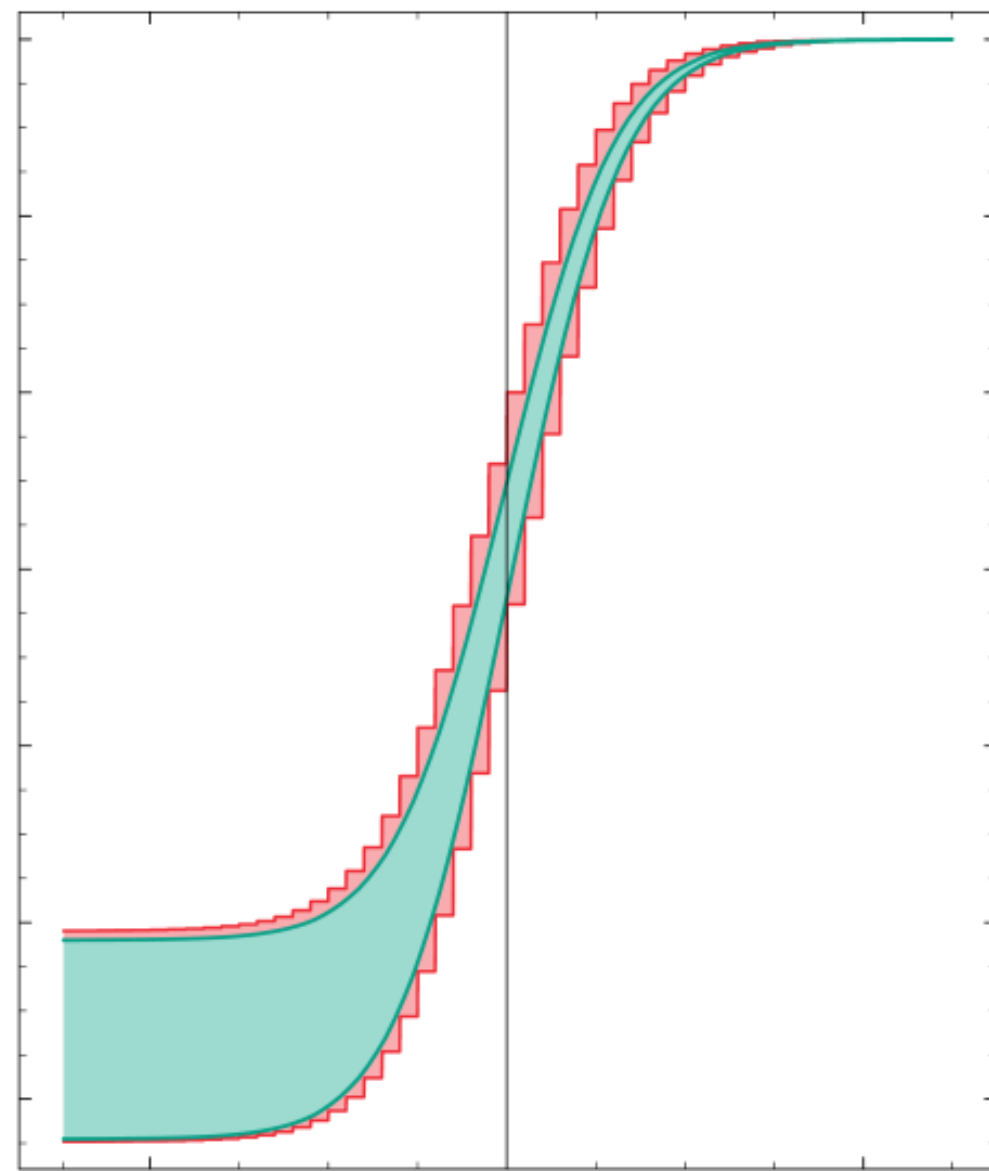
For a given curve in  $\mathbb{R}^n$ , compute an isotopic  
approximation to the curve.

**Katsamaki-Rouillier-Tsigaridas-Zafeirakopoulos 2023:**

For a parametric curve in  $\mathbb{R}^n$

**Burr-Byrd 2024:** Certified approximation for curves in  $\mathbb{R}^2$

# Interval arithmetic



[https://en.wikipedia.org/wiki/Interval\\_arithmetic](https://en.wikipedia.org/wiki/Interval_arithmetic)

**Interval arithmetic** executes reliable computation, mitigating rounding error

- Addition:

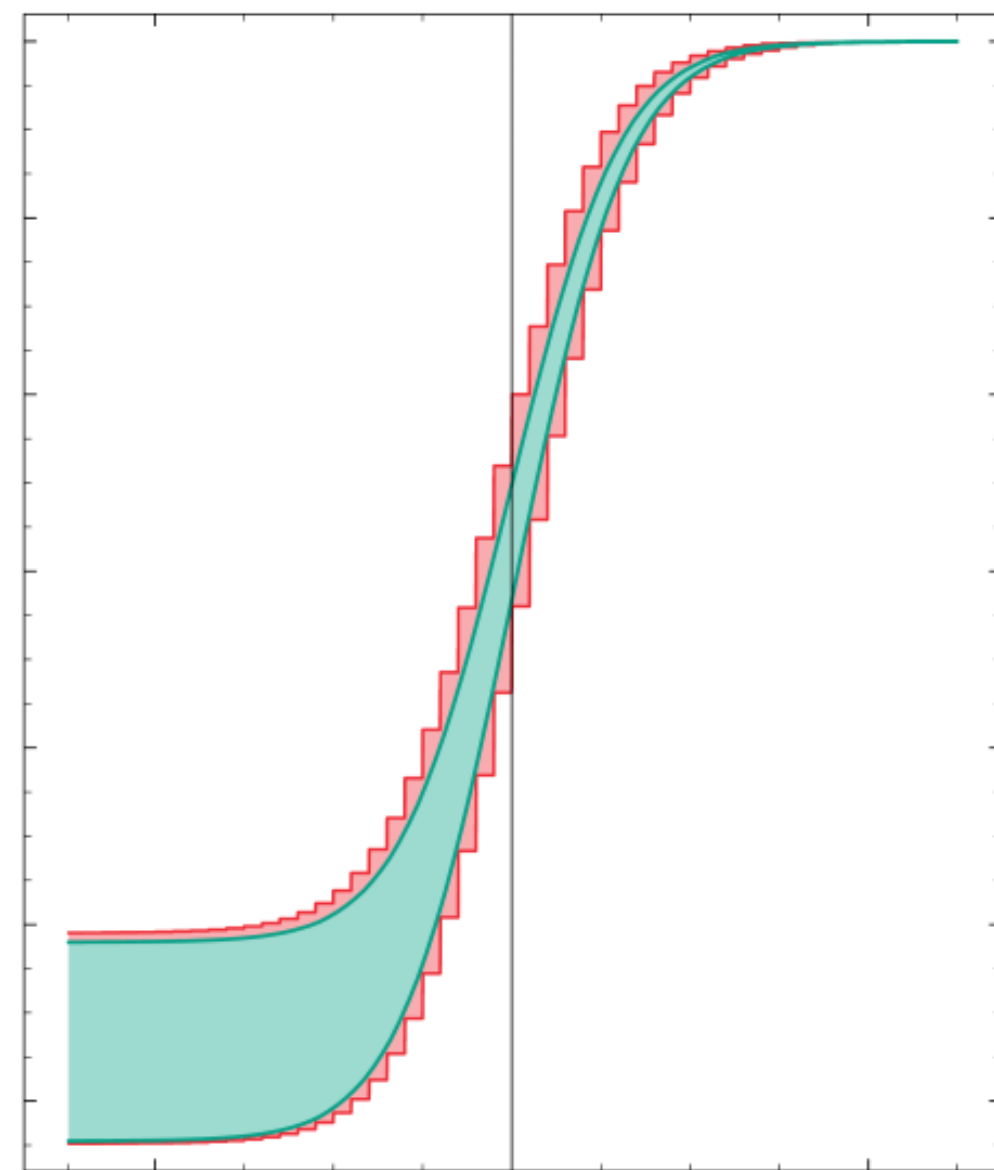
$$[x_1, x_2] + [y_1, y_2] = [x_1 + y_1, x_2 + y_2]$$

- Multiplication:

$$\begin{aligned} [x_1, x_2] \cdot [y_1, y_2] \\ = [\min\{x_1y_1, x_1y_2, x_2y_1, x_2y_2\}, \max\{x_1y_1, x_1y_2, x_2y_1, x_2y_2\}] \end{aligned}$$

**ex)**

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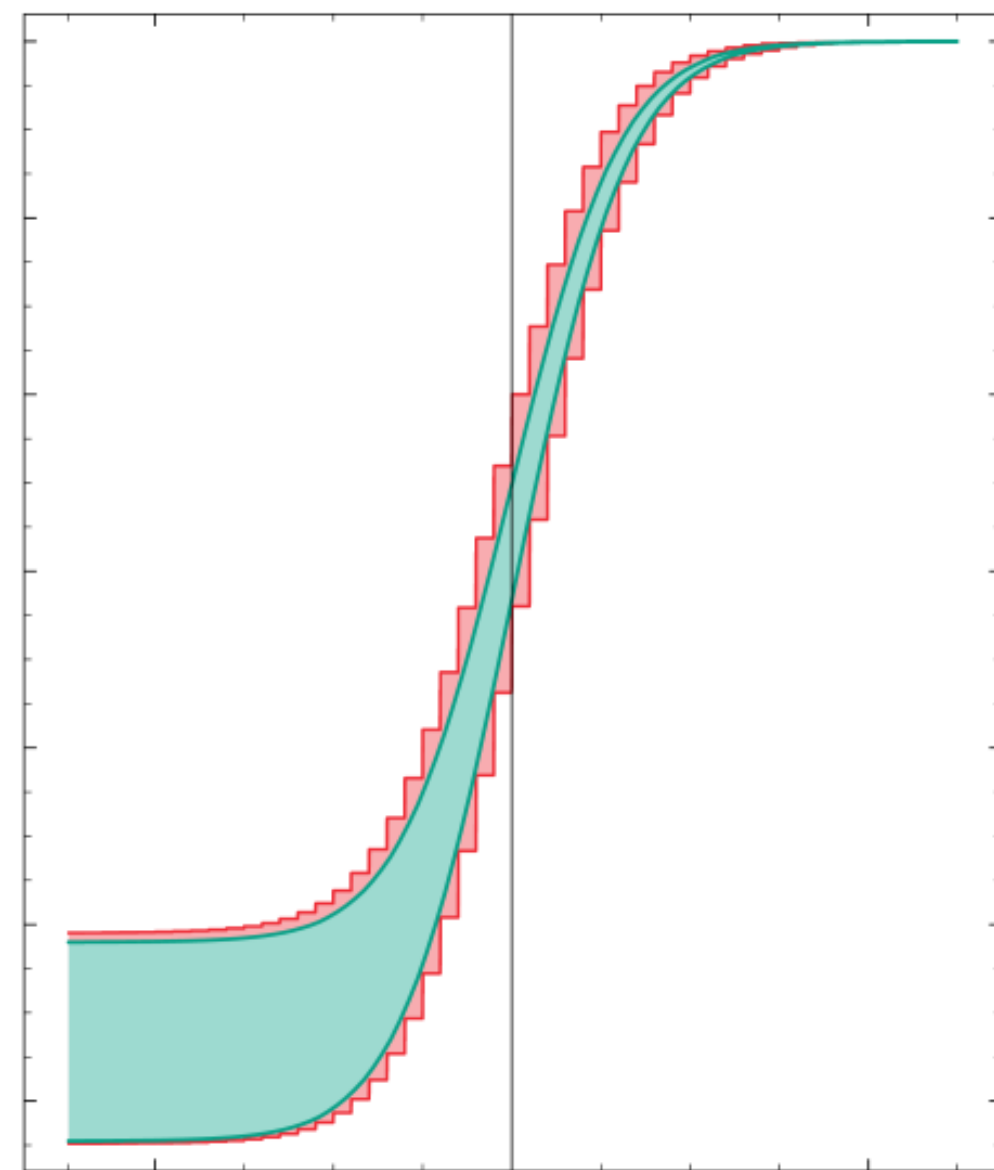
**ex)** Approximate  $\sqrt{2}$  using Newton's method:  $f(X) = X^2 - 2$ ,  $x_0 = 1$

$$x_1 = 1.5, \quad x_2 = 1.4167, \quad x_3 = 1.414213562$$

$$x_4 = 1.4142135623731$$

$$\sqrt{2} = 1.4142135623731\dots$$

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**How to prove the correctness of Newton iterations?**



# Krawczyk method

The Krawczyk method combines interval arithmetic and Newton's method. For a system  $f : \mathbb{R}^n \rightarrow \mathbb{R}^n$ , an interval box  $B := [-1, 1]^n$  with a point  $x$ , and an  $n \times n$  invertible matrix  $Y$ , define the **Krawczyk operator**

$$x - Yf(x) + \left( Id - Y \square \frac{df}{dx}(x + rB) \right) B$$

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**Theorem (Krawczyk 1969).** For a radius  $r$  and a constant  $\rho \in (0, 1)$ , if

$$x - Yf(x) + \left( Id - Y \square \frac{df}{dx}(x + rB) \right) B \subset \rho(x + rB),$$

Then, there is a unique solution  $x^\star$  to  $f$  in  $x + rB$ .

Also, Newton iteration applied at a point in  $x + rB$  converges to  $x^\star$ .

In this case, we say that  $x$  is a  **$\rho$ -approximate solution** to  $F$ .

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**Theorem (Krawczyk 1969).** For a radius  $r$  and a constant  $\rho \in (0, 1)$ , if

$$x - YH(x; T) + \left( Id - Y \square \frac{dH}{dx}(x + rB; T) \right) B \subset \rho(x + rB),$$

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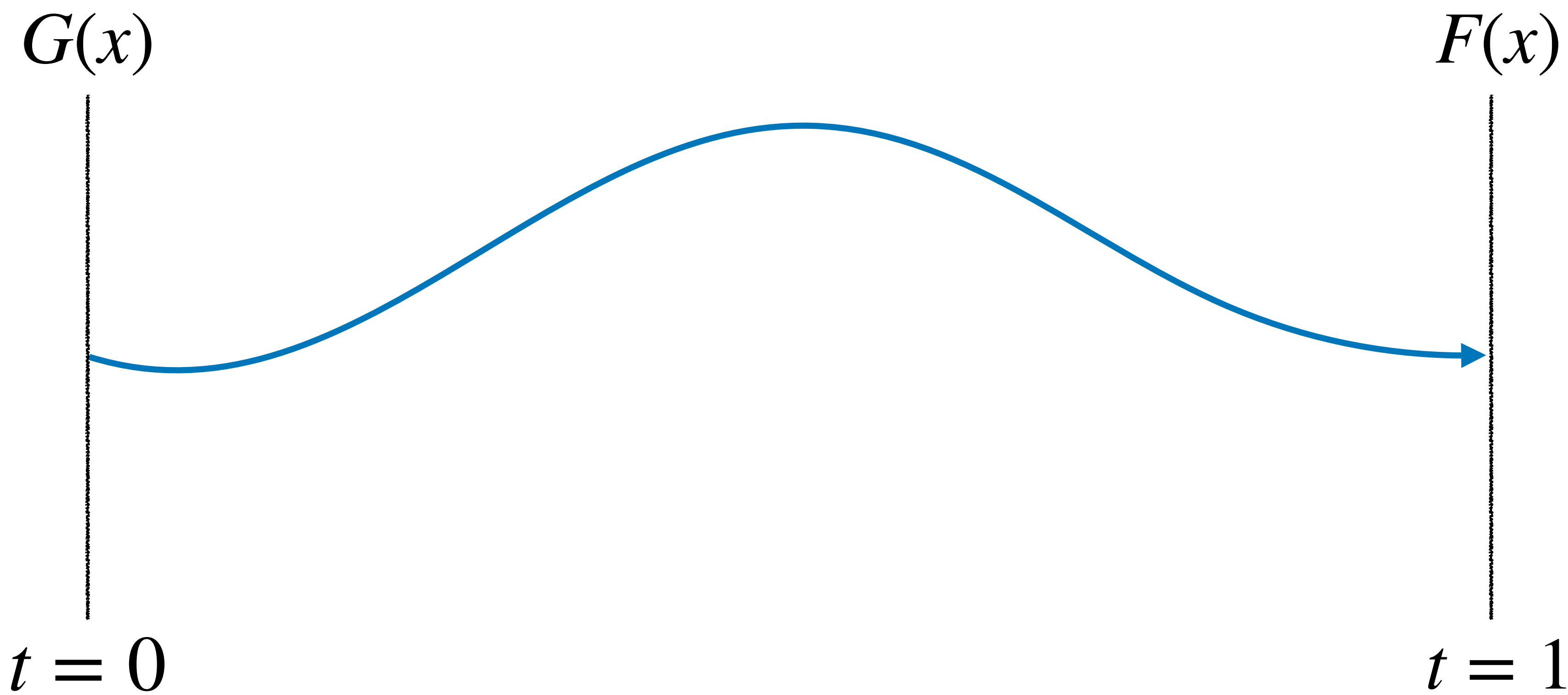
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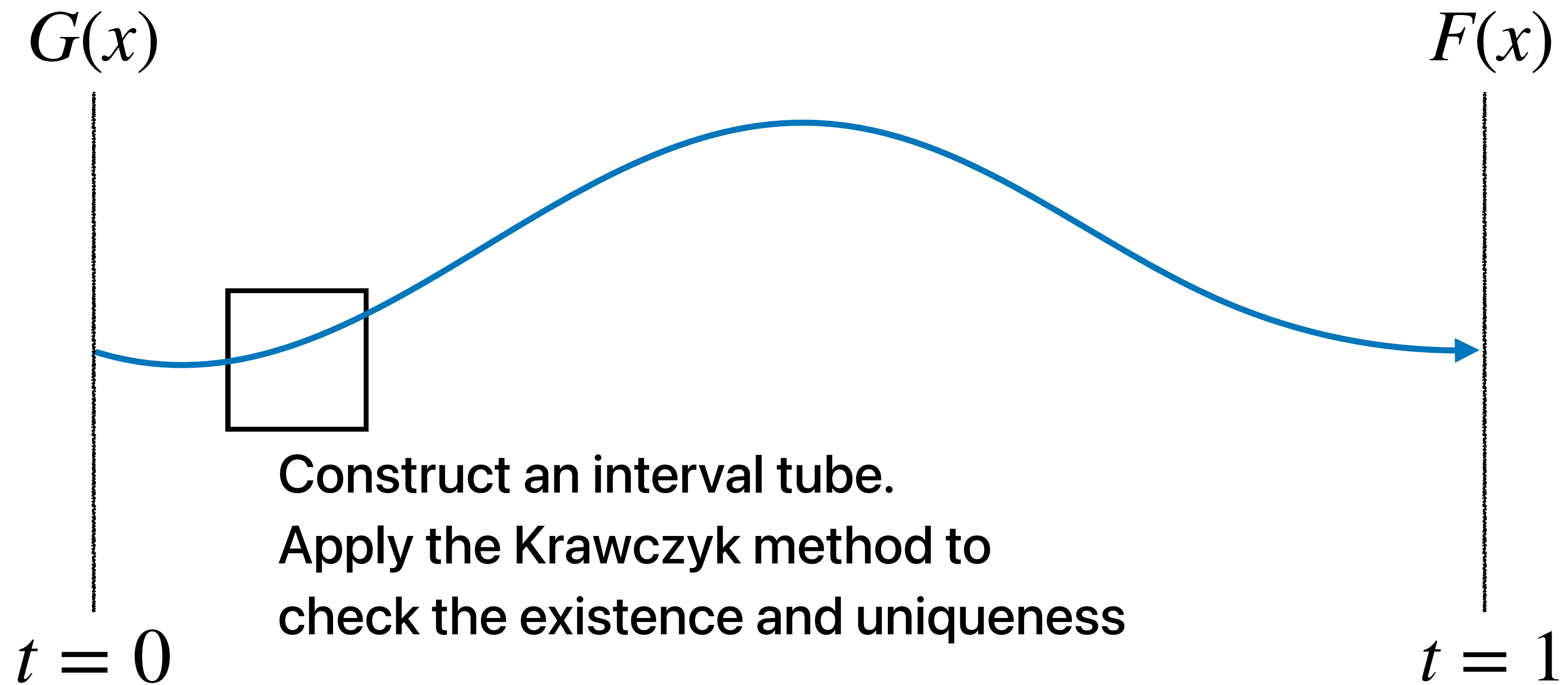
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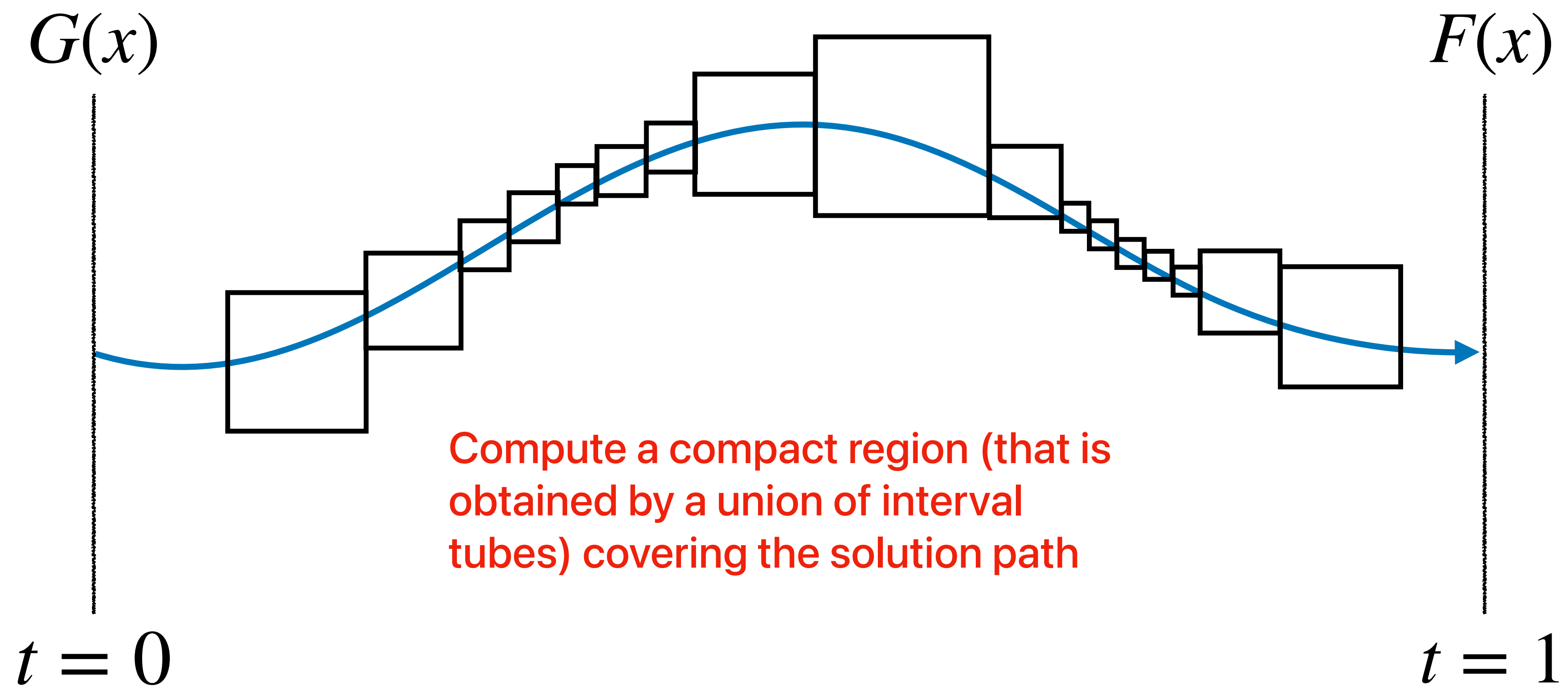
In this case, we say that  $x$  is a  **$\rho$ -approximate solution** to  $F$ .

In certified homotopy tracking, the Krawczyk method is applied to the homotopy  $H(X; t)$  over an interval  $T \subset [0, 1]$ .









# **Curve tracking:**

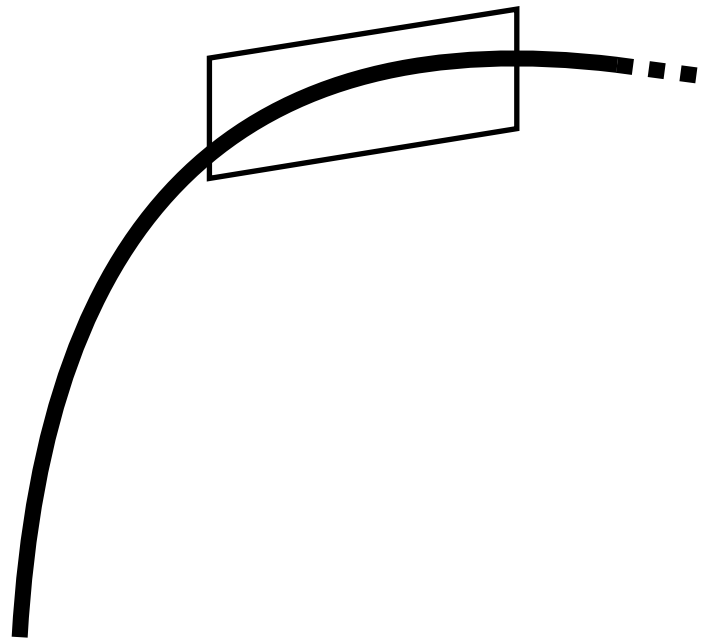
## **algorithm overview**

Choosing a direction to track

# Curve tracking:

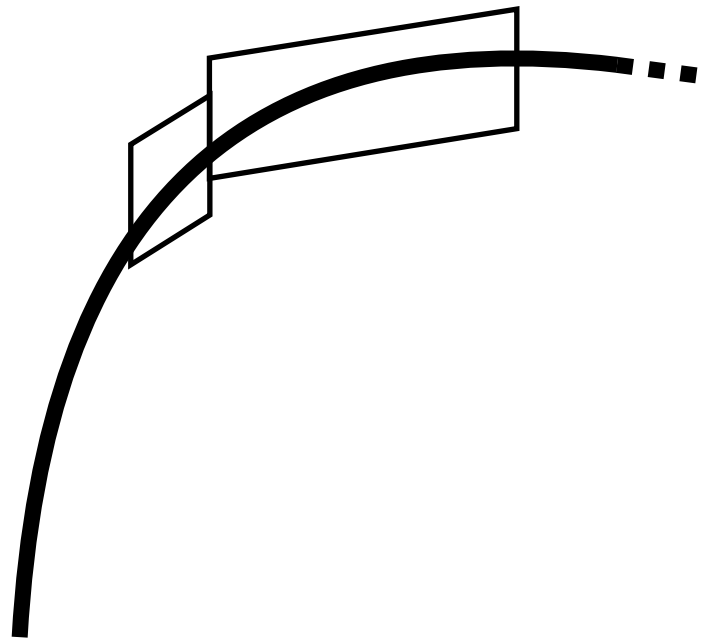
## algorithm overview

Choosing a direction to track



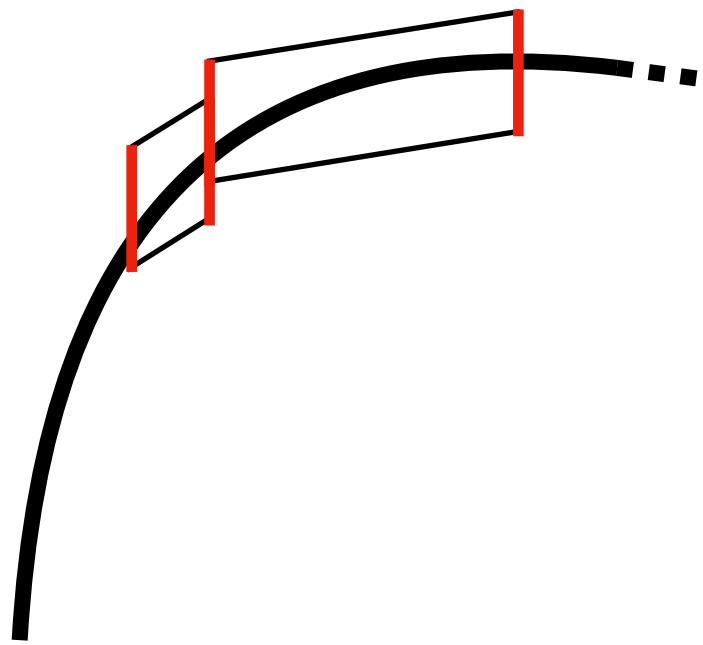
# Curve tracking: algorithm overview

Choosing a direction to track



# Curve tracking: algorithm overview

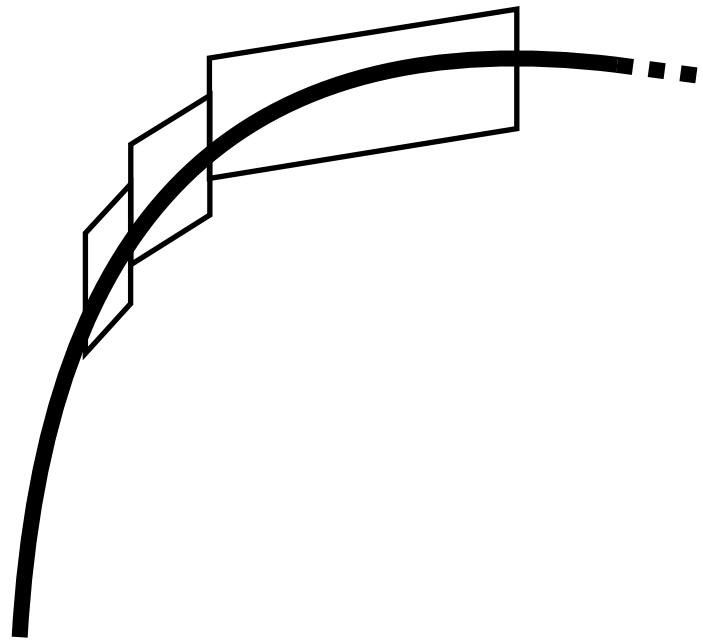
Choosing a direction to track



axis-aligned intervals

# Curve tracking: algorithm overview

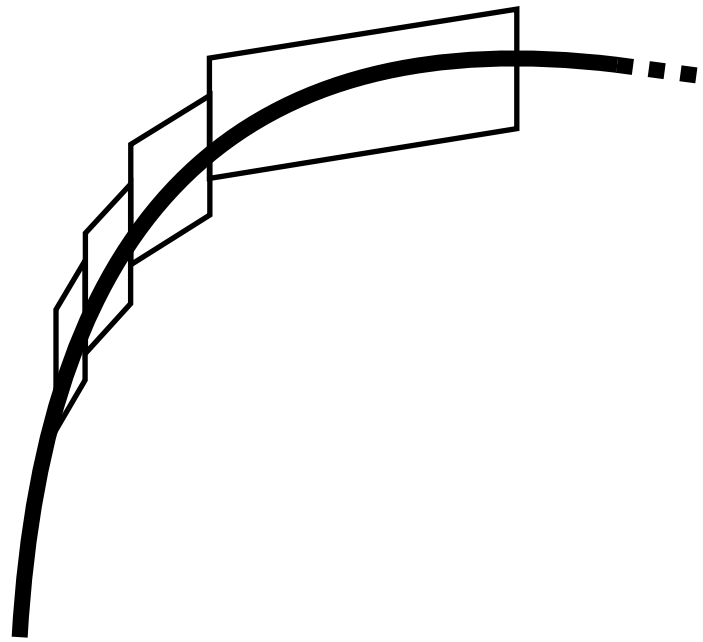
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# Curve tracking: algorithm overview

Choosing a direction to track



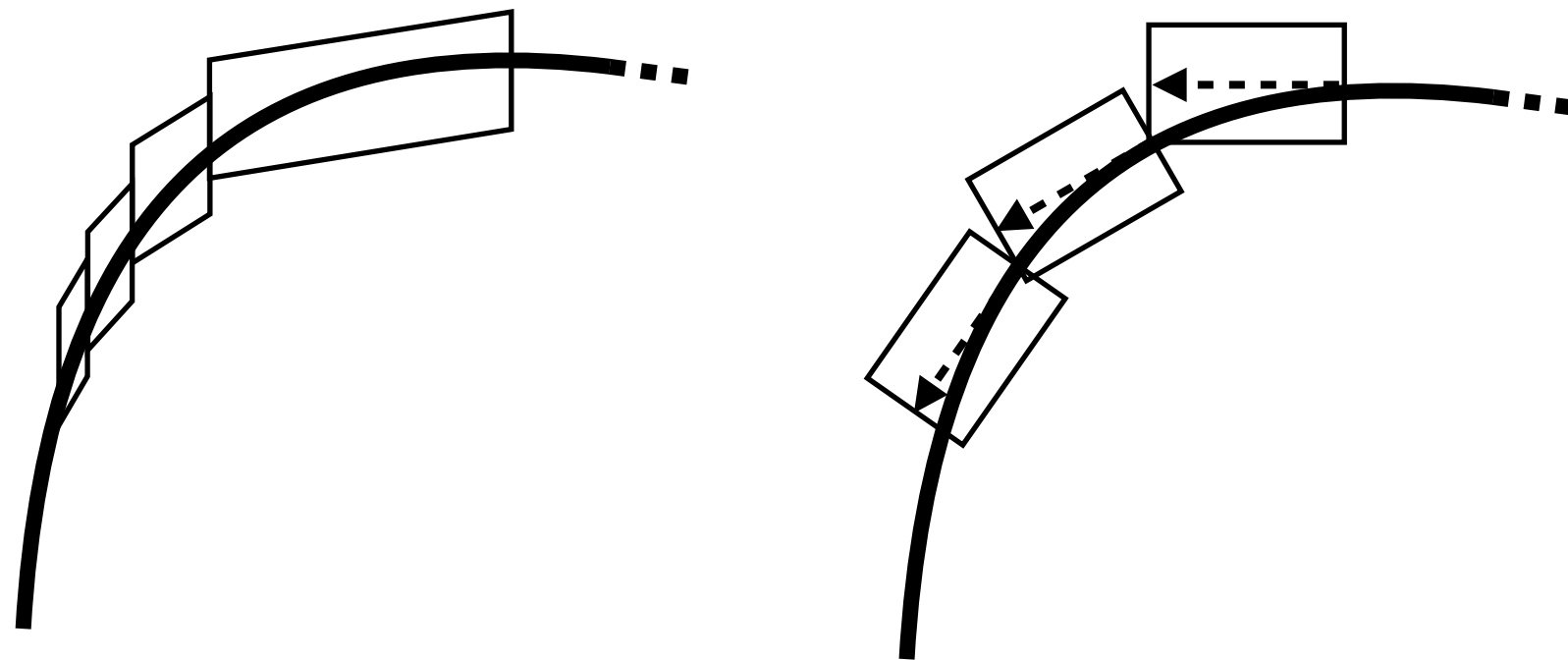
Tracking a curve with axis-aligned intervals  
is inefficient (or impossible).



# Curve tracking: algorithm overview

Choosing a direction to track

For effective tracking, apply a **unitary transformation** (rotation) at each step to align the tangent vector vertically.



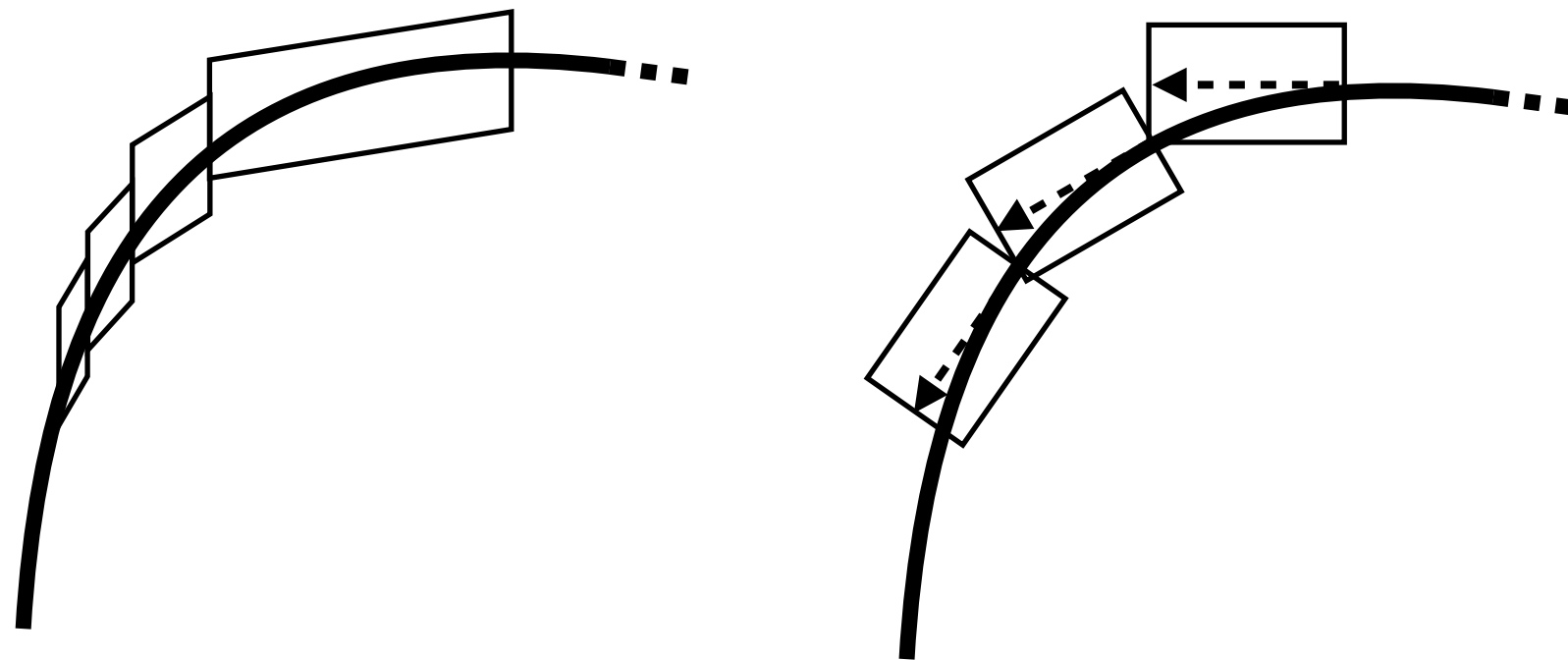
Rotating the curve  
enables efficient tracking

# Curve tracking: algorithm overview

## Unitary transformation

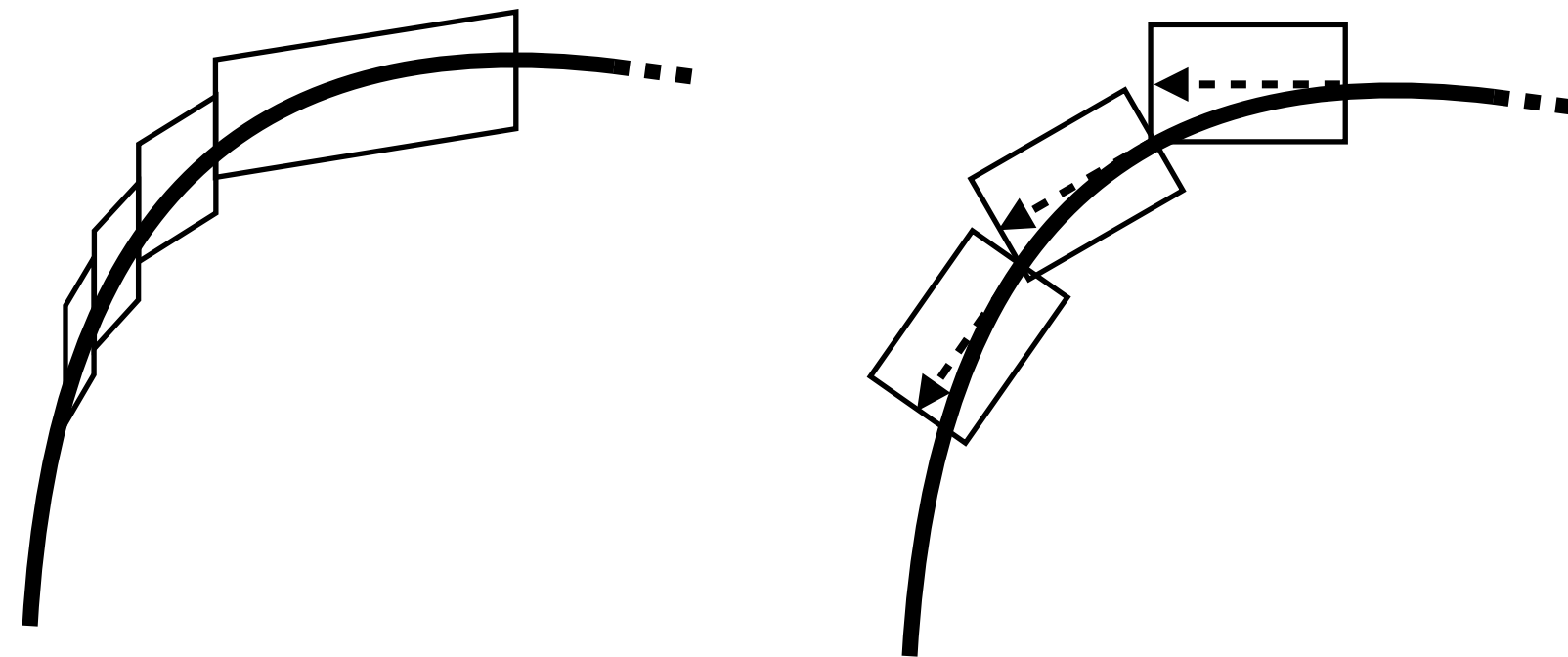
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A unitary transformation is computed via SVD using interval arithmetic.



# Curve tracking: algorithm overview

Unitary transformation



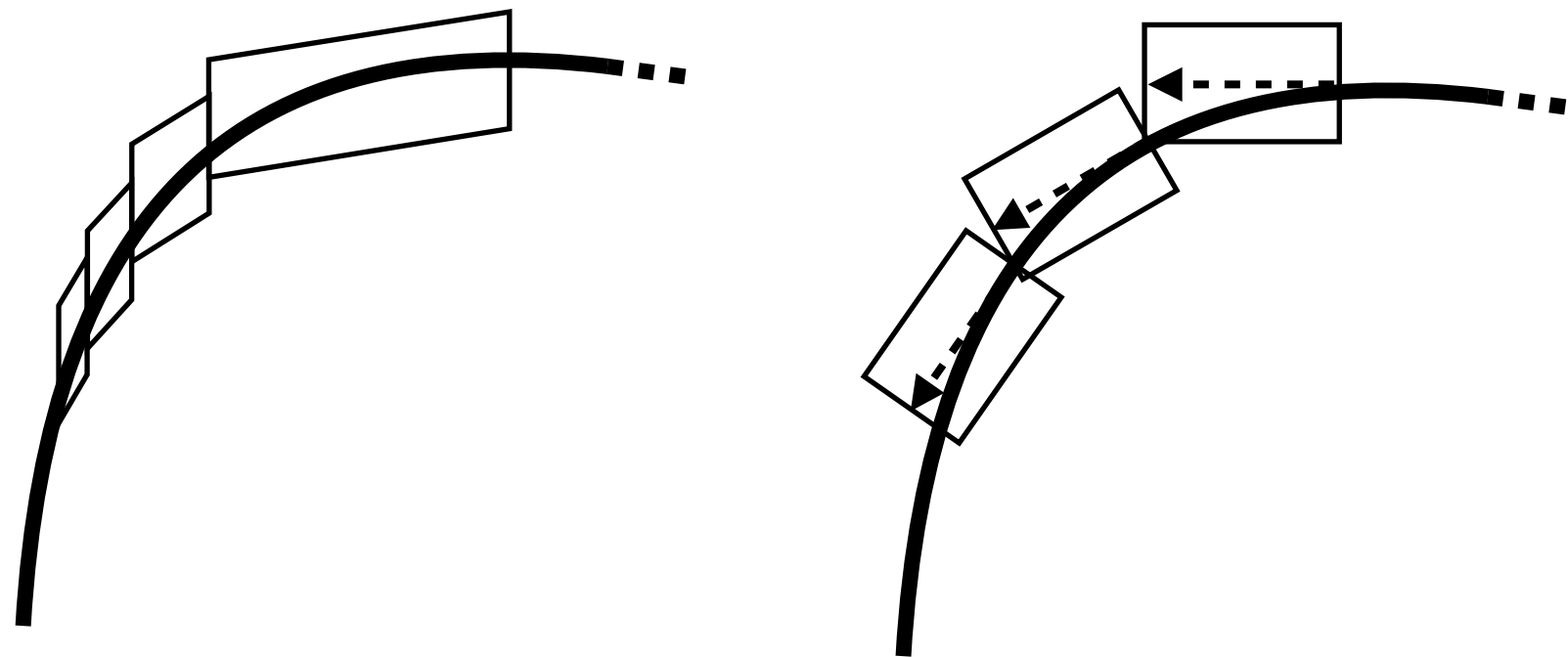
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For a curve  $C := \{c_1, \dots, c_{n-1}\} \subset \mathbb{R}[X_1, \dots, X_n]$  and a point  $x \in \mathbb{R}^n$ , compute the SVD  $J C(x) = U \Sigma V^\star$

# Curve tracking: algorithm overview

## Unitary transformation



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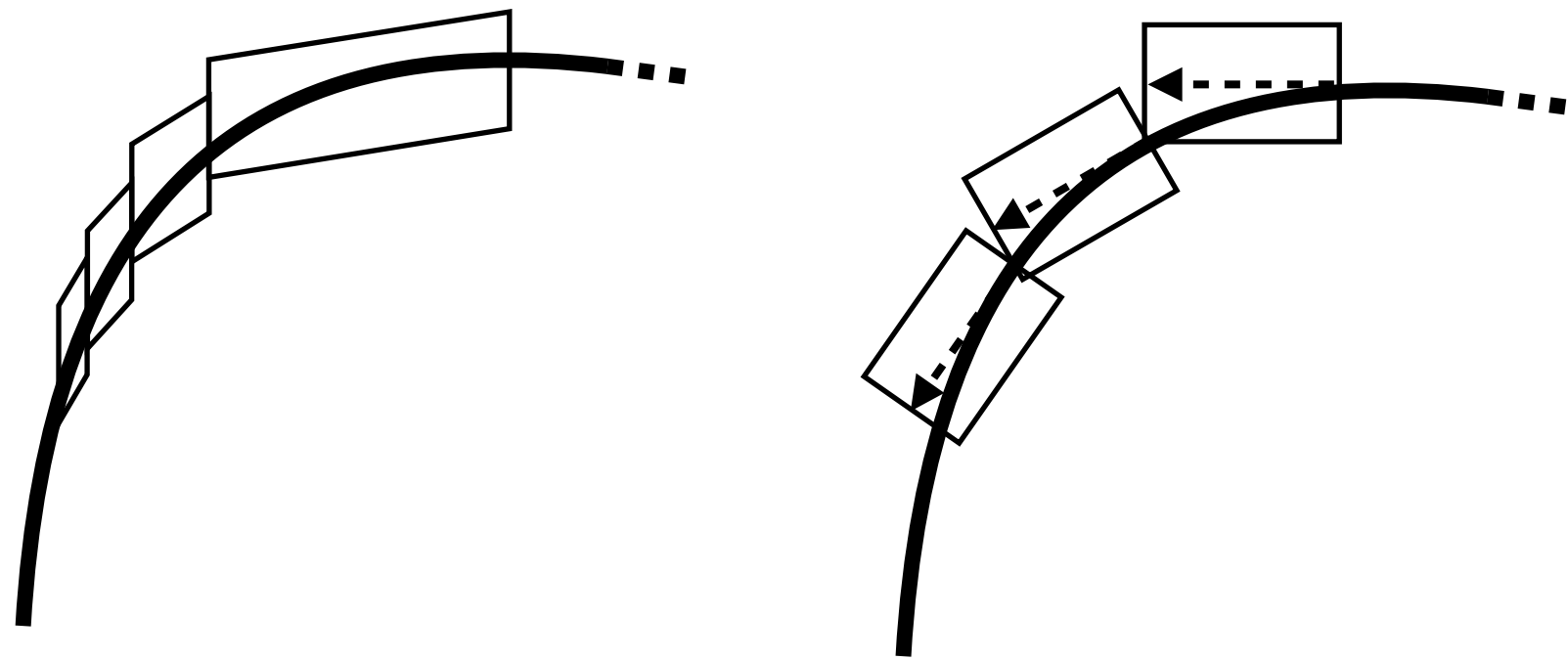
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compute the SVD  $JC(x) = U\Sigma V^\star$  and define  $\hat{x} := V^\star x$  and  $\hat{C}(X) := U^\star C(VX)$ .

# Curve tracking: algorithm overview

Unitary transformation



For effective tracking, apply a **unitary transformation** (rotation) at each step to align the tangent vector vertically.

A unitary transformation is computed via SVD using interval arithmetic.

For a curve  $C := \{c_1, \dots, c_{n-1}\} \subset \mathbb{R}[X_1, \dots, X_n]$  and a point  $x \in \mathbb{R}^n$ ,

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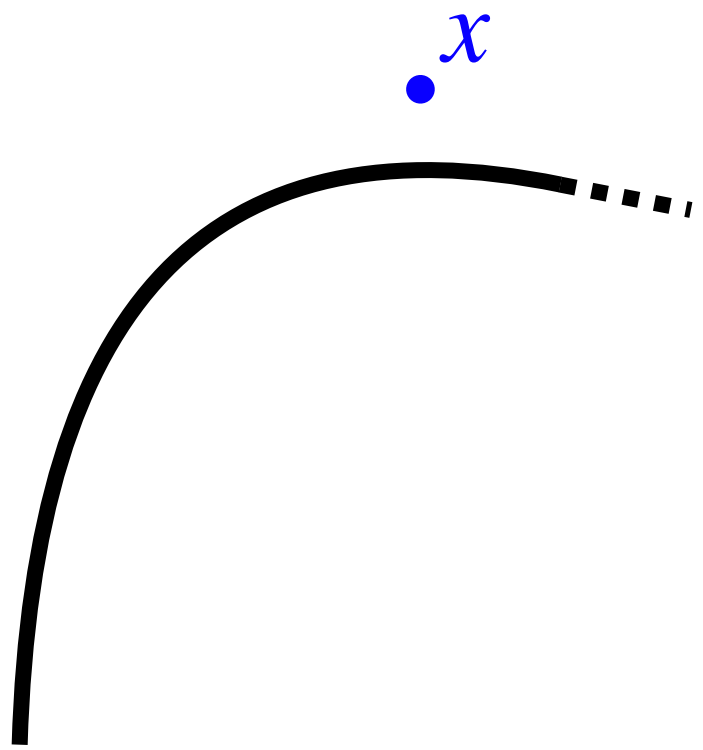
Then,  $J\hat{C}(\hat{x}) = U^\star JC(x)V = U^\star(U\Sigma V^\star)V = \Sigma$ .

Since  $\Sigma$  is an  $(n - 1) \times n$  matrix,  $\ker J\hat{C}(\hat{x}) = \langle e_n \rangle$ .

# Curve tracking: algorithm overview

Refine

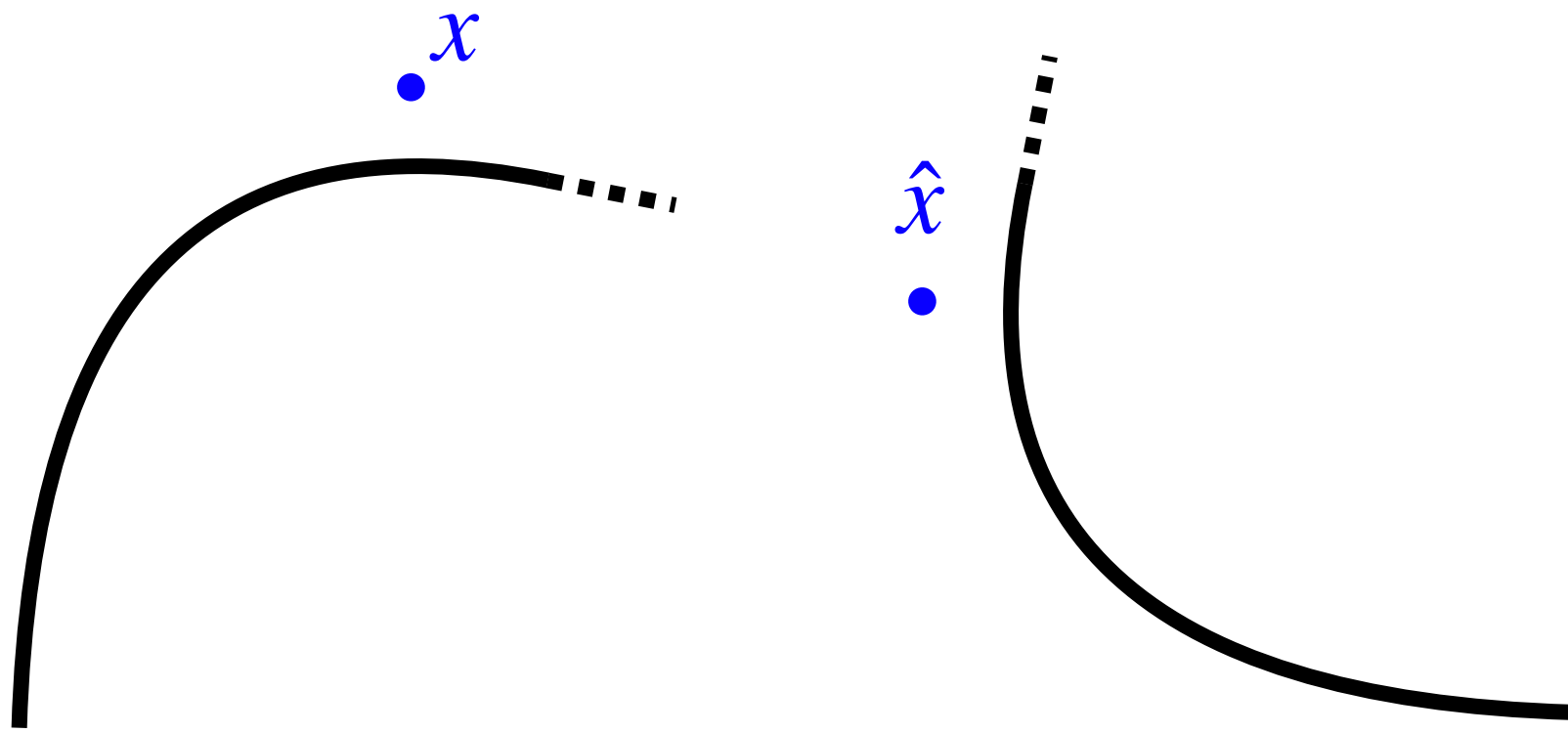
Apply a unitary transformation on  $C$  such that the rotated curve  $\hat{C}$  satisfies  $\ker J\hat{C}(\hat{x}) = \langle e_n \rangle$ .



# Curve tracking: algorithm overview

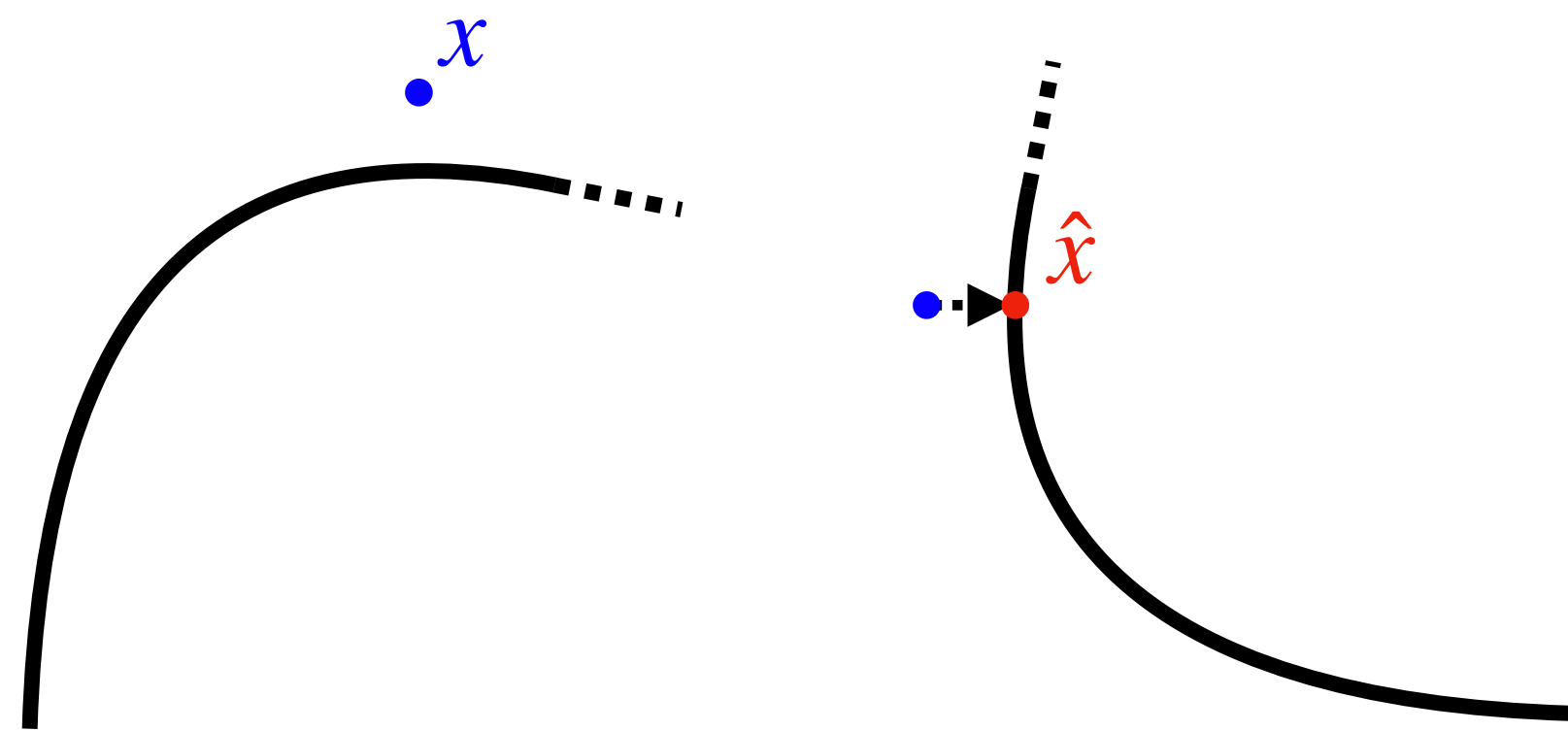
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# Curve tracking: algorithm overview

Refine



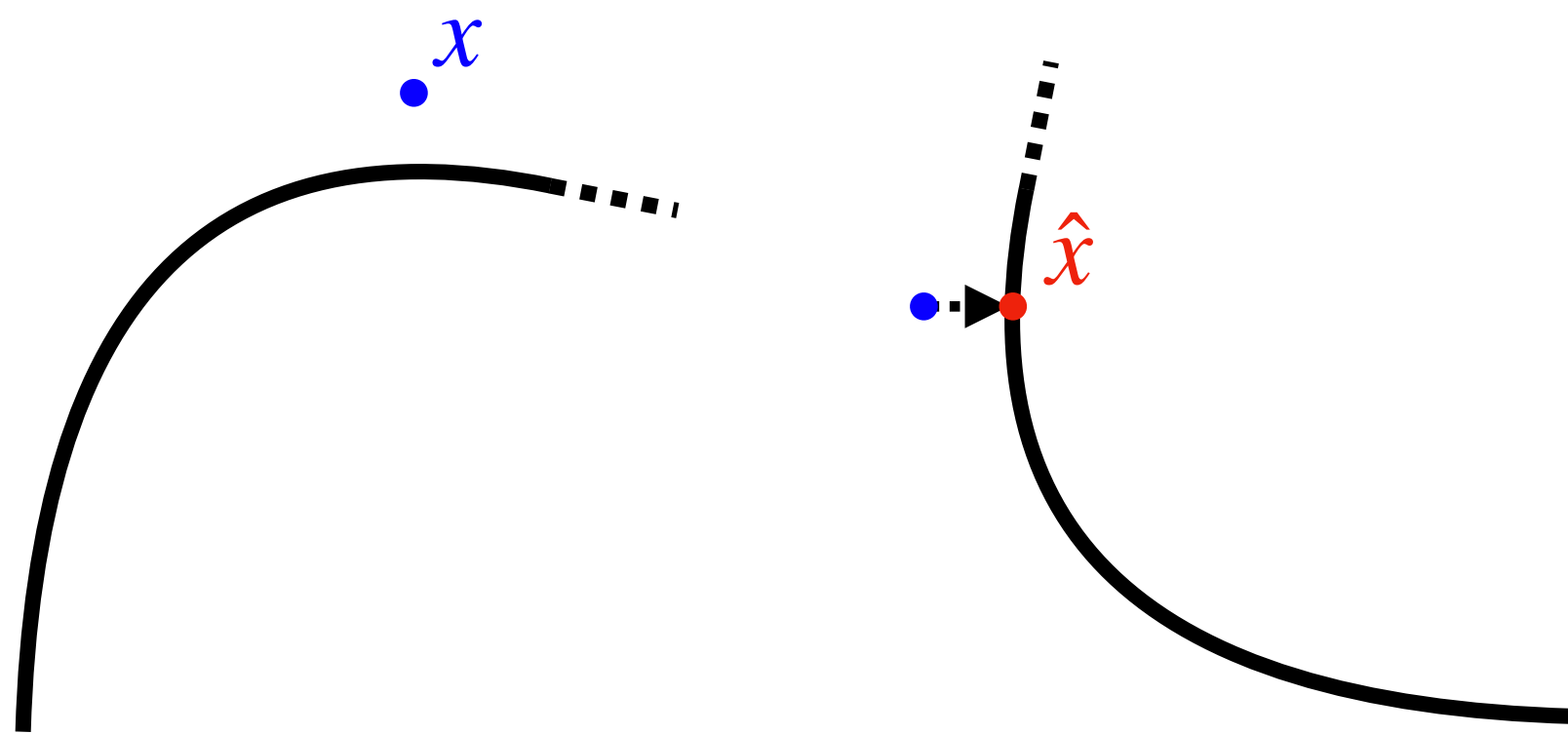
Apply a unitary transformation on  $C$  such that the rotated curve  $\hat{C}$  satisfies  $\ker J\hat{C}(\hat{x}) = \langle e_n \rangle$ .

Refine  $\hat{x}$  using the **Krawczyk method** to get a “nice” approximation ( $\rho$ -approximate solution) to  $\hat{C}$ .



# Curve tracking: algorithm overview

Refine



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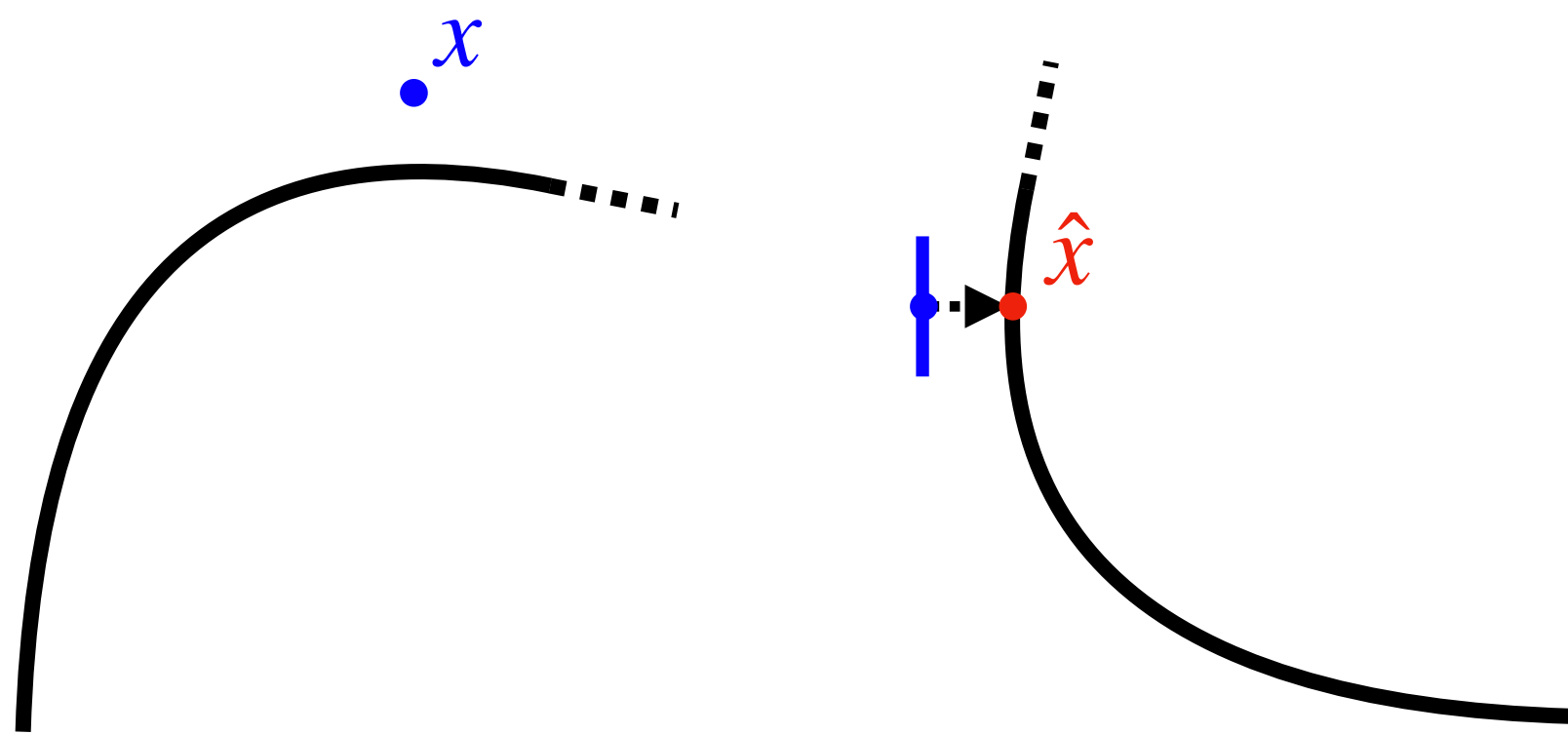
Refine  $\hat{x}$  using the **Krawczyk method** to get a "nice" approximation ( $\rho$ -approximate solution) to  $\hat{C}$ .

Refine does two things:

- Obtains a good approximation to track.
- Computes a radius  $r$  for the Krawczyk method.

# Curve tracking: algorithm overview

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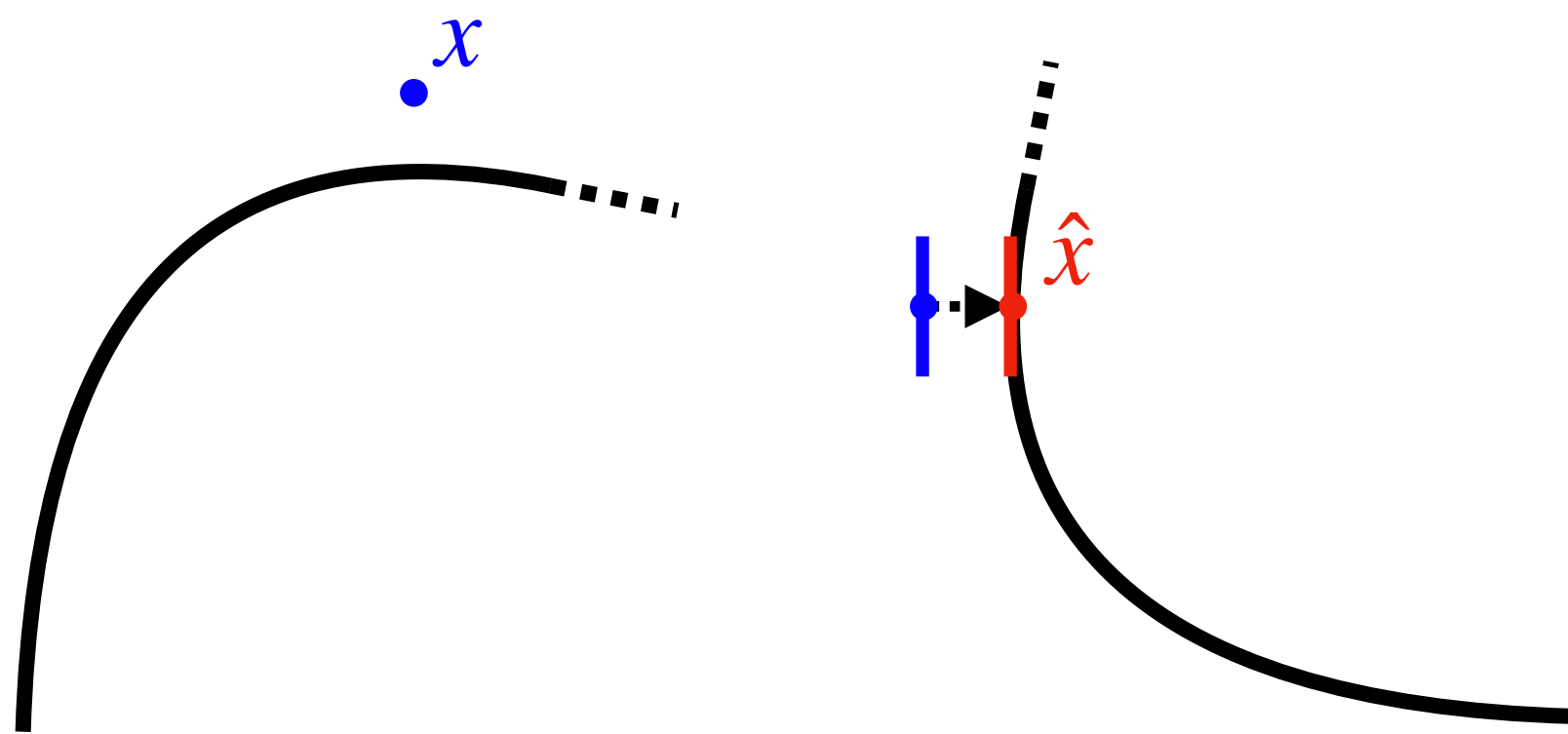
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Caveat: We should refine  $x + (\mathbf{0}_{n-1} \times r[-1,1])$ .

# Curve tracking: algorithm overview

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# Curve tracking: algorithm overview

Curve predictor

Starting from a “nice” approximation,  
compute a predictor for the curve.

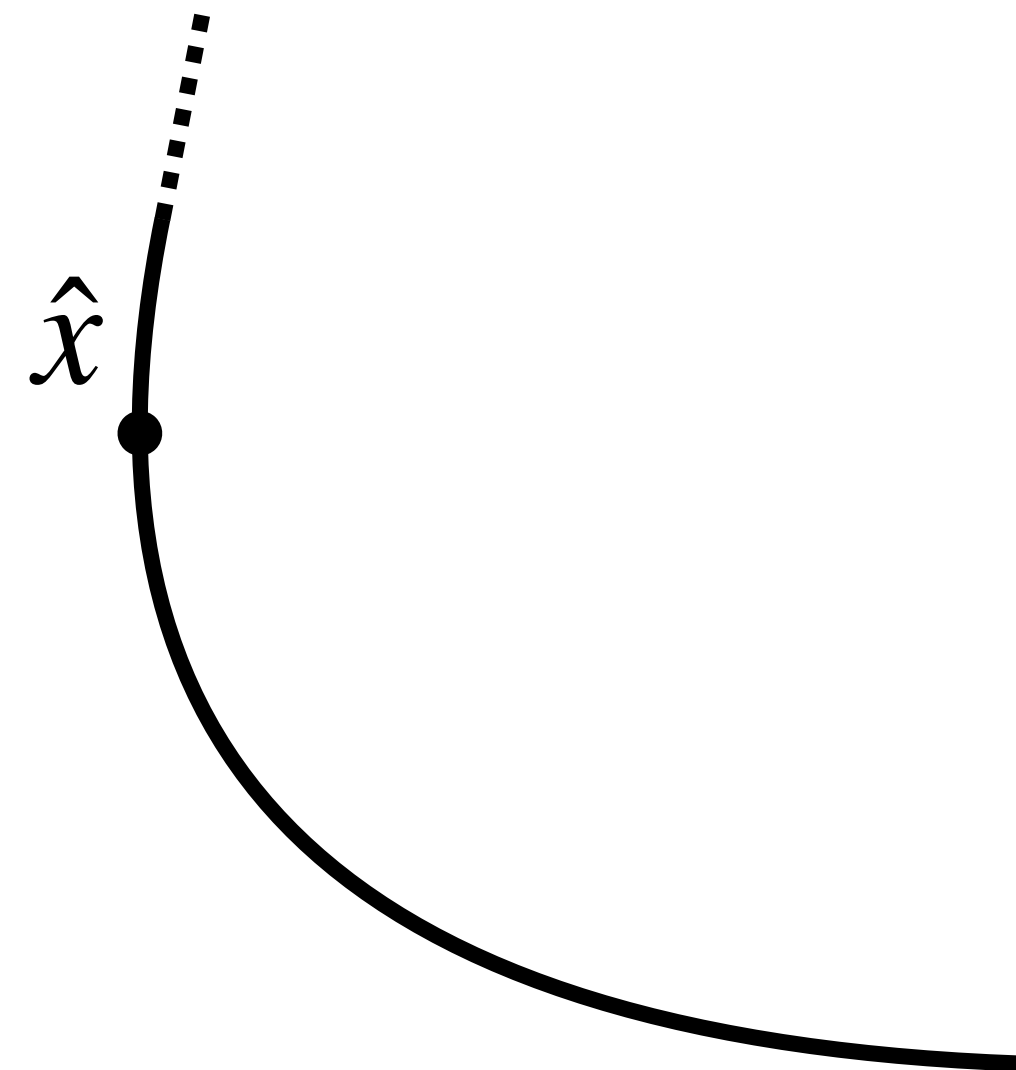
Construct an interval tube (i.e. track the curve)  
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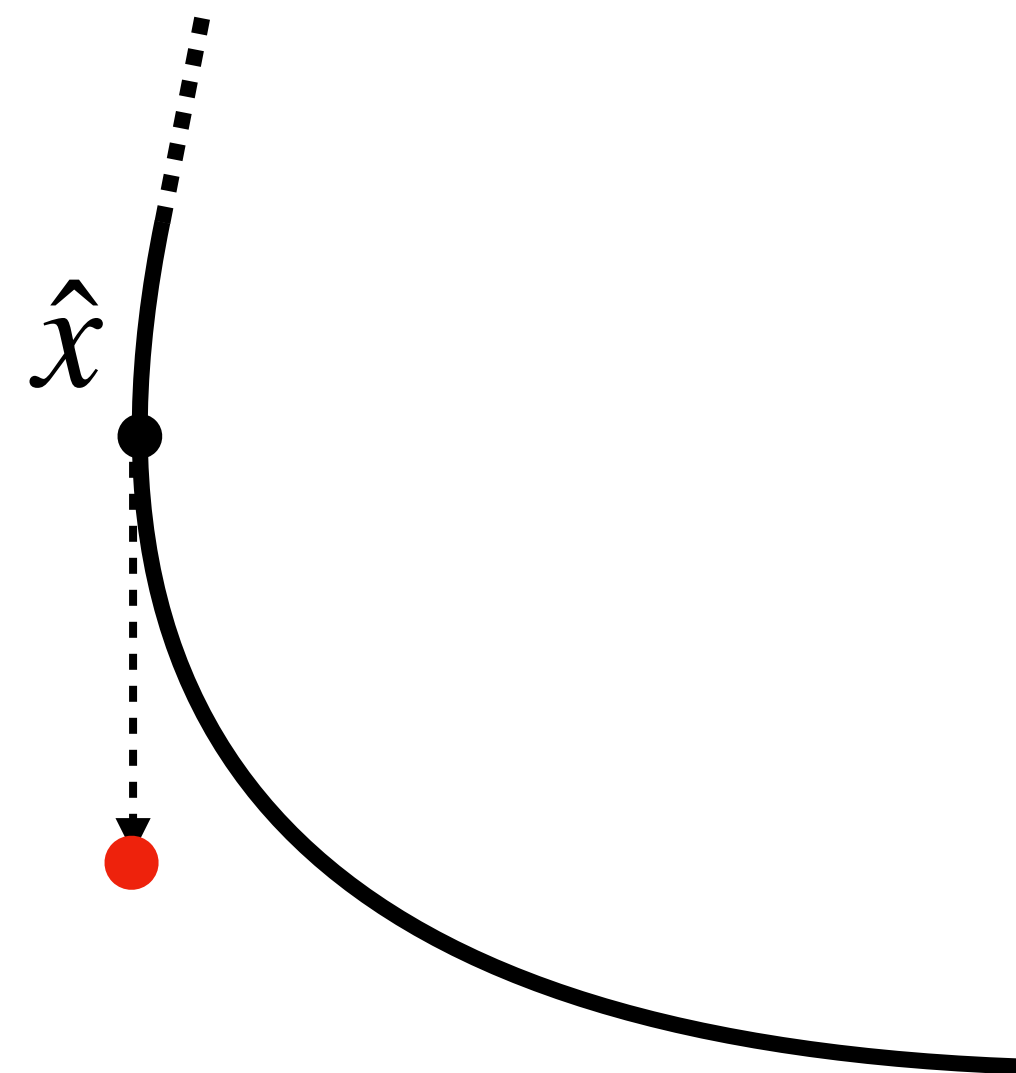


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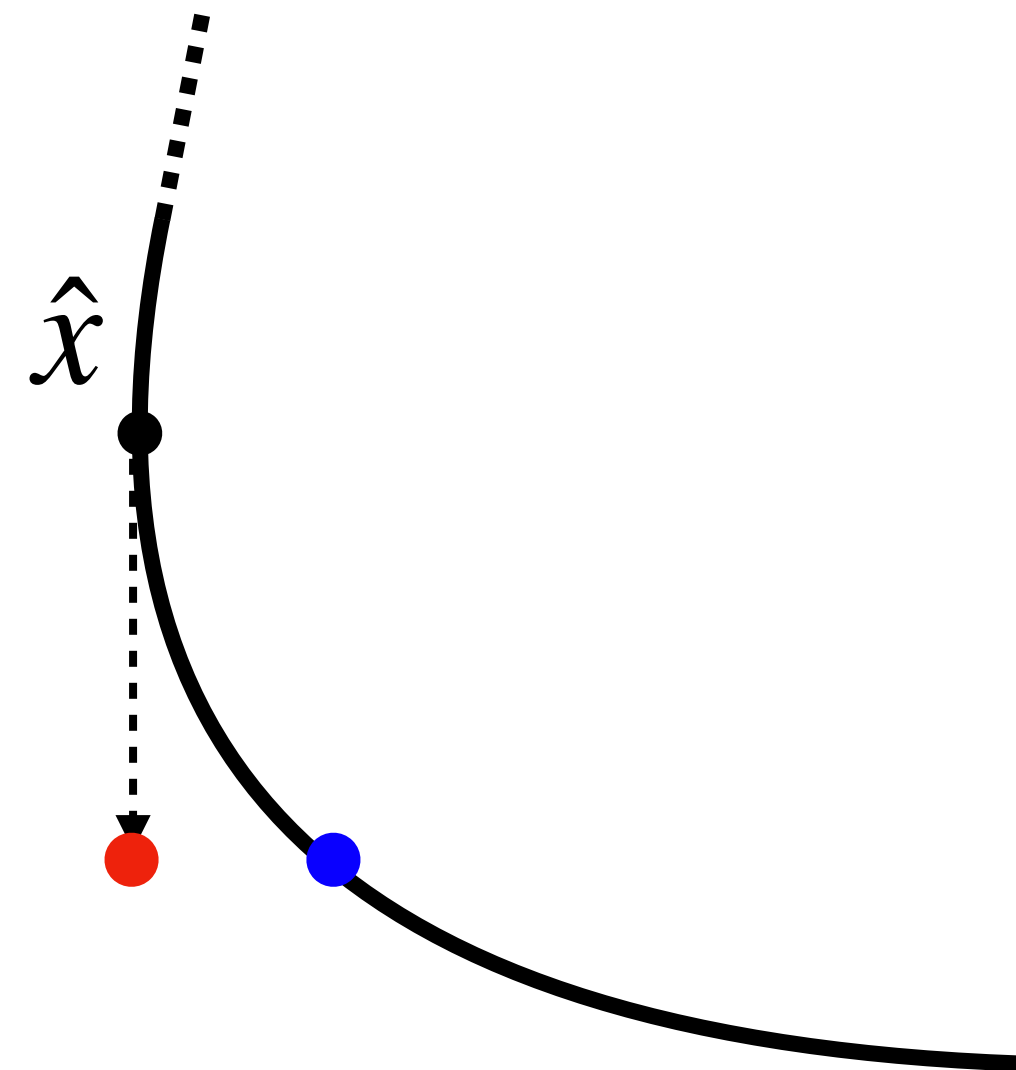


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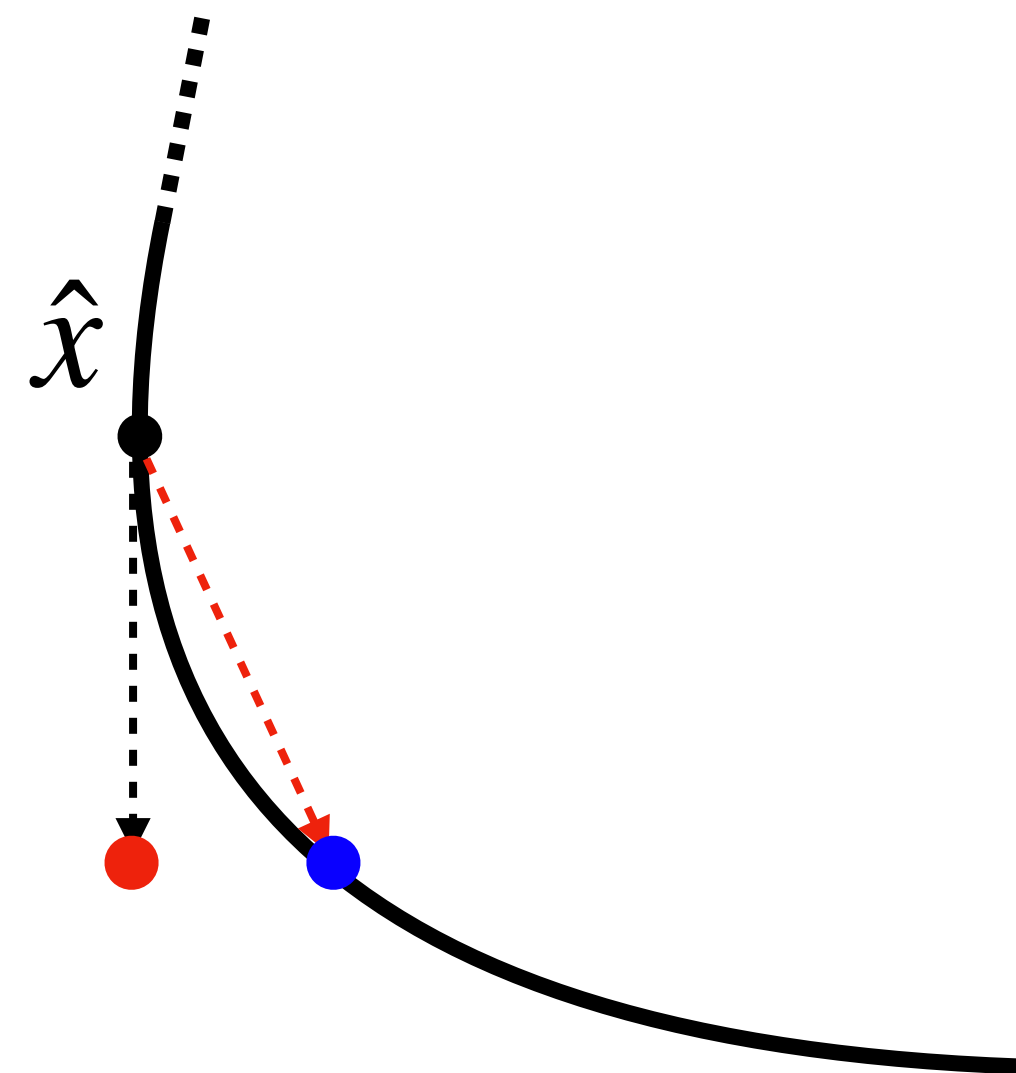


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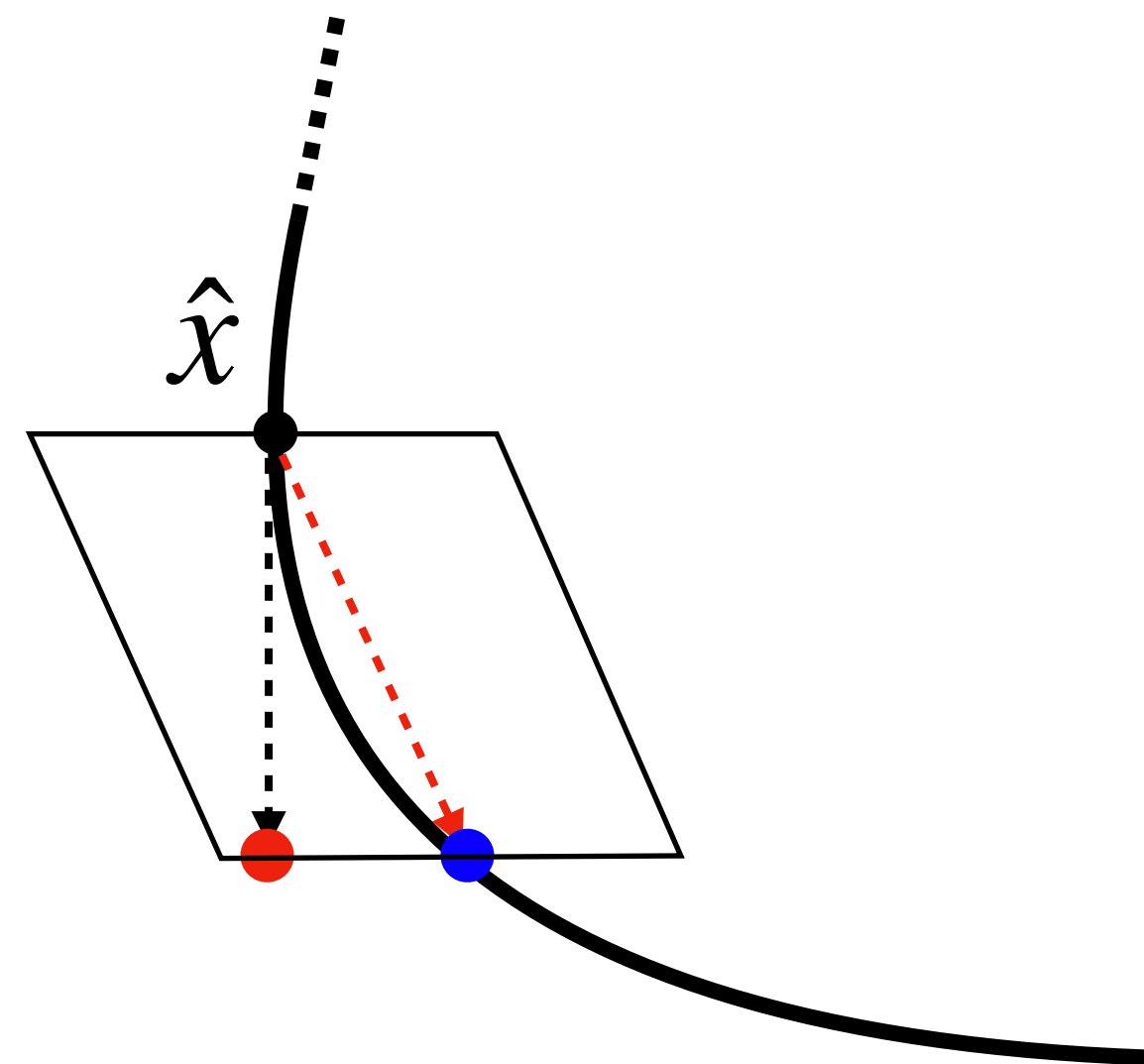


# Curve tracking: algorithm overview

Curve predictor

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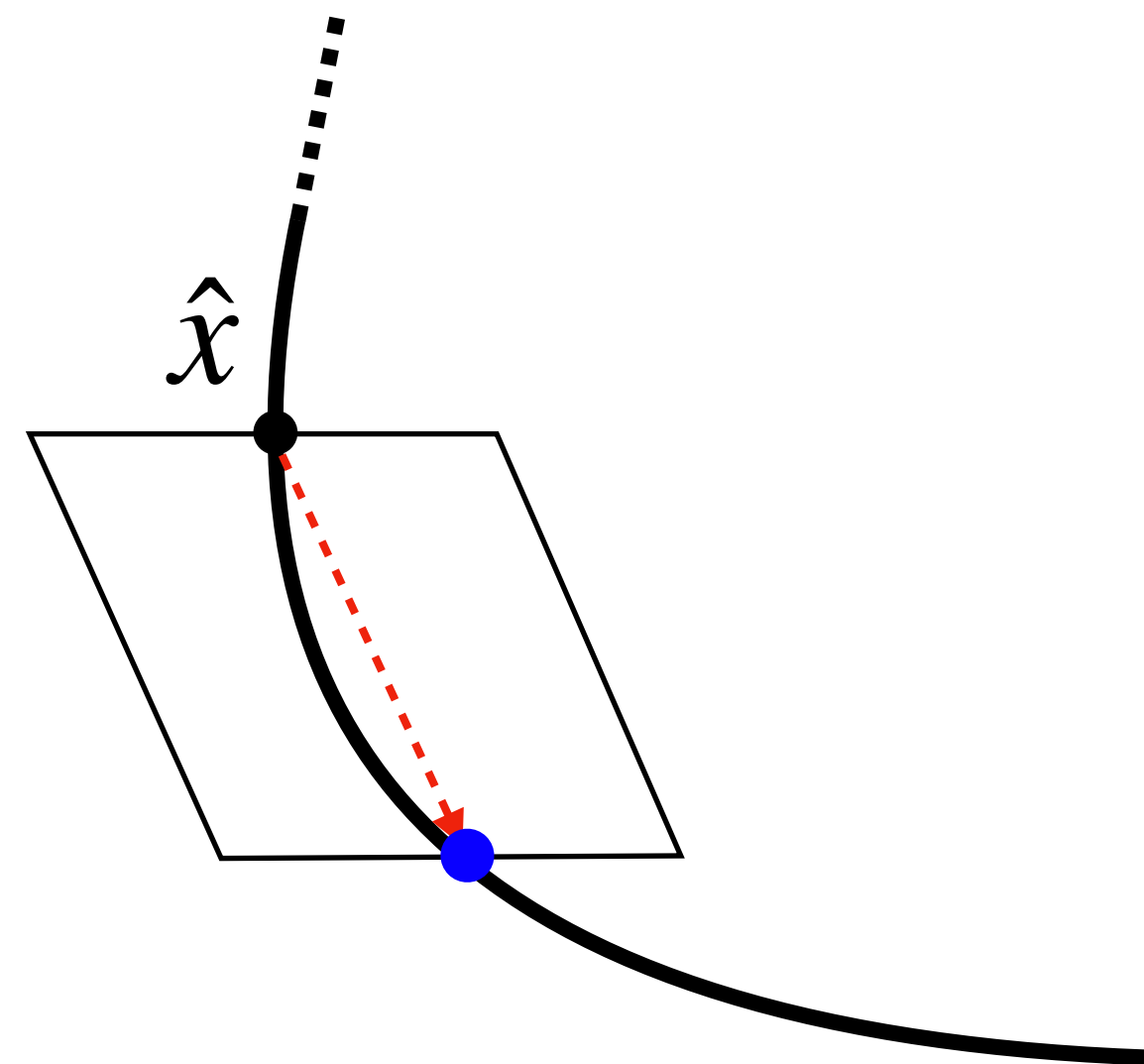
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# Curve tracking: algorithm overview

Curve predictor

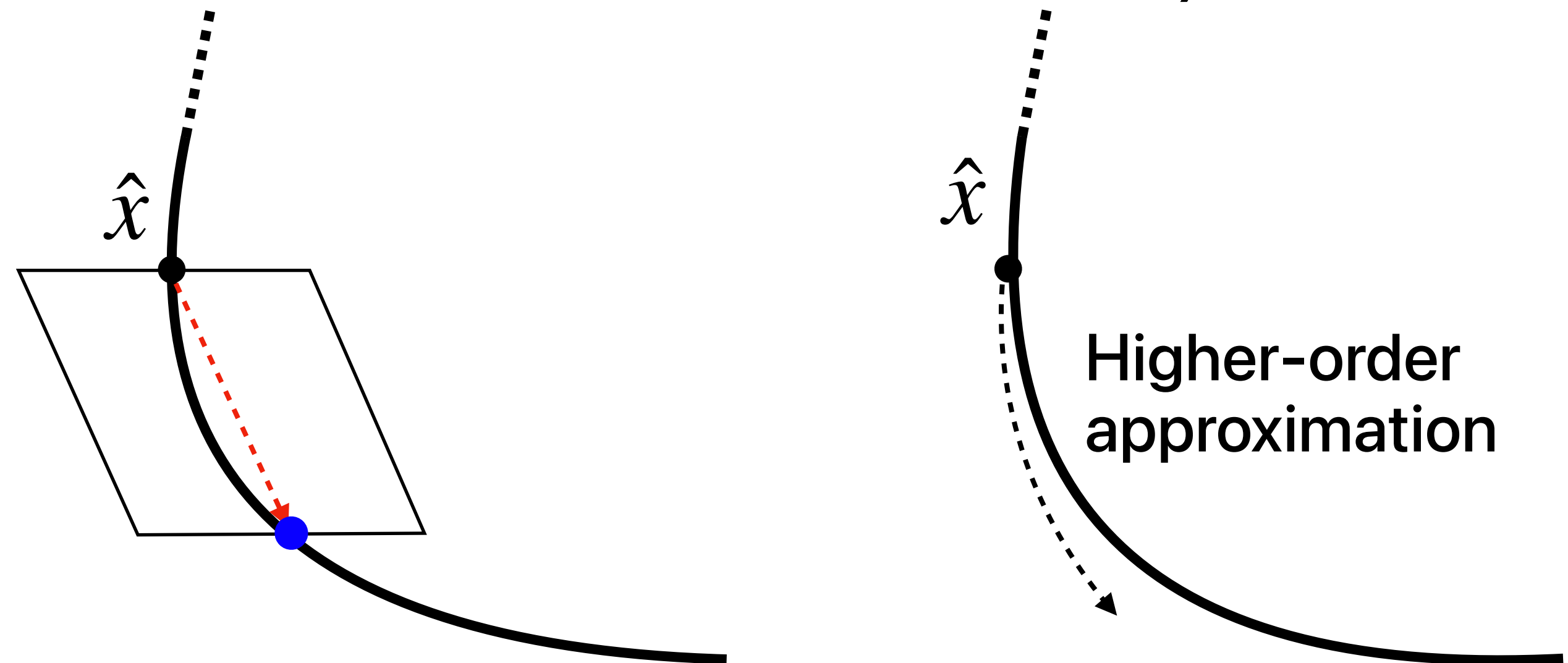
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# Curve tracking: algorithm overview

Curve predictor

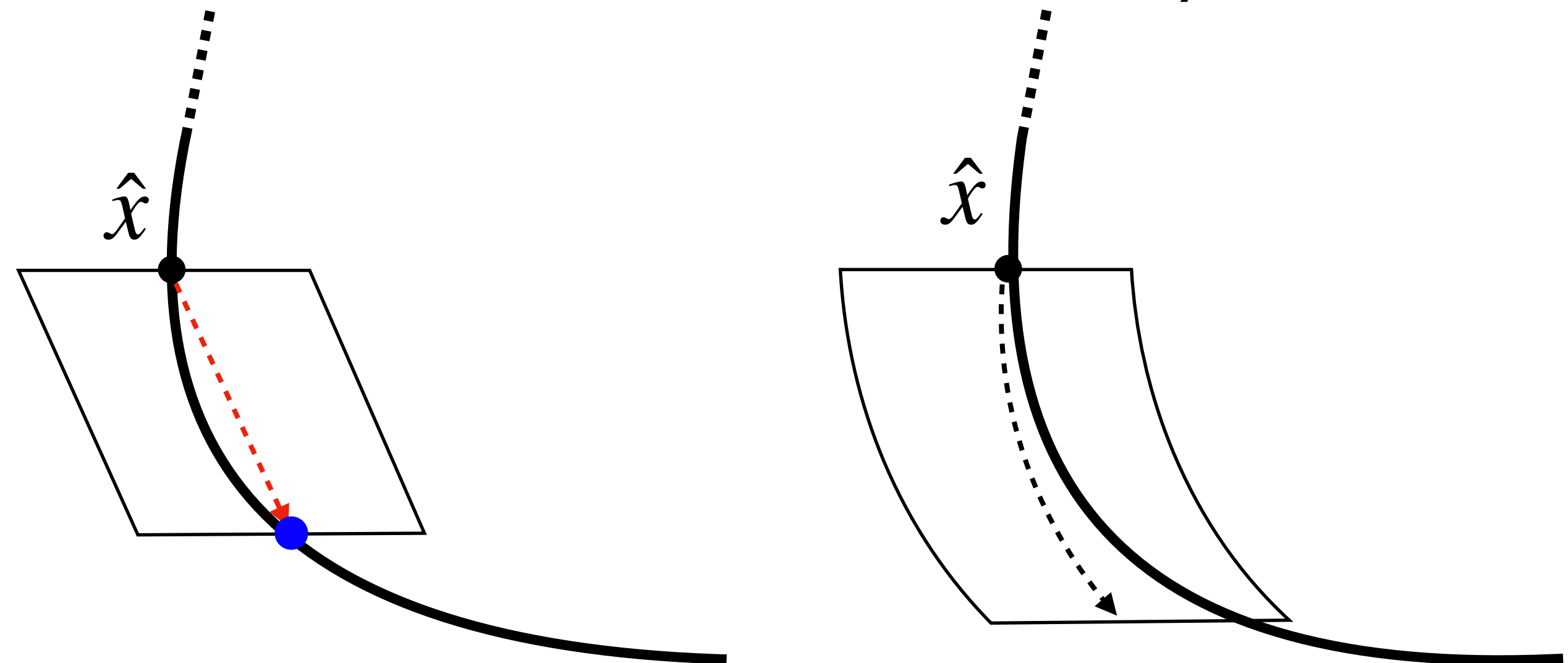
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# Curve tracking: algorithm overview

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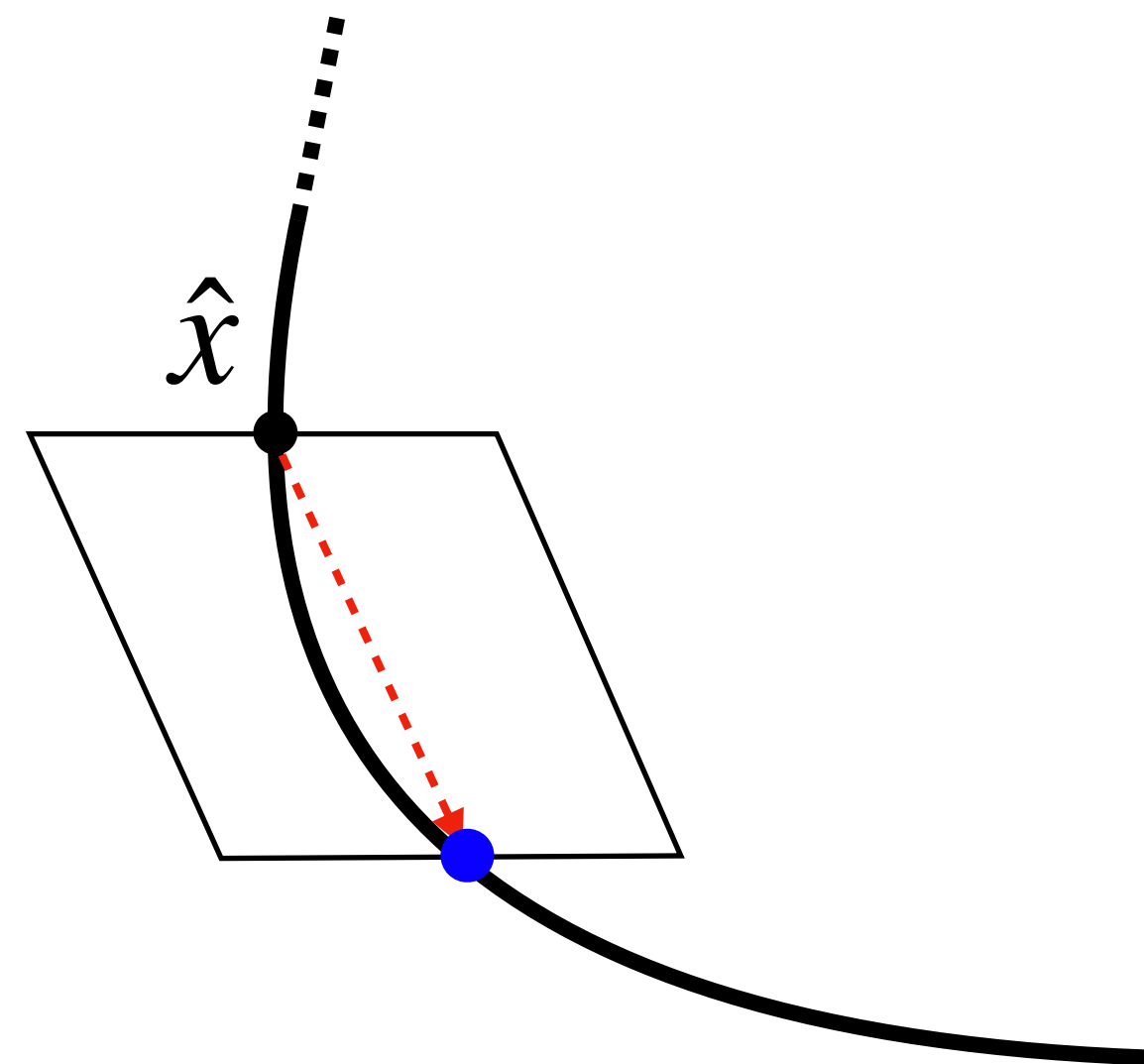
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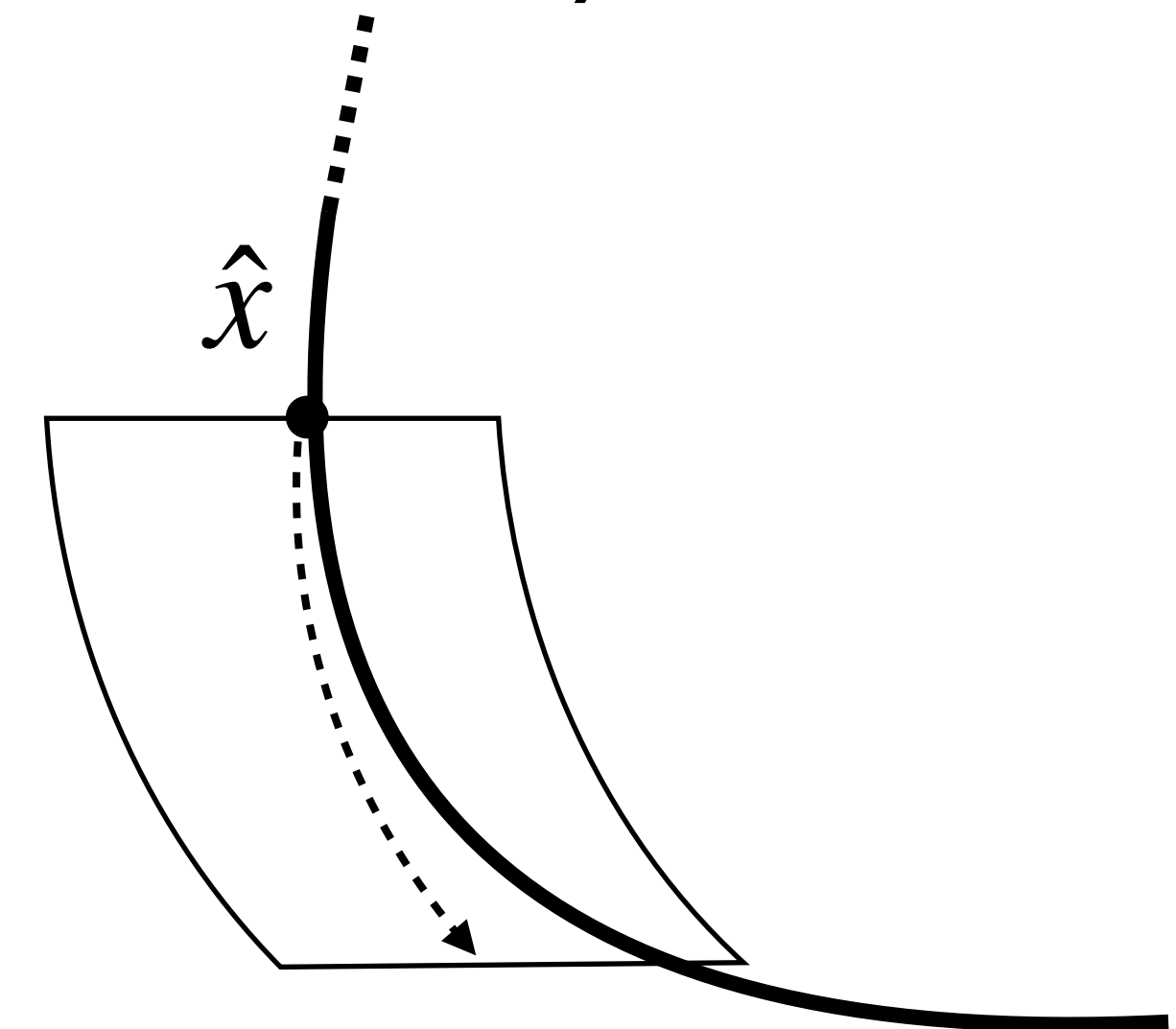
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Duff-Lee 2024

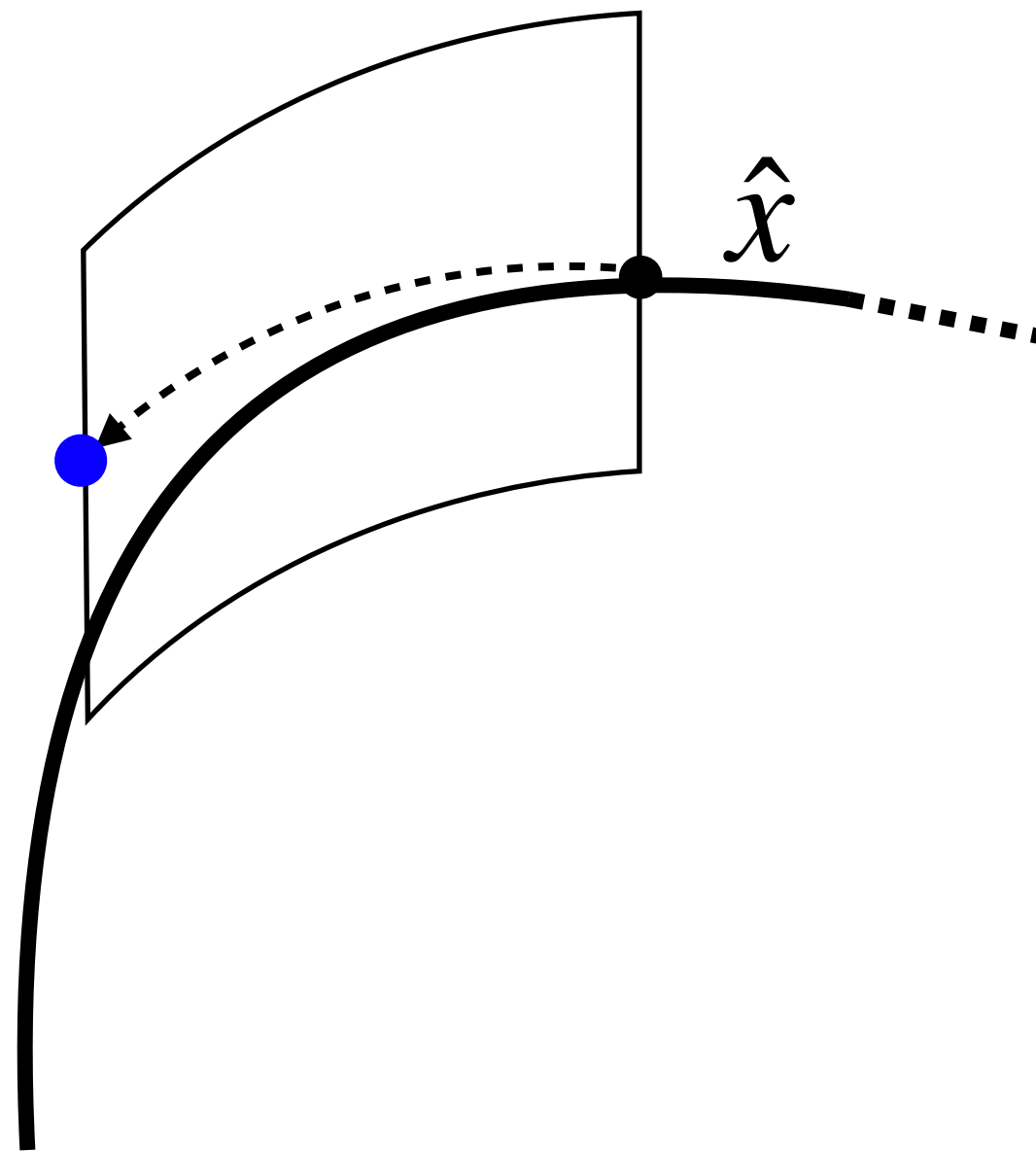


Guillemot-Lairez 2024

# Curve tracking: algorithm overview

Curve predictor

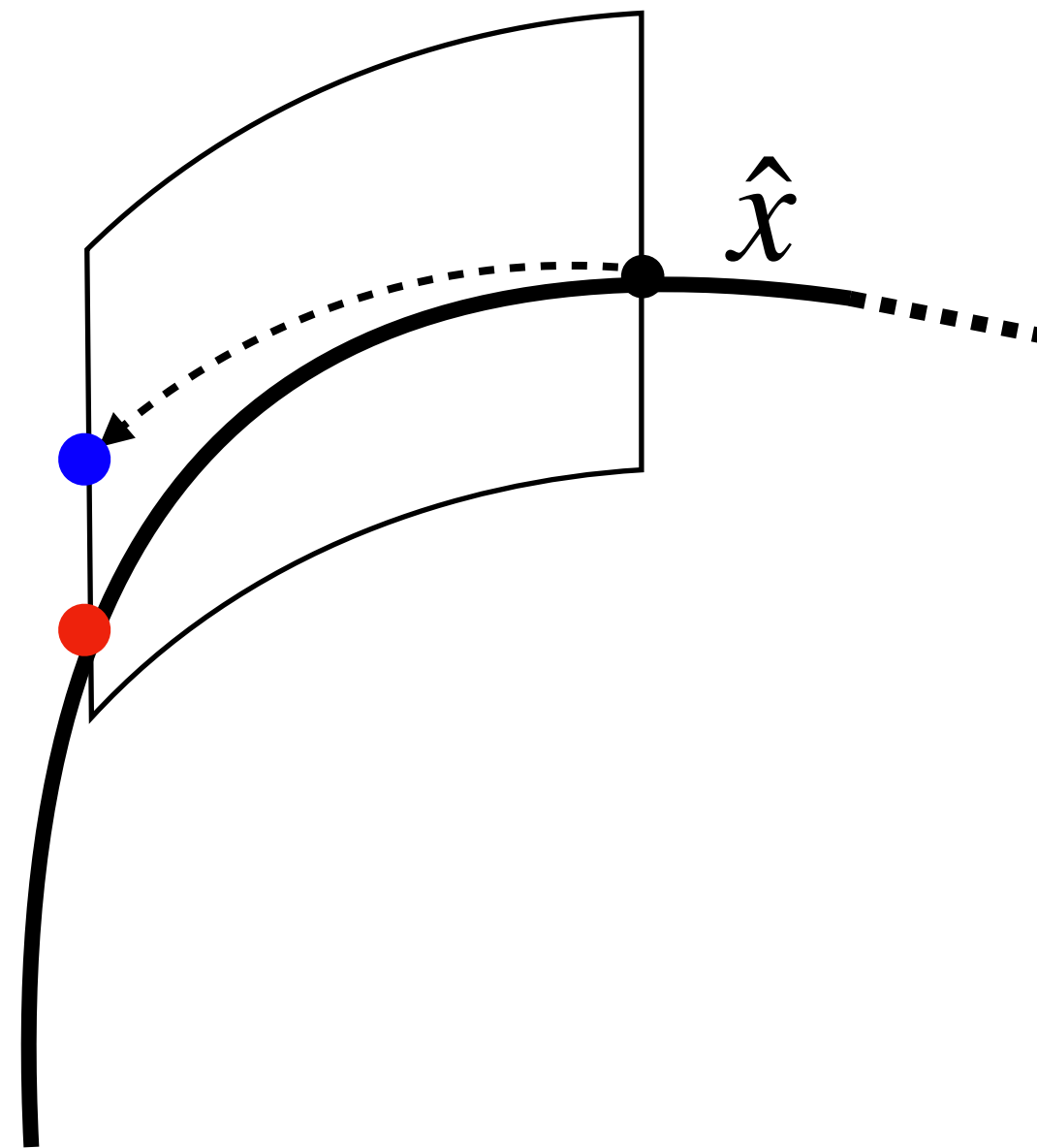
Rotate the curve back (from  $\hat{C}$  to  $C$ ) and refine an endpoint to a  $\rho$ -approximate solution to  $C$ .



# Curve tracking: algorithm overview

Curve predictor

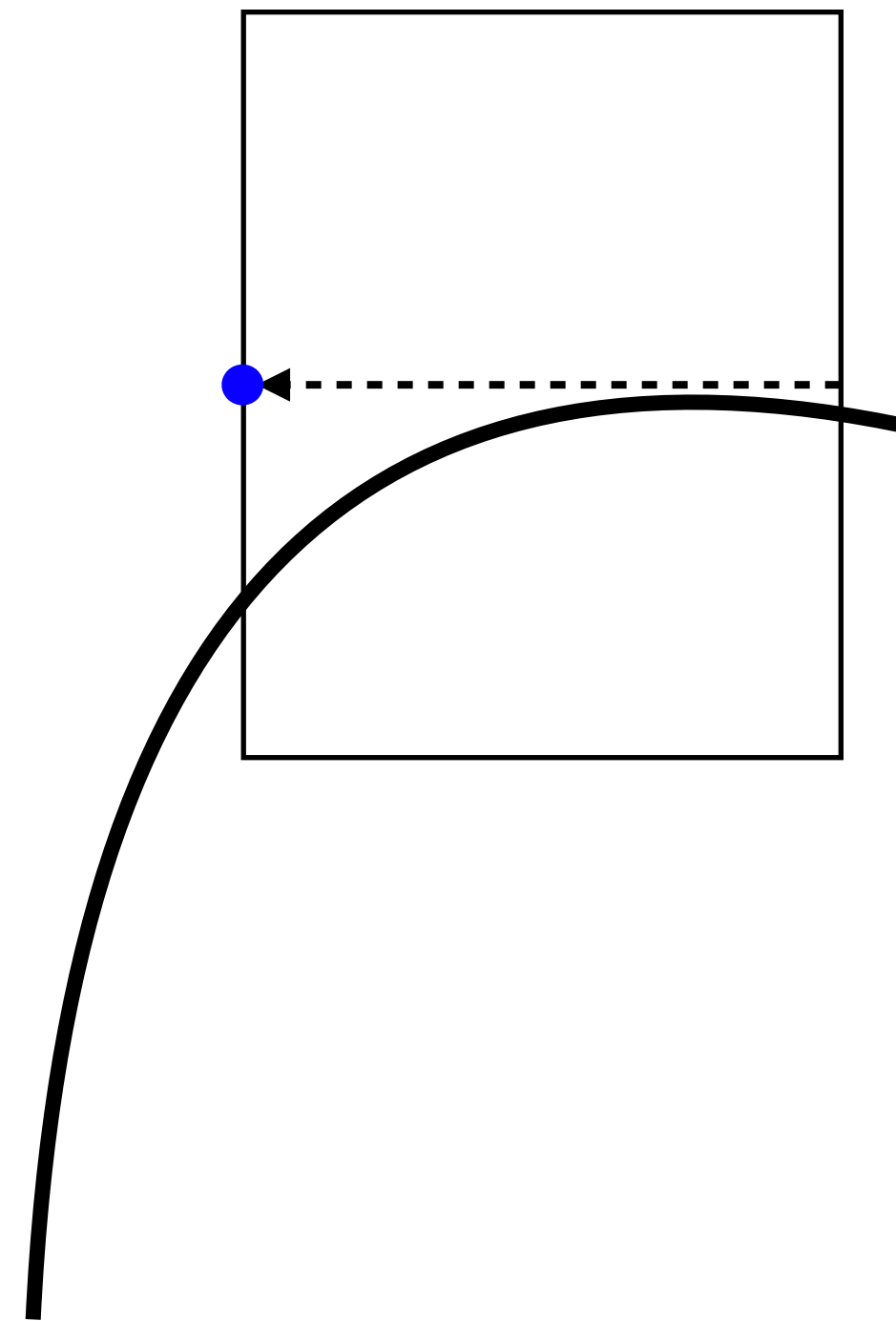
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# Curve tracking: algorithm overview

Caveats in refinement

Two possible bad scenarios:

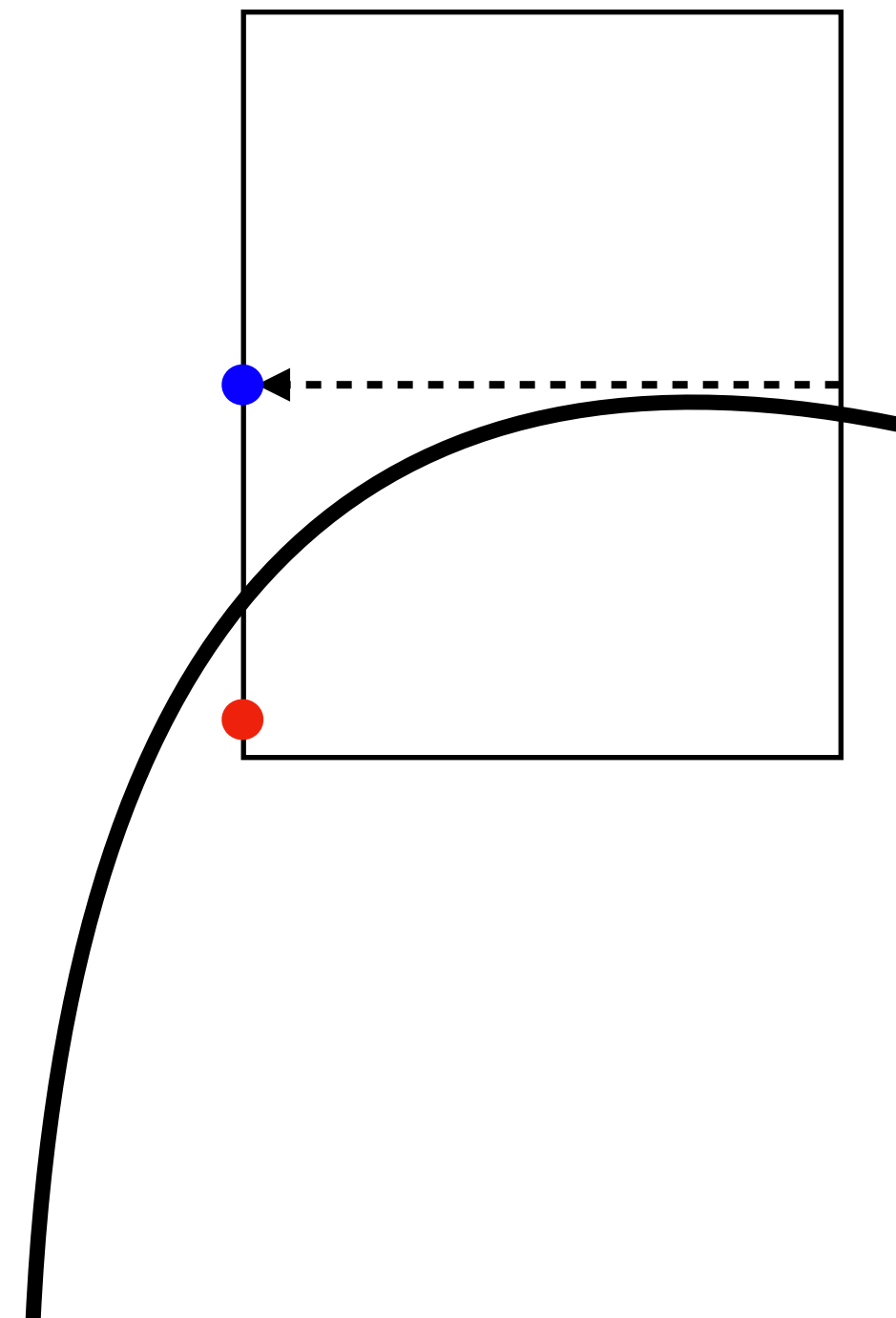




# Curve tracking: algorithm overview

Caveats in refinement

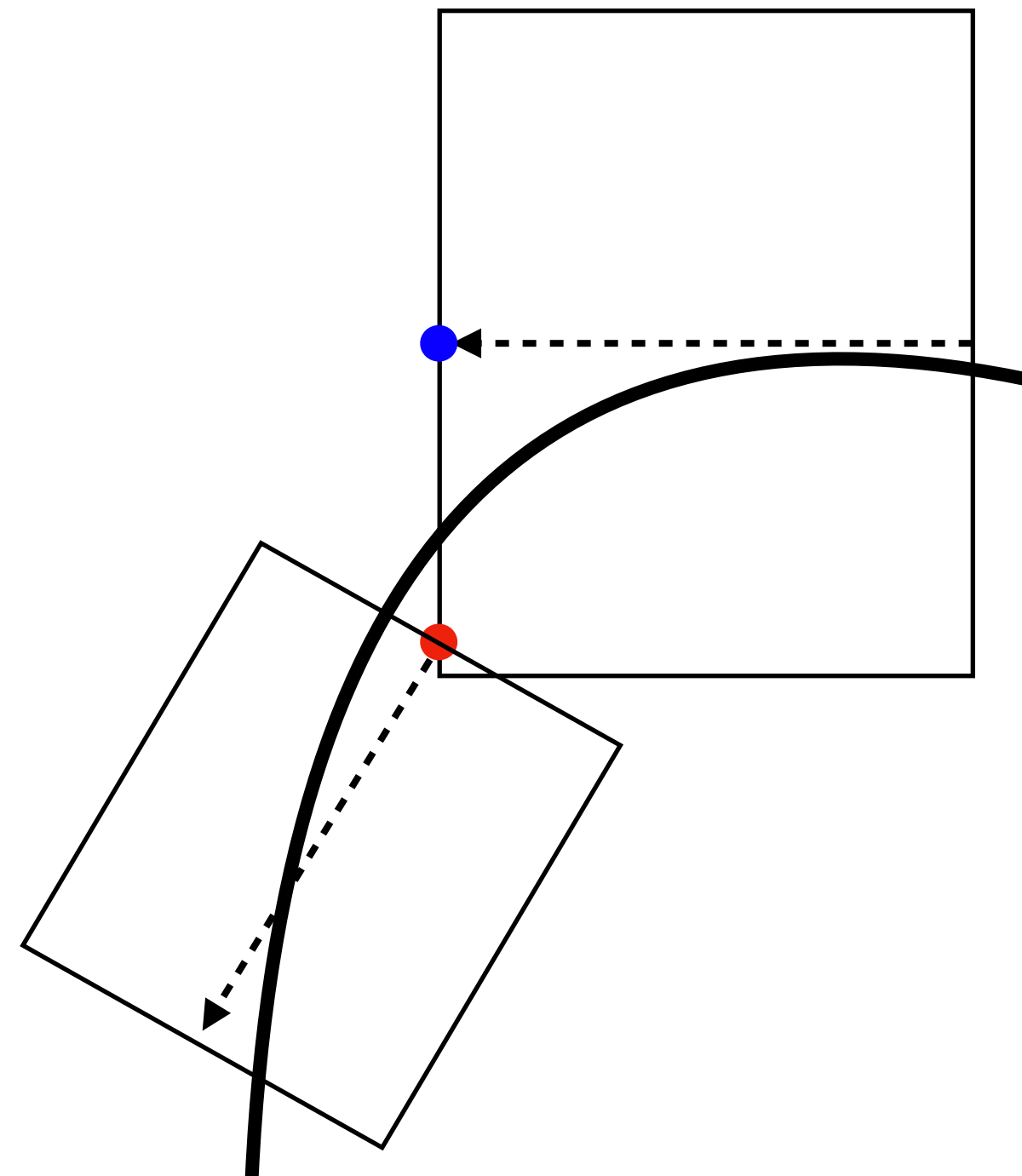
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# Curve tracking: algorithm overview

Caveats in refinement

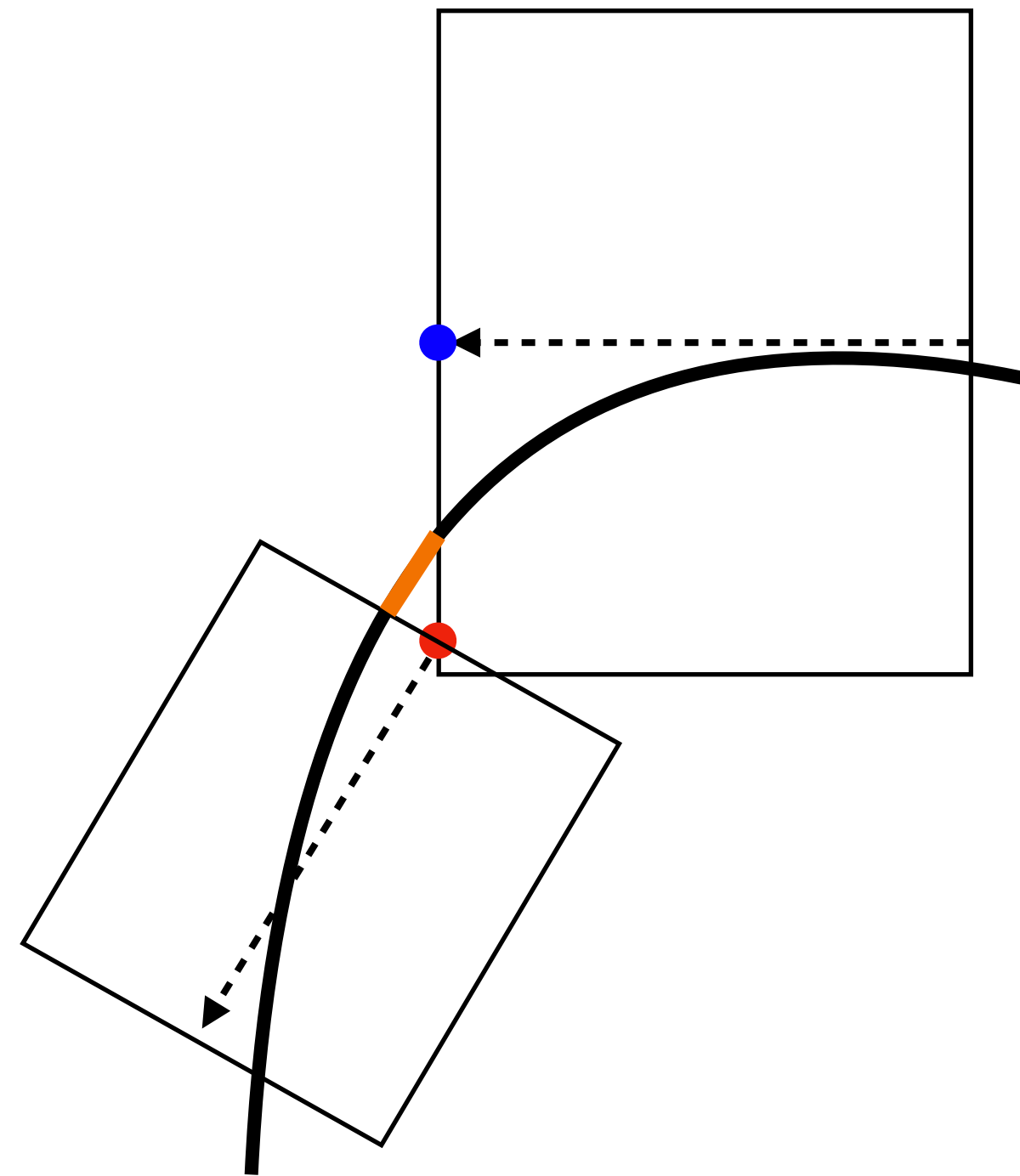
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# Curve tracking: algorithm overview

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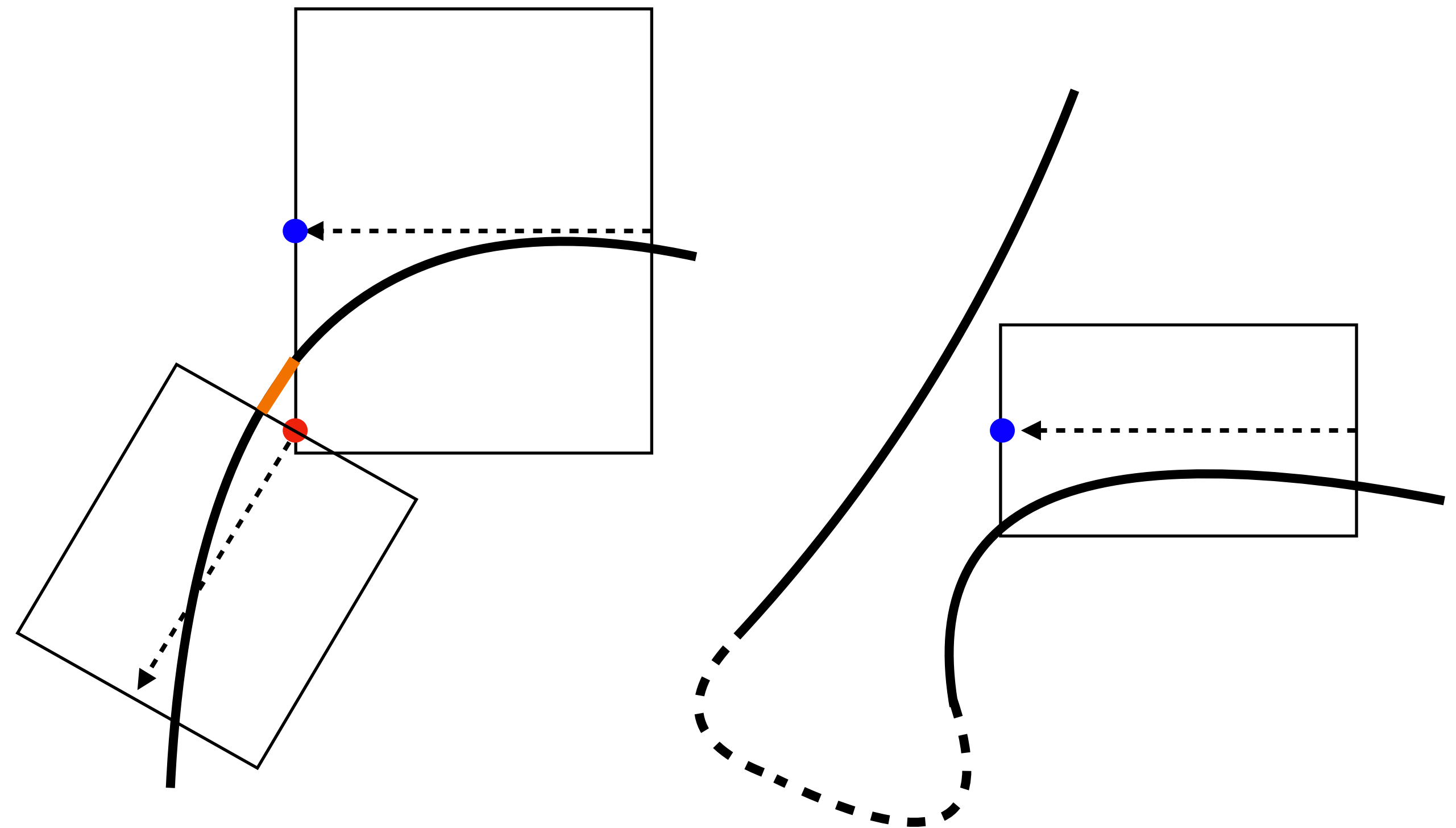
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# Curve tracking: algorithm overview

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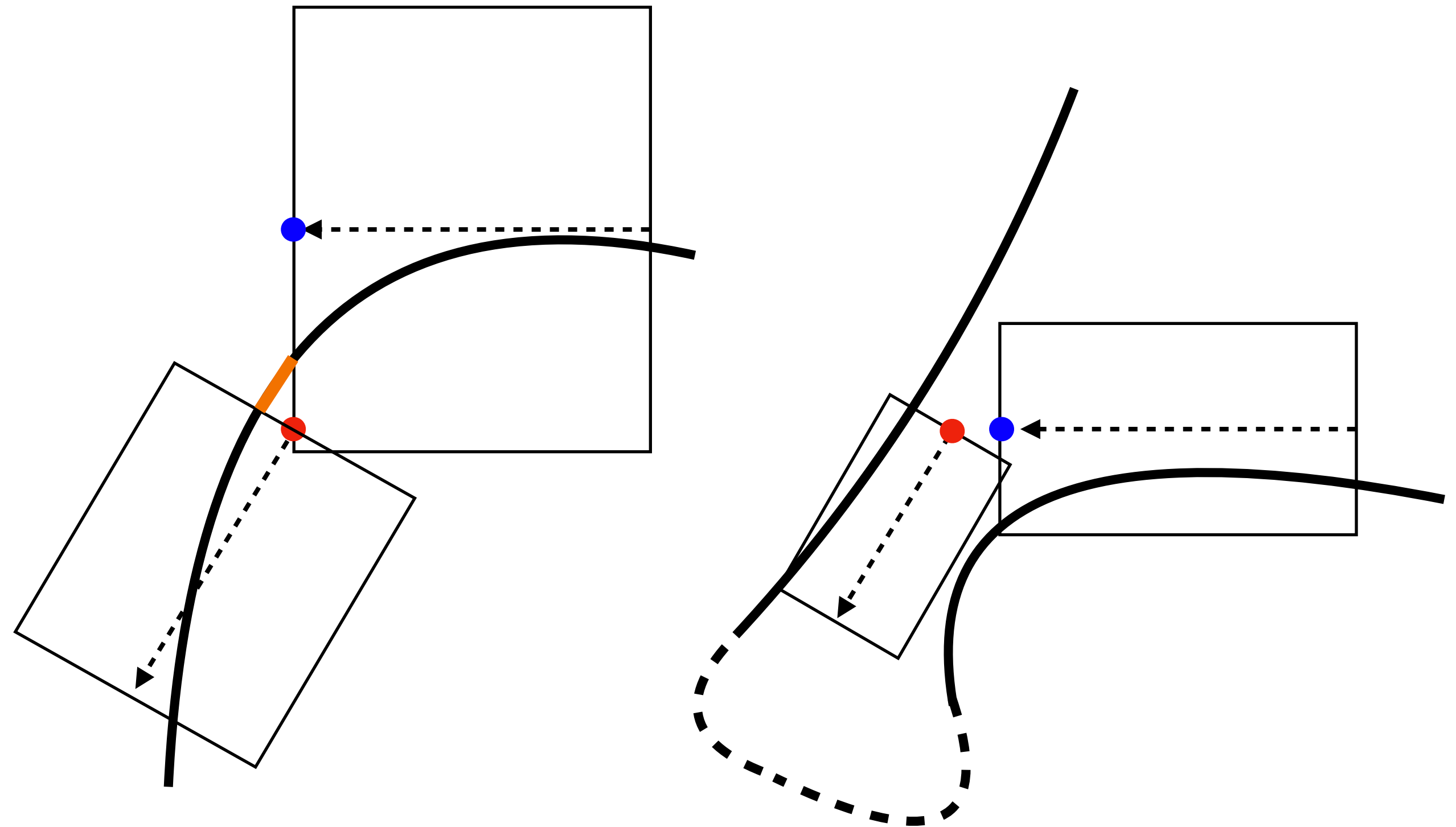
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# Curve tracking: algorithm overview

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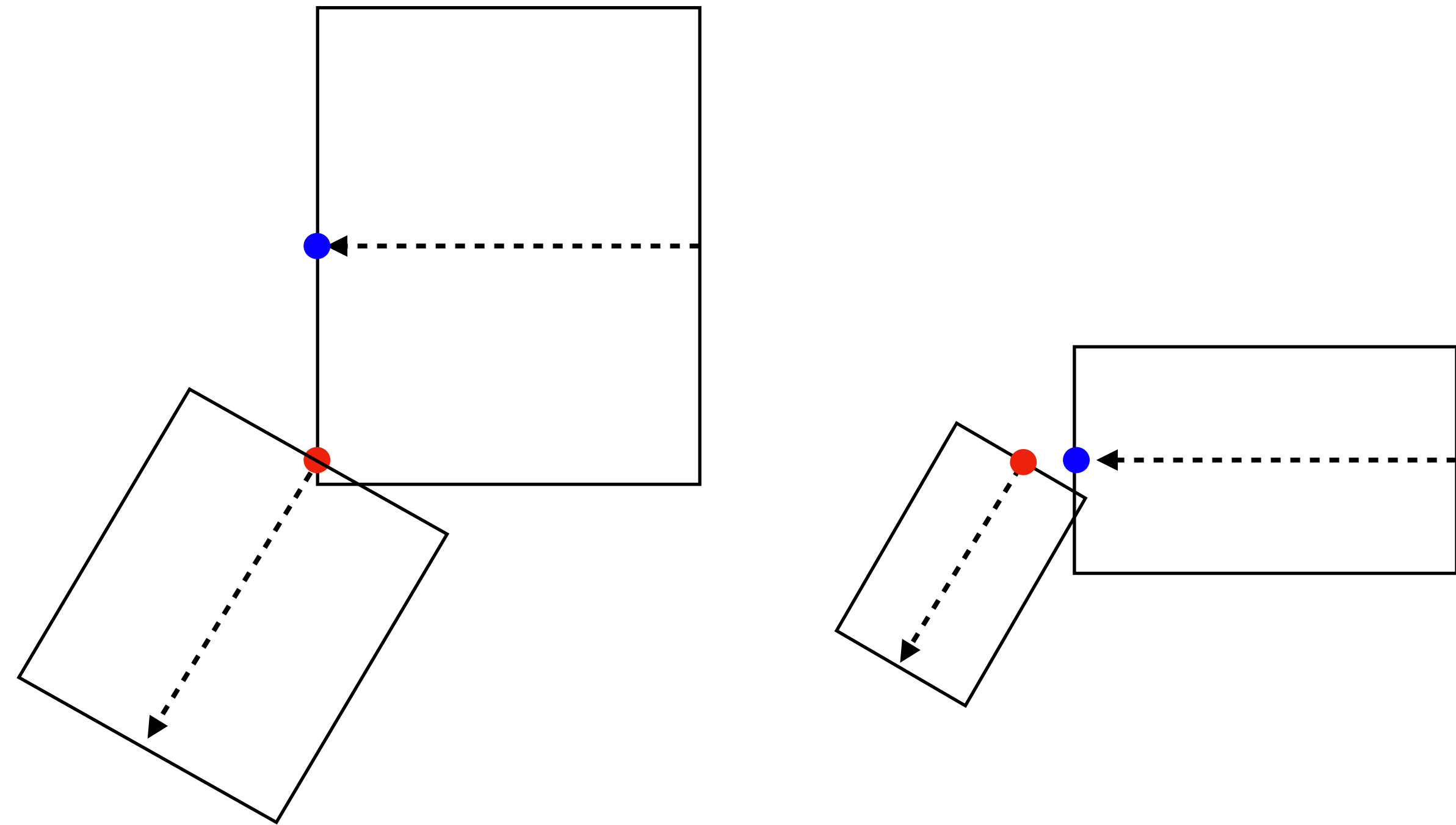
Two possible bad scenarios:



# Curve tracking: algorithm overview

Caveats in refinement

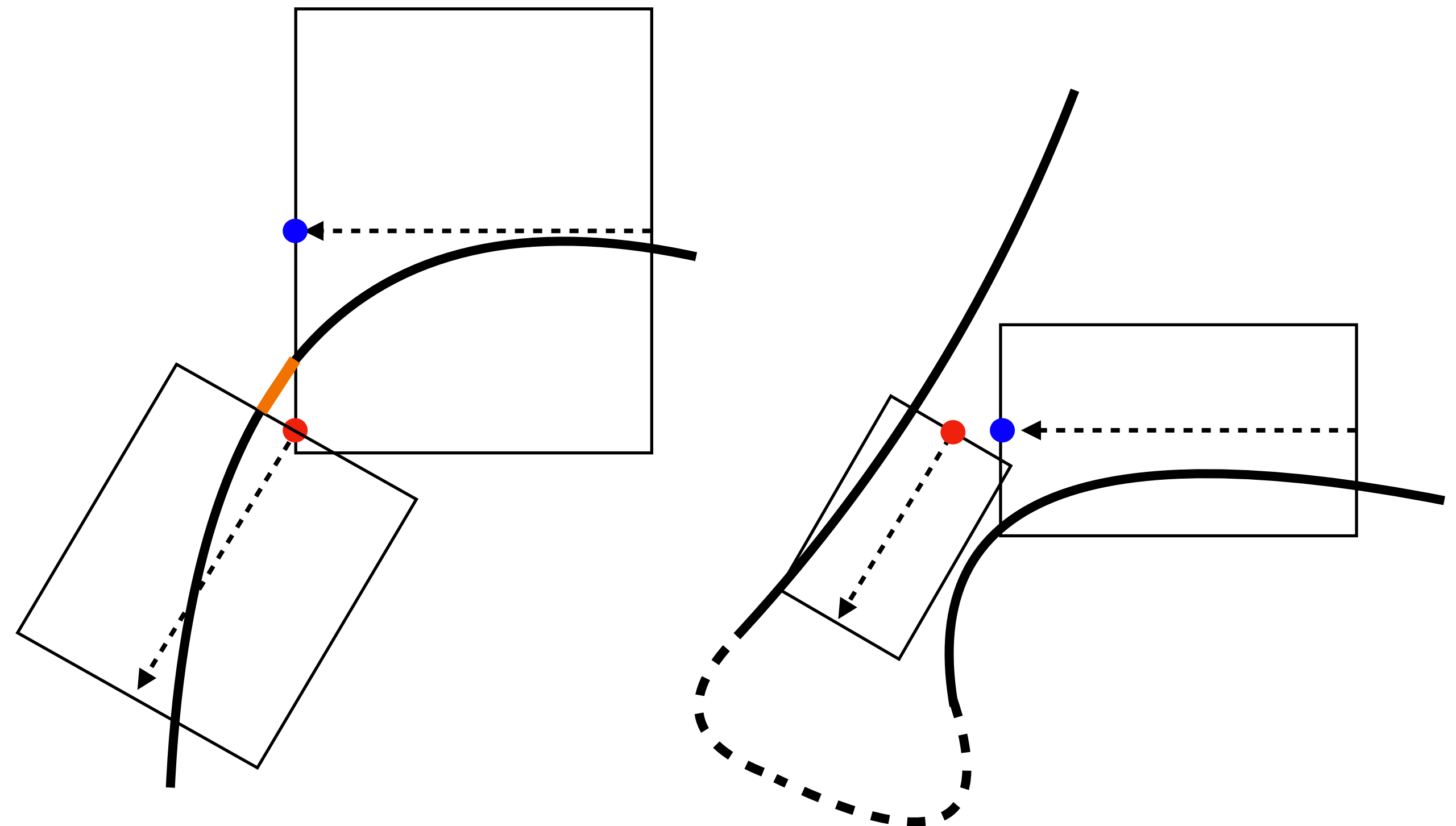
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# Curve tracking: algorithm overview

Caveats in refinement

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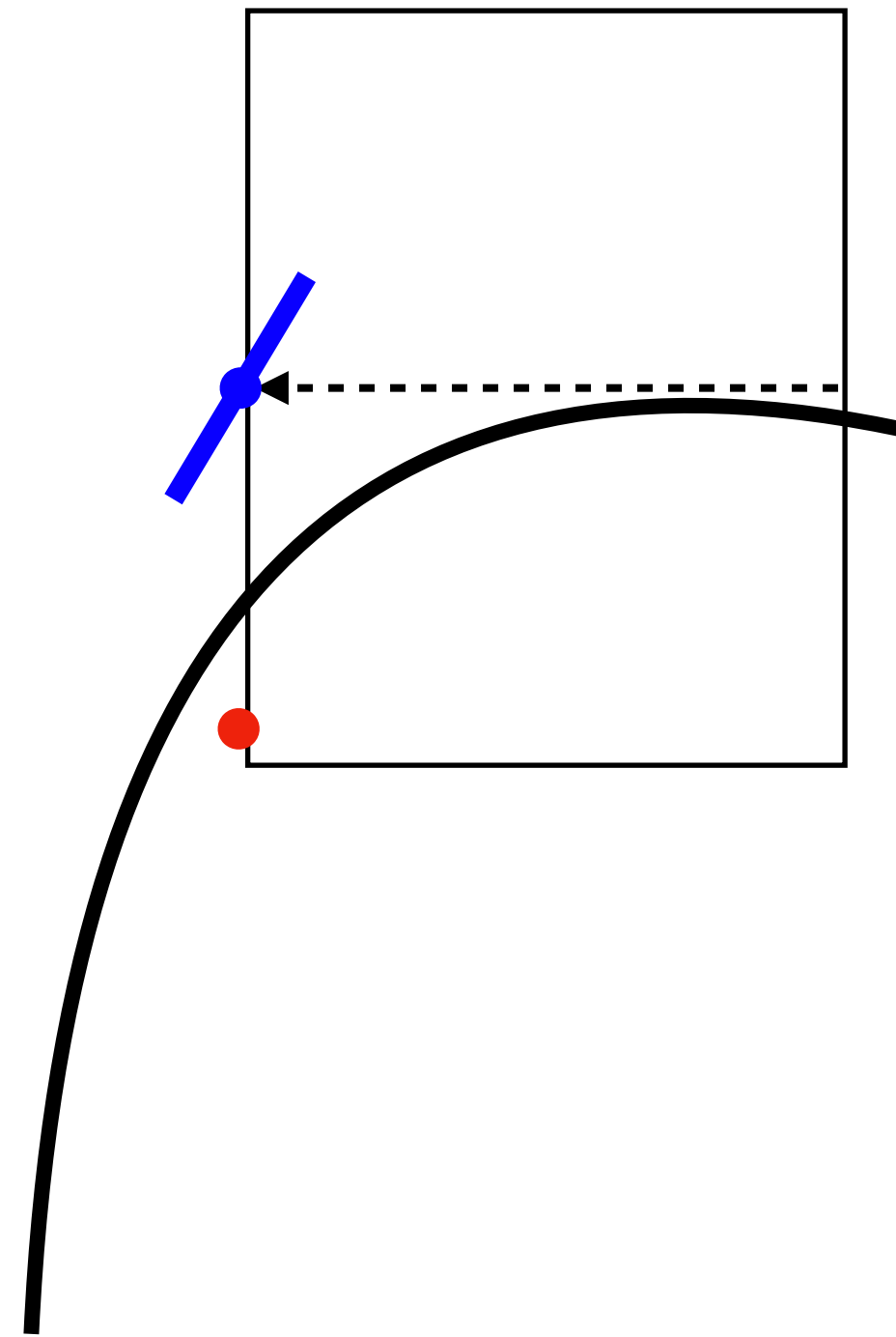


# Curve tracking: algorithm overview

Caveats in refinement

Two possible bad scenarios:

Refine  $x + (\mathbf{0}_{n-1} \times r[-1,1])$



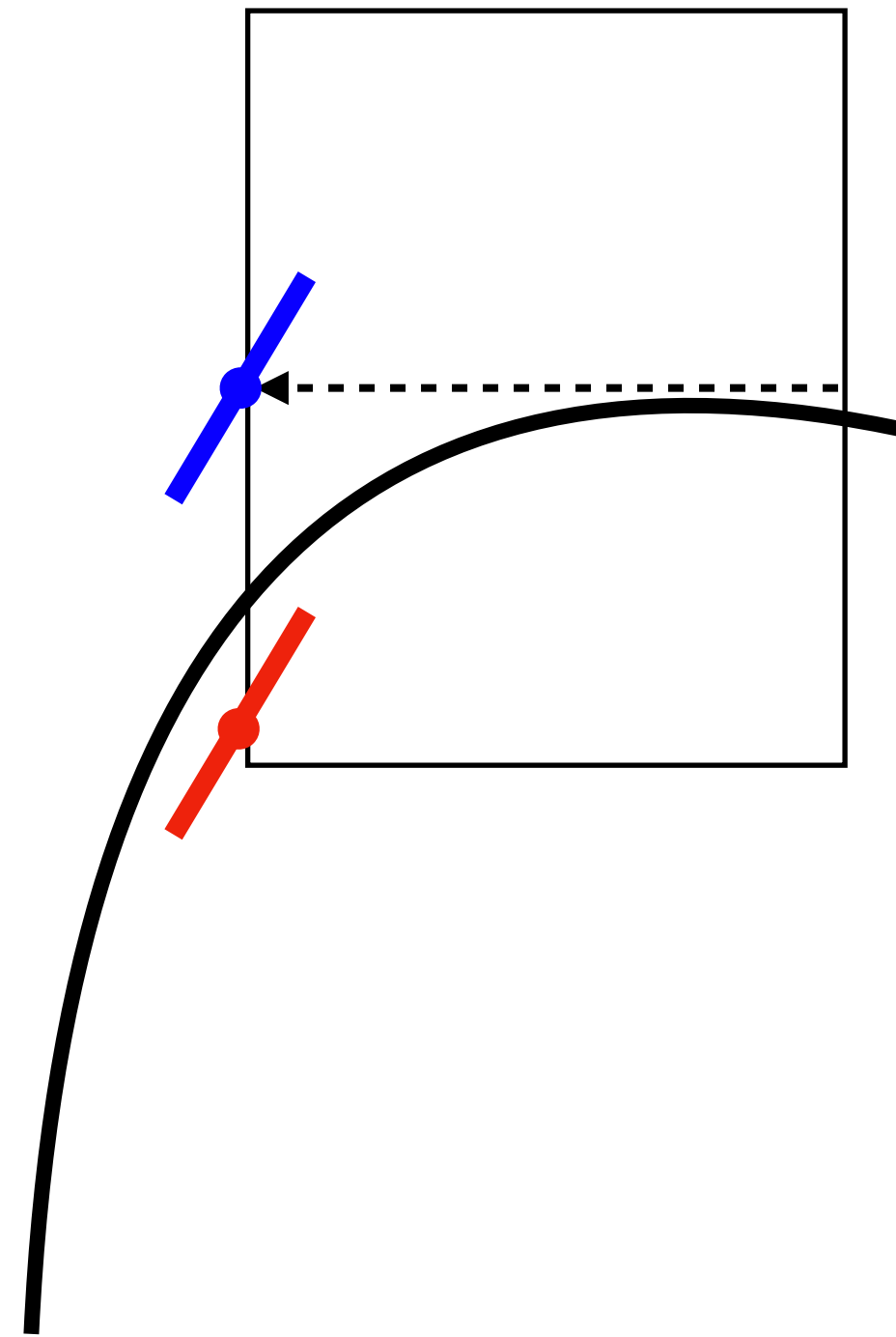


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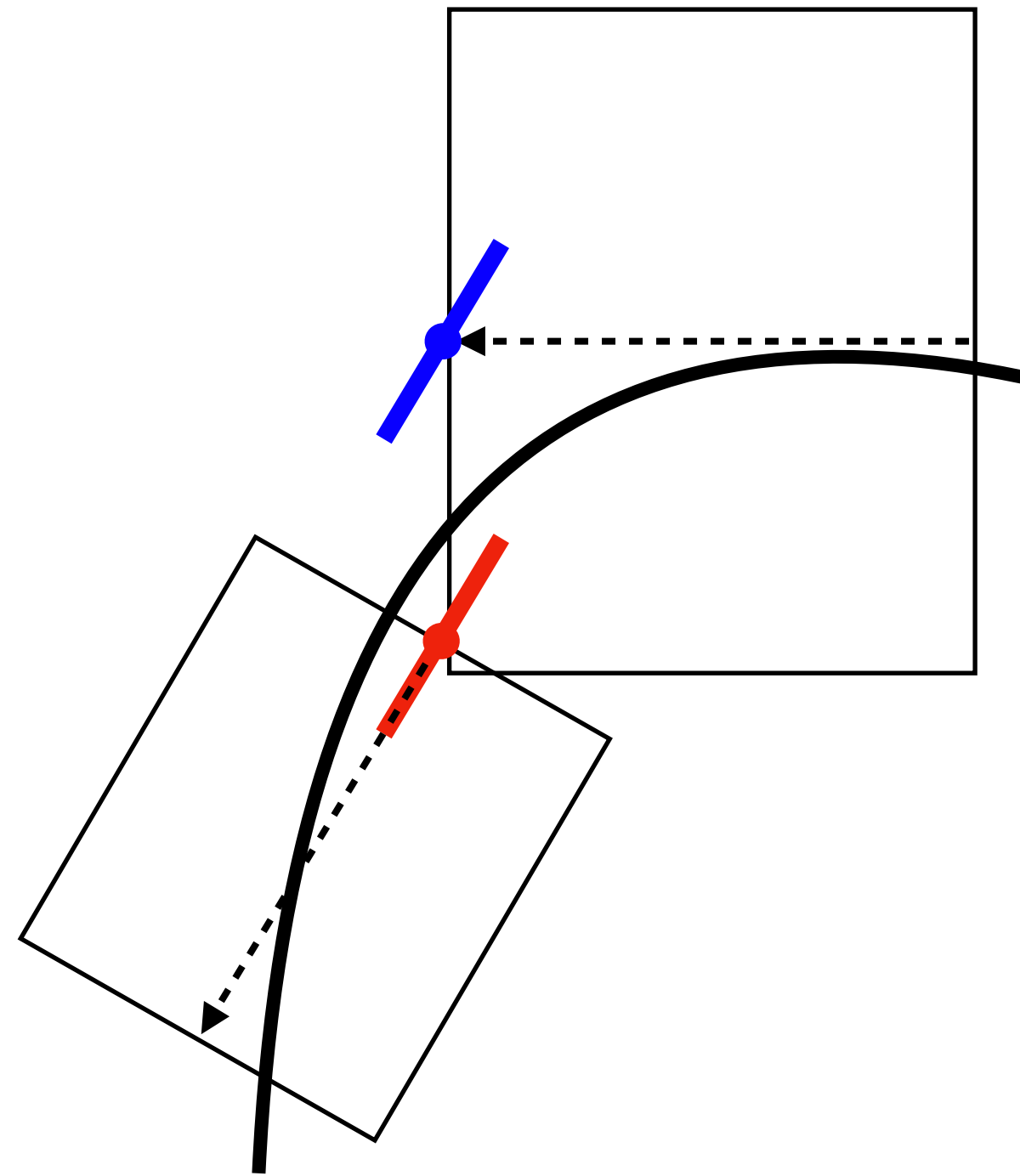


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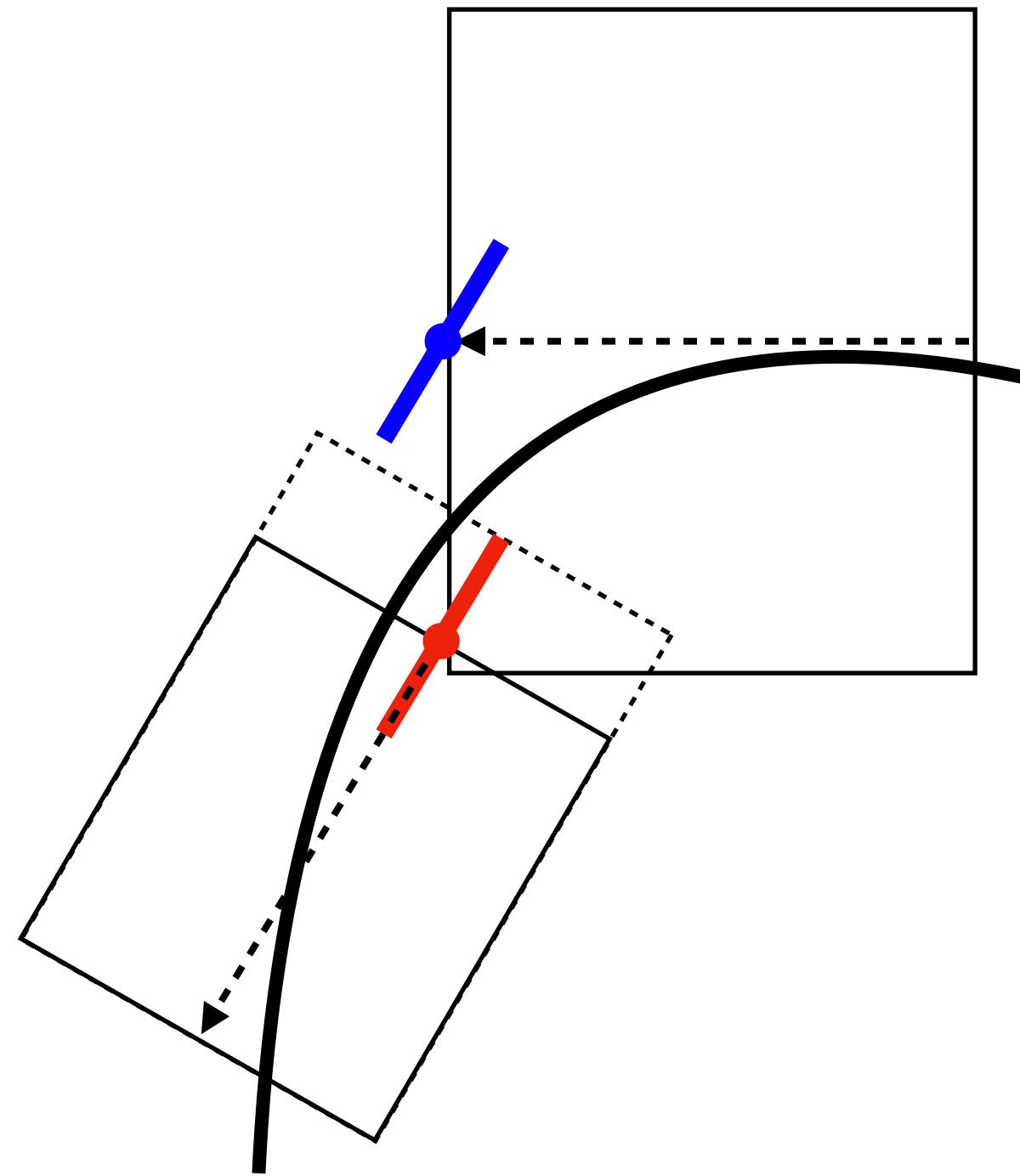


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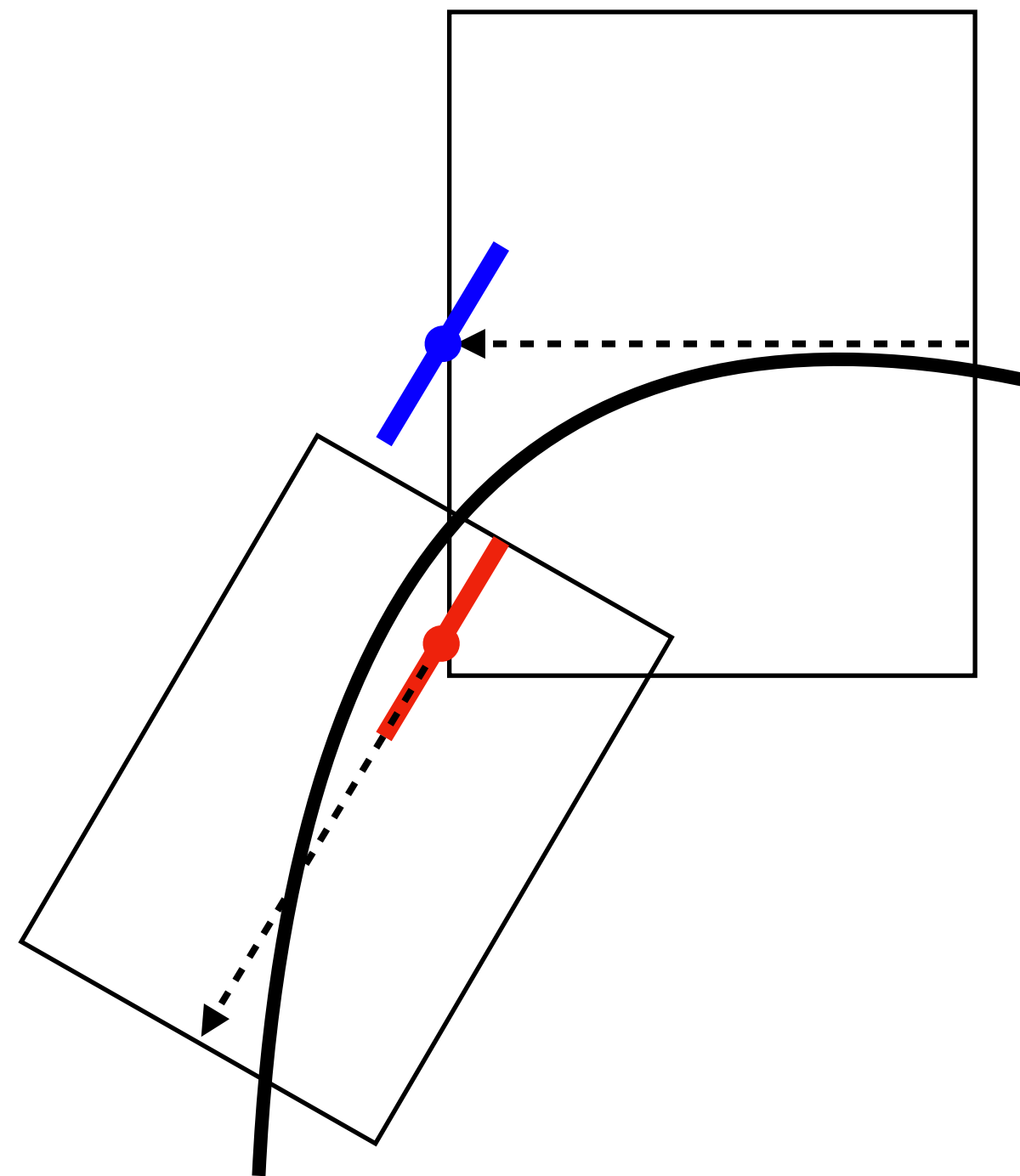


# Curve tracking: algorithm overview

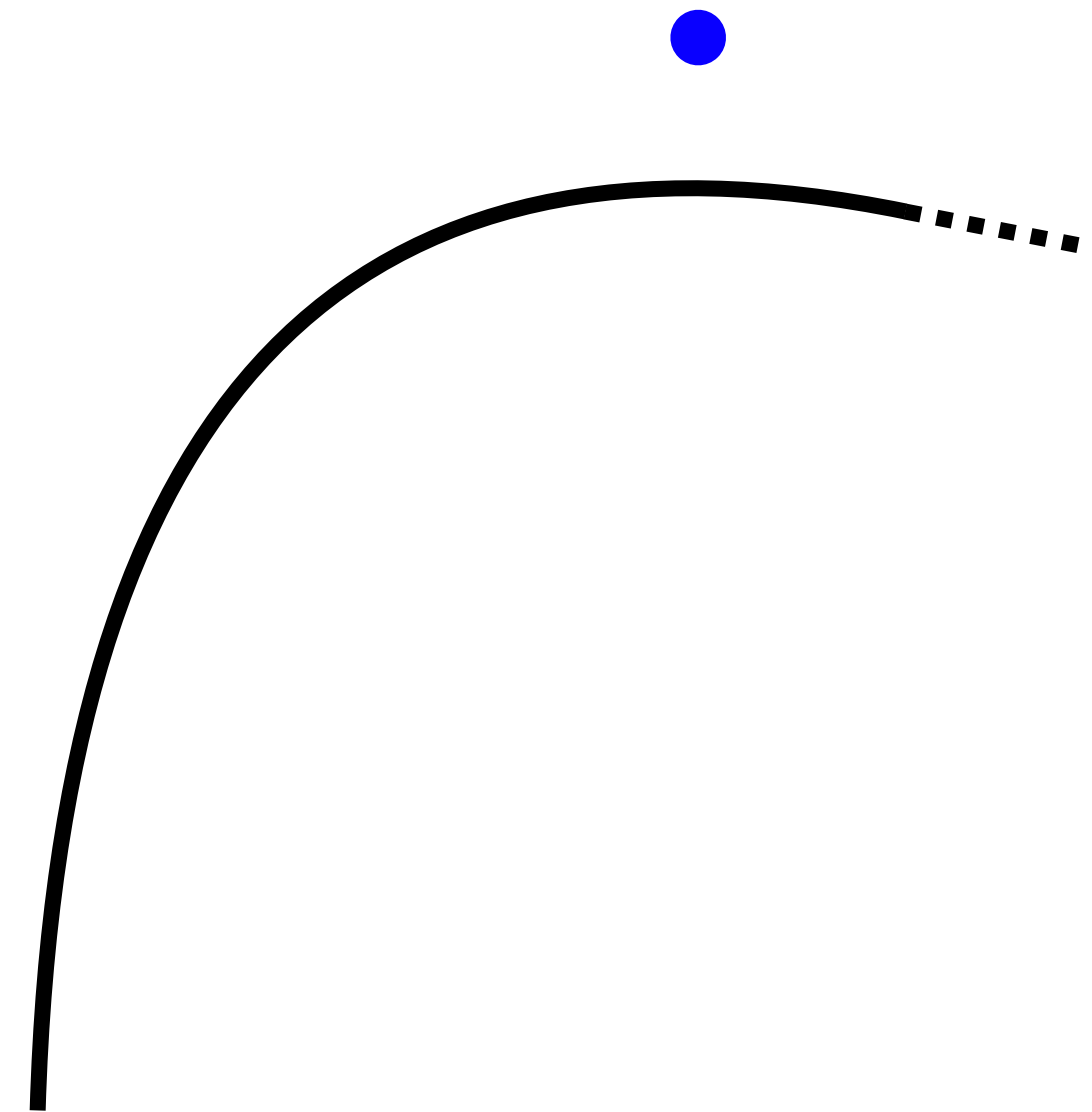
Caveats in refinement

Two possible bad scenarios:

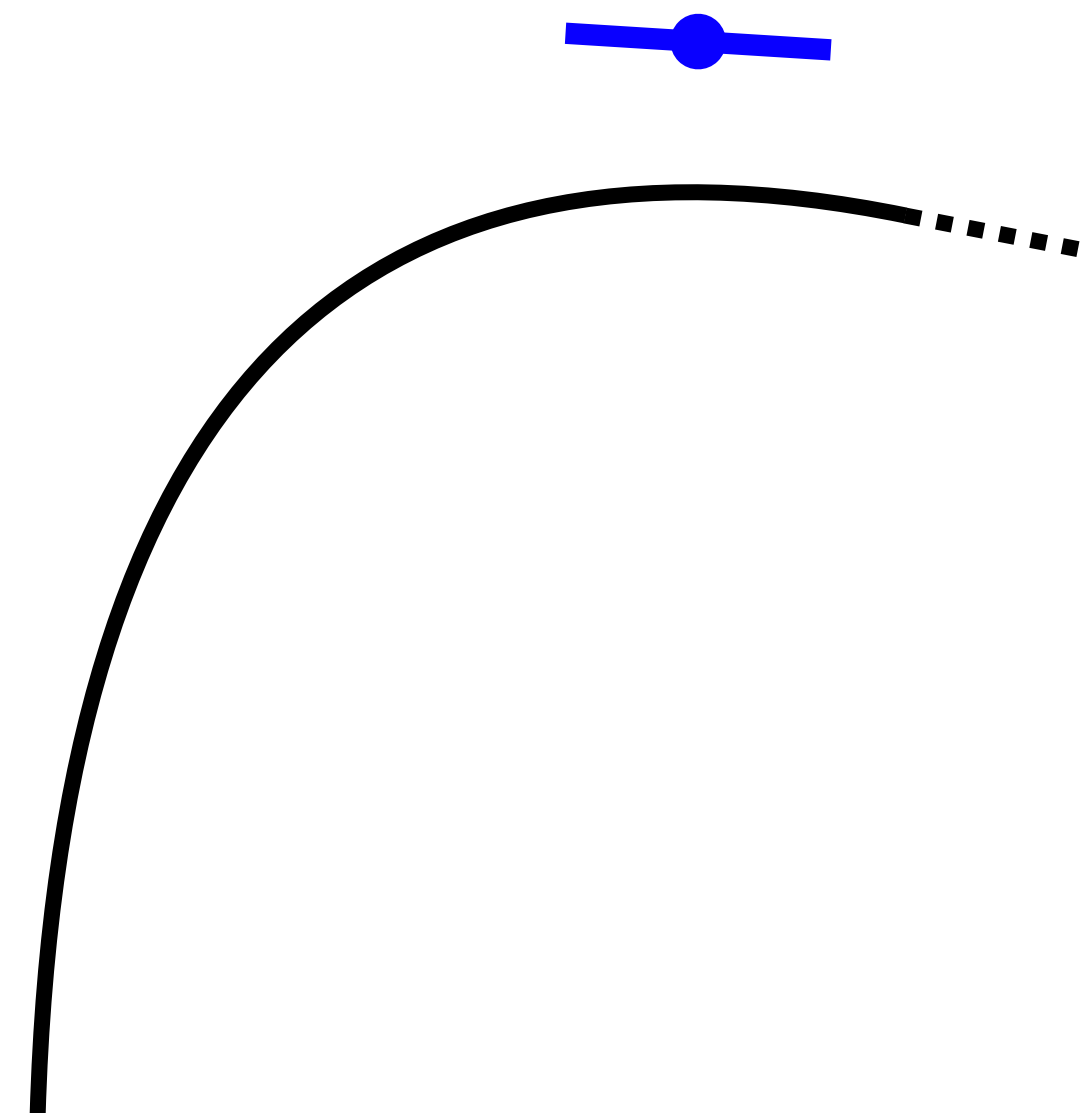
Refine  $x + (\mathbf{0}_{n-1} \times r[-1,1])$



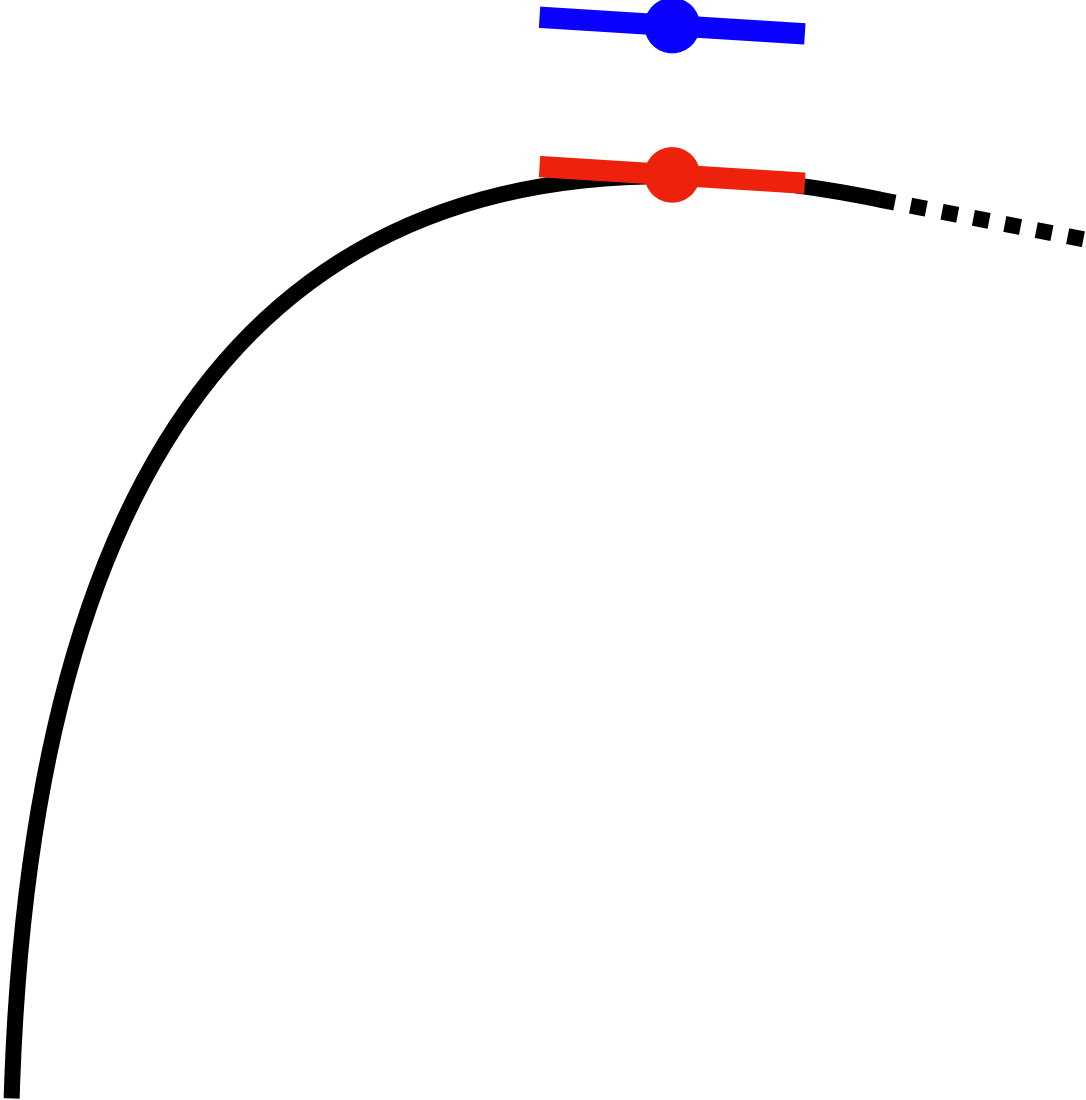
# Algorithm outlines



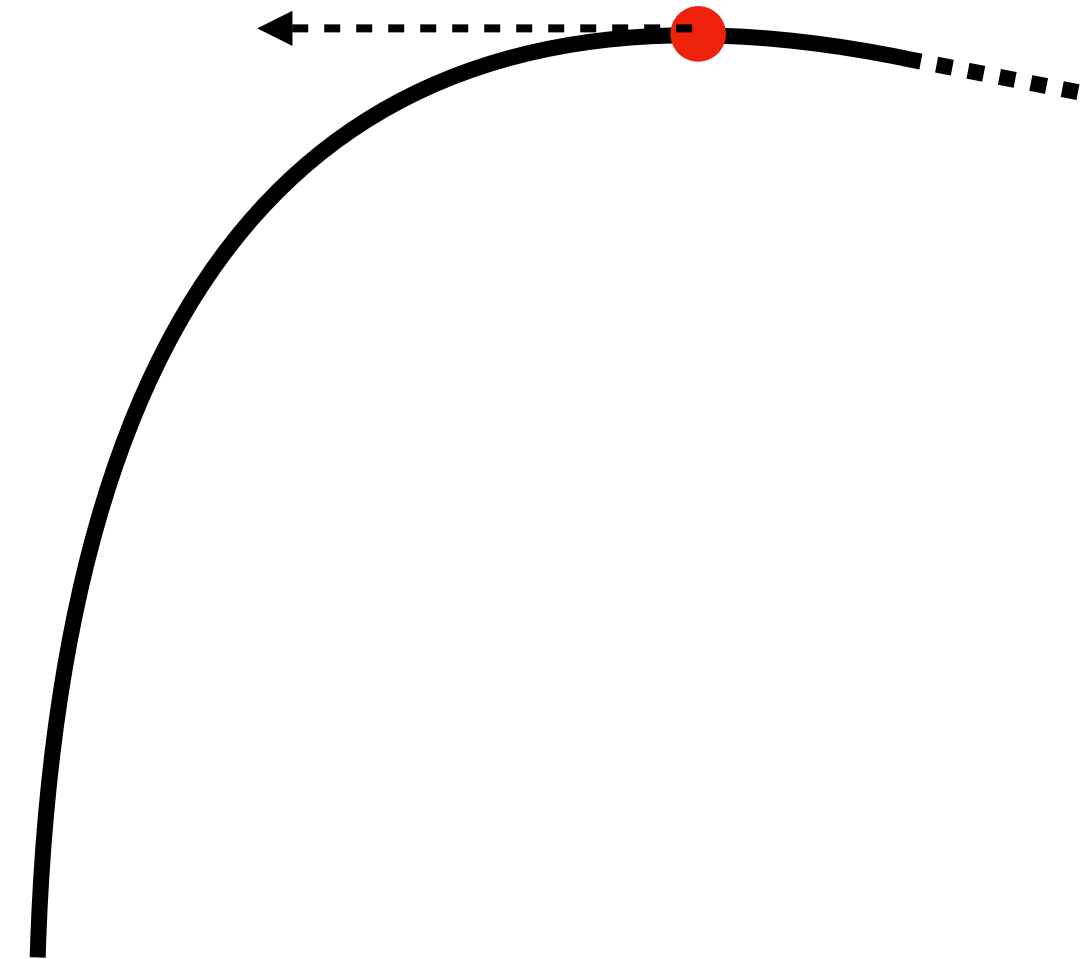
# Algorithm outlines



# Algorithm outlines

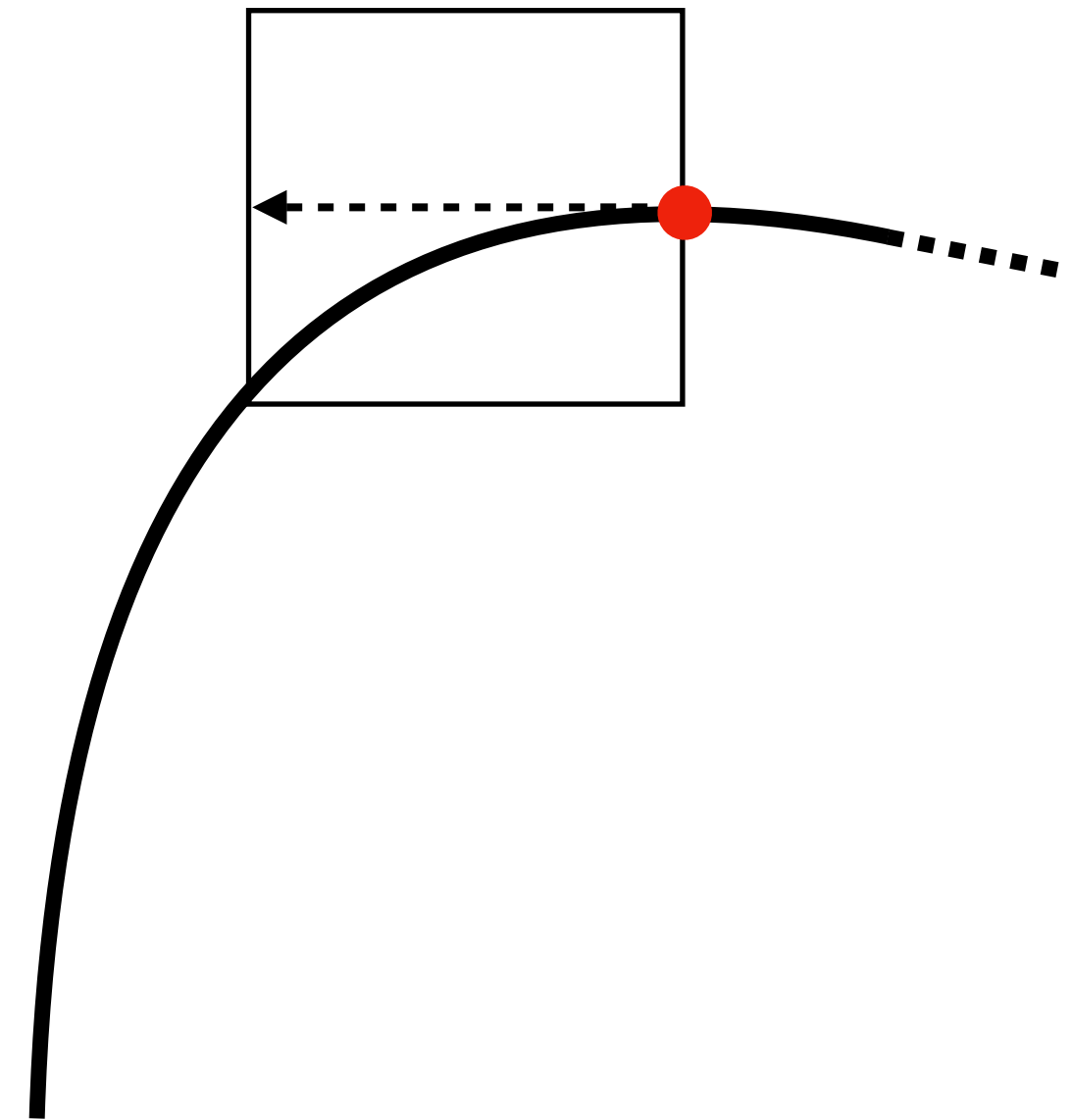


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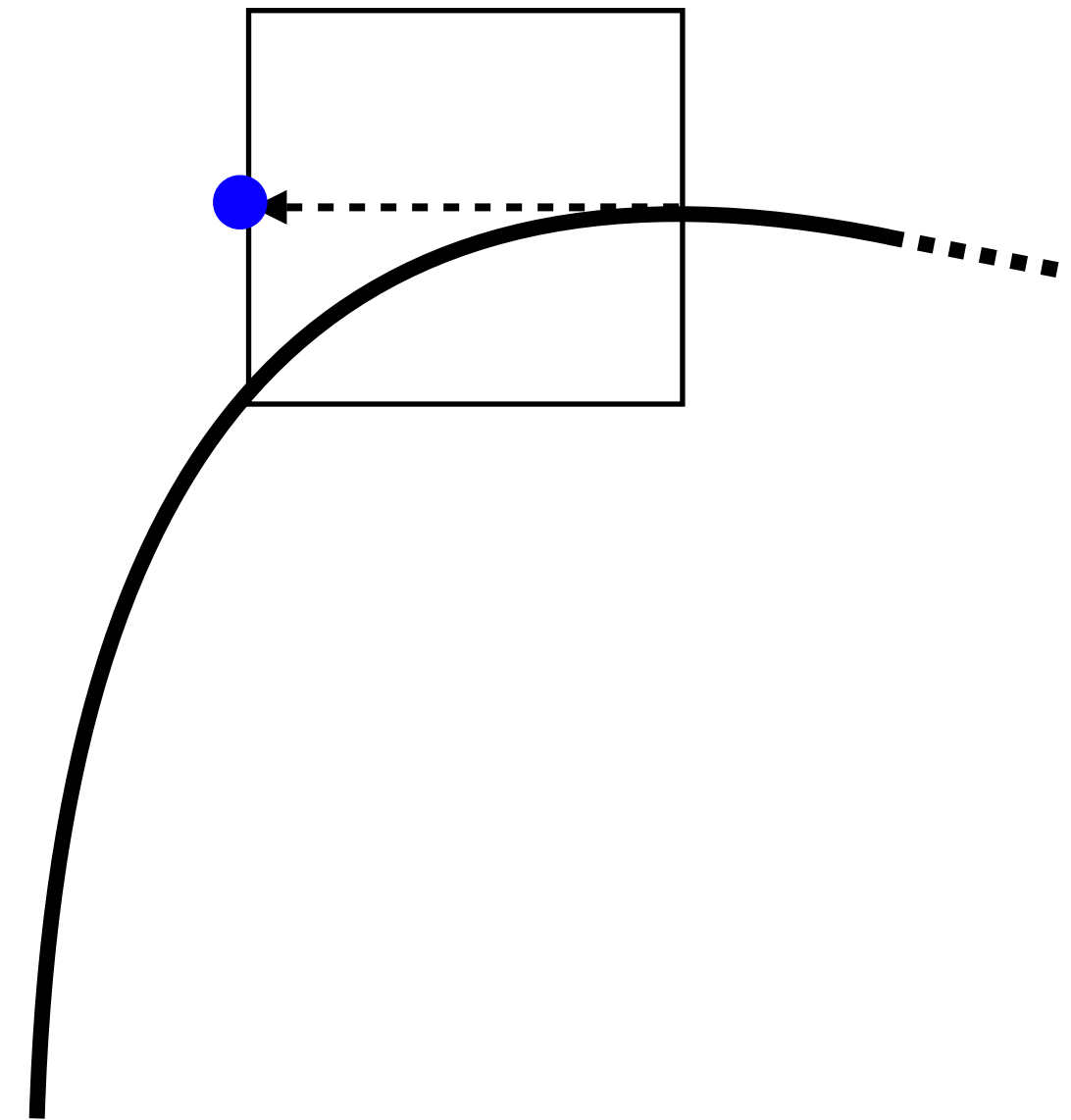




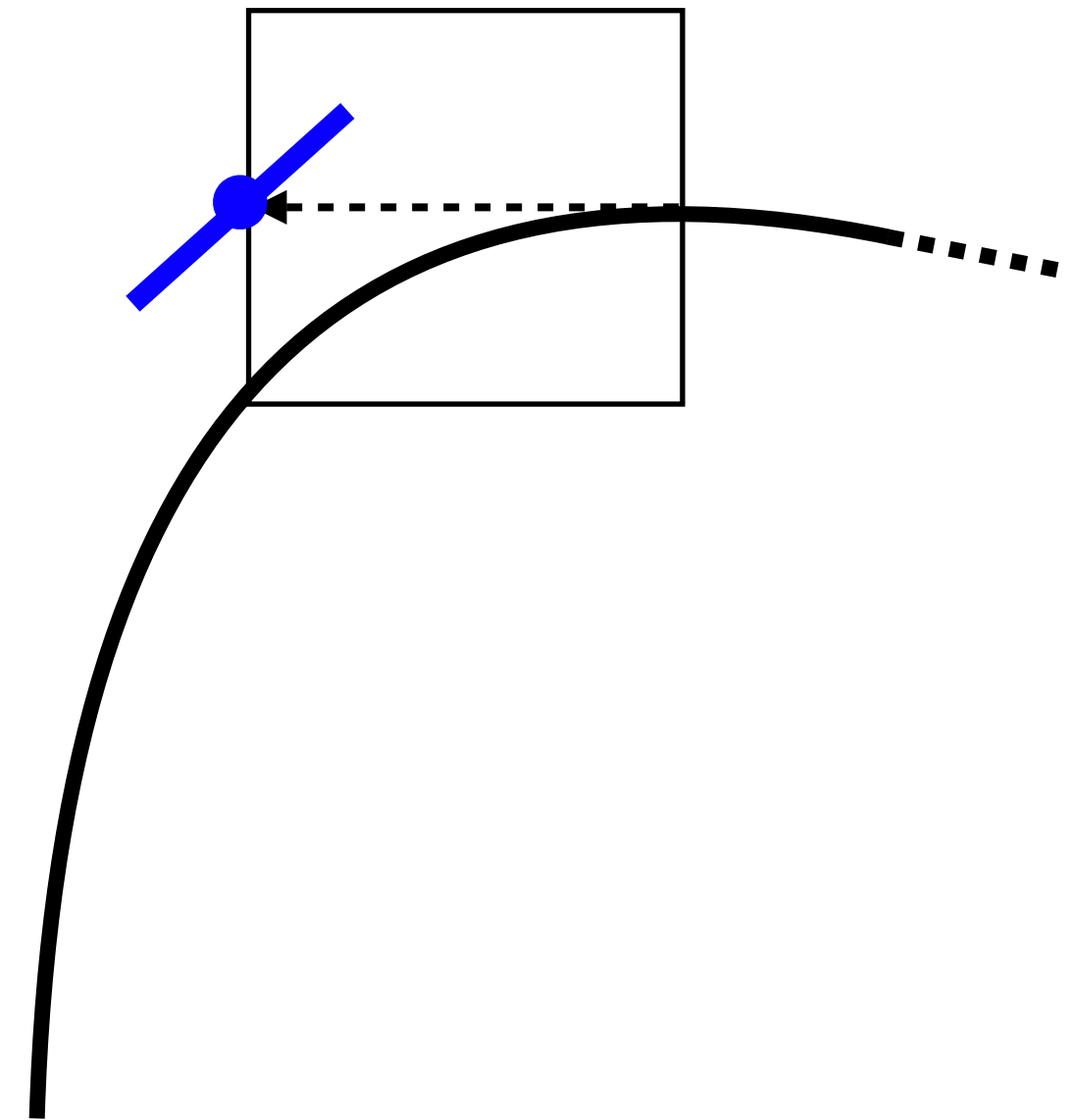
# Algorithm outlines



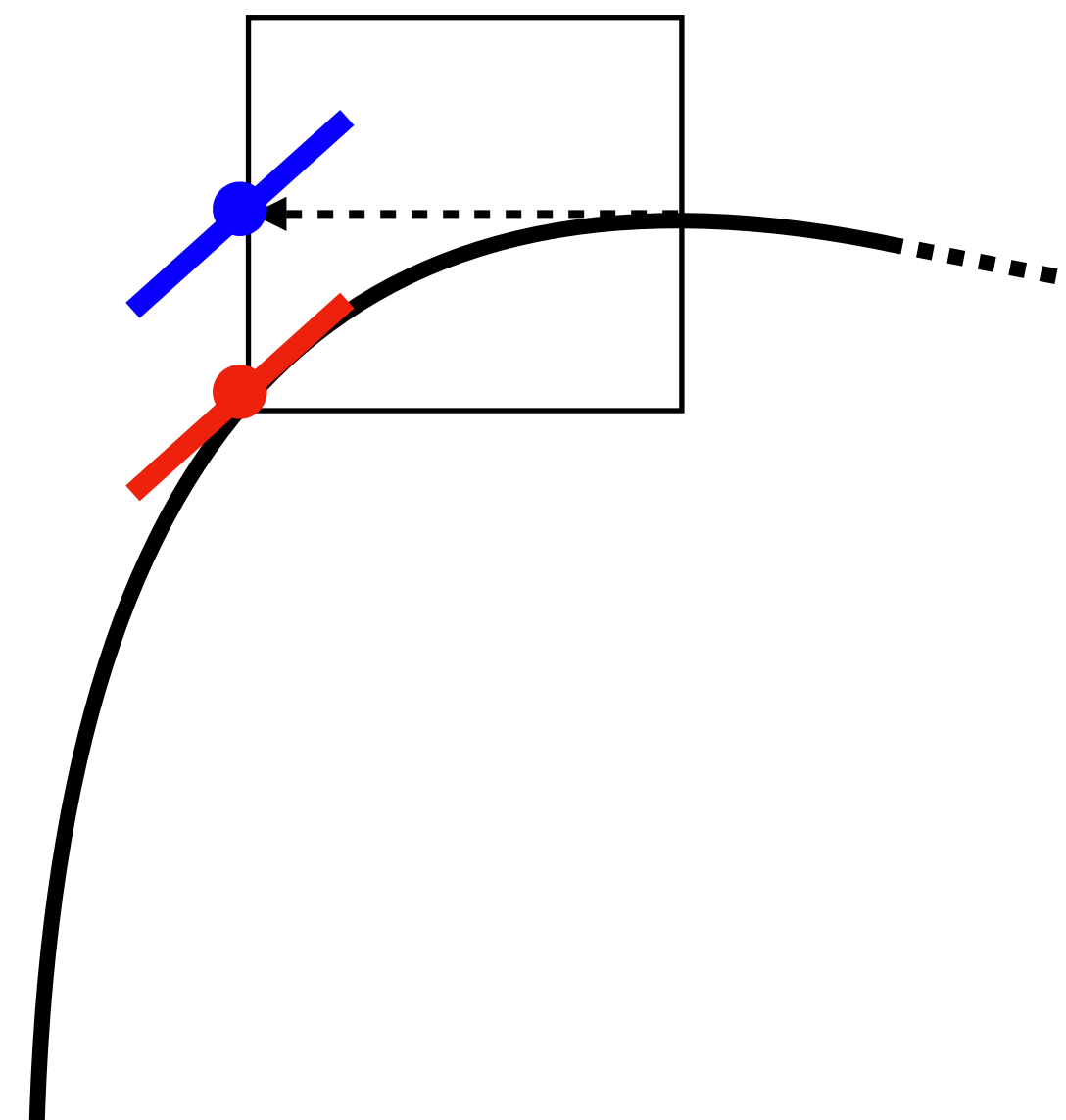
# Algorithm outlines



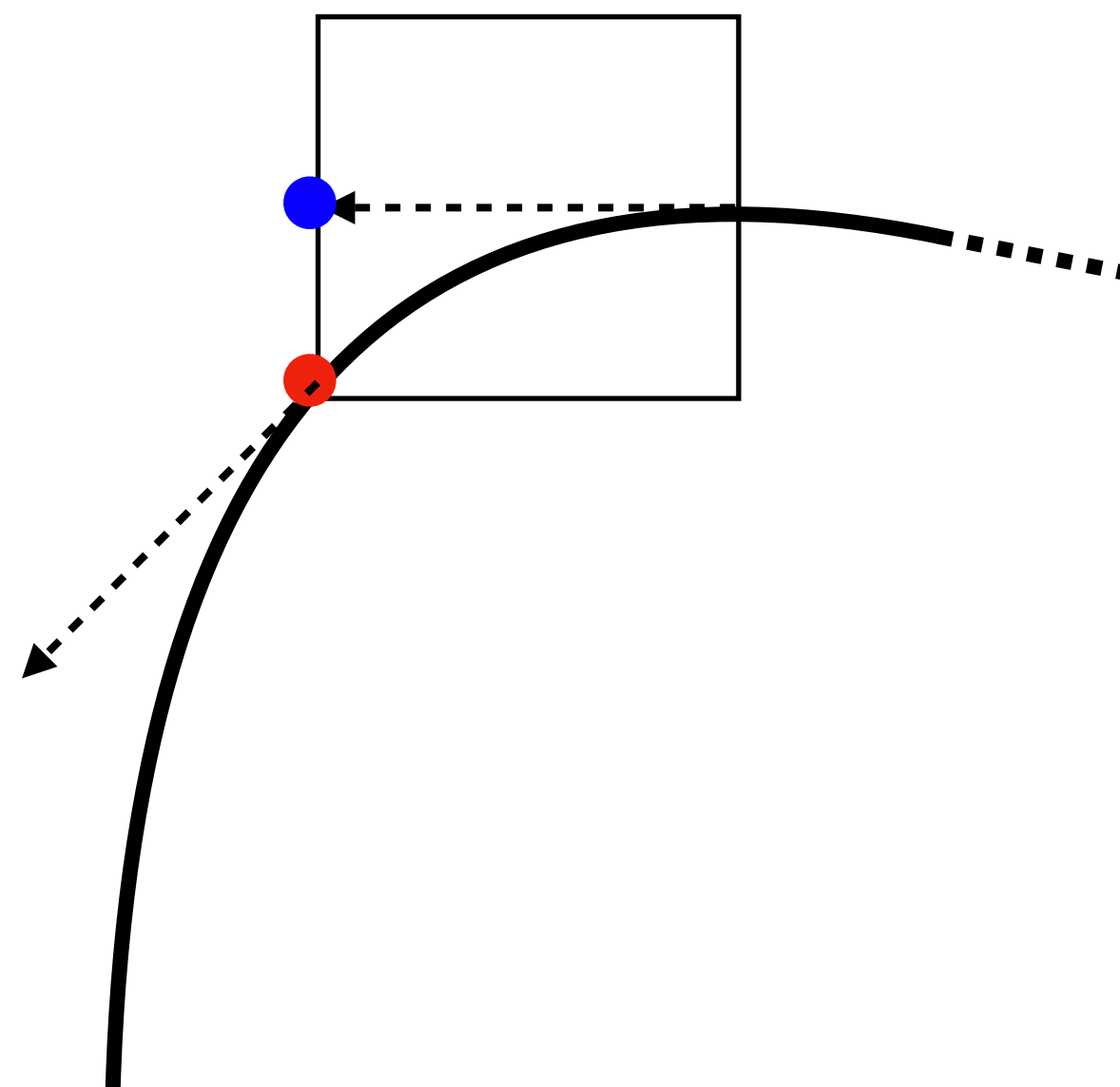
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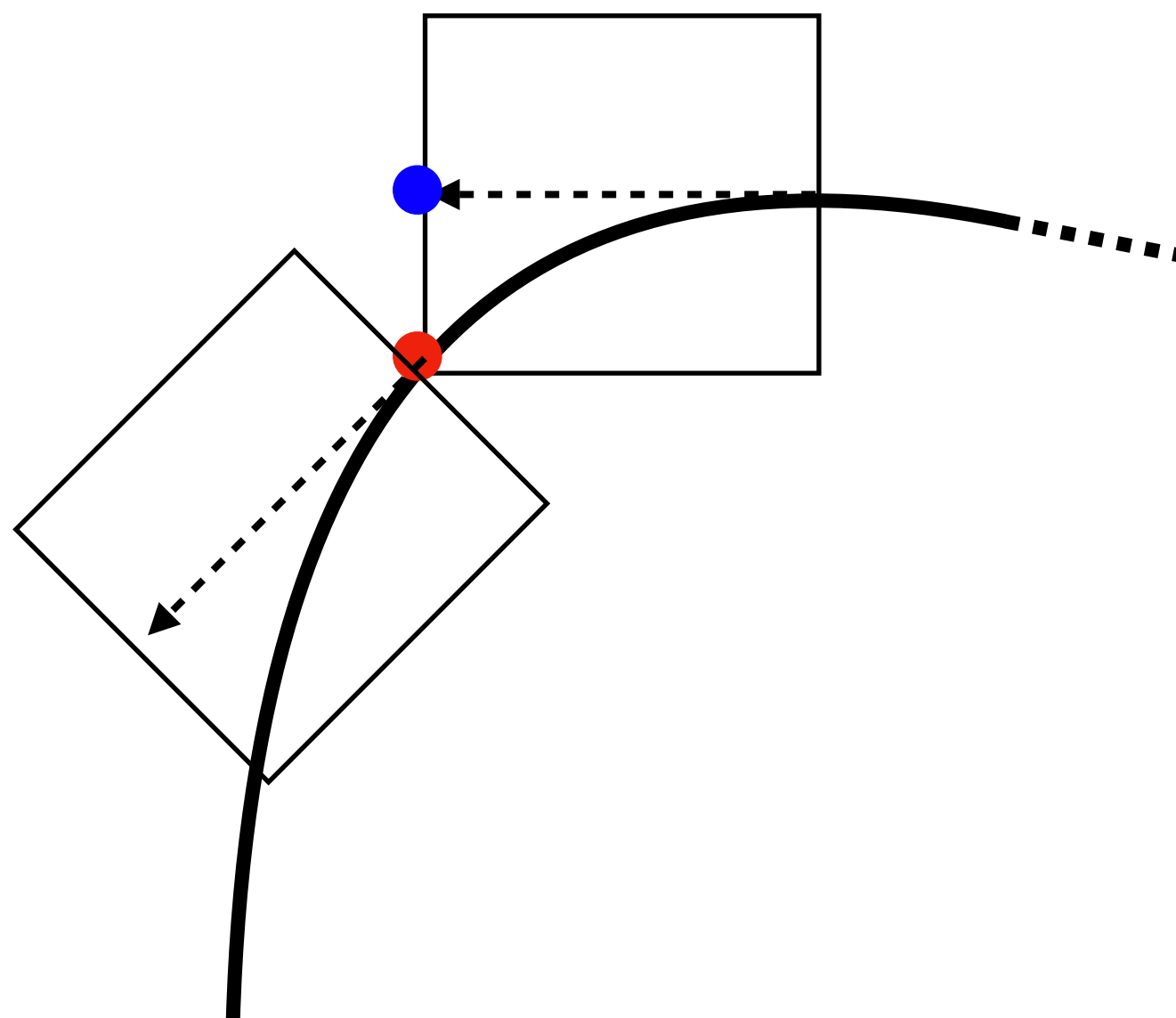
# Algorithm outlines



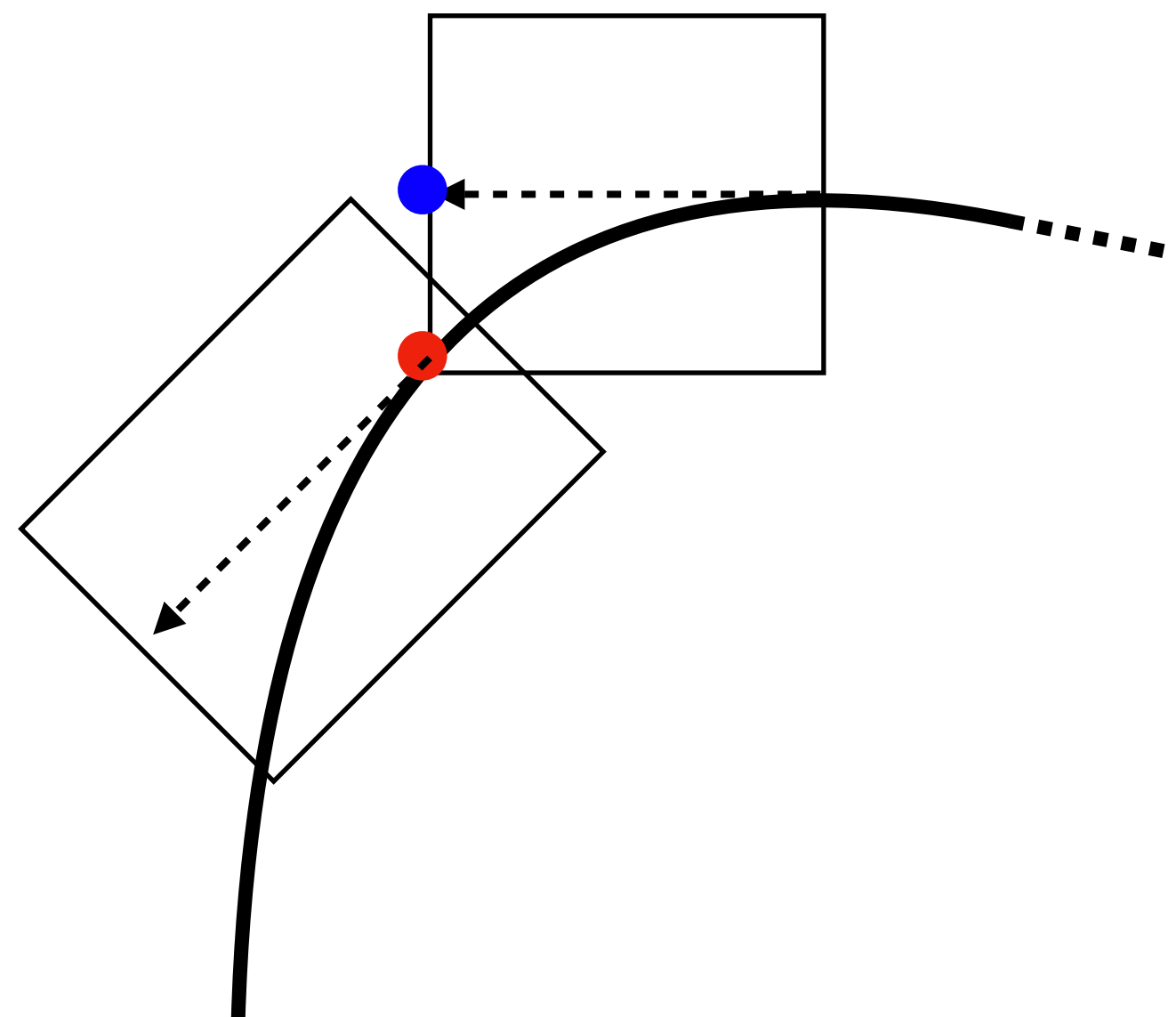
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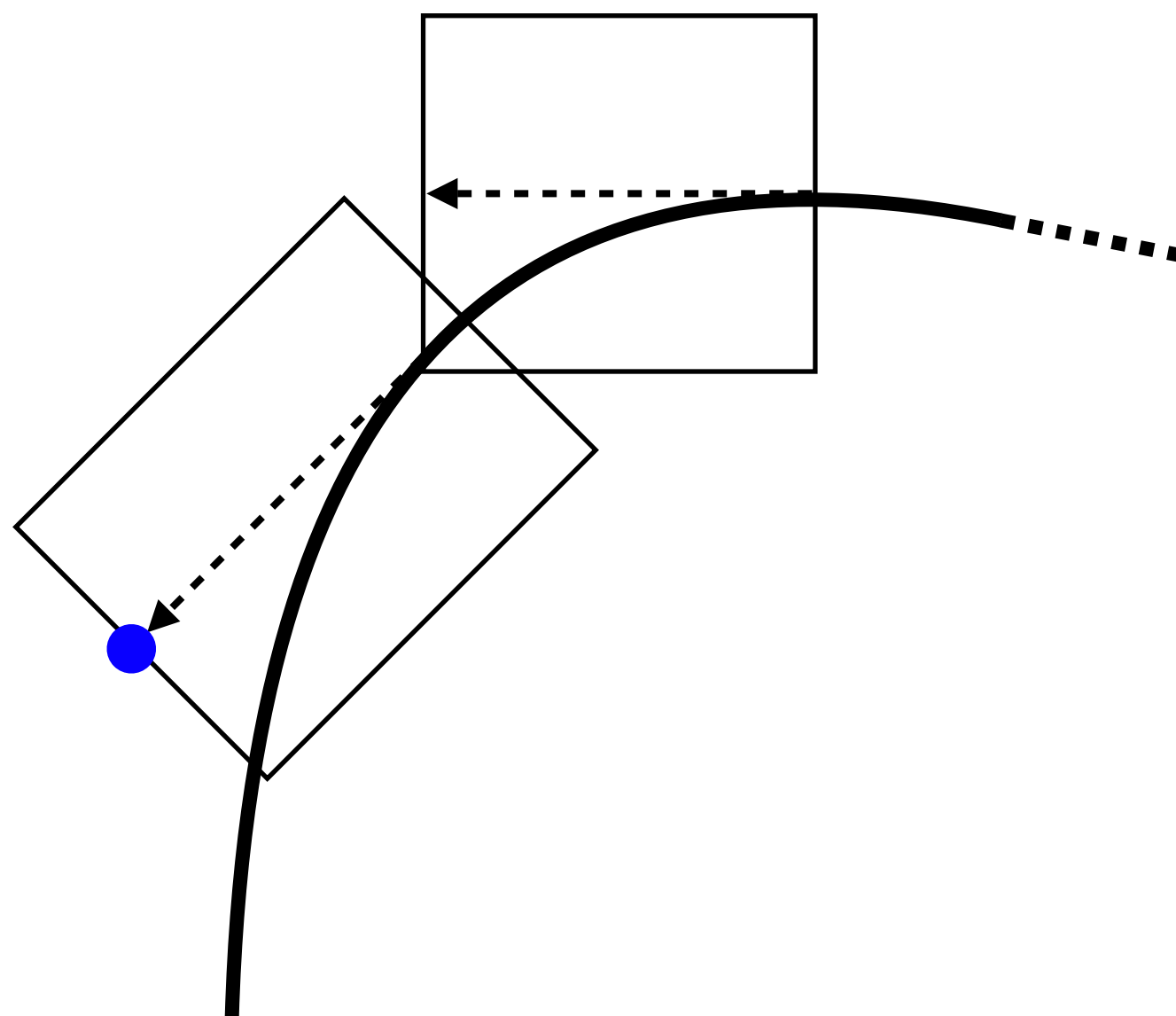
# Algorithm outlines



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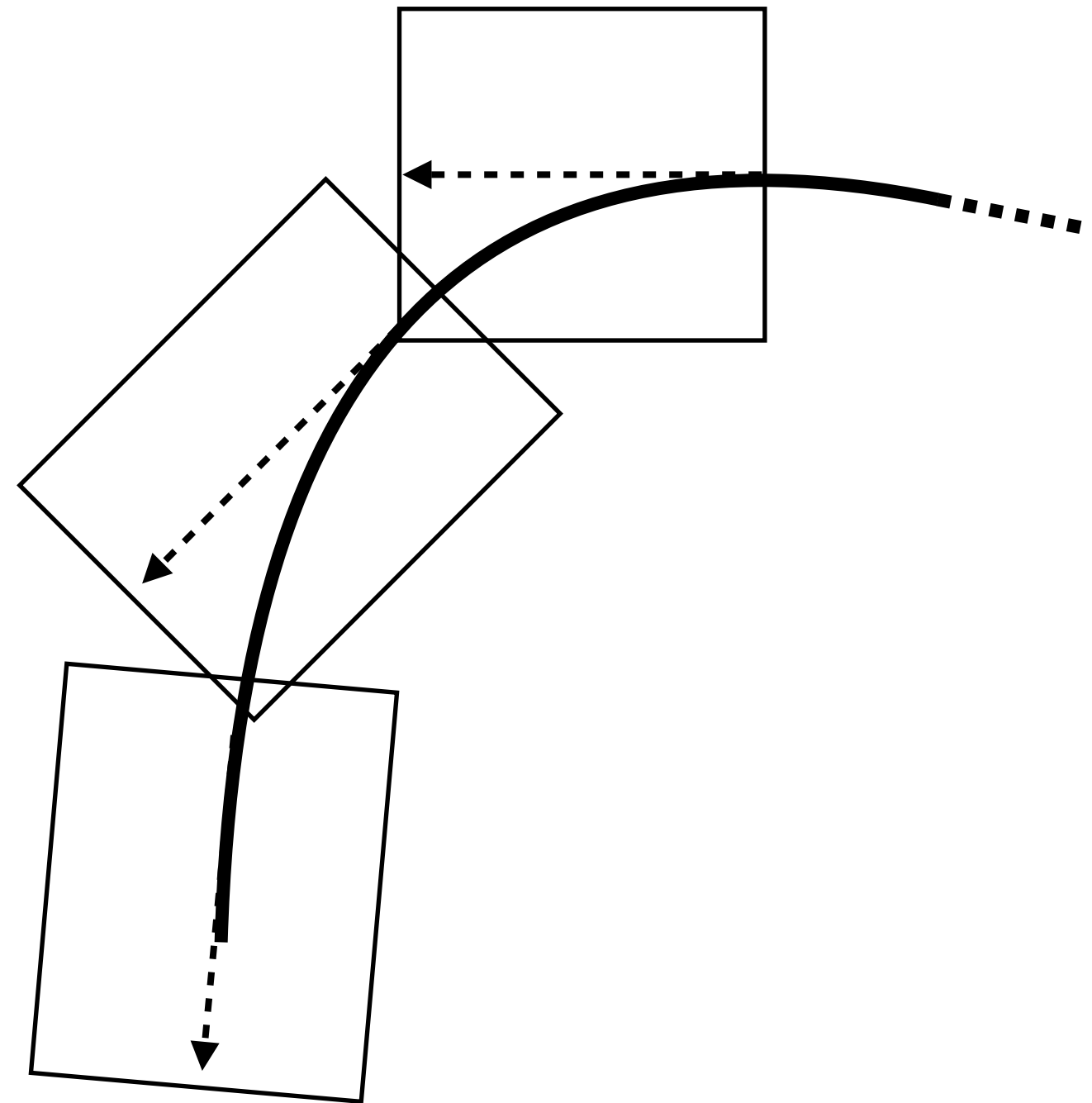


# Algorithm outlines

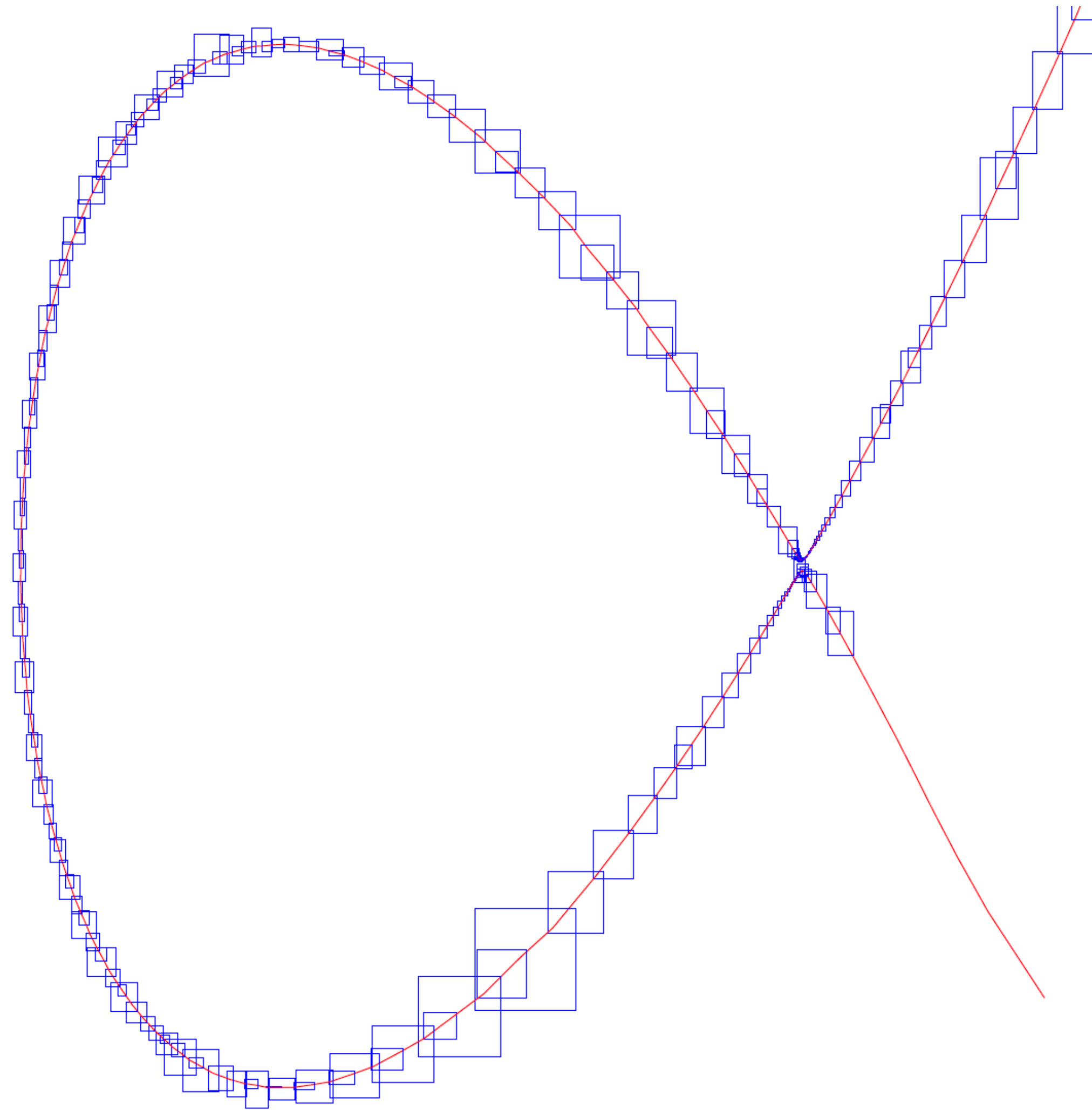




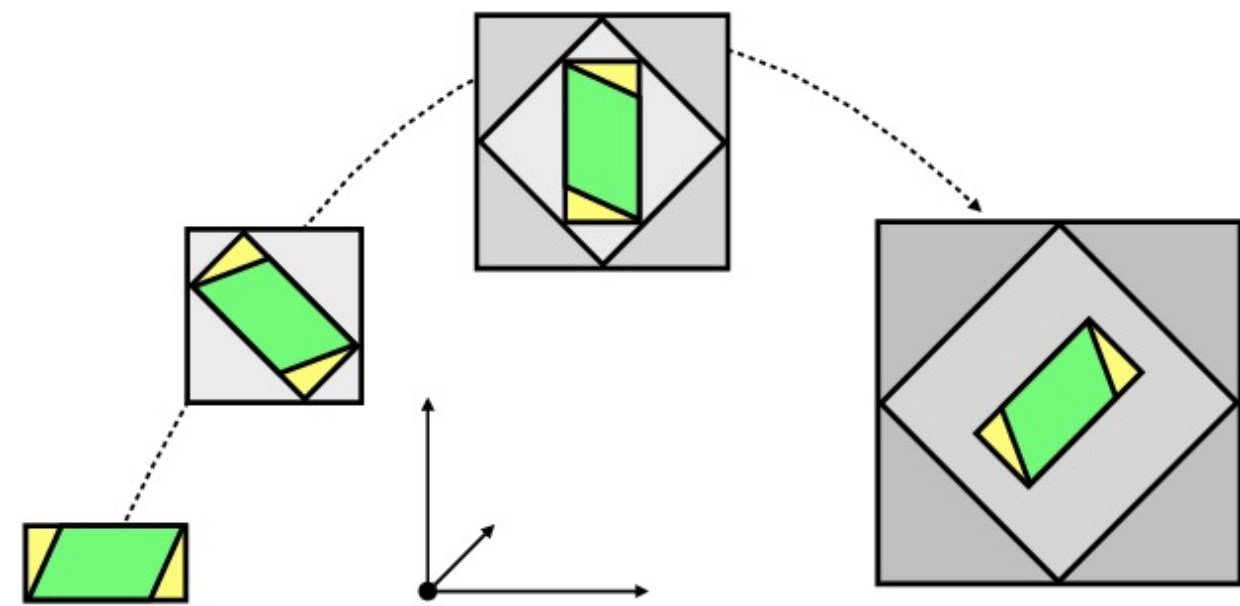
# Algorithm outlines



A **tubular neighborhood** of  
 $C = \{x^3 - 2.9995x - y^2 + 2\}$

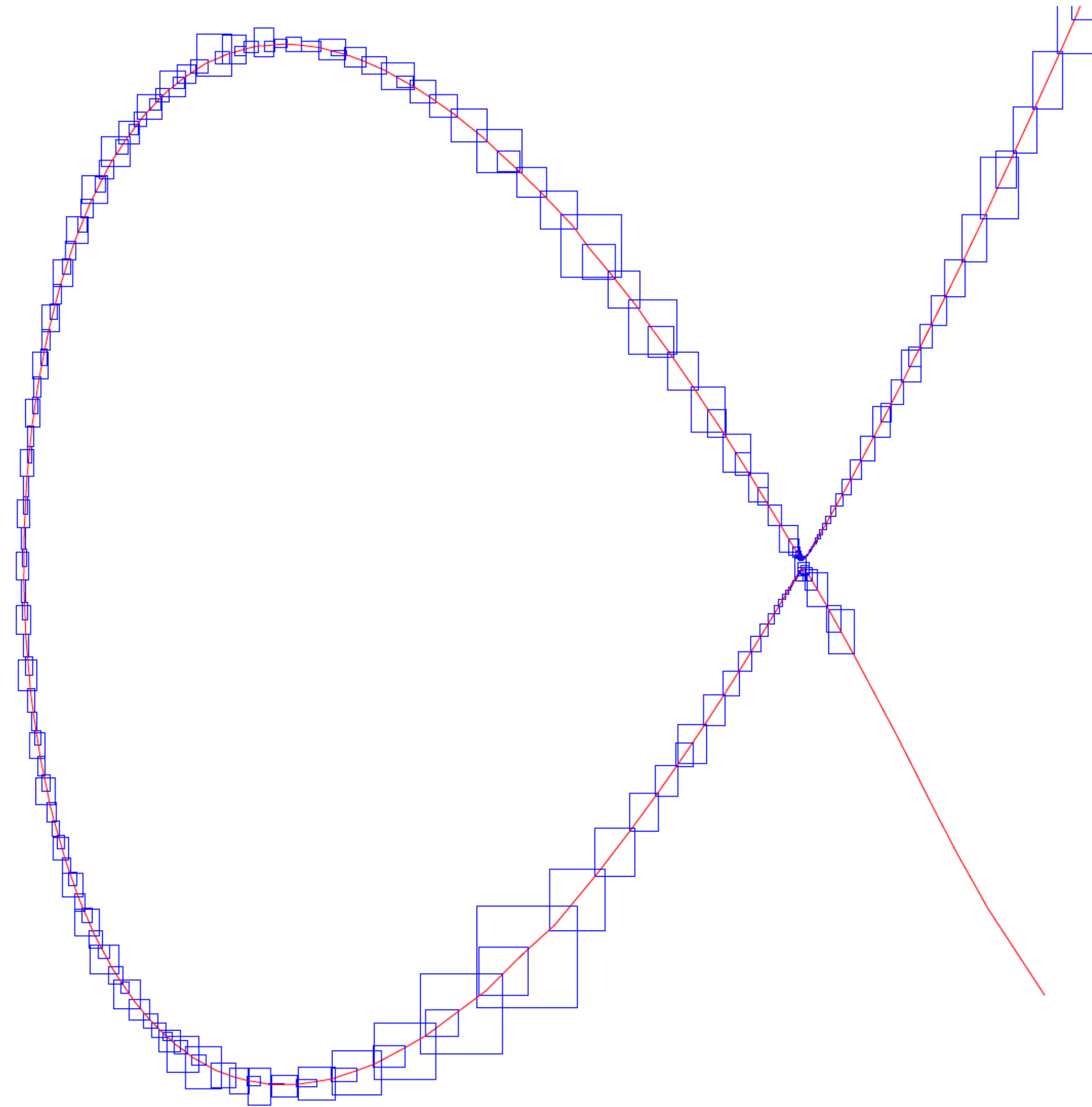


A **tubular neighborhood** of  
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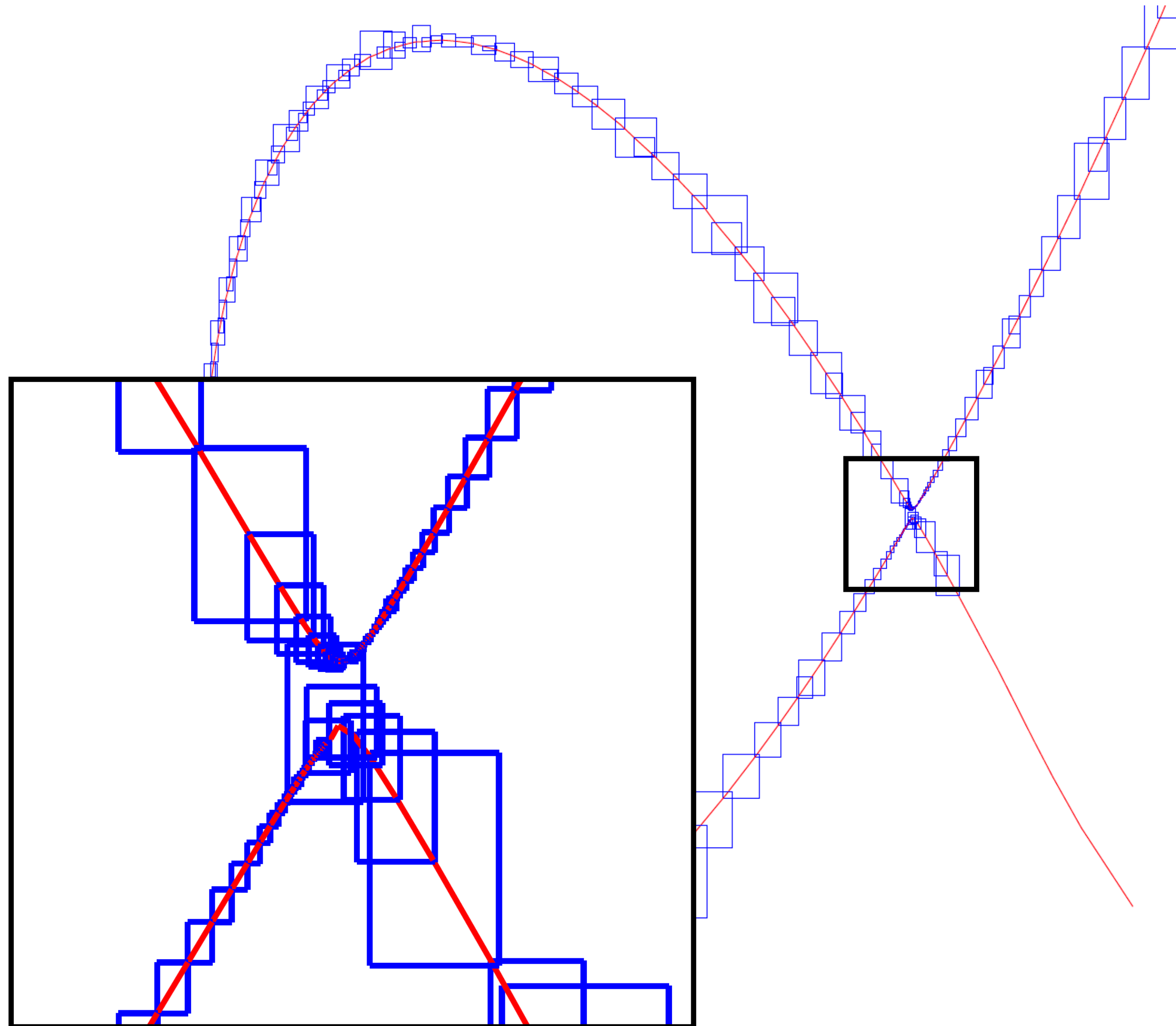


Neubauer-Grosu 2022

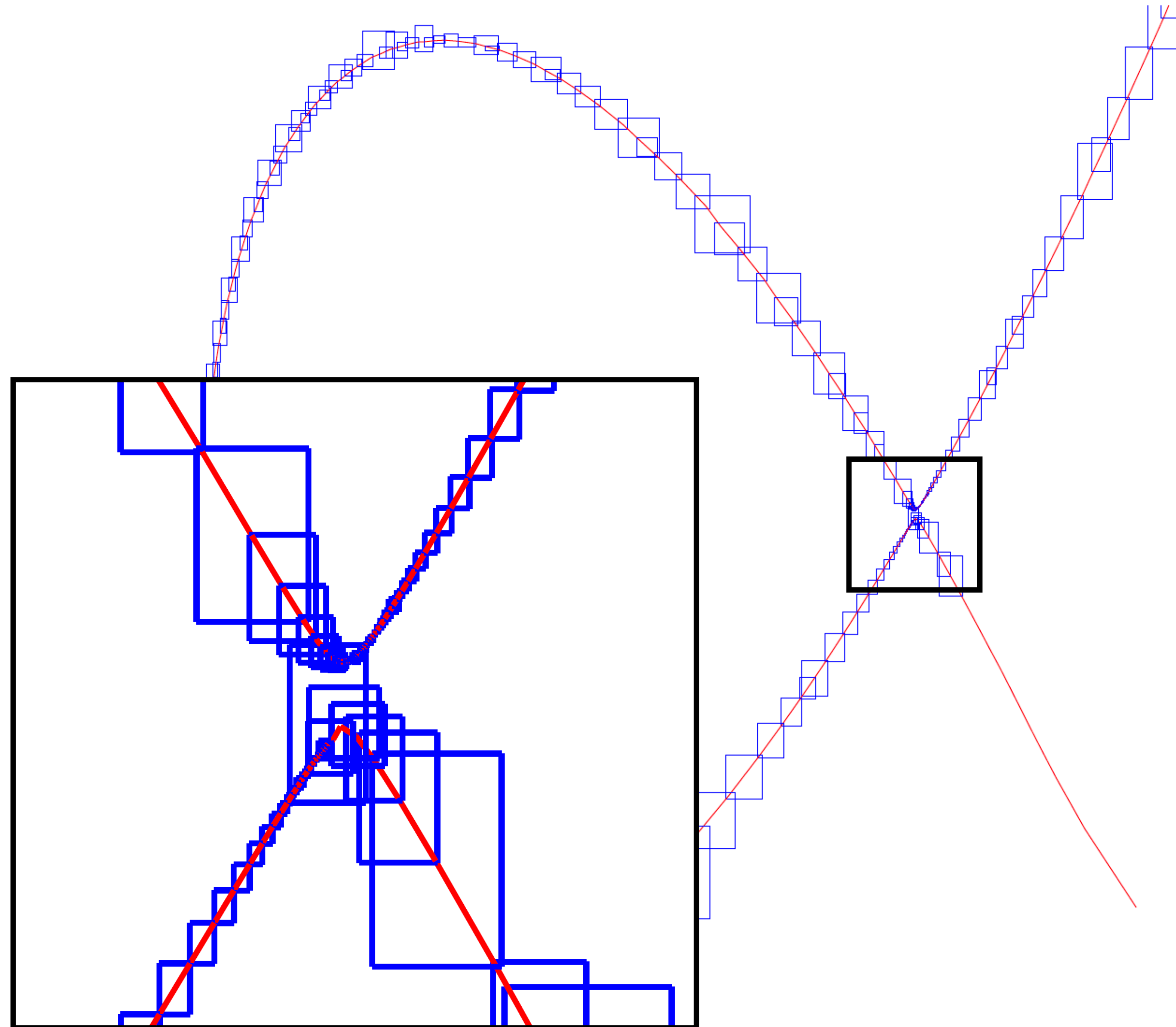
Rectangular boxes are generated  
 due to the wrapping effect.



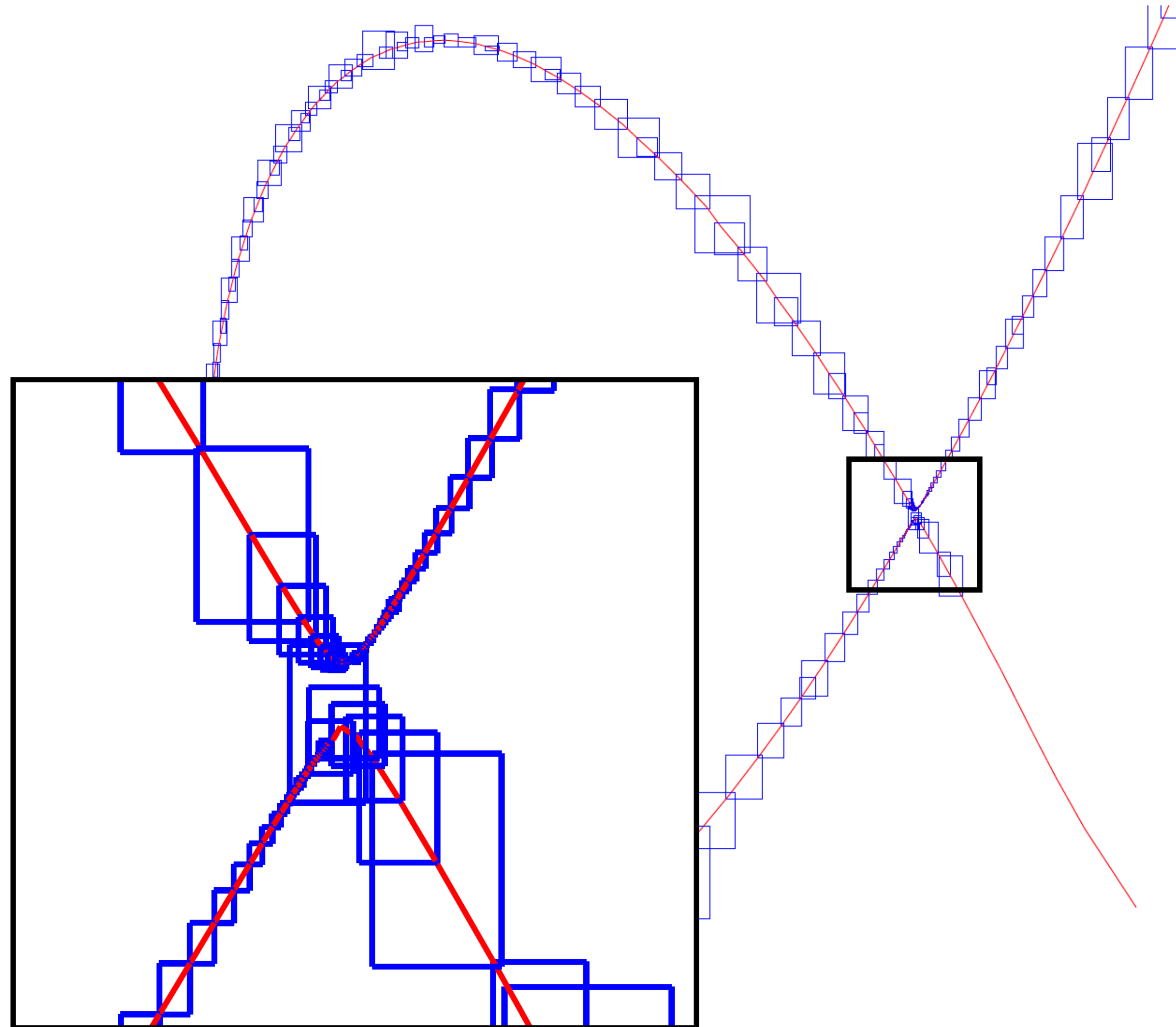
A **tubular neighborhood** of  
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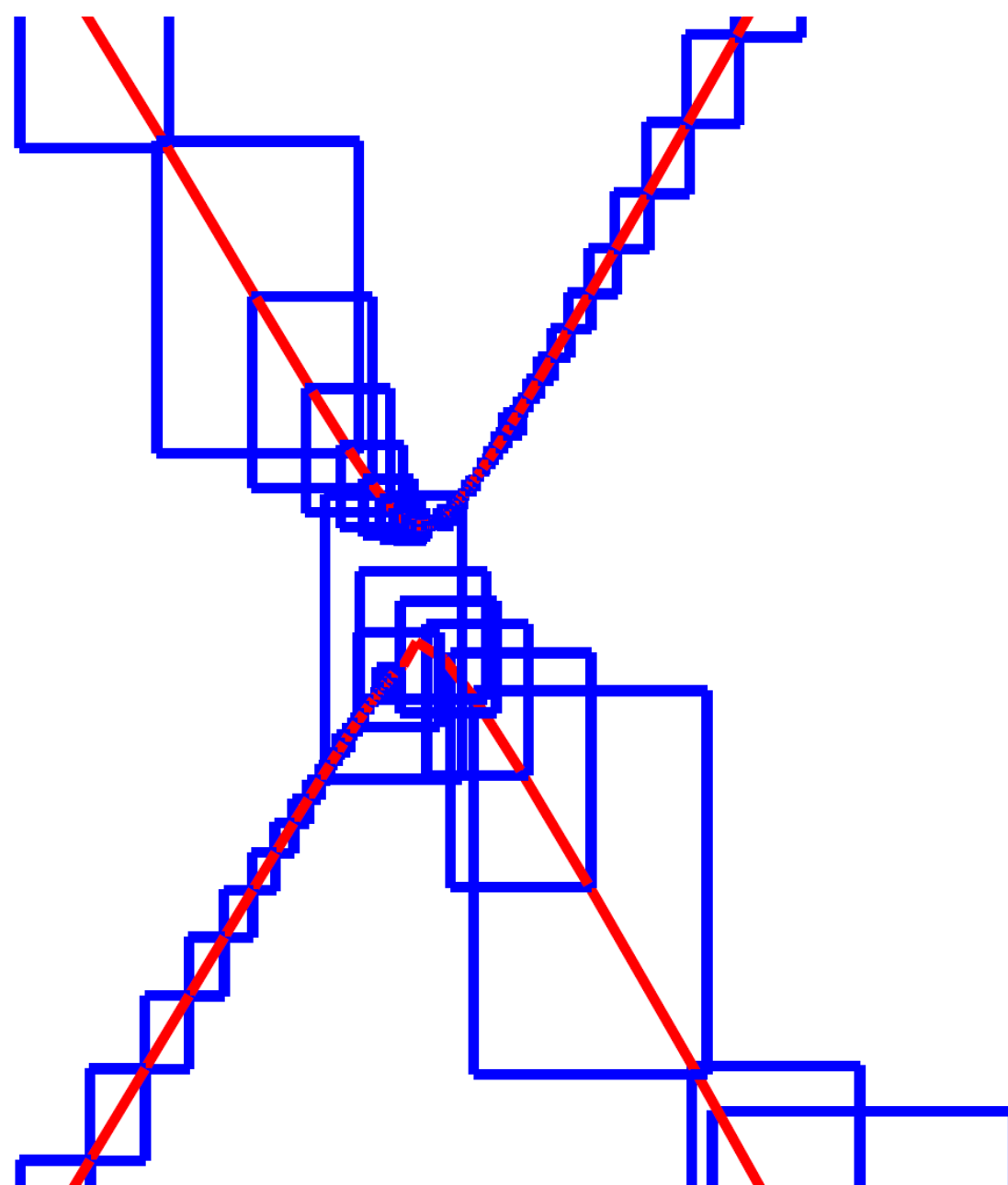


A **tubular neighborhood** of  
 $C = \{x^3 - 2.9995x - y^2 + 2\}$   
How to remove the overlapped  
intervals in the tubular  
neighborhood?

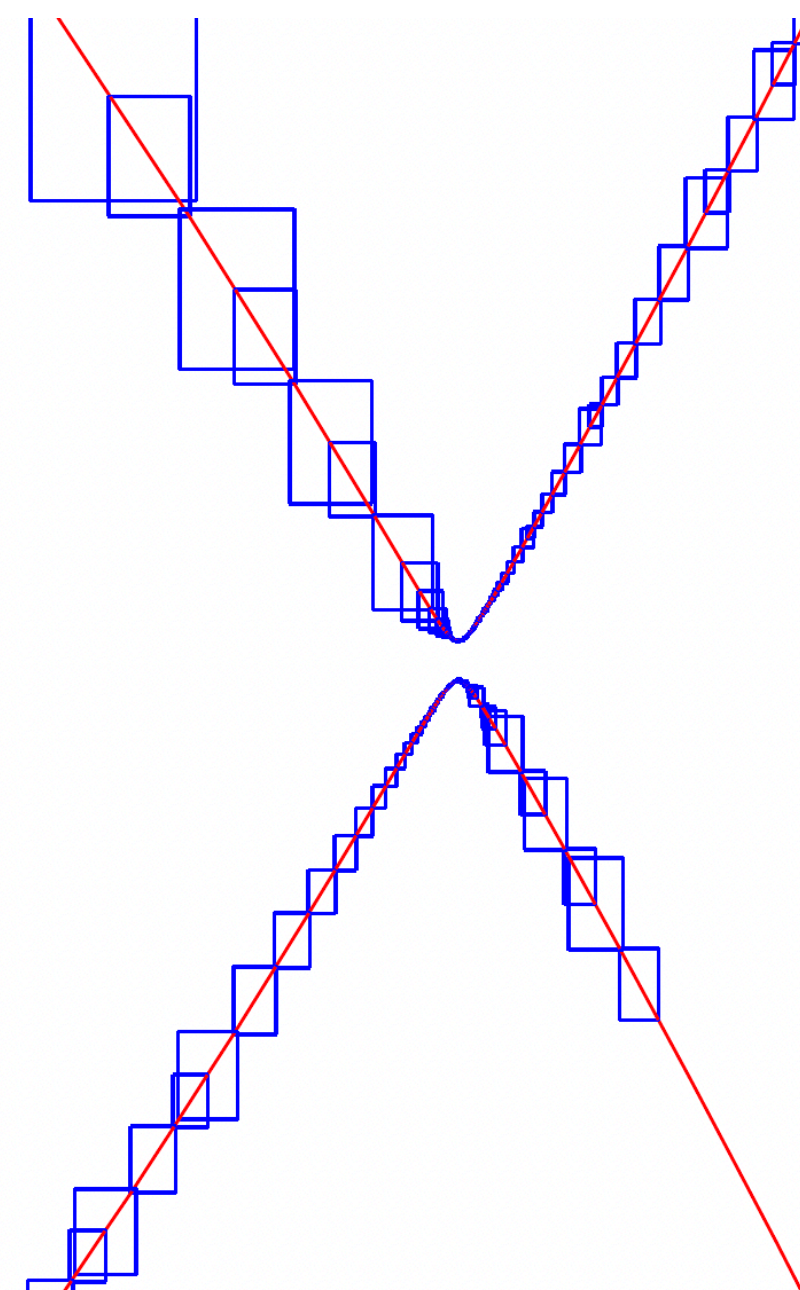


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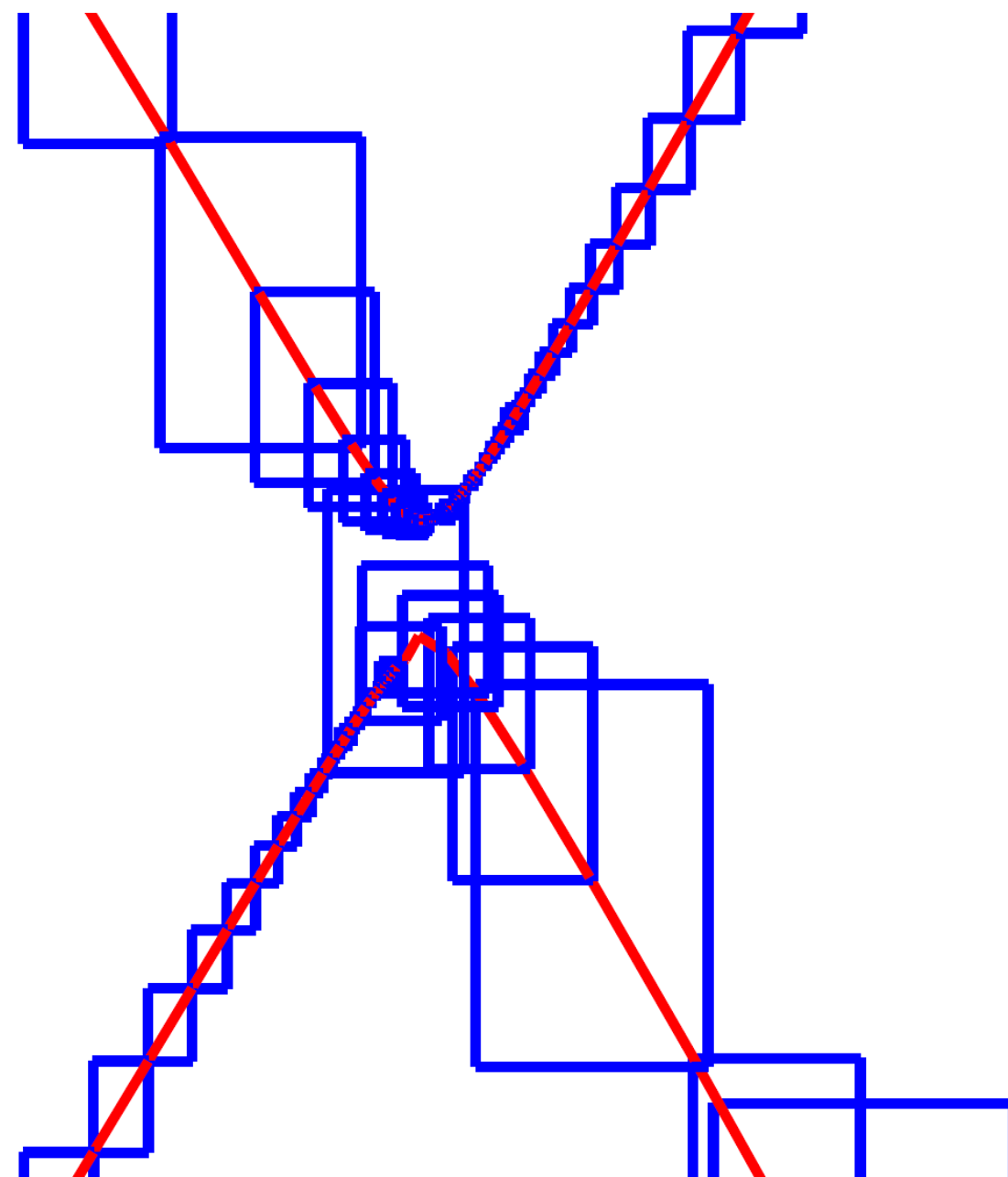
Refined with  $\rho = \frac{1}{8}$



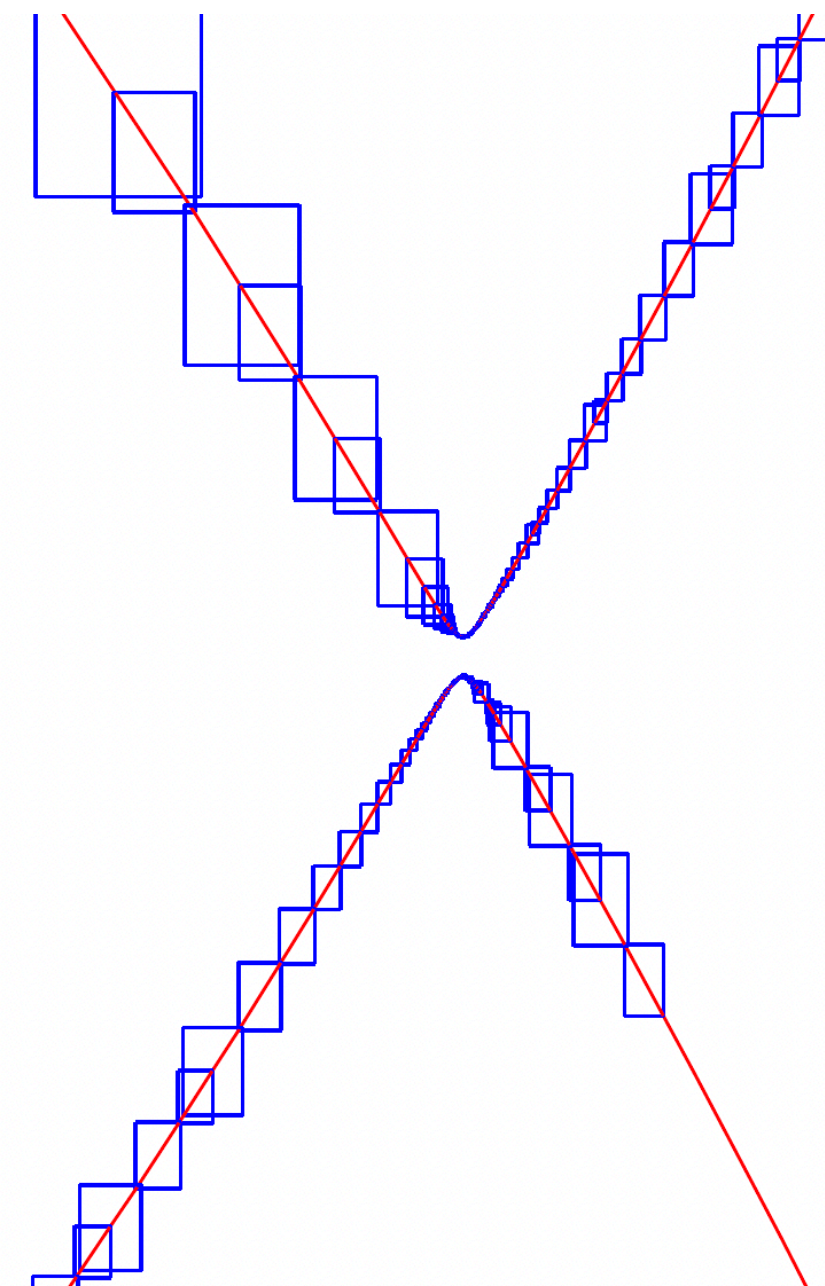
Refined with  $\rho = \frac{1}{16}$



See MS64, **Michael Byrd's** talk  
at 2:30, for further details about  
finalizing the approximation.



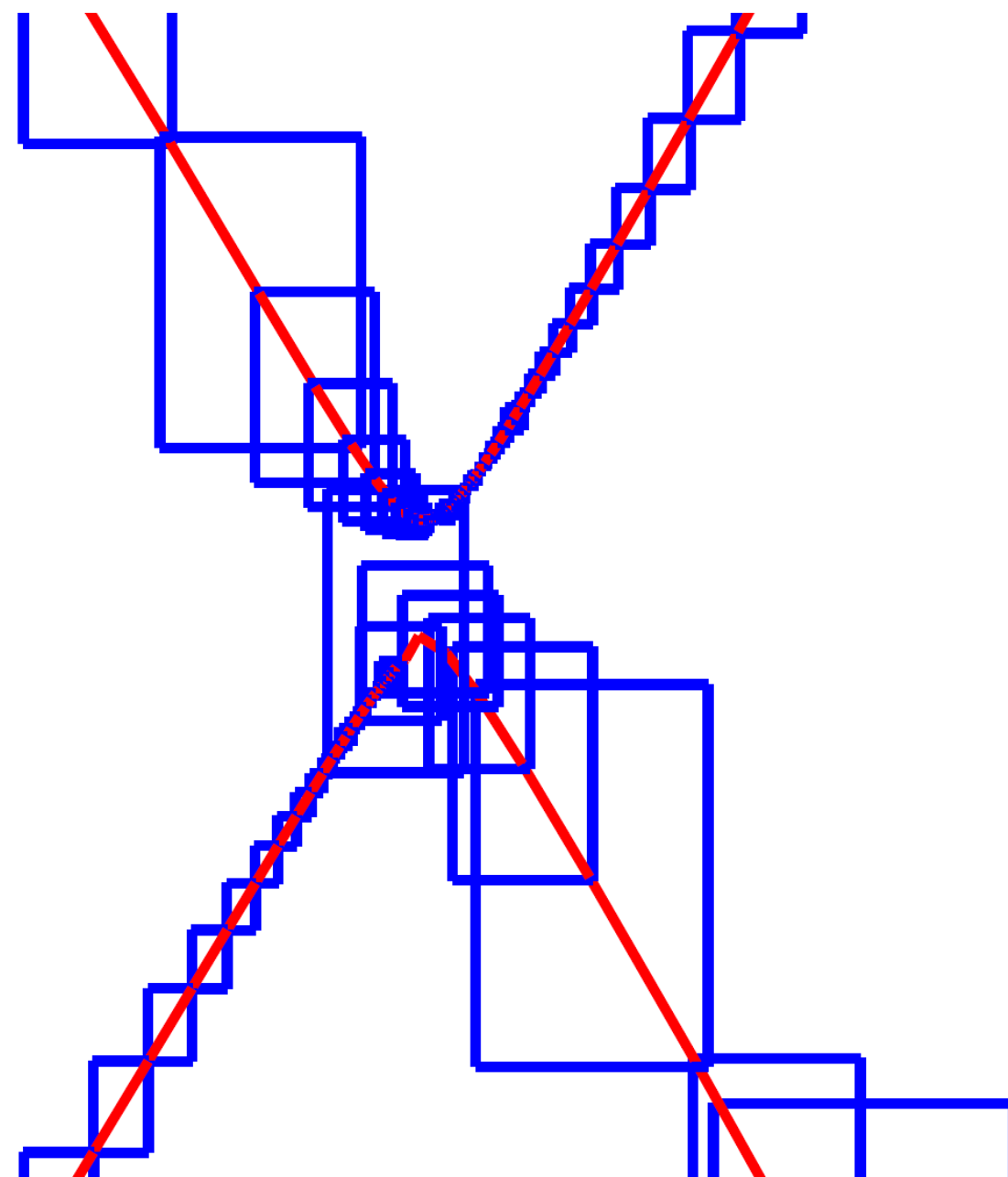
Refined with  $\rho = \frac{1}{8}$



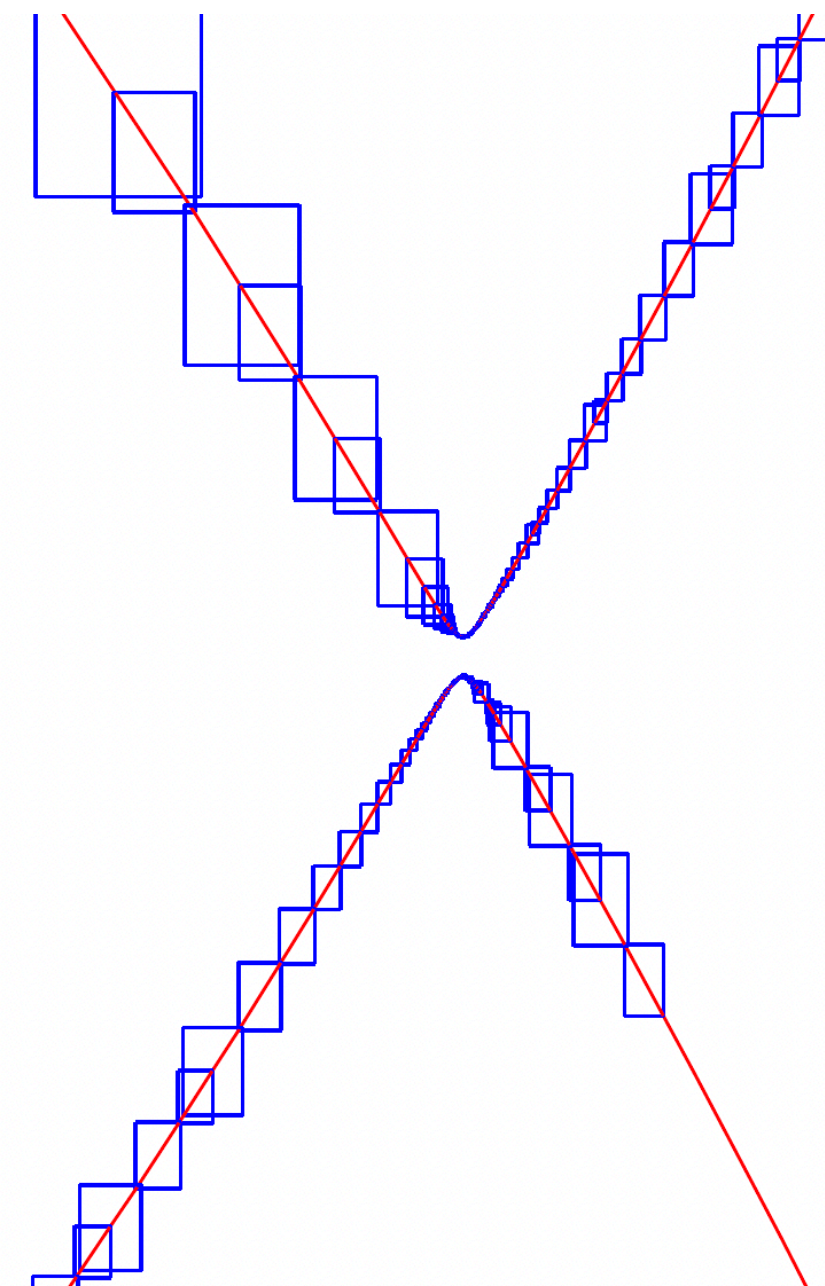
Refined with  $\rho = \frac{1}{16}$



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Refined with  $\rho = \frac{1}{8}$



Refined with  $\rho = \frac{1}{16}$

# Future directions

- Homotopy tracking with curve rotation
  - Can this improve tracking efficiency?
  - Will it have larger predictor steps?
- Certified approximation in higher dimensions
- Complexity analysis
  - The number of iterations to terminate?
  - The minimal step-size?



# Thank you for your attention!

Registration and travel support for this presentation was provided by the National Science Foundation.