

ASEN 5070  
Statistical Orbit Determination I  
Fall 2012



Professor Jeffrey S. Parker  
Professor George H. Born

Lecture 15: Numerical Compensations



University of Colorado  
Boulder

# Announcements

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- ▶ Exam 1
- ▶ Final Project
- ▶ Homework 5 is not graded...neither is the test.
- ▶ Homework 6 due Today
- ▶ Homework 7 due next week (Tuesday!)
  - It's okay to use Matlab to compute partials and to output them. But verify them.
- ▶ Concept Quizzes to resume Monday!
- ▶ Guest lecturer next week 10/25



# Exam 1 Debrief

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- ▶ Too easy?
- ▶ Too short?



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# Exam 1 Debrief

1. 20% Answer the following and be sure to provide sufficient explanation. If you need more space, use an additional sheet of paper.

- a. If one column of the  $\tilde{H}_i$  matrix is zero, the information matrix  $H_i^T H_i$  will always be singular. T or F

If it is always singular, provide an explanation for why.

If it is not always singular, provide an example when it is not.



# Exam 1 Debrief

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- a. If one column of the  $\tilde{H}_i$  matrix is zero, the information matrix  $H_i^T H_i$  will always be singular. T or F

If it is always singular, provide an explanation for why.

If it is not always singular, provide an example when it is not.

True. The transpose of a column of zeros becomes a row of zeros. That row propagates through the whole system and the HTH matrix becomes singular.



# Exam 1 Debrief

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- b. If one row of the  $H_i$  matrix is zero, the information matrix  $H_i^T H_i$  will always be singular. T or F

If it is always singular, provide an explanation for why.  
If it is not always singular, provide an example when it is not.



# Exam 1 Debrief

- b. If one row of the  $\tilde{H}_i$  matrix is zero, the information matrix  $\tilde{H}_i^T \tilde{H}_i$  will always be singular. T or F

If it is always singular, provide an explanation for why.  
If it is not always singular, provide an example when it is not.

False. Consider  $\tilde{H} = [a, b, 0]^T$  and  $\Phi = 1$ .



# Exam 1 Debrief

2. 15% A random variable has the probability density function given by:

$$f(x) = \begin{cases} 1/3, & 0 \leq x < 1, \\ k(x^3 + 1/4), & 1 \leq x \leq 2, \\ 0, & \text{elsewhere} \end{cases}$$

Find  $k$



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# Exam 1 Debrief

$$\int_{-\infty}^{\infty} f(x)dx = 1$$

$$\int_0^1 \frac{1}{3} dx + \int_1^2 k \left( x^3 + \frac{1}{4} \right) dx = 1$$

$$\frac{1}{3} [x]_0^1 + k \left[ \frac{x^4}{4} + \frac{1}{4}x \right]_1 = 1$$

$$\frac{1}{3} + k(4 + \frac{1}{2} - \frac{1}{4} - \frac{1}{4}) = 1$$

$$\frac{1}{3} + 4k = 1$$

$$k = \frac{2/3}{4} = \frac{1}{6}$$



# Exam 1 Debrief

3. 15% Given the observation-state equation

$$y(t_i) = (a + bx_0)t_i^2 + (c + x_1)t_i + d \quad i = 1, 2, \dots, l$$

where  $a, b, c, d, x_0$ , and  $x_1$  are constants and  $t_i$  are given. Which of the following state vectors are observable (assuming that the constants that aren't in the state are known perfectly):

1.  $\mathbf{X} = \begin{bmatrix} x_0 \\ x_1 \end{bmatrix}$
2.  $\begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix}$
3.  $\begin{bmatrix} b \\ x_0 \\ d \end{bmatrix}$
4.  $\begin{bmatrix} a \\ c \\ x_1 \end{bmatrix}$
5.  $\begin{bmatrix} b \\ c \\ d \end{bmatrix}$



# Exam 1 Debrief

3. 15% Given the observation-state equation

$$y(t_i) = (a + bx_0)t_i^2 + (c + x_1)t_i + d \quad i = 1, 2, \dots, l$$

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- |  |   |  |  |  |
|--|---|--|--|--|
| 1. $\mathbf{X} = \begin{bmatrix} x_0 \\ x_1 \end{bmatrix}$ | 2. $\begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix}$ | 3. $\begin{bmatrix} b \\ x_0 \\ d \end{bmatrix}$ | 4. $\begin{bmatrix} a \\ c \\ x_1 \end{bmatrix}$ | 5. $\begin{bmatrix} b \\ c \\ d \end{bmatrix}$ |
|--|---|--|--|--|

answer: 1 and 5 are observable



# Exam 1 Debrief

4. 10% Which weighting matrix is the most appropriate to use given two observations and the second is twice as accurate as the first:

1.  $W = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$
2.  $W = \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}$
3.  $W = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}$
4.  $W = \begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix}$
5.  $W = \begin{bmatrix} 2 & 1 \\ 2 & 1 \end{bmatrix}$



# Exam 1 Debrief

4. 10% Which weighting matrix is the most appropriate to use given two observations and the second is twice as accurate as the first:

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3.  $W = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}$
4.  $W = \begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix}$
5.  $W = \begin{bmatrix} 2 & 1 \\ 2 & 1 \end{bmatrix}$

answer: 3



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# Exam 1 Debrief

5. 40% The equation of motion for a particular system is given by

$$\ddot{z} - 3z^2\dot{z} = c$$

and we wish to estimate the state

$$\mathbf{X} = [ z(t_0), \dot{z}(t_0), c ]^T$$

using observations of  $z(t)$ . That is,  $z(t)$  is the observation. Answer the following.

- a. 5% What is the degree and order of this equation of motion?



# Exam 1 Debrief

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using observations of  $z(t)$ . That is,  $z(t)$  is the observation. Answer the following.

- a. 5% What is the degree and order of this equation of motion?

a. 1<sup>st</sup> degree, 2<sup>nd</sup> order



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- b. 5% Is this equation of motion linear or nonlinear?



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**b. Nonlinear**



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using observations of  $z(t)$ . That is,  $z(t)$  is the observation. Answer the following.

- c. 10% What is the  $A(t)$  matrix?



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using observations of  $z(t)$ . That is,  $z(t)$  is the observation. Answer the following.

- c. 10% What is the  $A(t)$  matrix?

$$\dot{\mathbf{X}}(t) = \begin{bmatrix} \dot{z} \\ \ddot{z} \\ \dot{c} \end{bmatrix} = \begin{bmatrix} \dot{z} \\ 3z^2\dot{z} + c \\ 0 \end{bmatrix}$$

$$\begin{aligned} A(t) &= \frac{\partial \dot{\mathbf{X}}}{\partial \mathbf{X}} = \begin{bmatrix} \frac{\partial \dot{z}}{\partial z} & \frac{\partial \dot{z}}{\partial \dot{z}} & \frac{\partial \dot{z}}{\partial c} \\ \frac{\partial \ddot{z}}{\partial z} & \frac{\partial \ddot{z}}{\partial \dot{z}} & \frac{\partial \ddot{z}}{\partial c} \\ \frac{\partial \dot{c}}{\partial z} & \frac{\partial \dot{c}}{\partial \dot{z}} & \frac{\partial \dot{c}}{\partial c} \end{bmatrix} \\ &= \begin{bmatrix} 0 & 1 & 0 \\ 6z\dot{z} & 3z^2 & 1 \\ 0 & 0 & 0 \end{bmatrix} \end{aligned}$$



# Exam 1 Debrief

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using observations of  $z(t)$ . That is,  $z(t)$  is the observation. Answer the following.

- d. 5% What is the  $\tilde{H}$  matrix?



# Exam 1 Debrief

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using observations of  $z(t)$ . That is,  $z(t)$  is the observation. Answer the following.

d. 5% What is the  $\tilde{H}$  matrix?

$$\begin{aligned}\tilde{H}(t) &= \frac{\partial z}{\partial \mathbf{X}} = \left[ \begin{array}{ccc} \frac{\partial z}{\partial z} & \frac{\partial z}{\partial \dot{z}} & \frac{\partial z}{\partial c} \end{array} \right] \\ &= \left[ \begin{array}{ccc} 1 & 0 & 0 \end{array} \right]\end{aligned}$$

$G(t) = z(t)$  is the simplest representation of the observation.  
 If you solve the EOM for  $z$ , you can also get partials wrt  $z$ -dot and  $c$ , which is more information (optional).



# Exam 1 Debrief

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using observations of  $z(t)$ . That is,  $z(t)$  is the observation. Answer the following.

- e. 10% How would you determine the state transition matrix,  $\Phi(t_i, t_0)$ ? Note: don't attempt to compute it. Can you use a Laplace Transform? Why/why not?



# Exam 1 Debrief

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- e. 10% How would you determine the state transition matrix,  $\Phi(t_i, t_0)$ ? Note: don't attempt to compute it. Can you use a Laplace Transform? Why/why not?

$$\begin{aligned}\dot{\Phi}(t, t_0) &= A(t)\Phi(t, t_0) \\ &= \begin{bmatrix} 0 & 1 & 0 \\ 6z\dot{z} & 3z^2 & 1 \\ 0 & 0 & 0 \end{bmatrix} \Phi(t, t_0)\end{aligned}$$

Can't use a Laplace  
Transform because A(t)  
is time-dependent!



# Exam 1 Debrief

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using observations of  $z(t)$ . That is,  $z(t)$  is the observation. Answer the following.

- f. 5% What is the minimum number of independent observations of  $z$  that is required to generate an estimate of the state?



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- f. 5% What is the minimum number of independent observations of  $z$  that is required to generate an estimate of the state?

f. 3



# Topics coming up

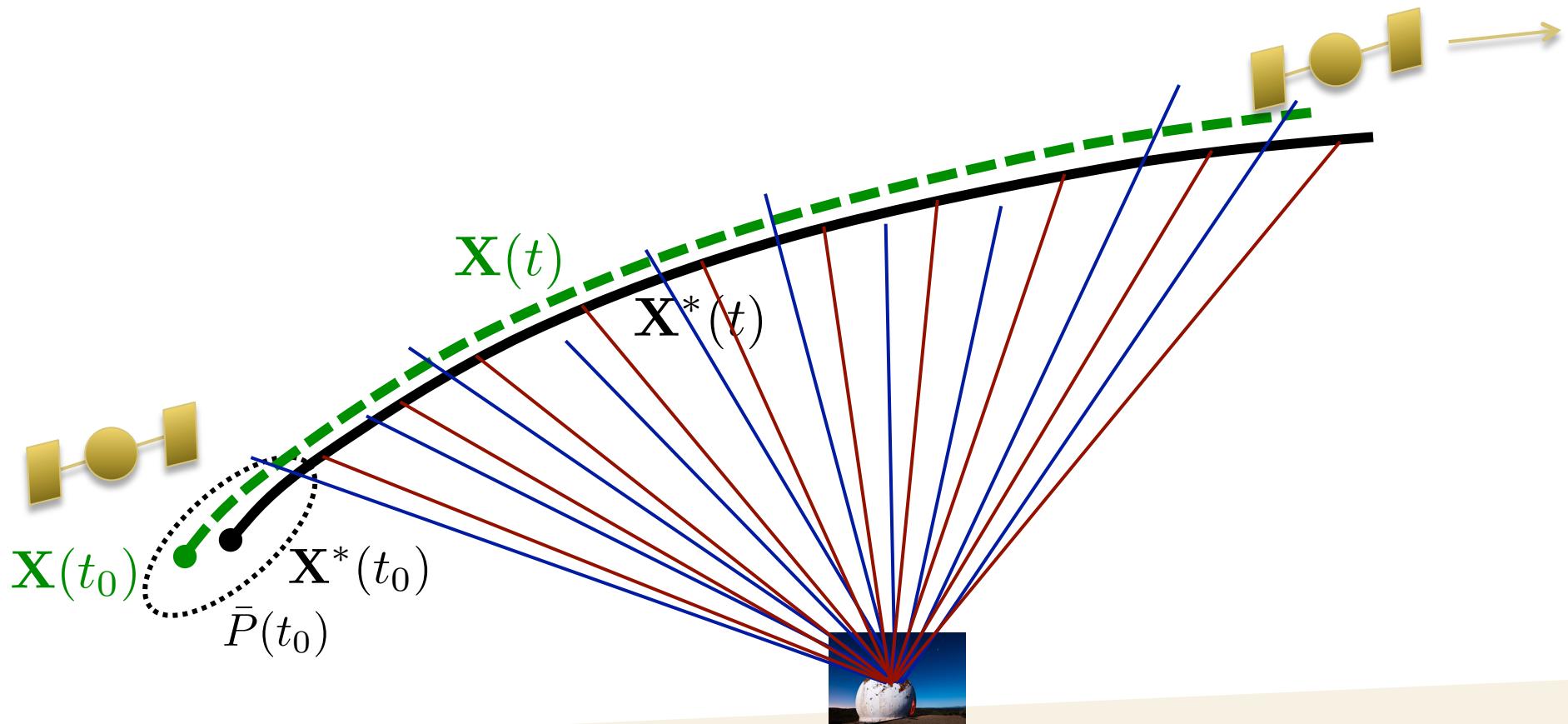
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- ▶ Conventional Kalman Filter (CKF)
- ▶ Extended Kalman Filter (EKF)
  
- ▶ Numerical Issues
  - Machine precision
  - Covariance collapse
  
- ▶ Numerical Compensation
  - Joseph, Potter, Cholesky, Square-root free, unscented, Givens, orthogonal transformation, SVD
  - State Noise Compensation, Dynamical Model Compensation



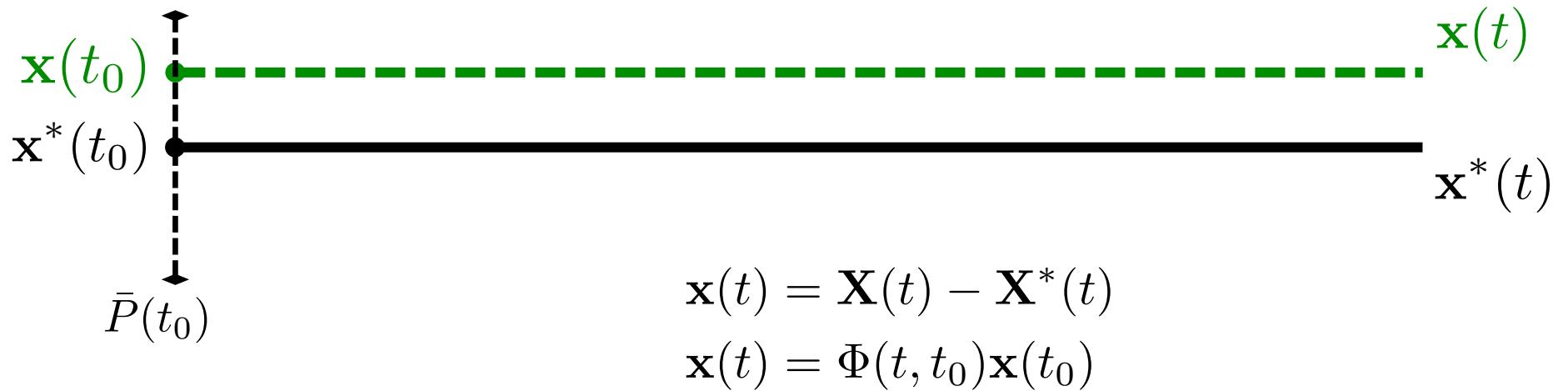
# Stat OD Conceptualization

- ▶ Full, nonlinear system:



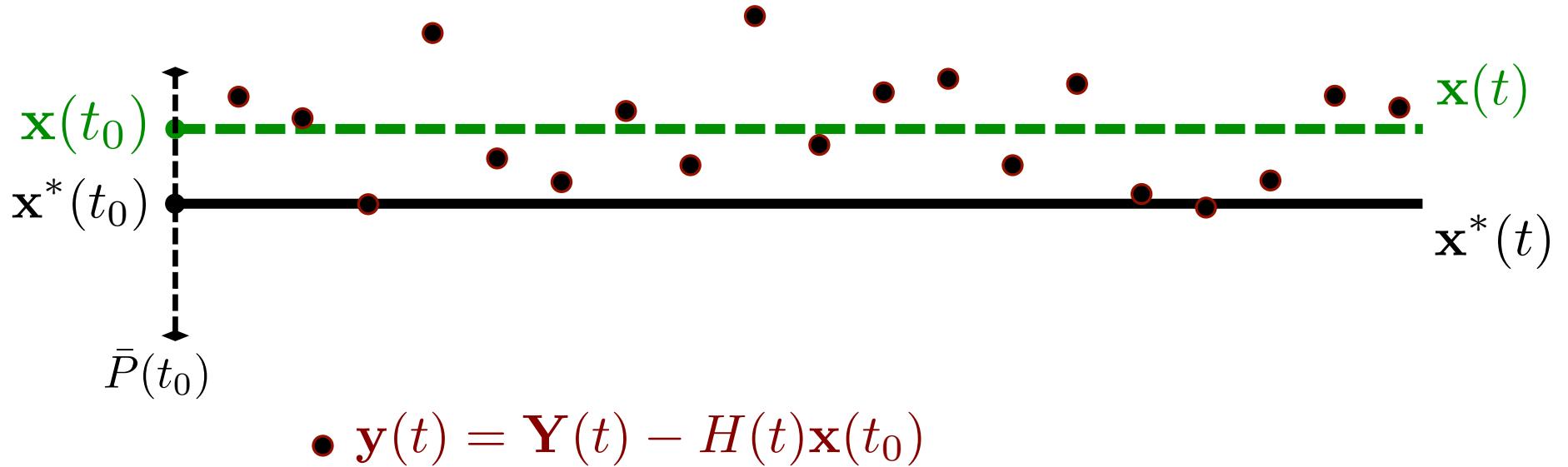
# Stat OD Conceptualization

## ► Linearization



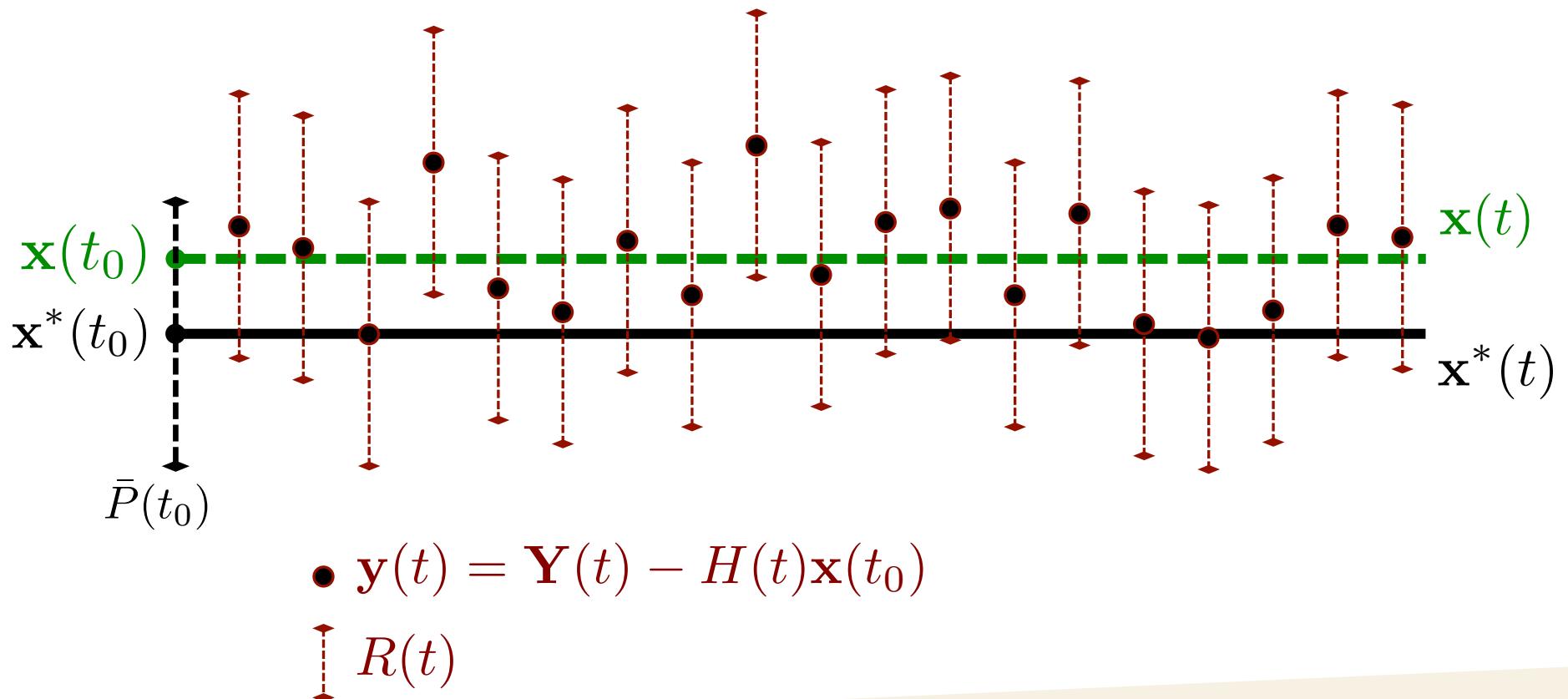
# Stat OD Conceptualization

## ► Observations



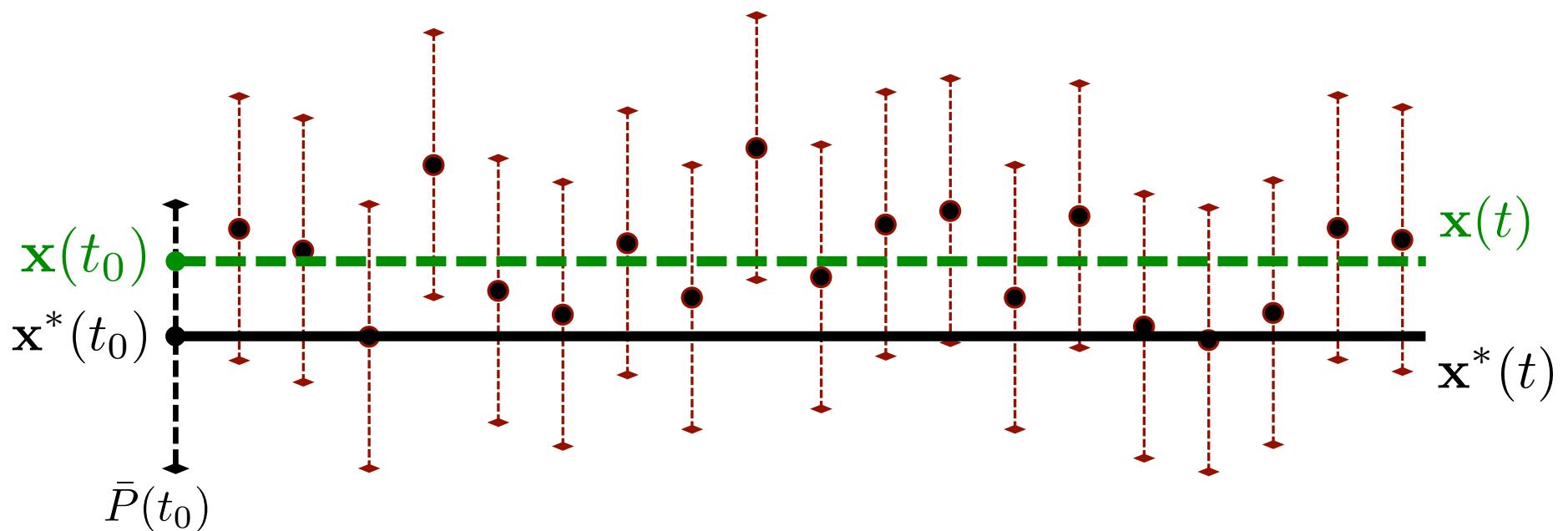
# Stat OD Conceptualization

## ► Observation Uncertainties



# Stat OD Conceptualization

## ► Least Squares (Batch)

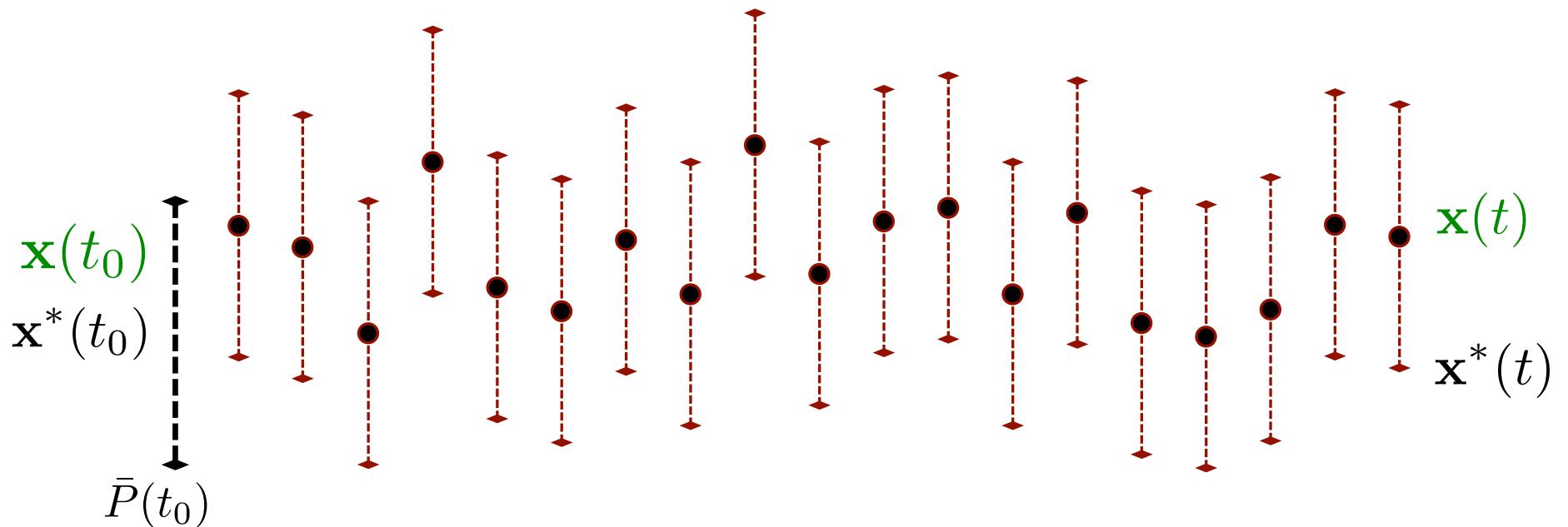


$$\hat{\mathbf{x}}(t_0) = (H^T R^{-1} H + \bar{P}^{-1})^{-1} (H^T R^{-1} \mathbf{y} + \bar{P}^{-1} \bar{\mathbf{x}})$$



# Stat OD Conceptualization

## ► Least Squares (Batch)

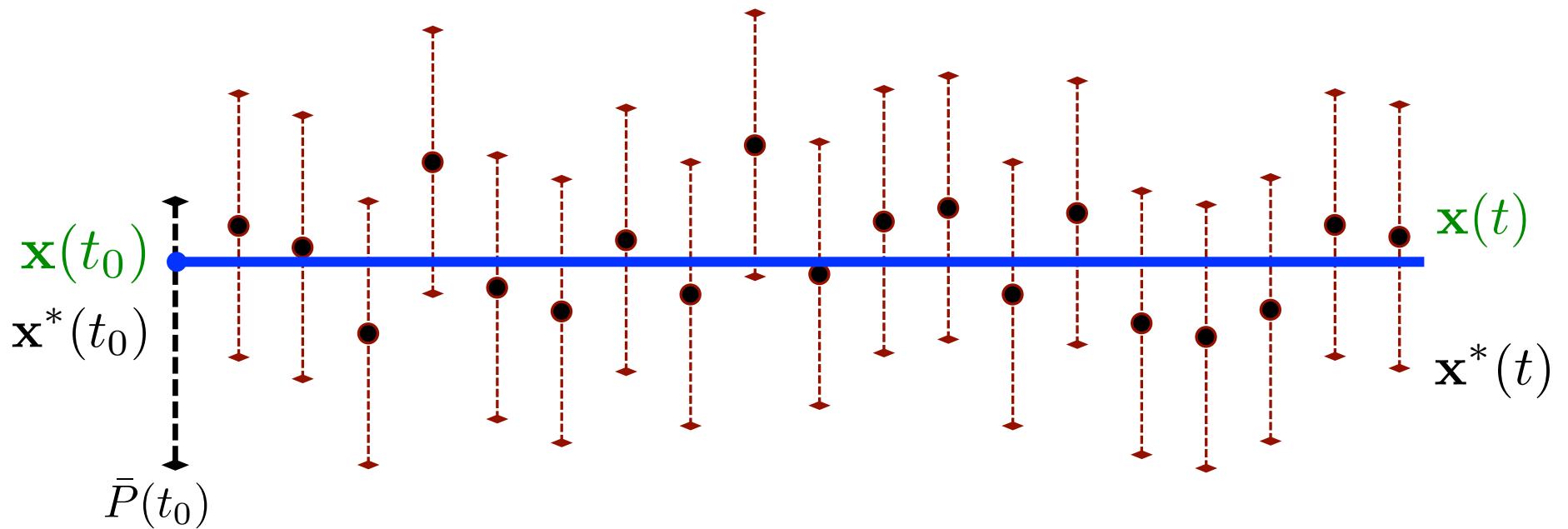


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# Stat OD Conceptualization

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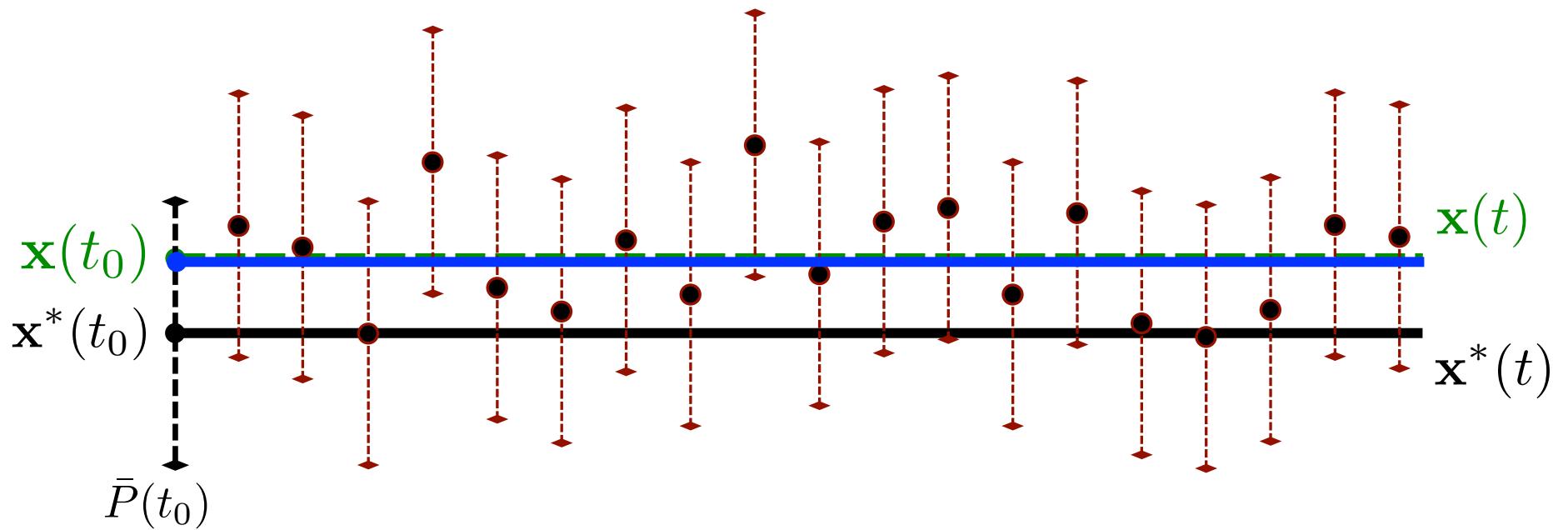


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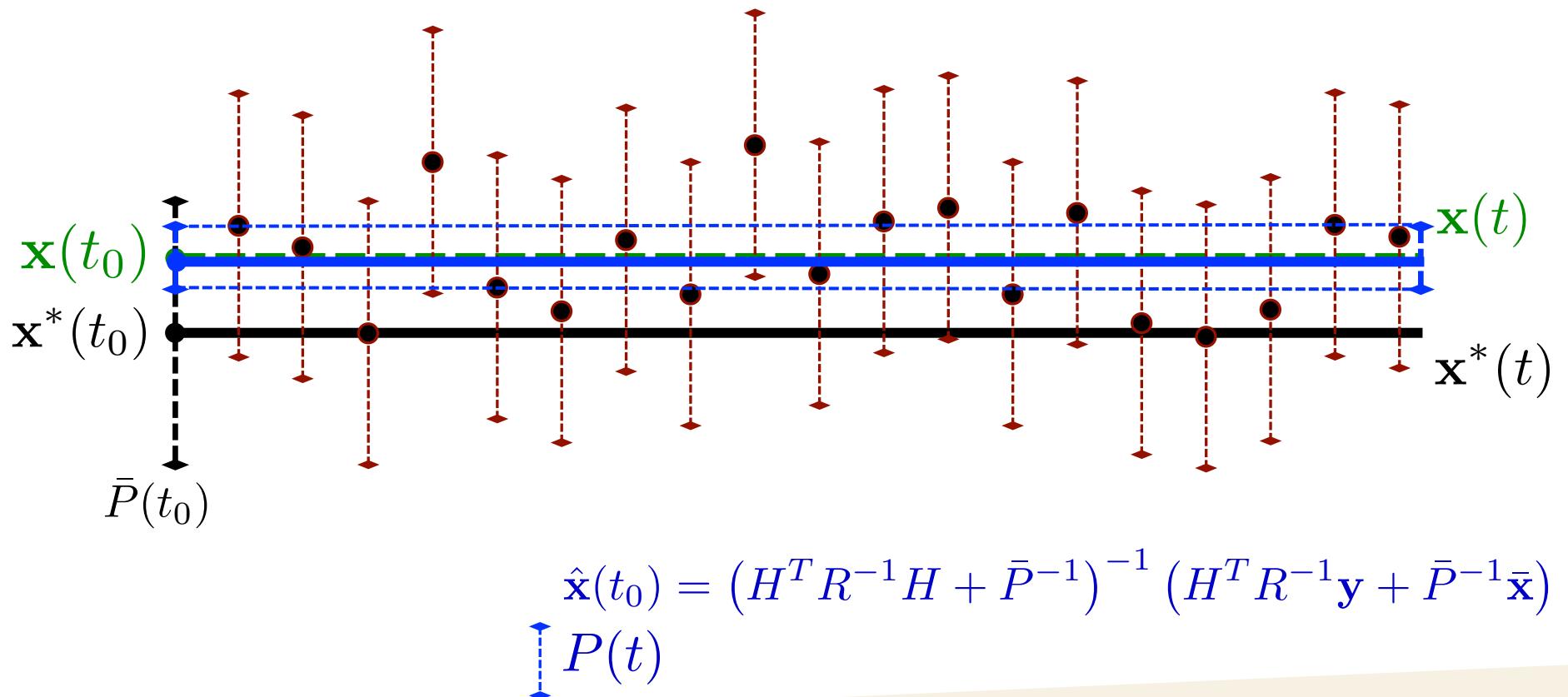


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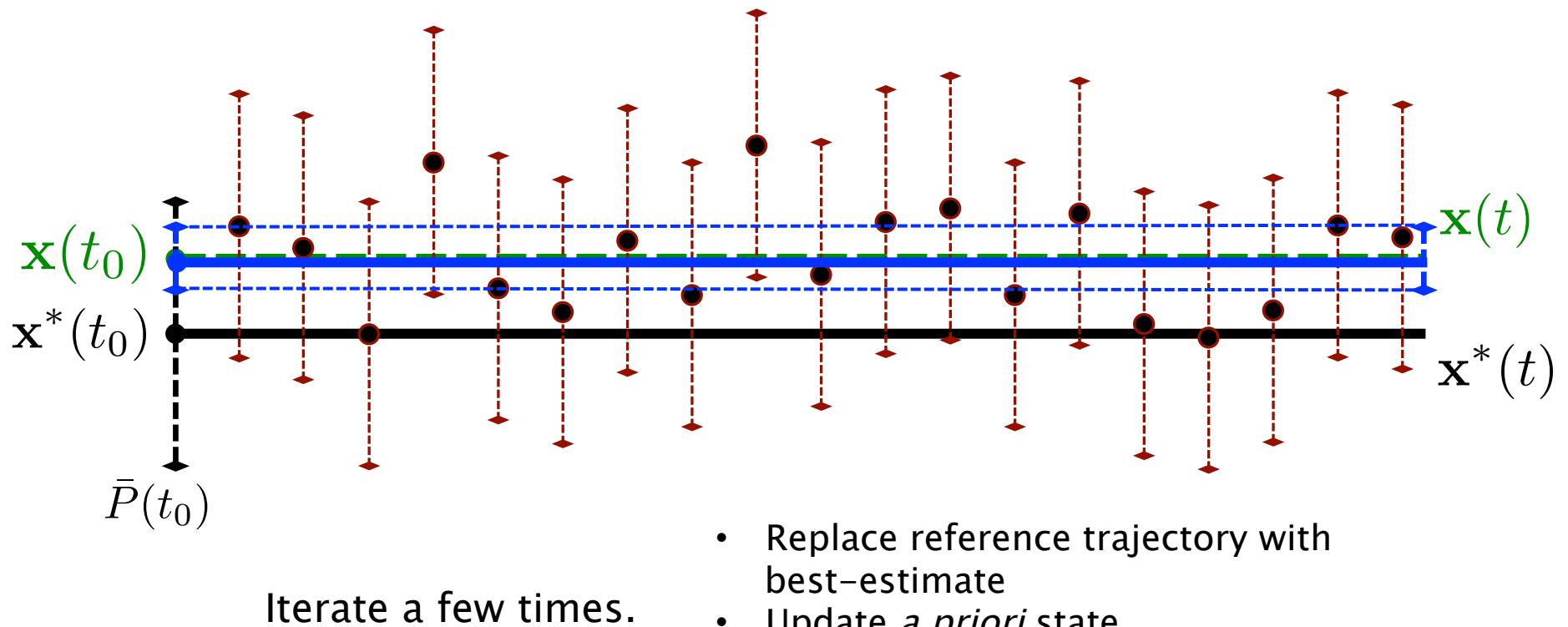
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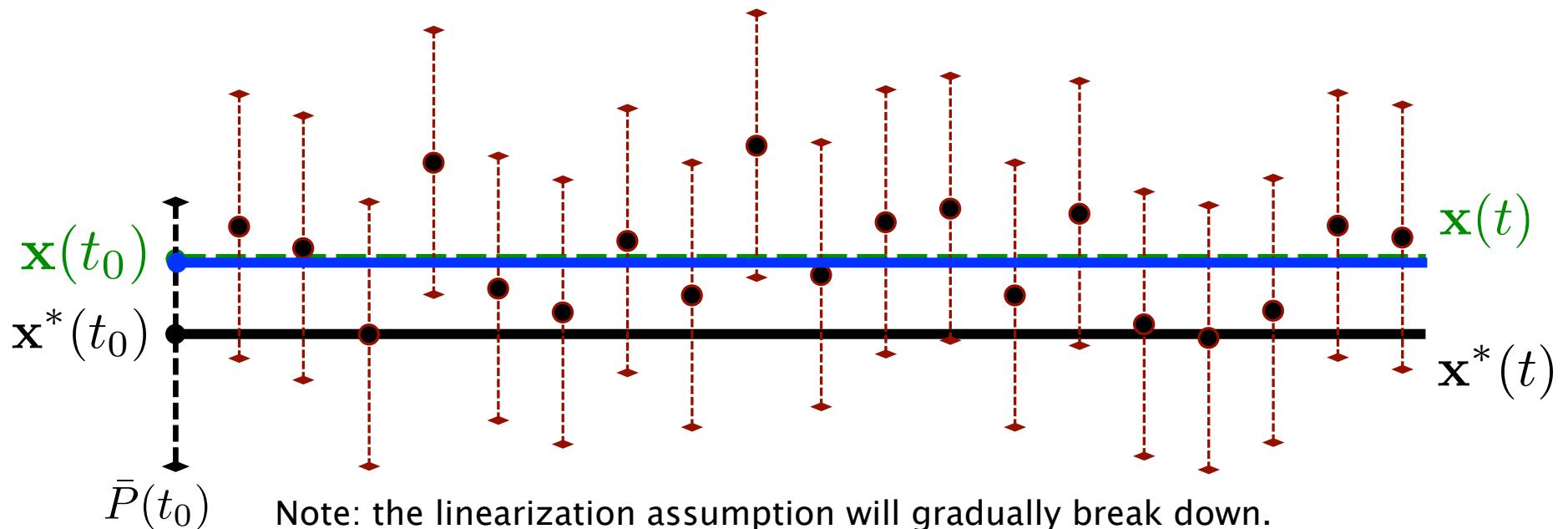
# Stat OD Conceptualization

## ► Least Squares (Batch)



# Stat OD Conceptualization

## ► Least Squares (Batch)



Select a measurement interval that balances the number of observations with the length of time being used.

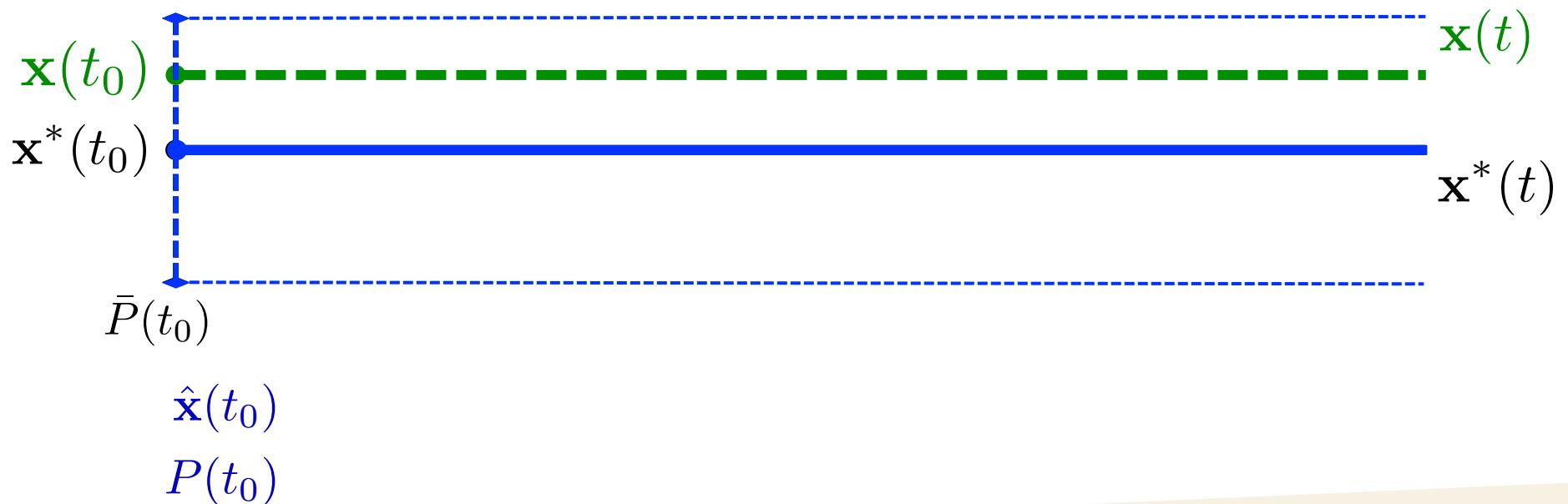


- ▶ Conceptualization of the Conventional Kalman Filter (Sequential Filter)



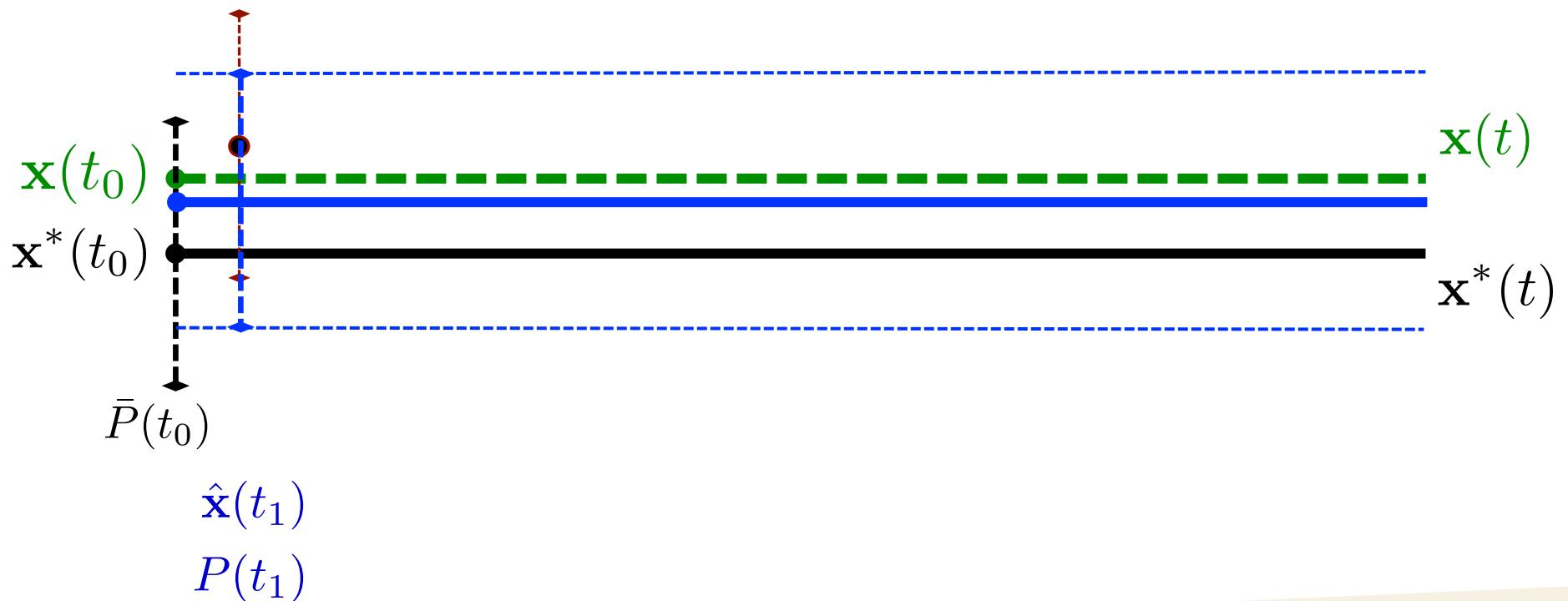
# Stat OD Conceptualization

## ► Conventional Kalman



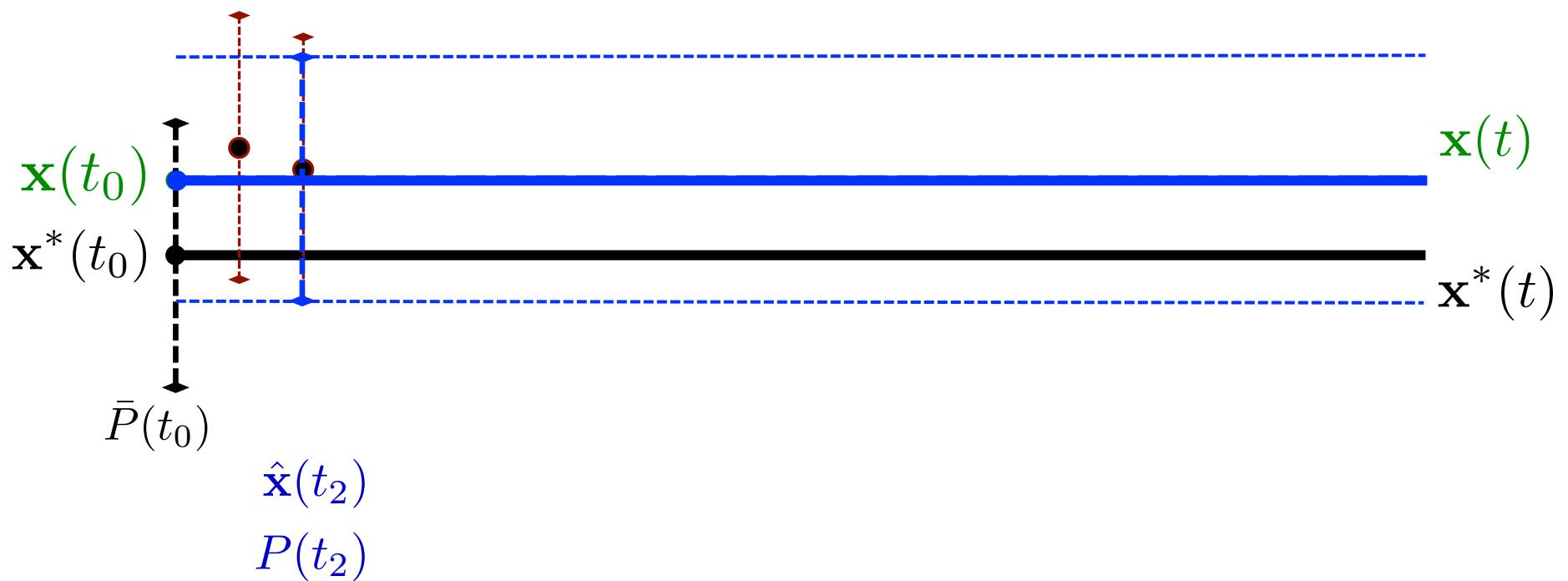
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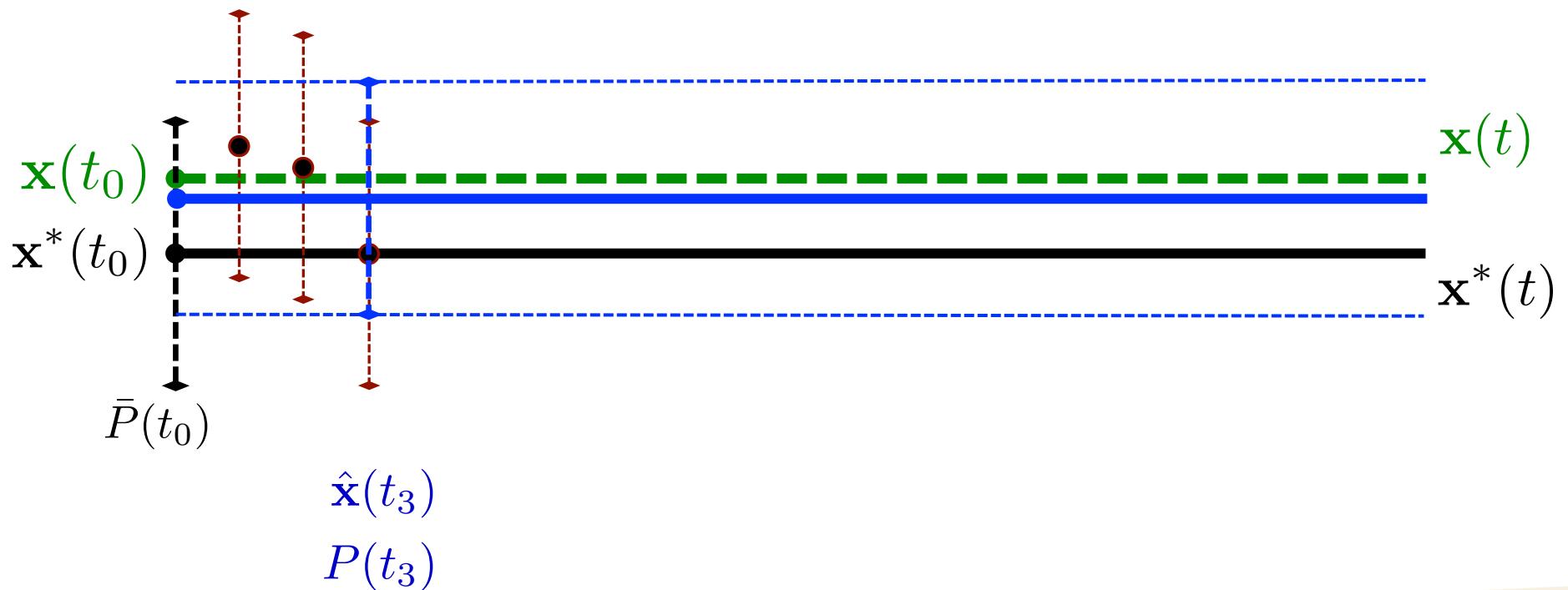
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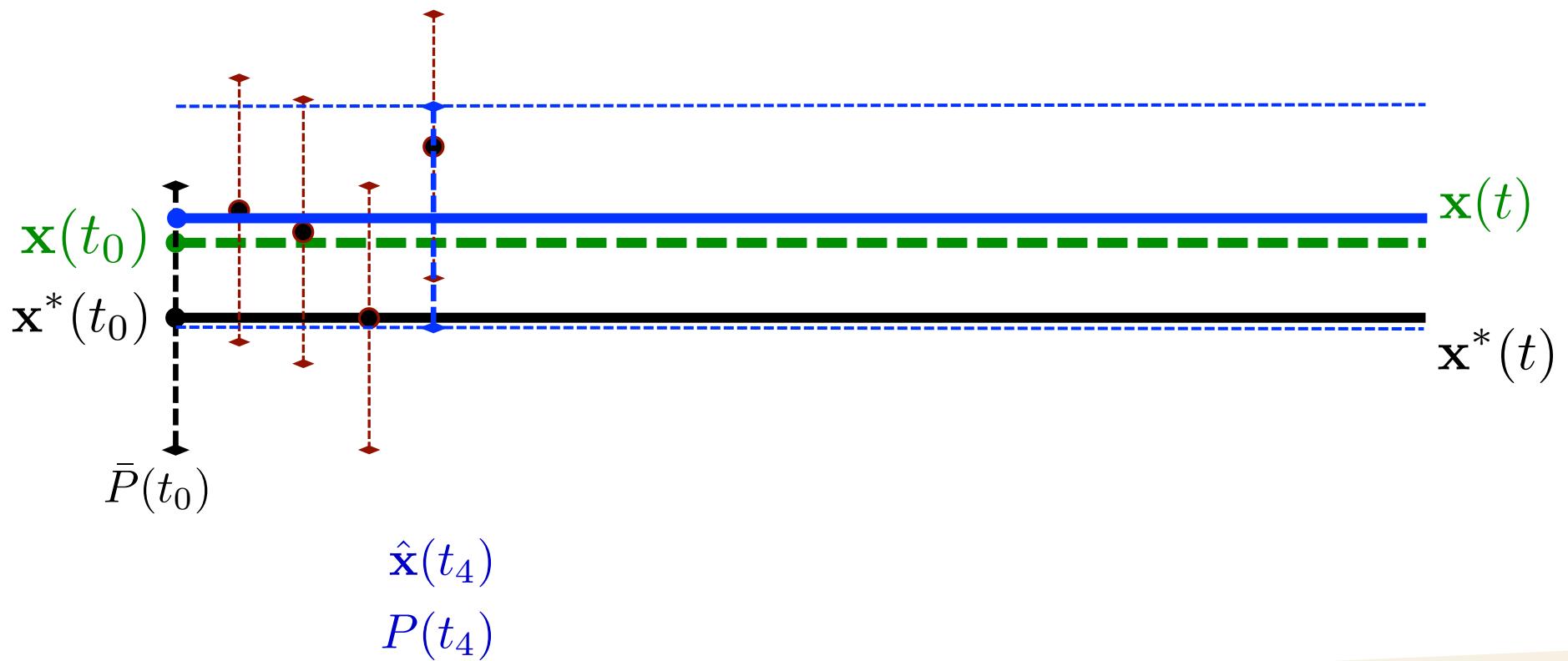
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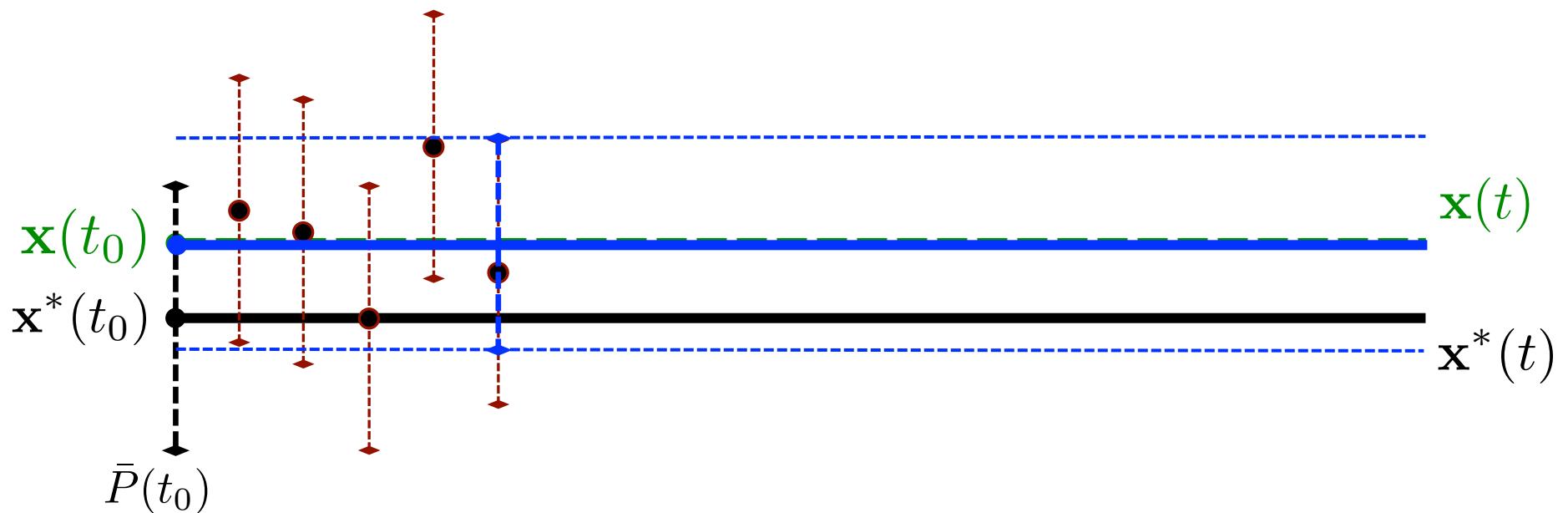
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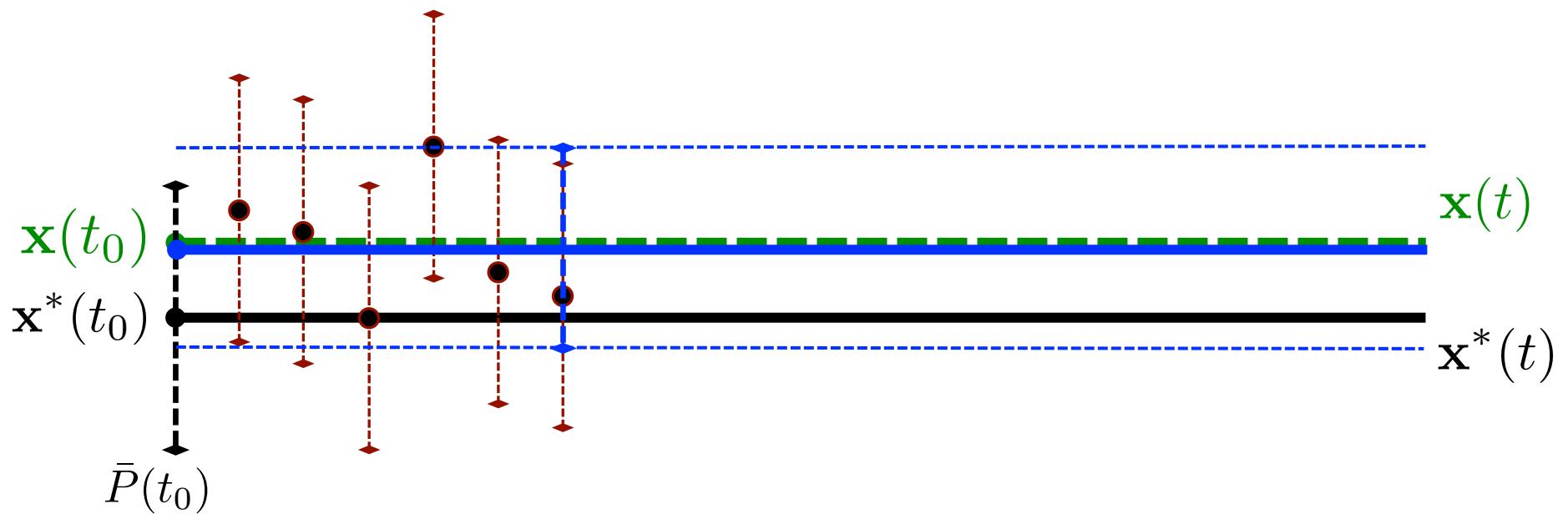
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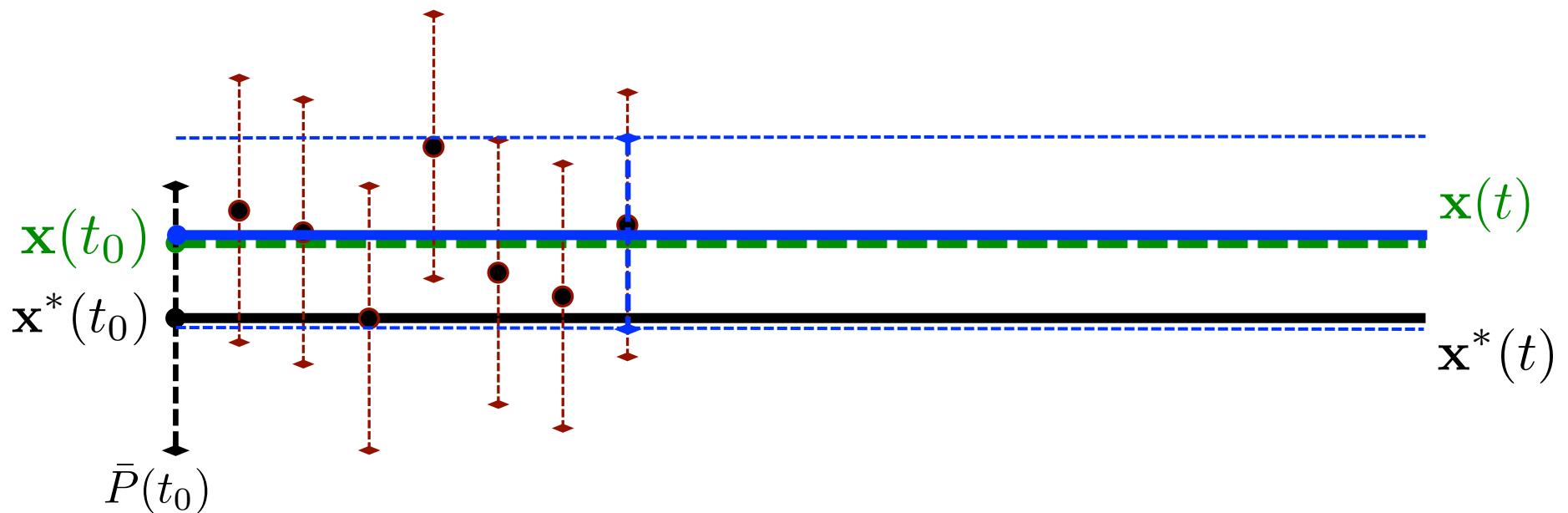
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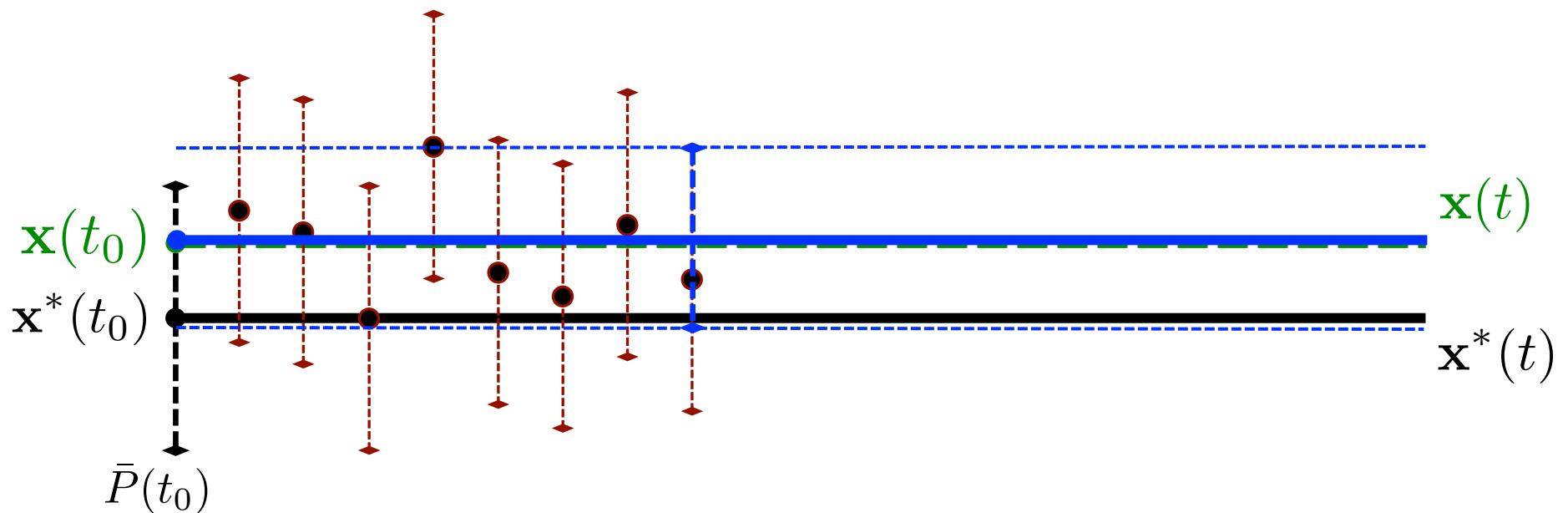
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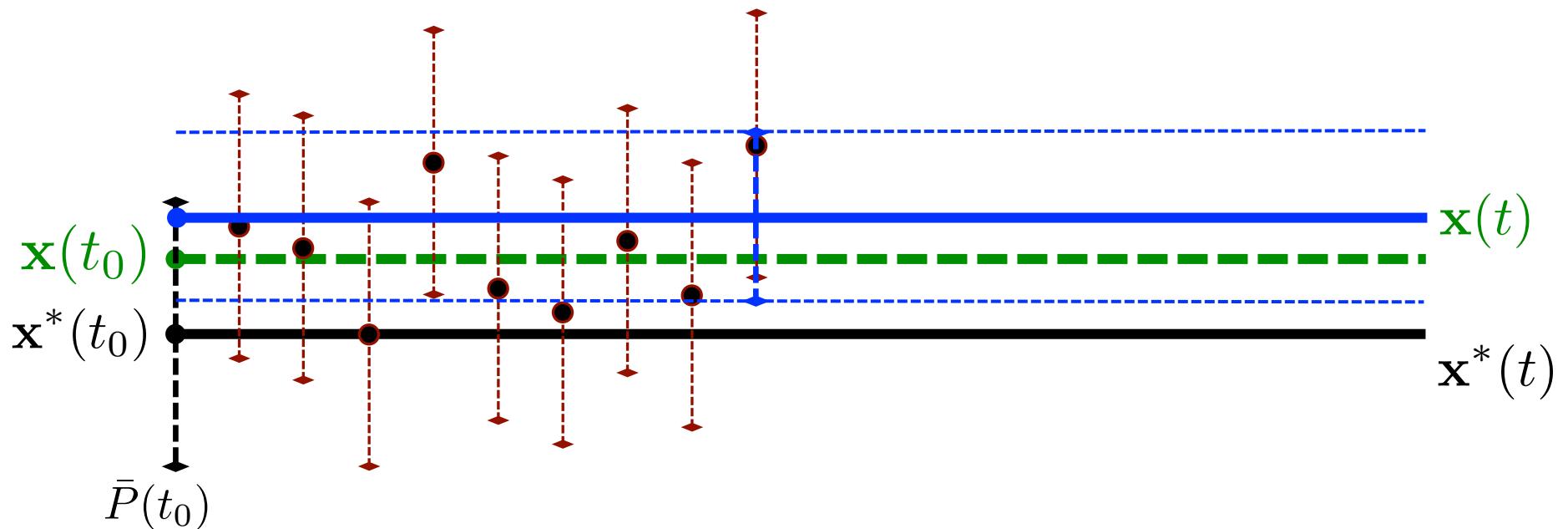
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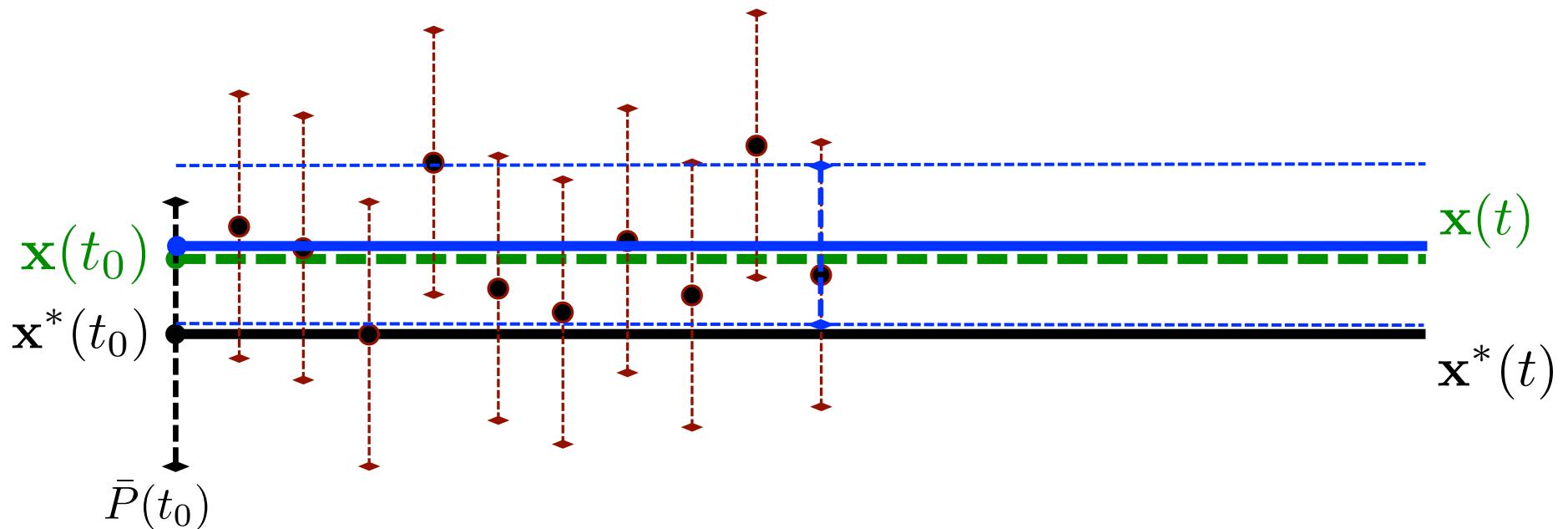
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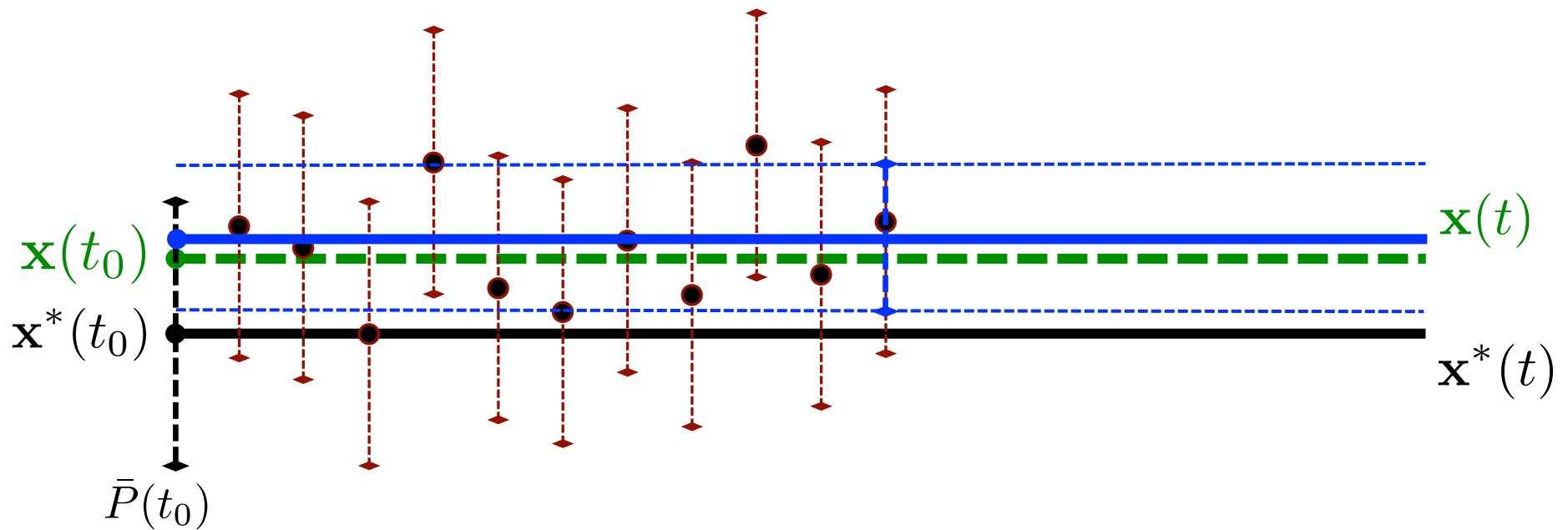
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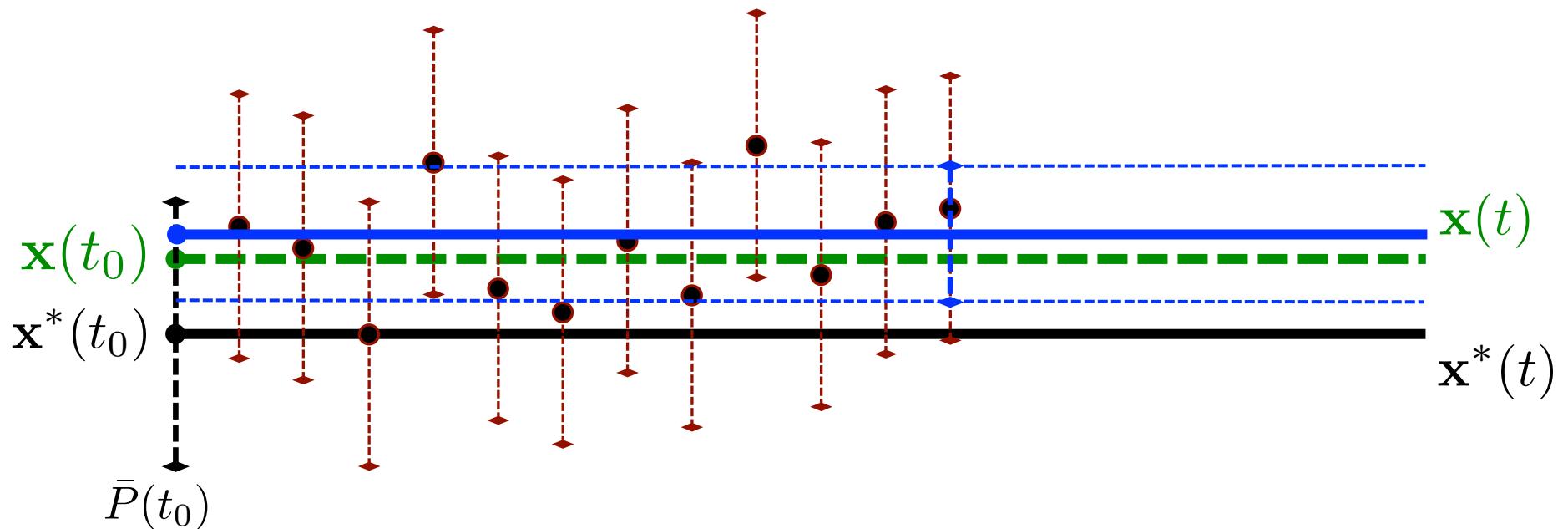
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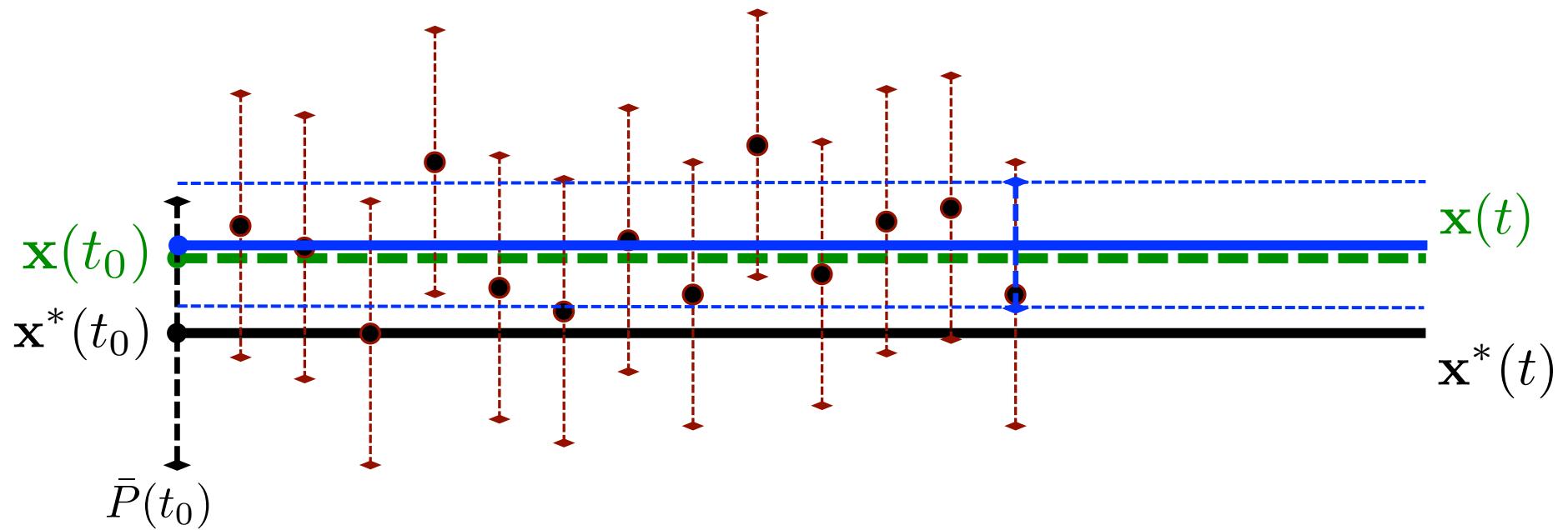
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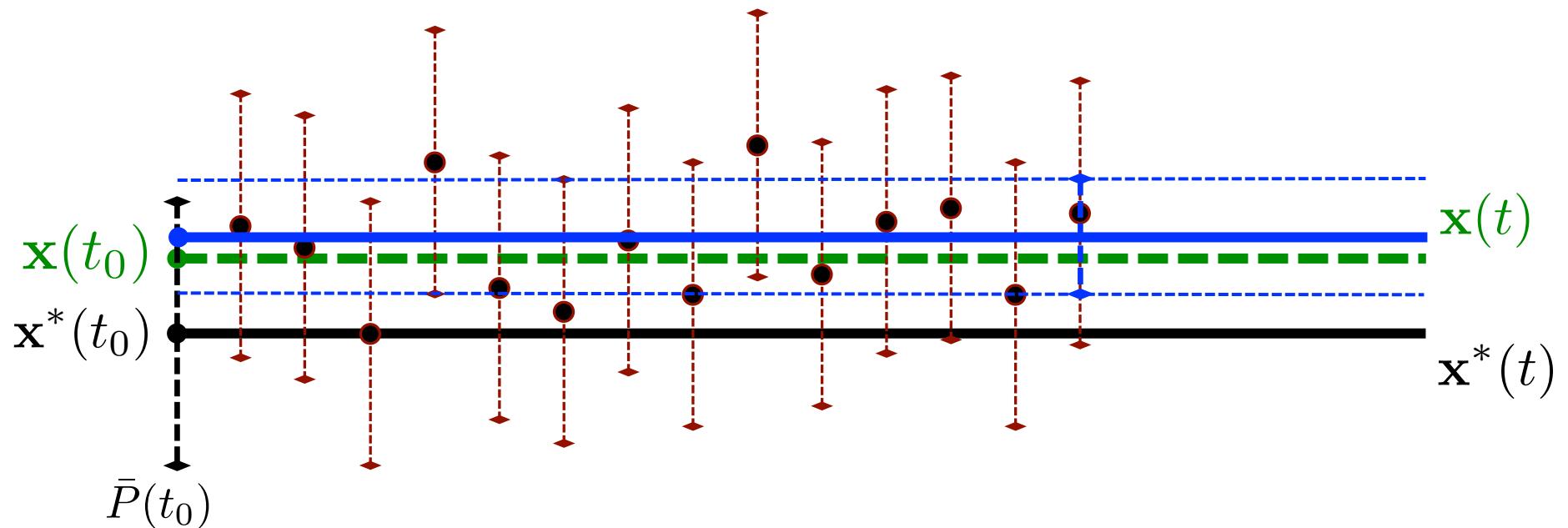
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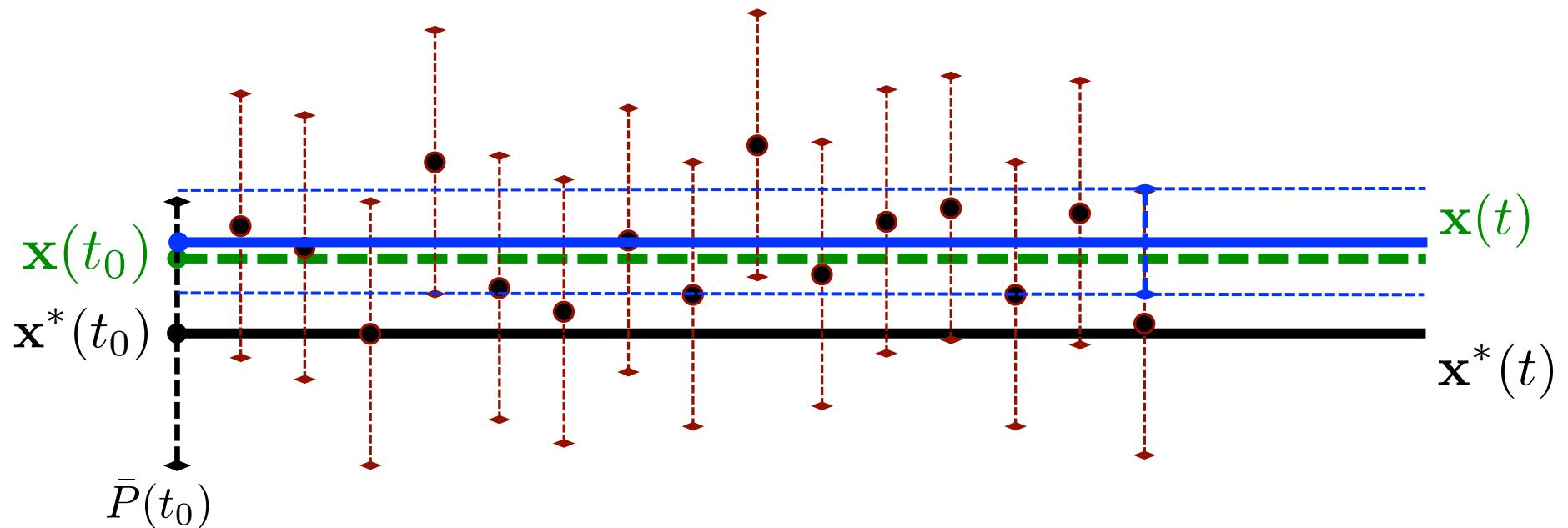
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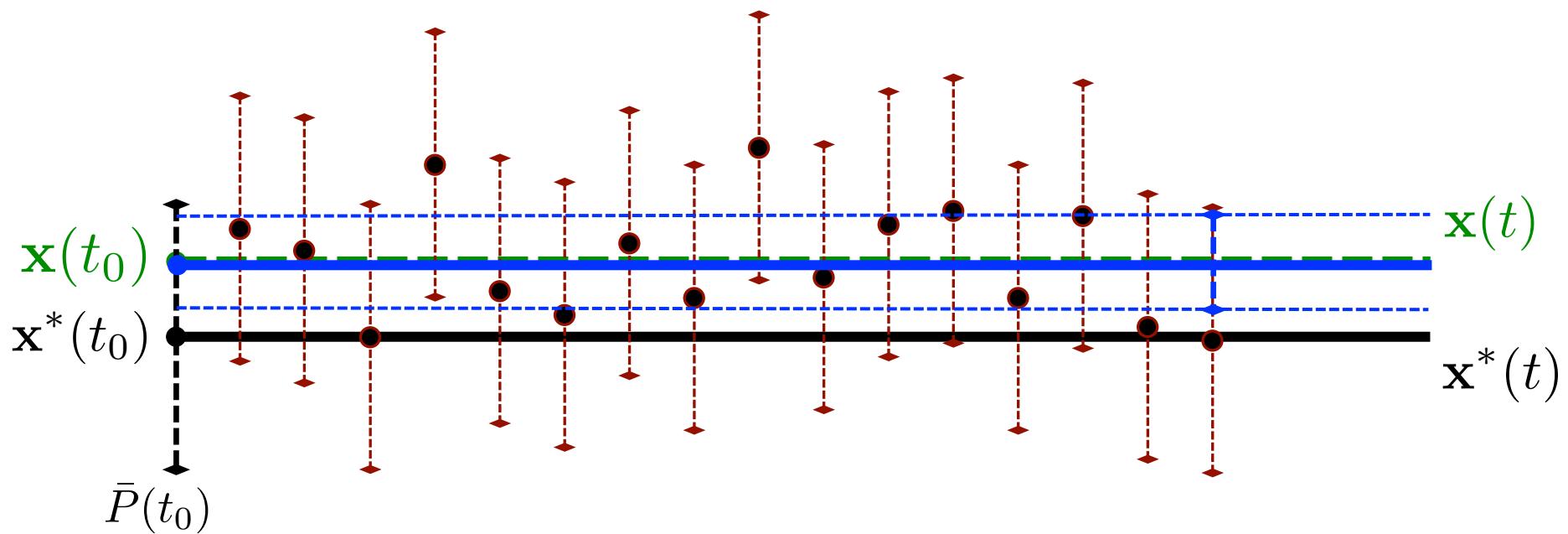
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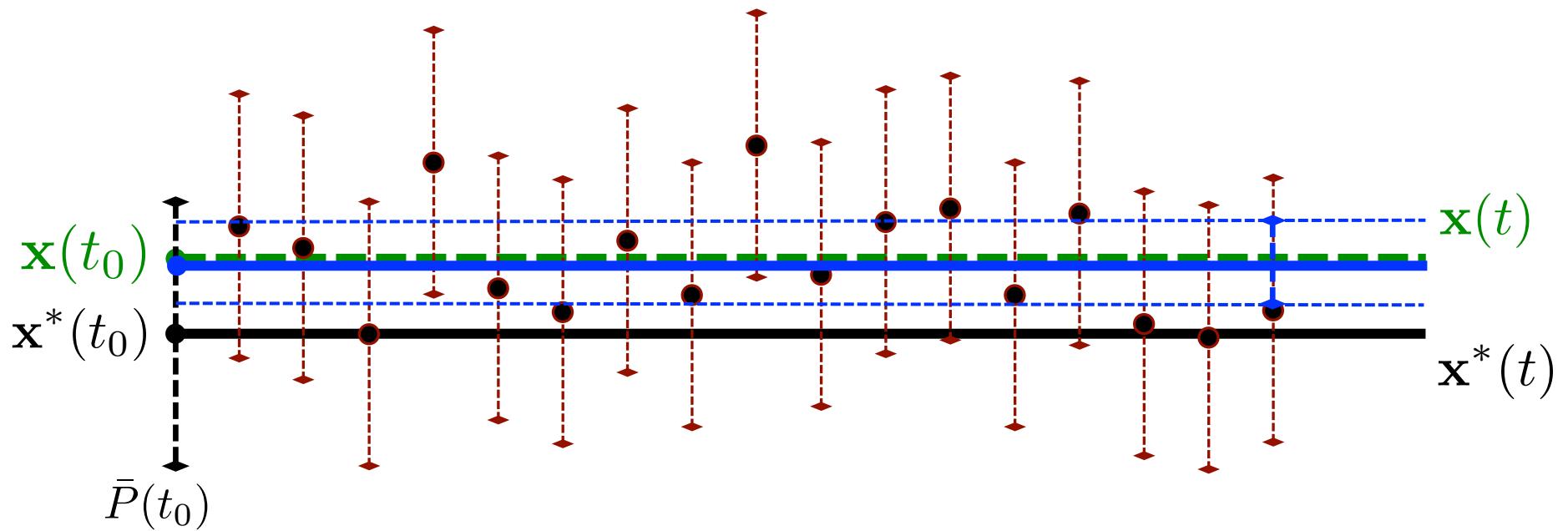
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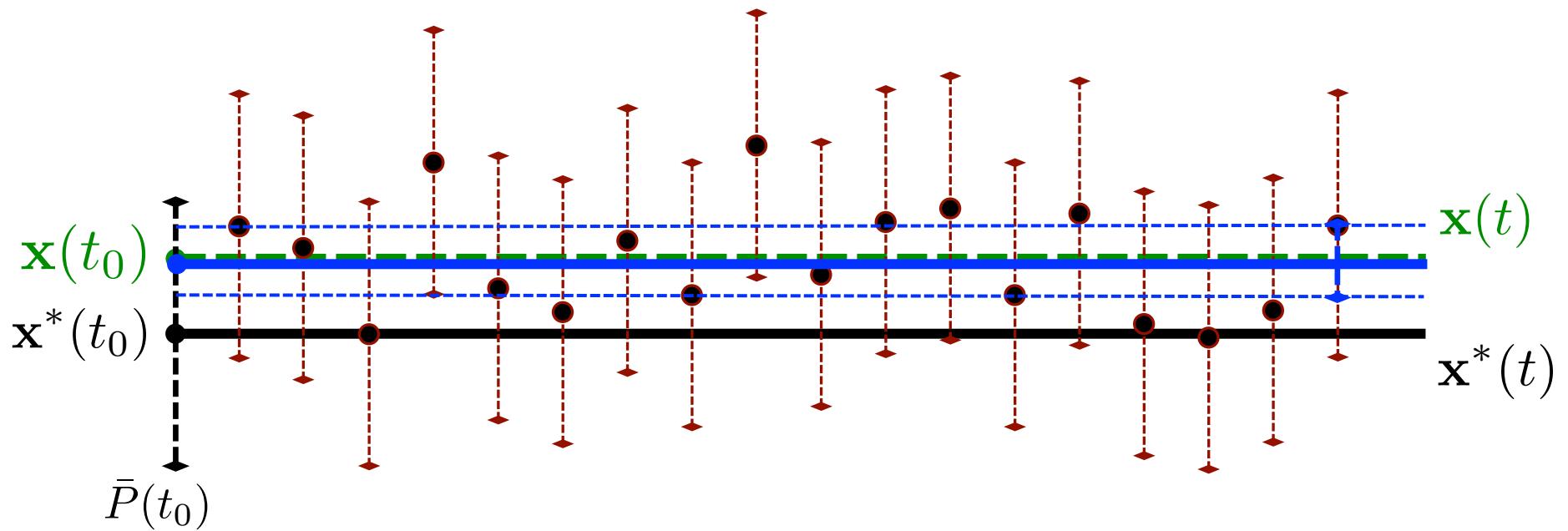
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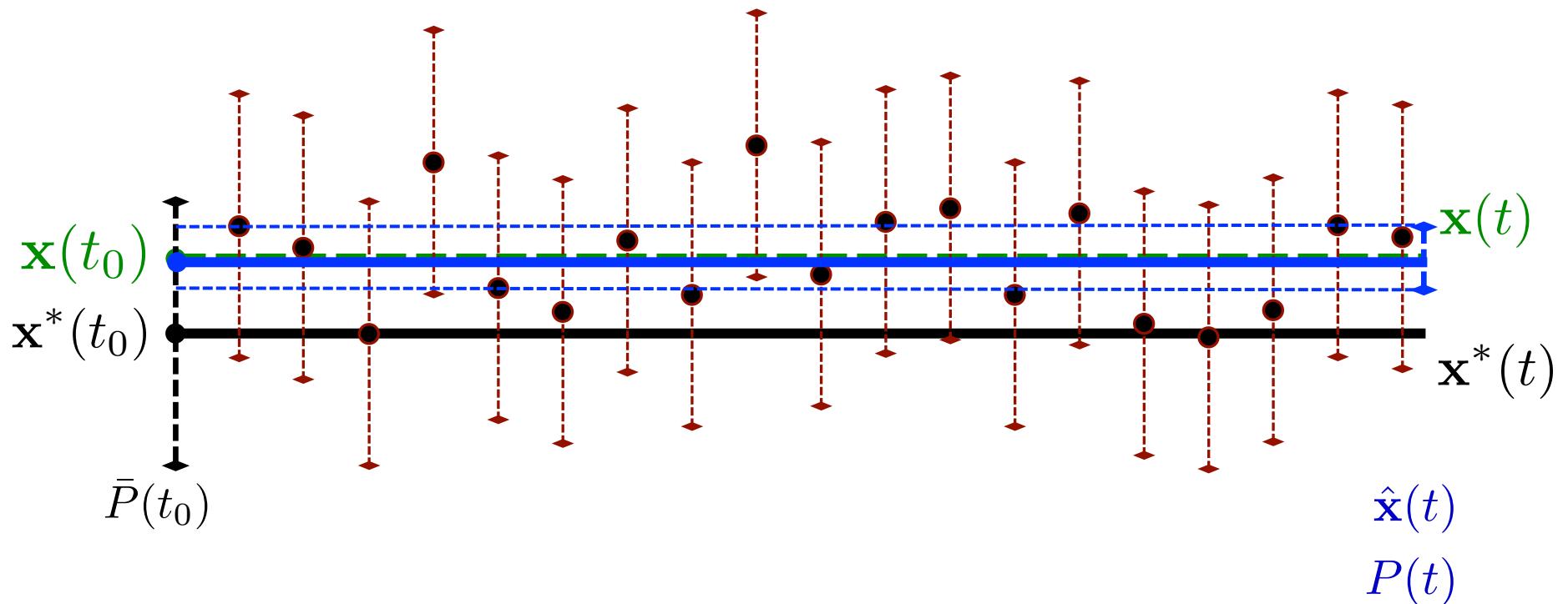
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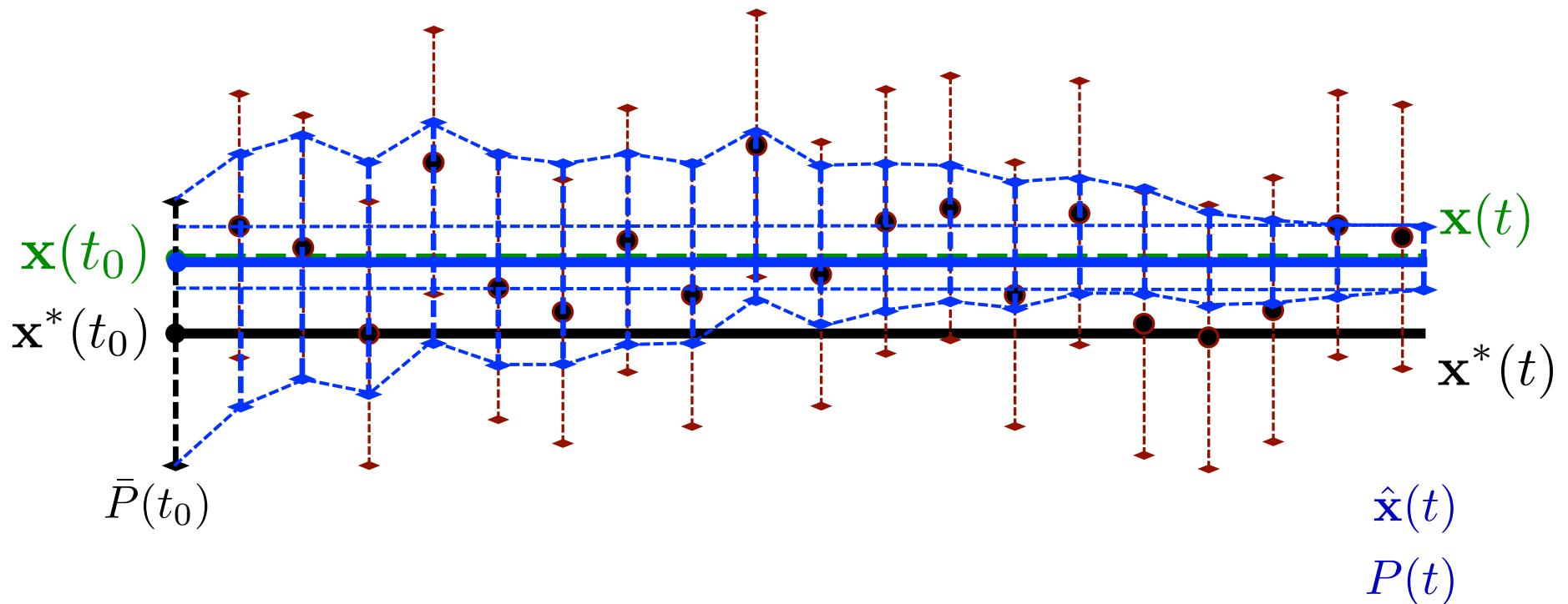
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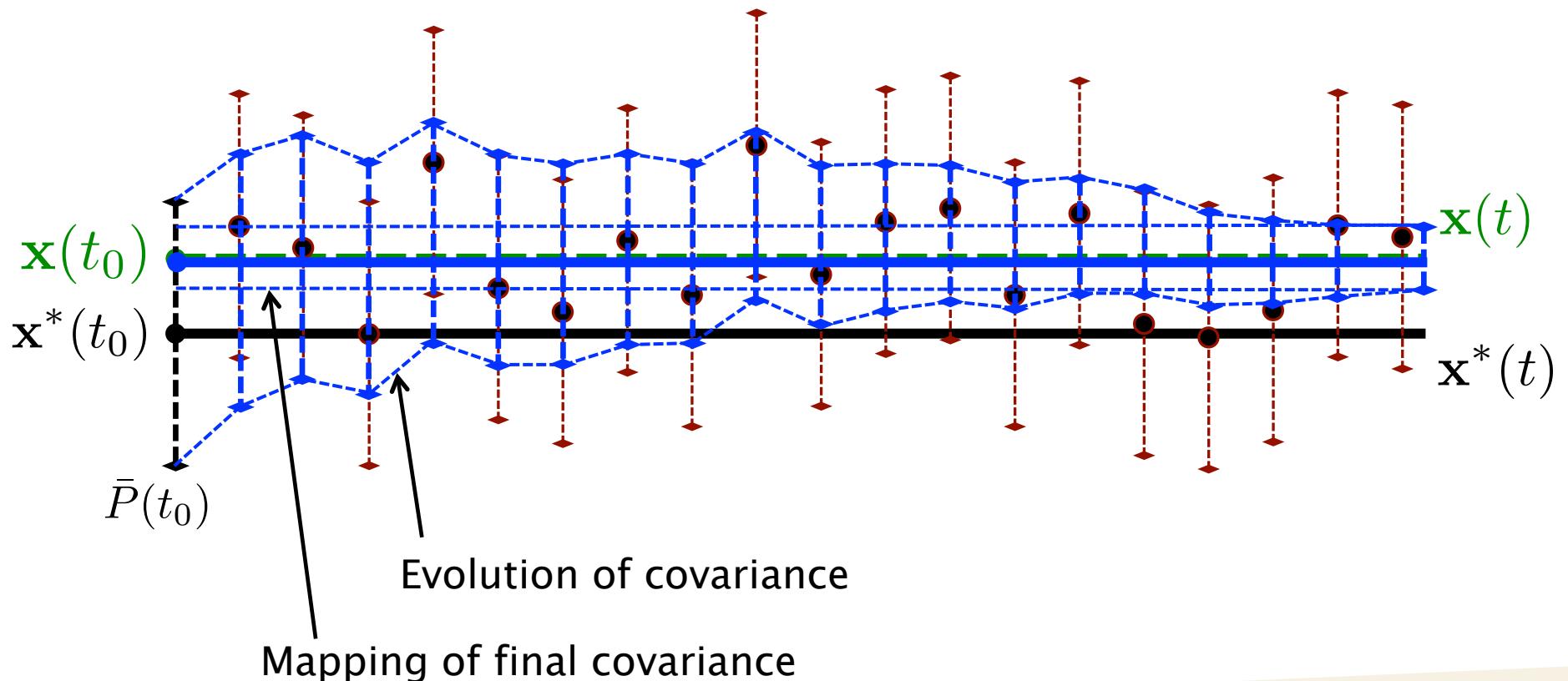
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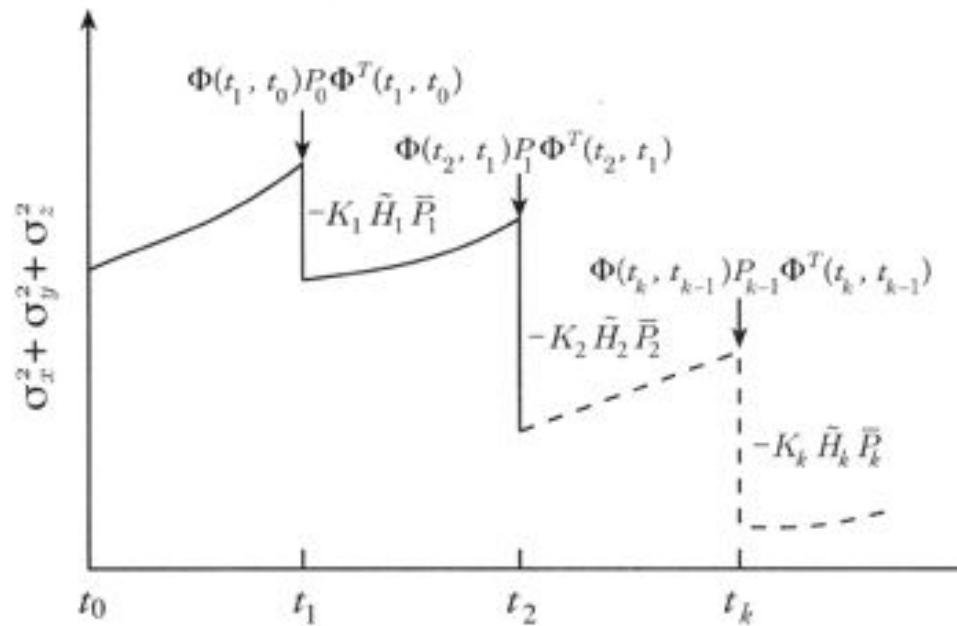
# Stat OD Conceptualization

## ► Conventional Kalman



# Evolution of Covariance Matrix

- ▶ Evolution of the covariance matrix as observations are processed.

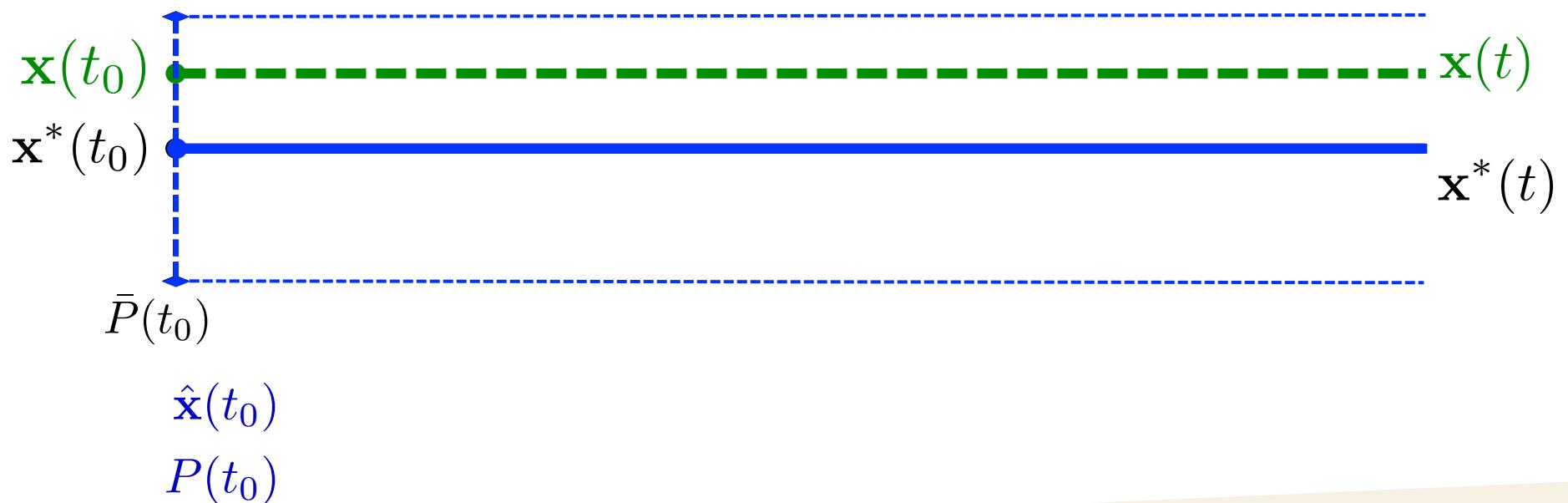


- ▶ Conceptualization of the Extended Kalman Filter (EKF)
- ▶ Major change: the reference trajectory is updated by the best estimate after every measurement.



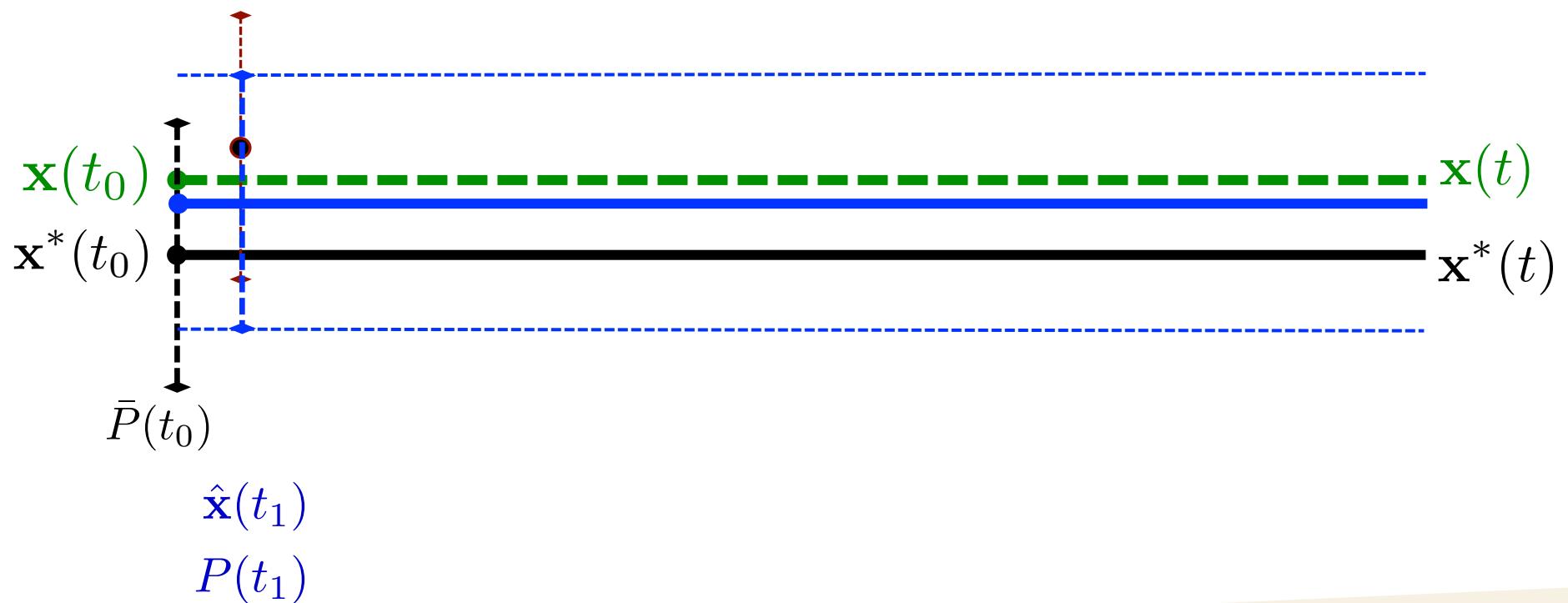
# Stat OD Conceptualization

## ► EKF



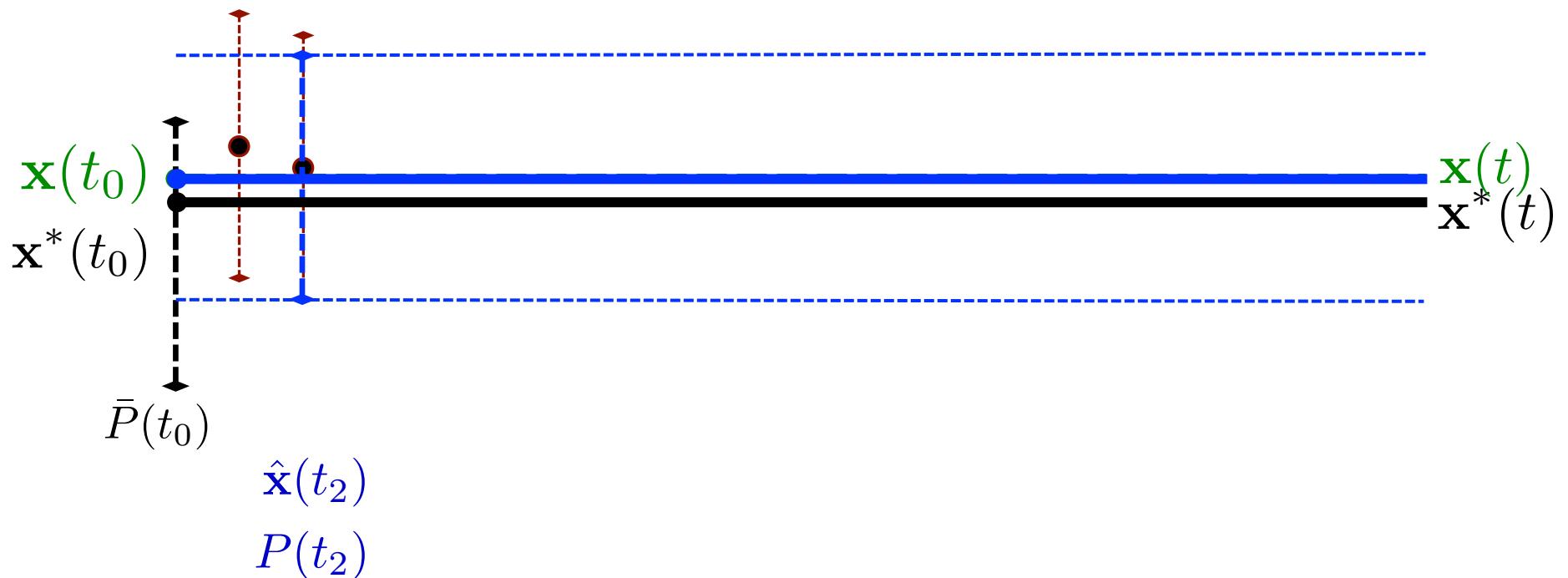
# Stat OD Conceptualization

## ► EKF



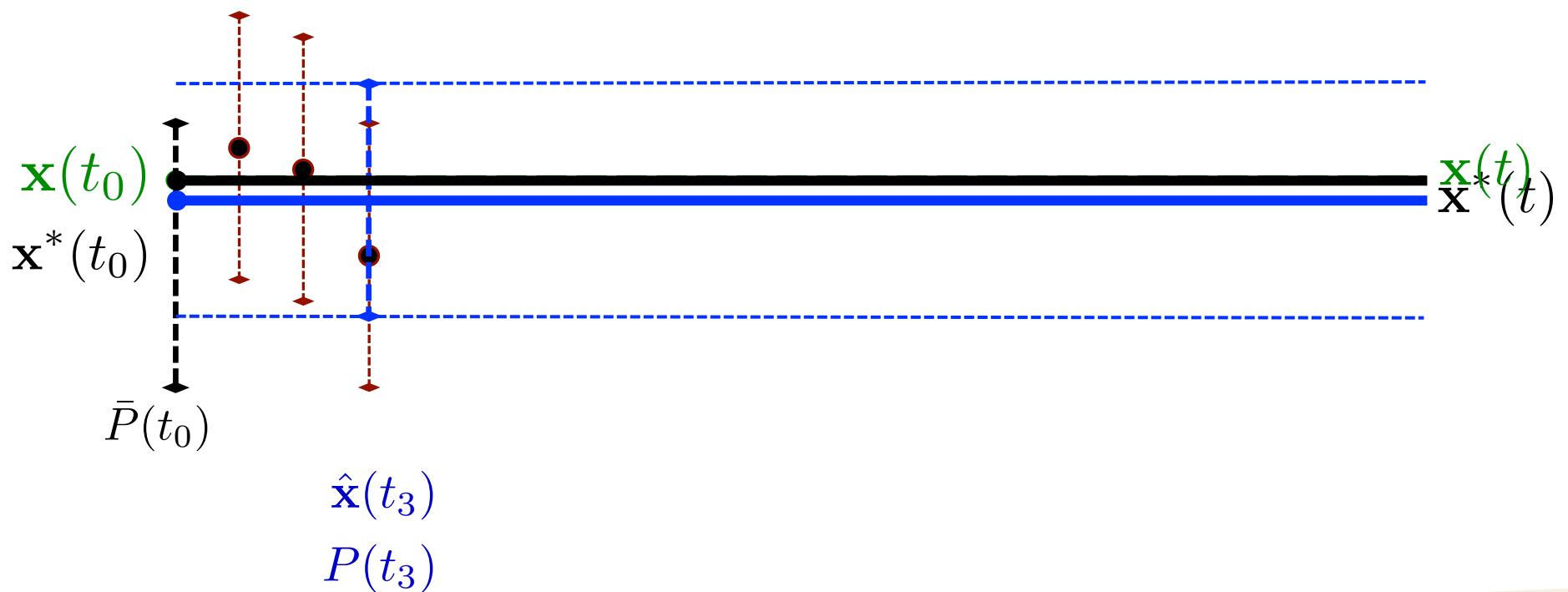
# Stat OD Conceptualization

## ► EKF



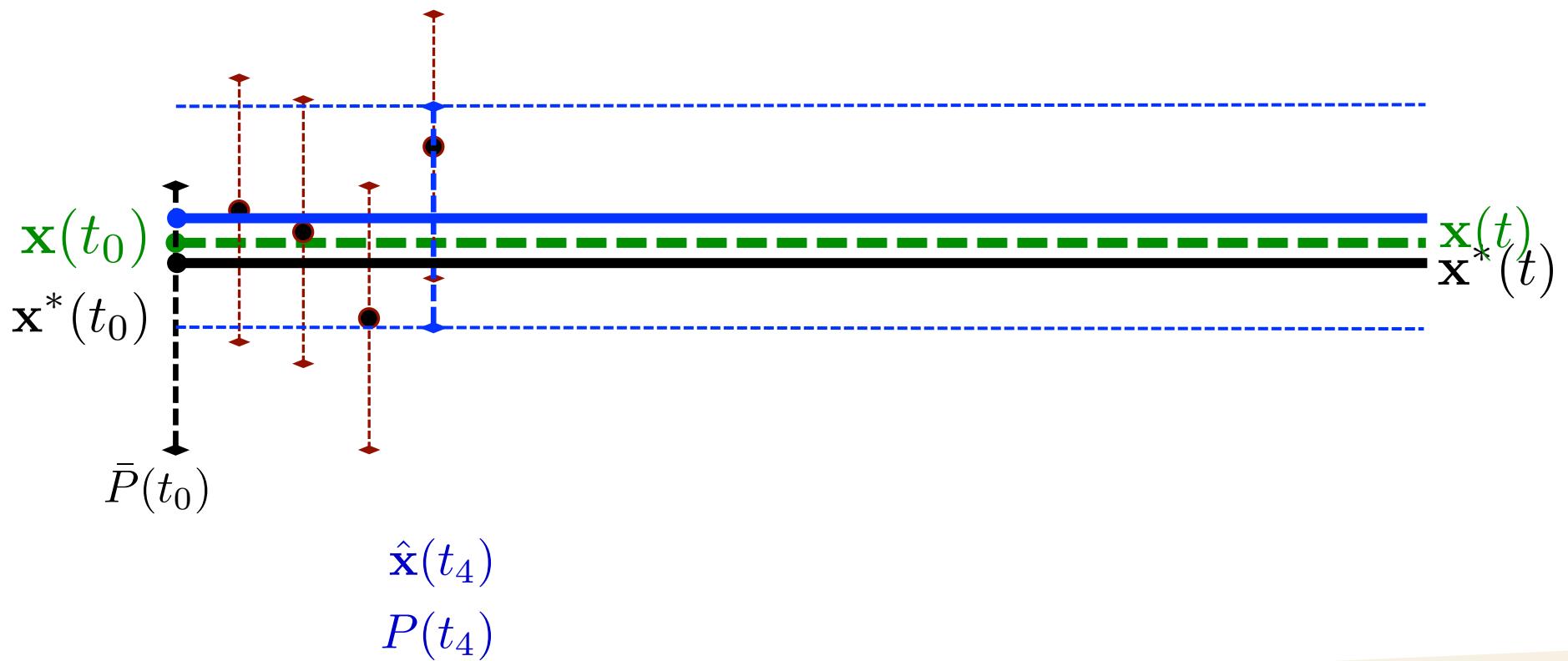
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## ► EKF



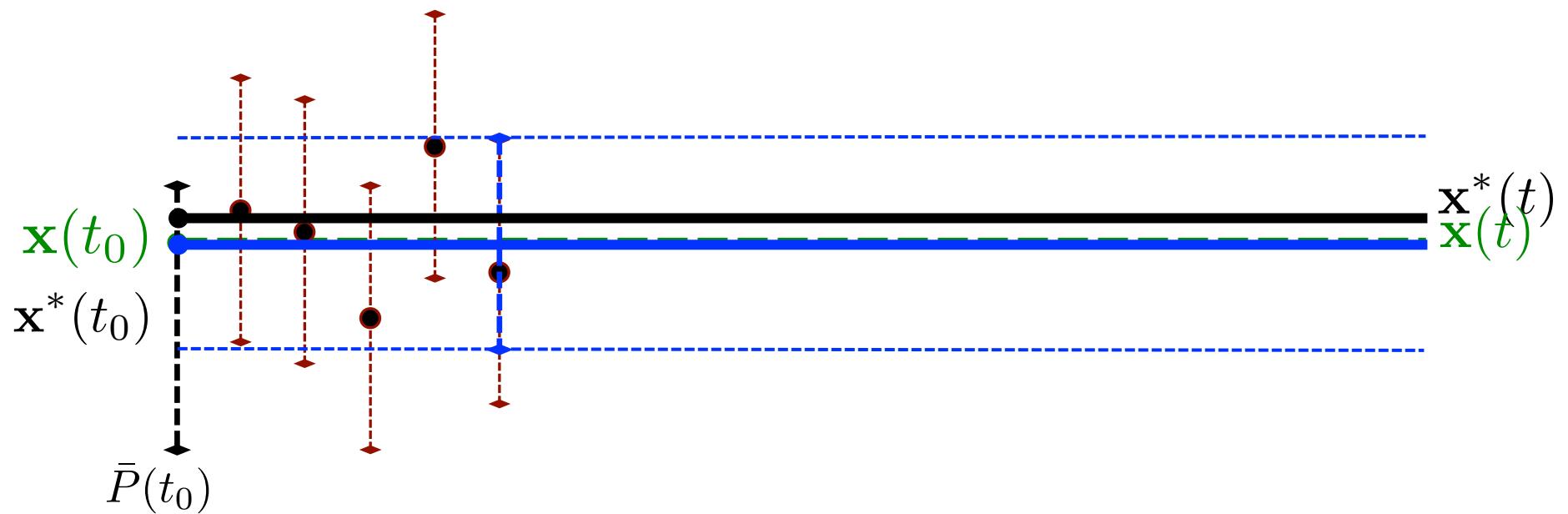
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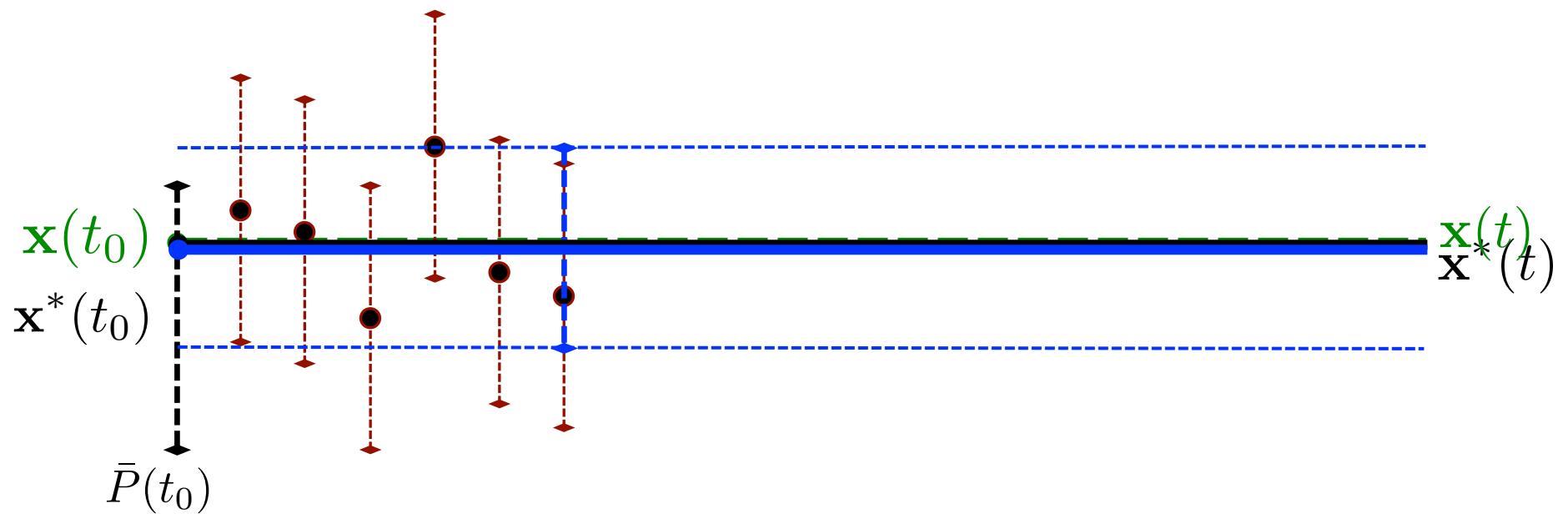
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## ► EKF



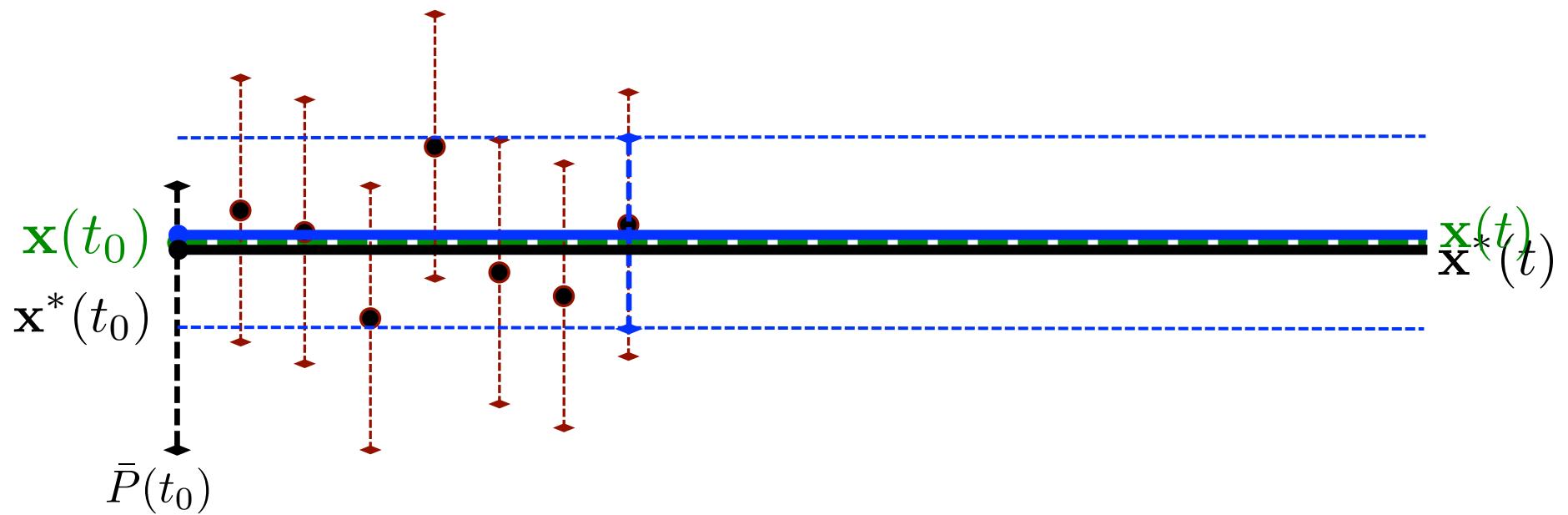
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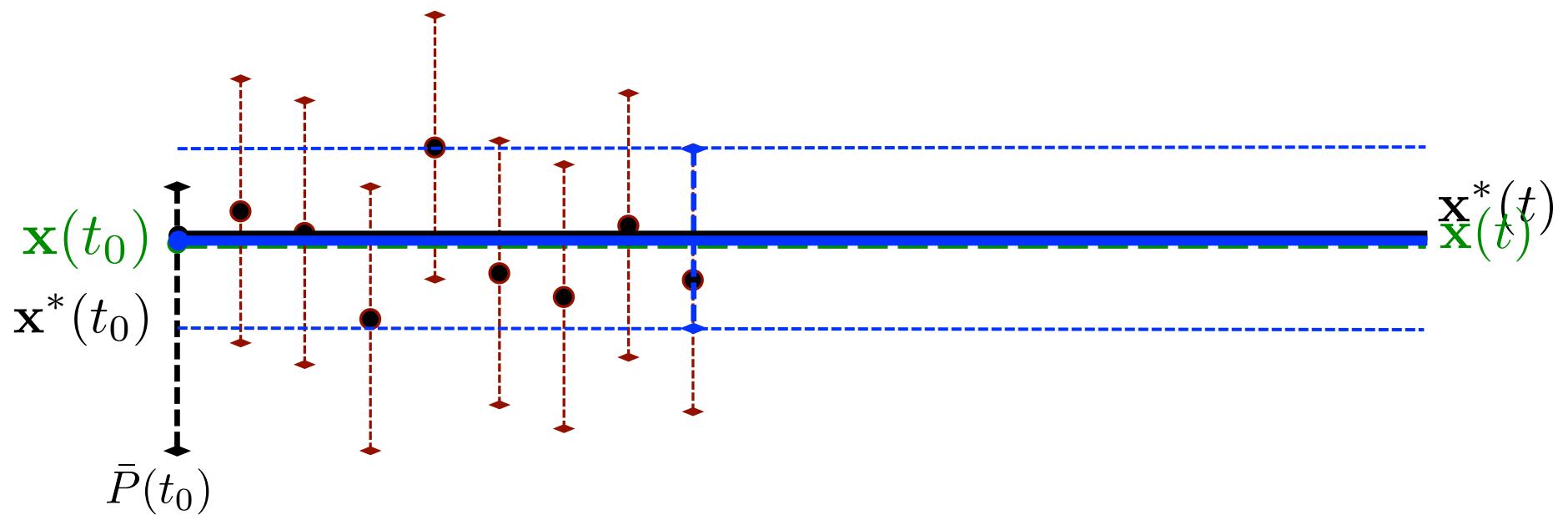
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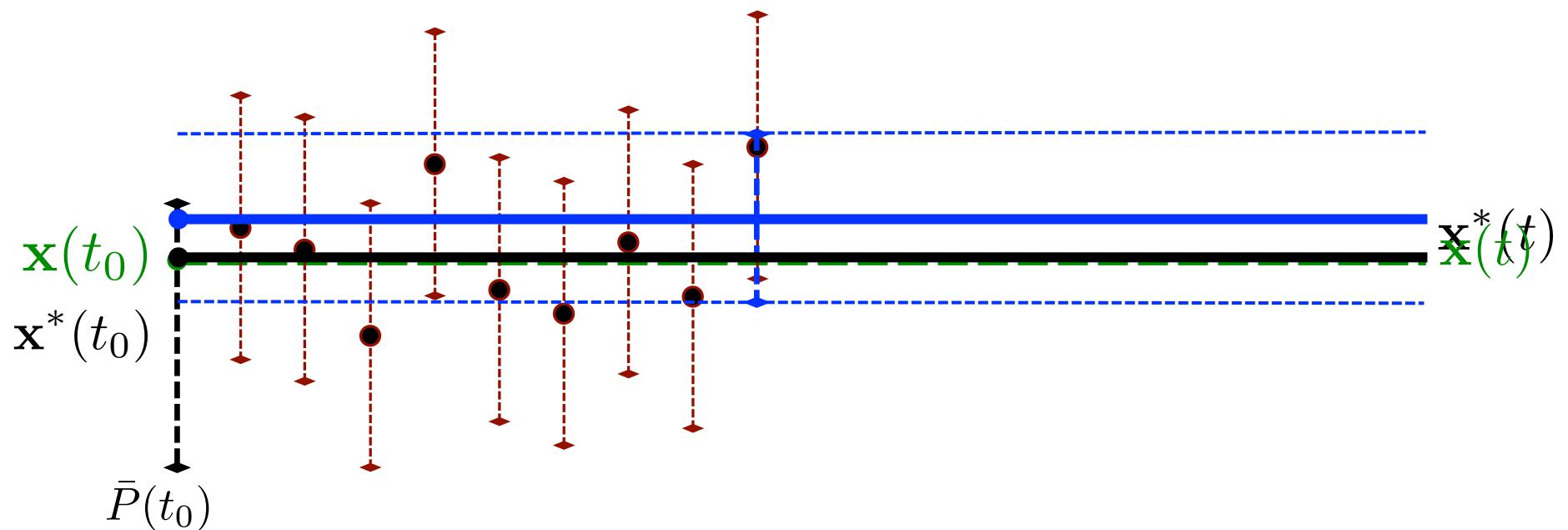
# Stat OD Conceptualization

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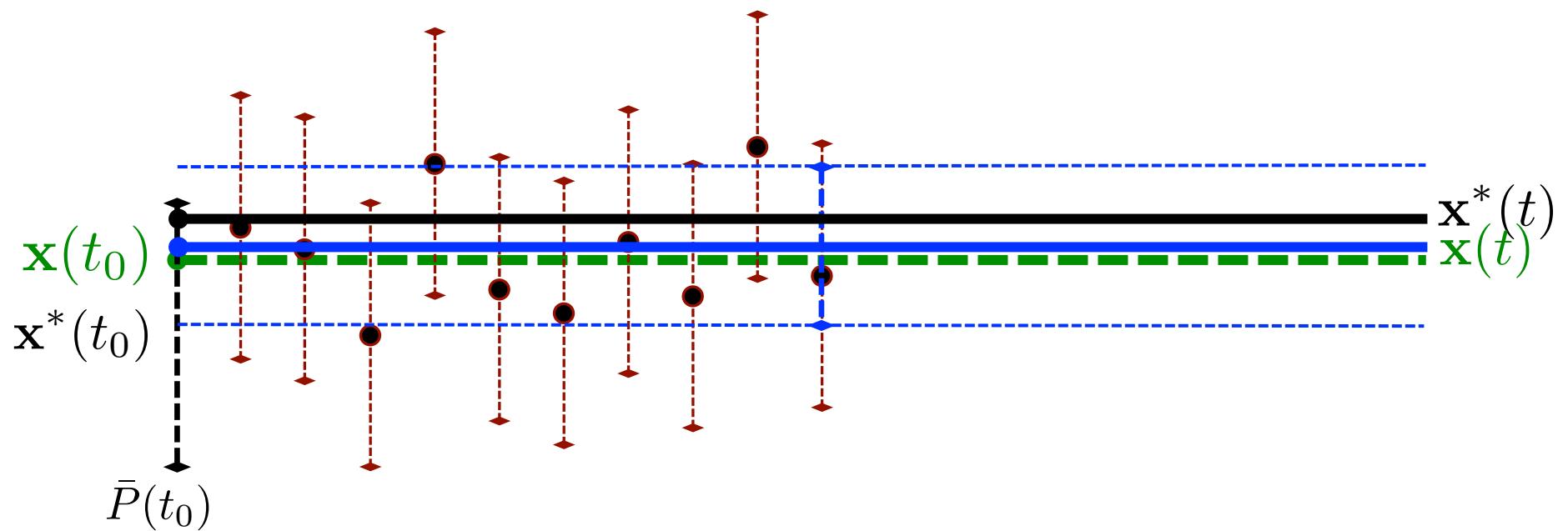
# Stat OD Conceptualization

## ► EKF



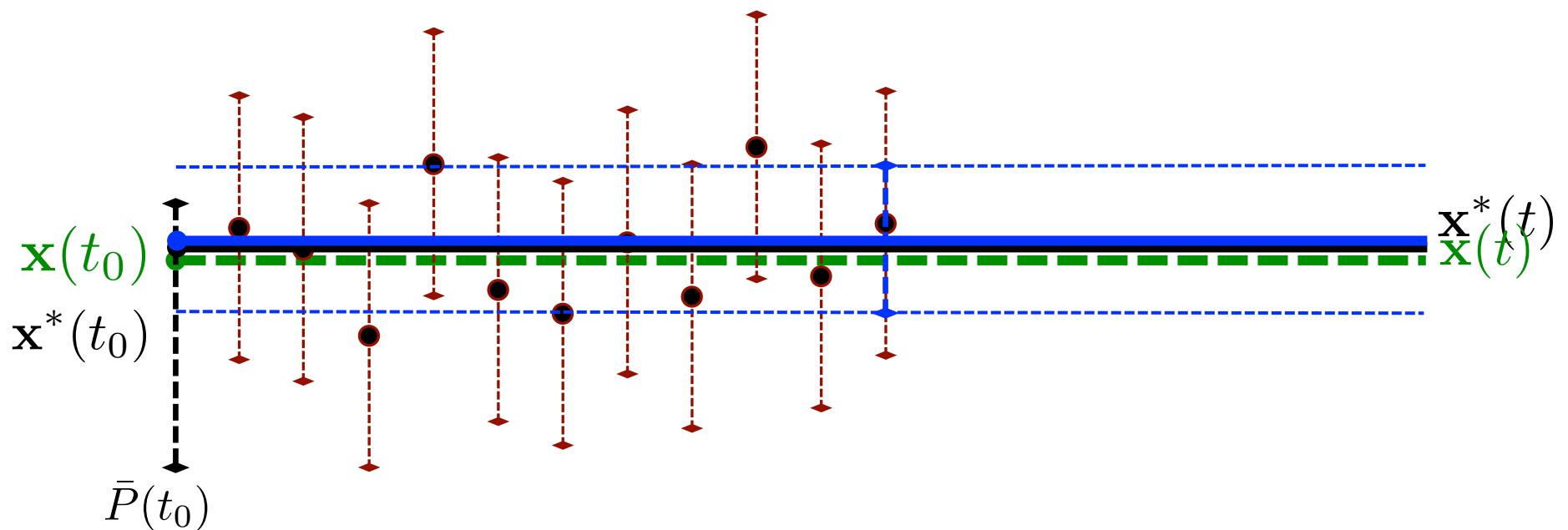
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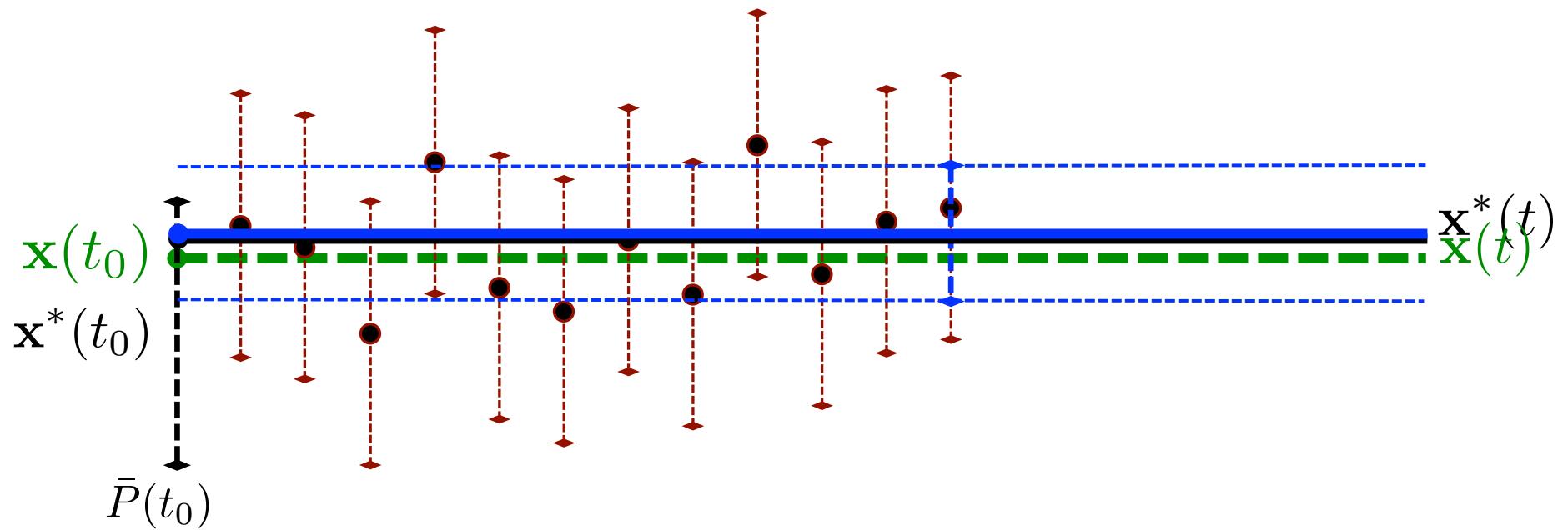
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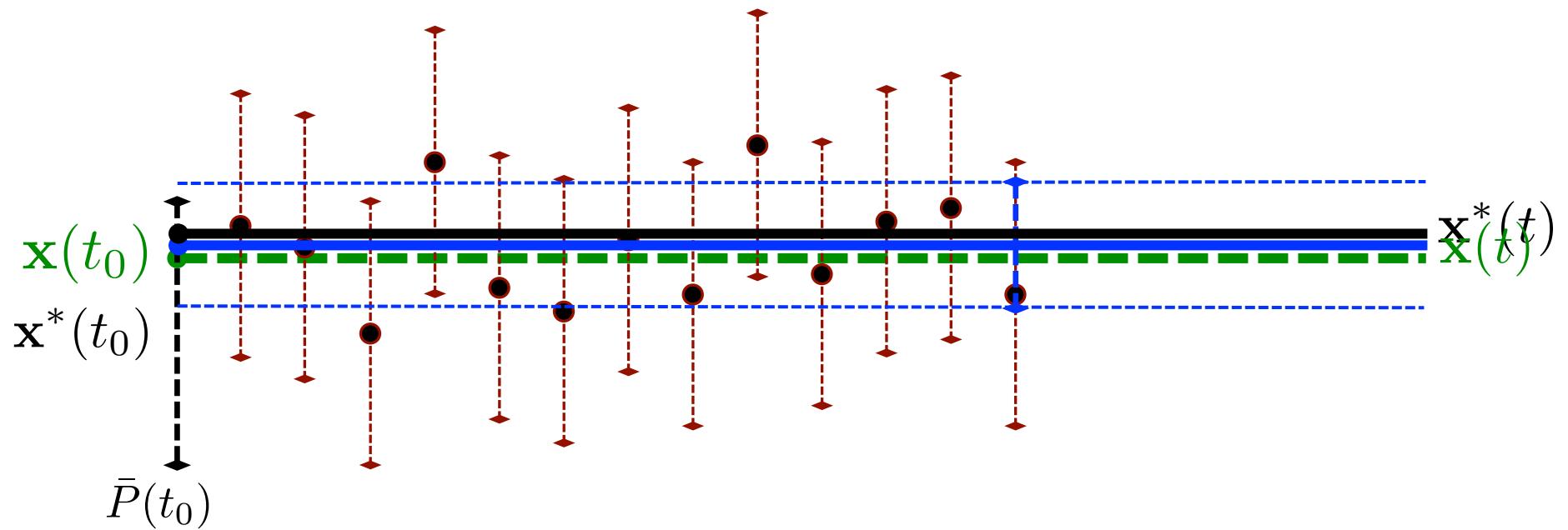
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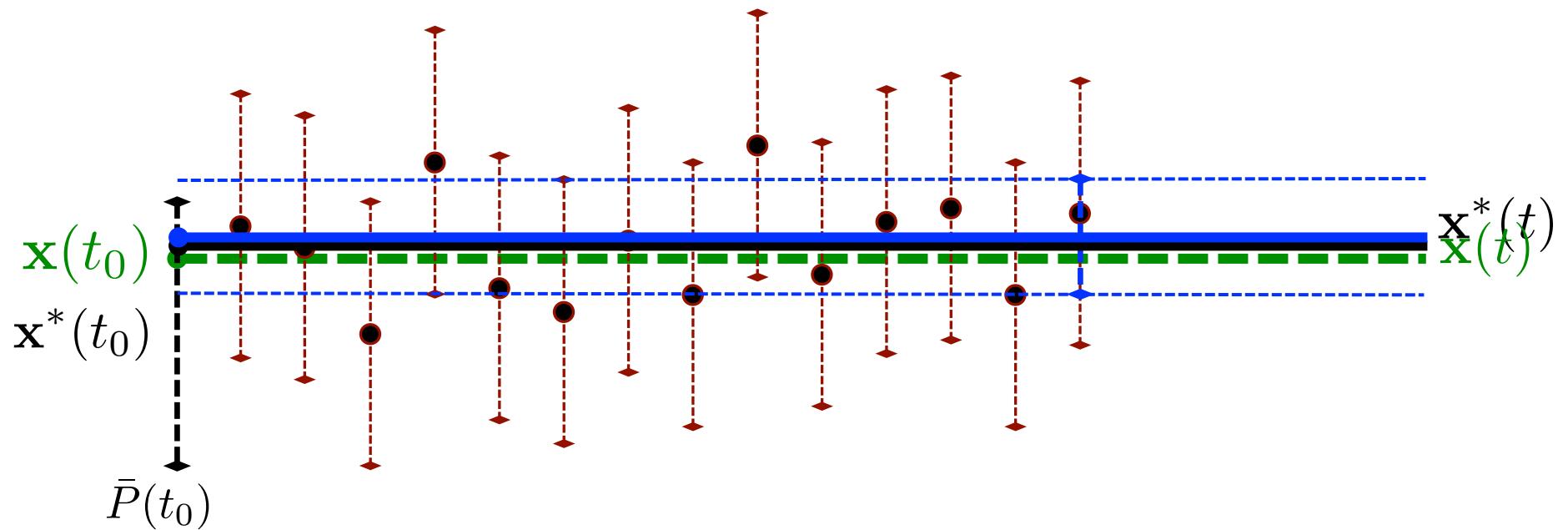
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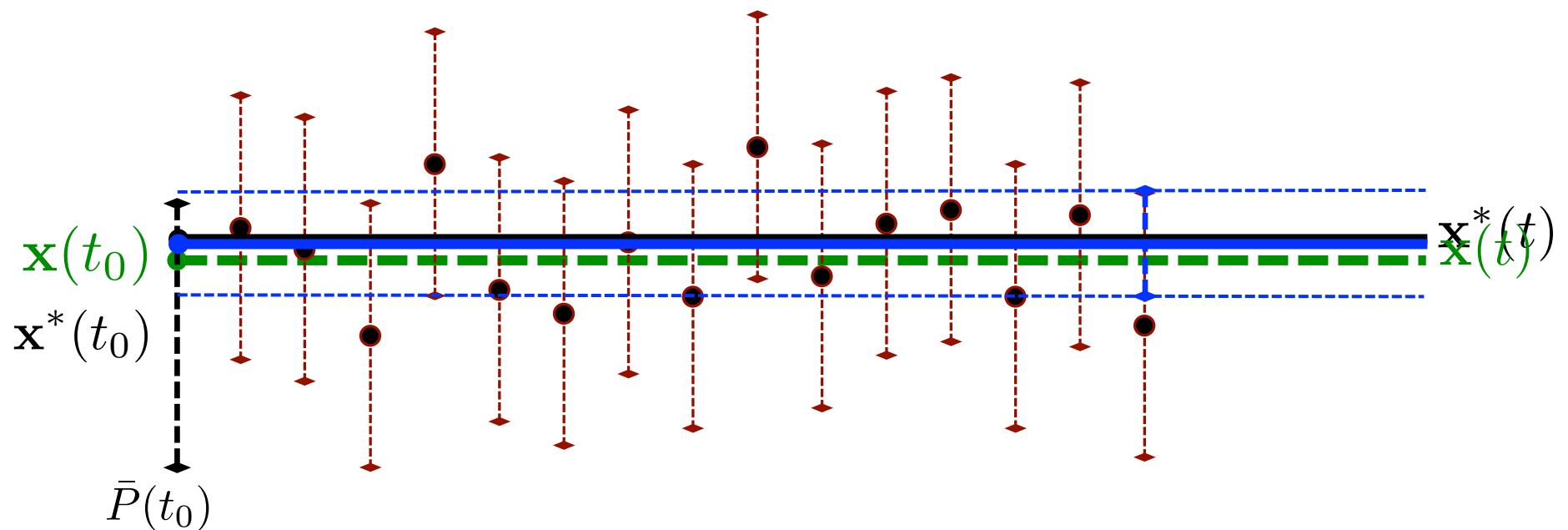
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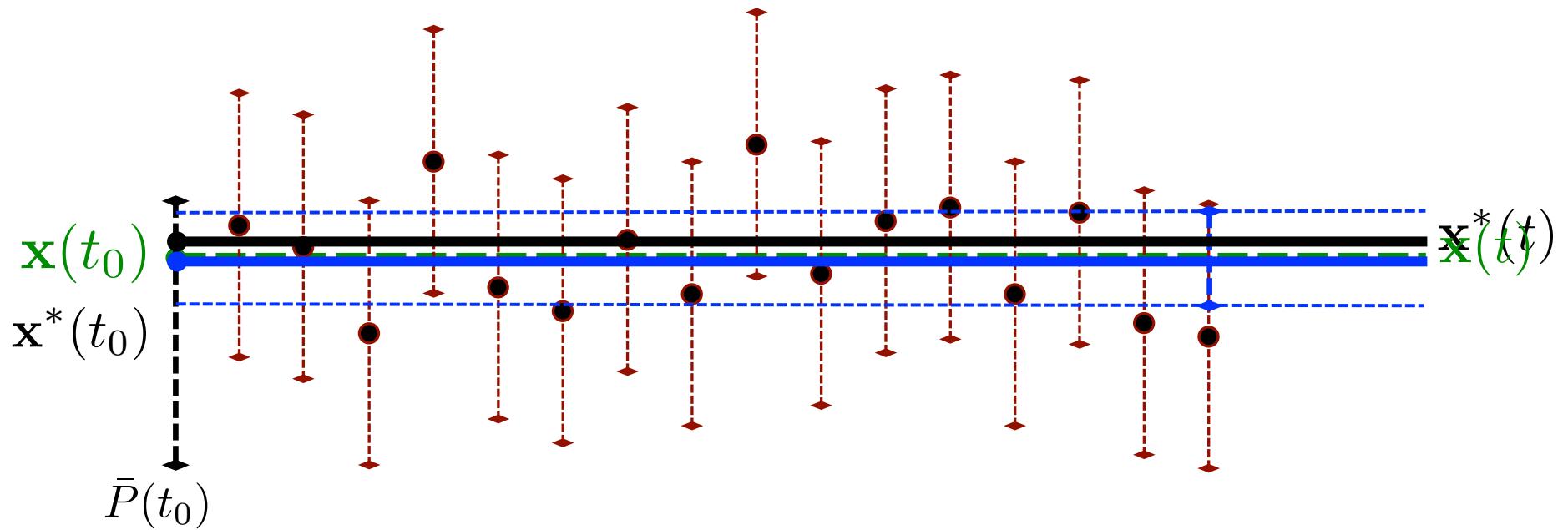
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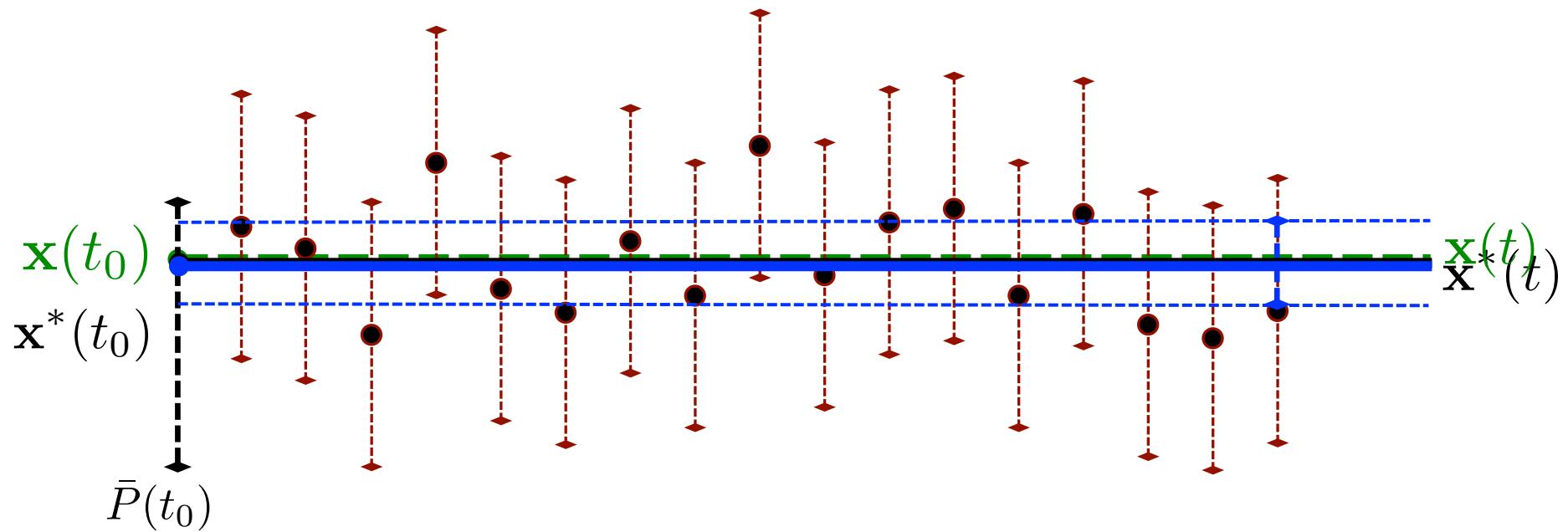
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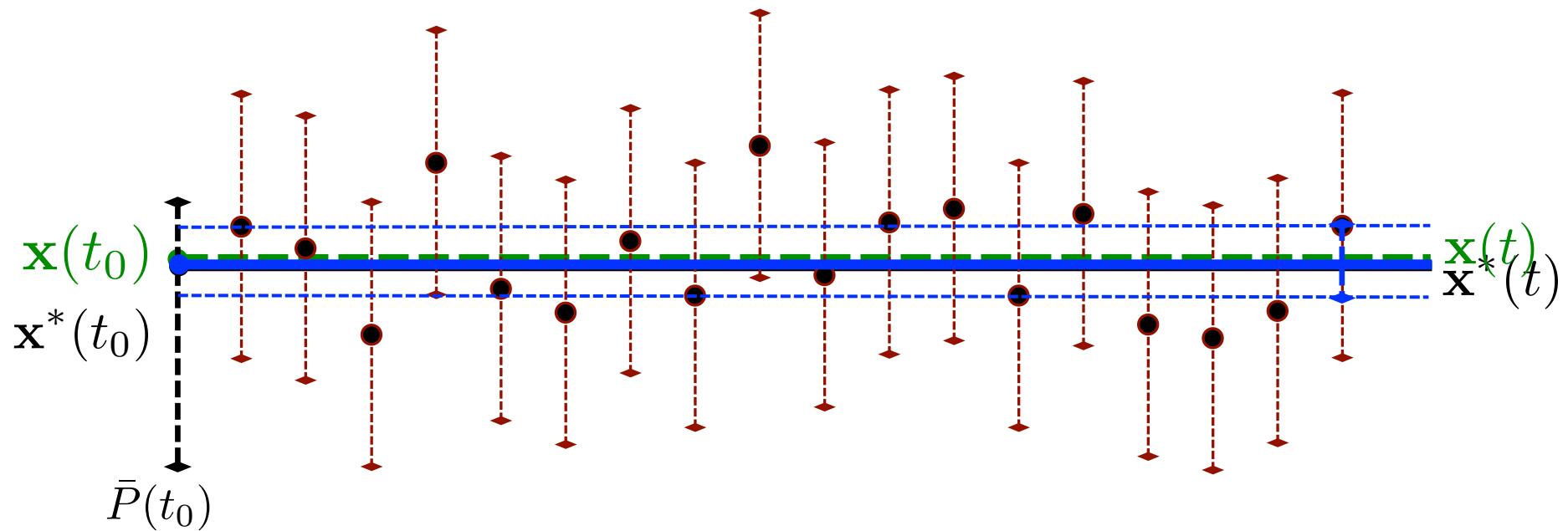
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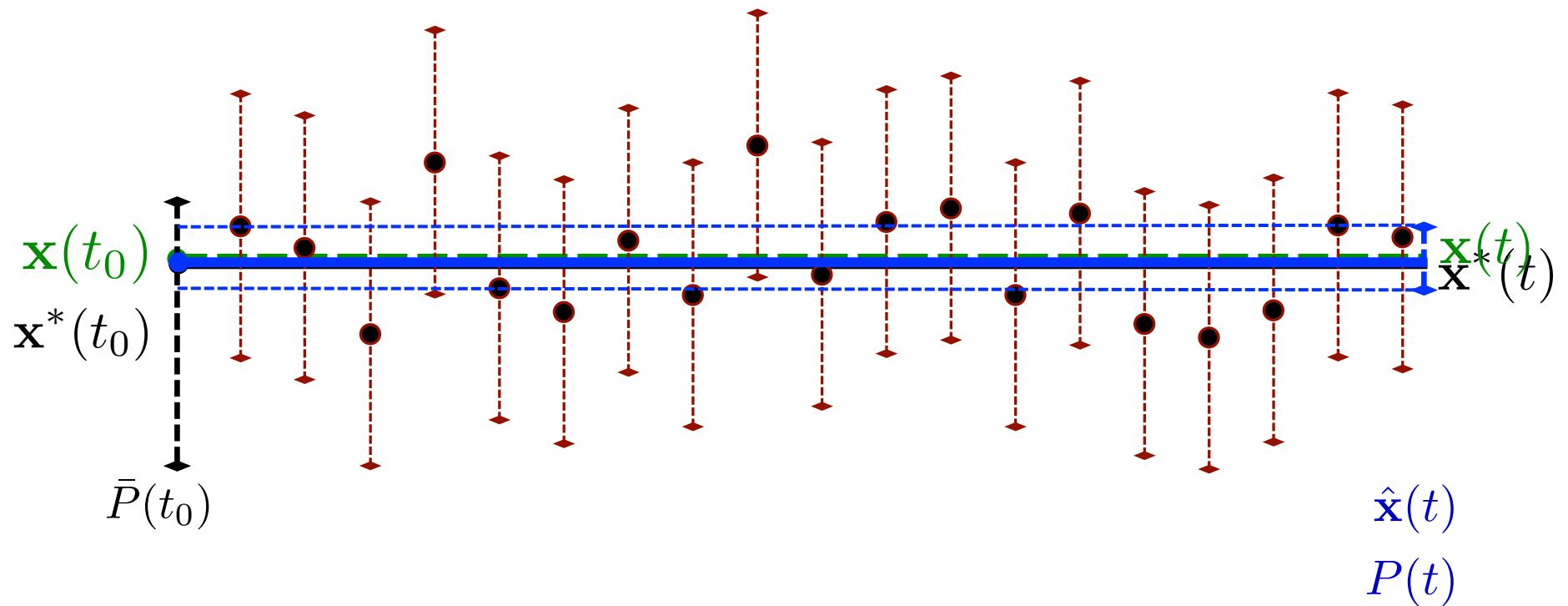
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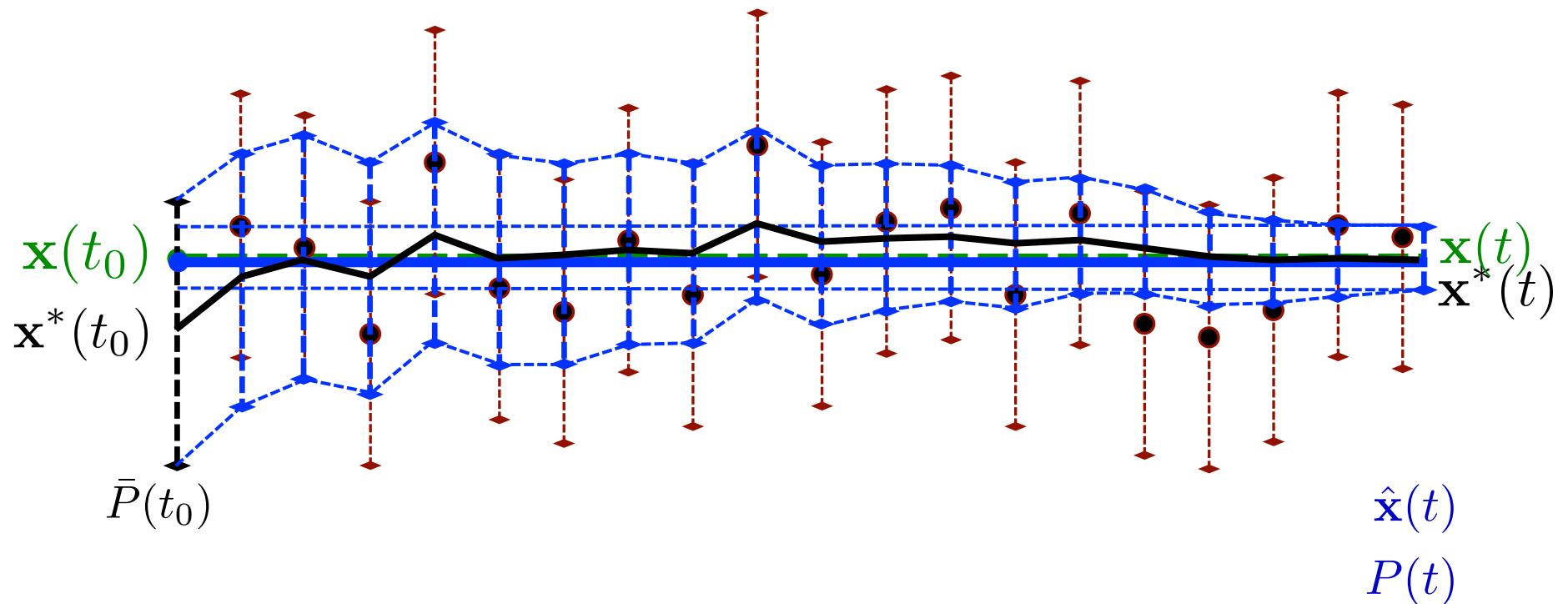
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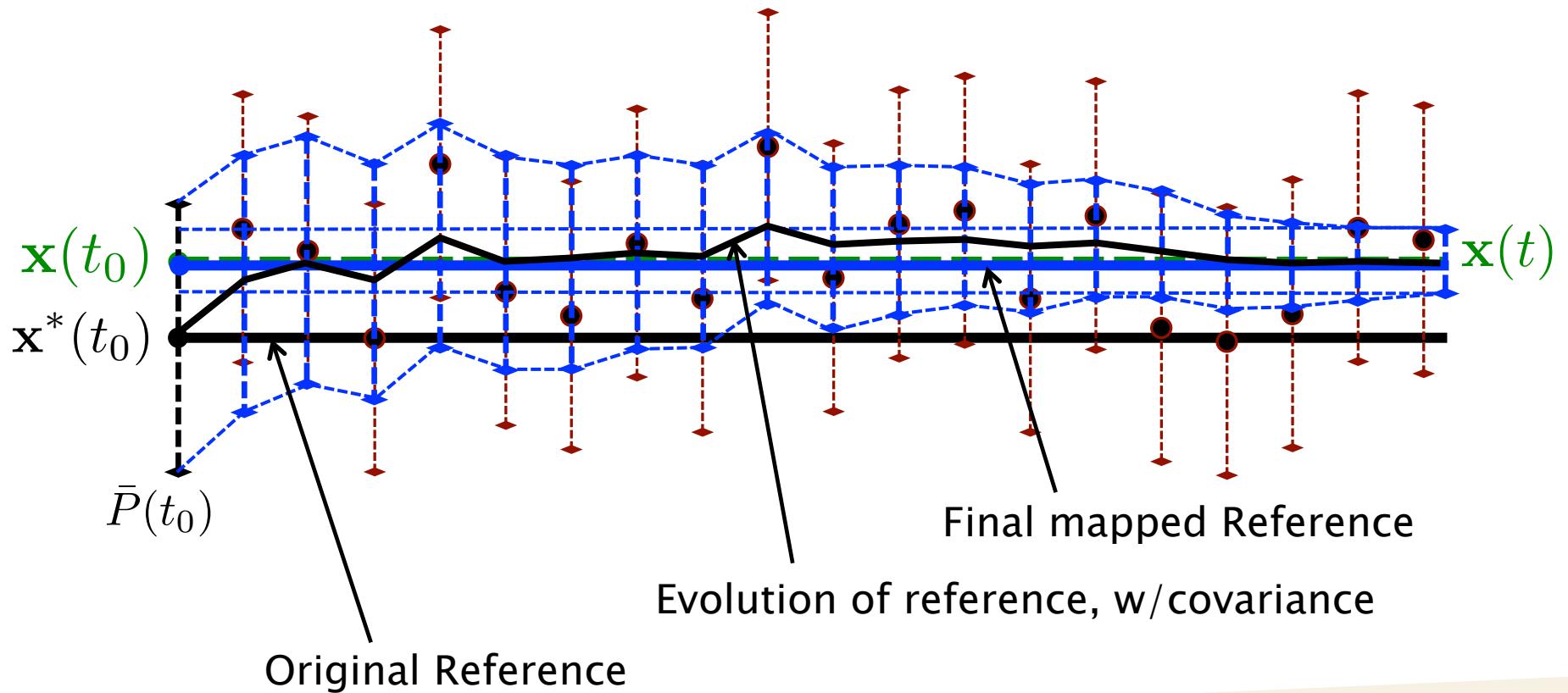
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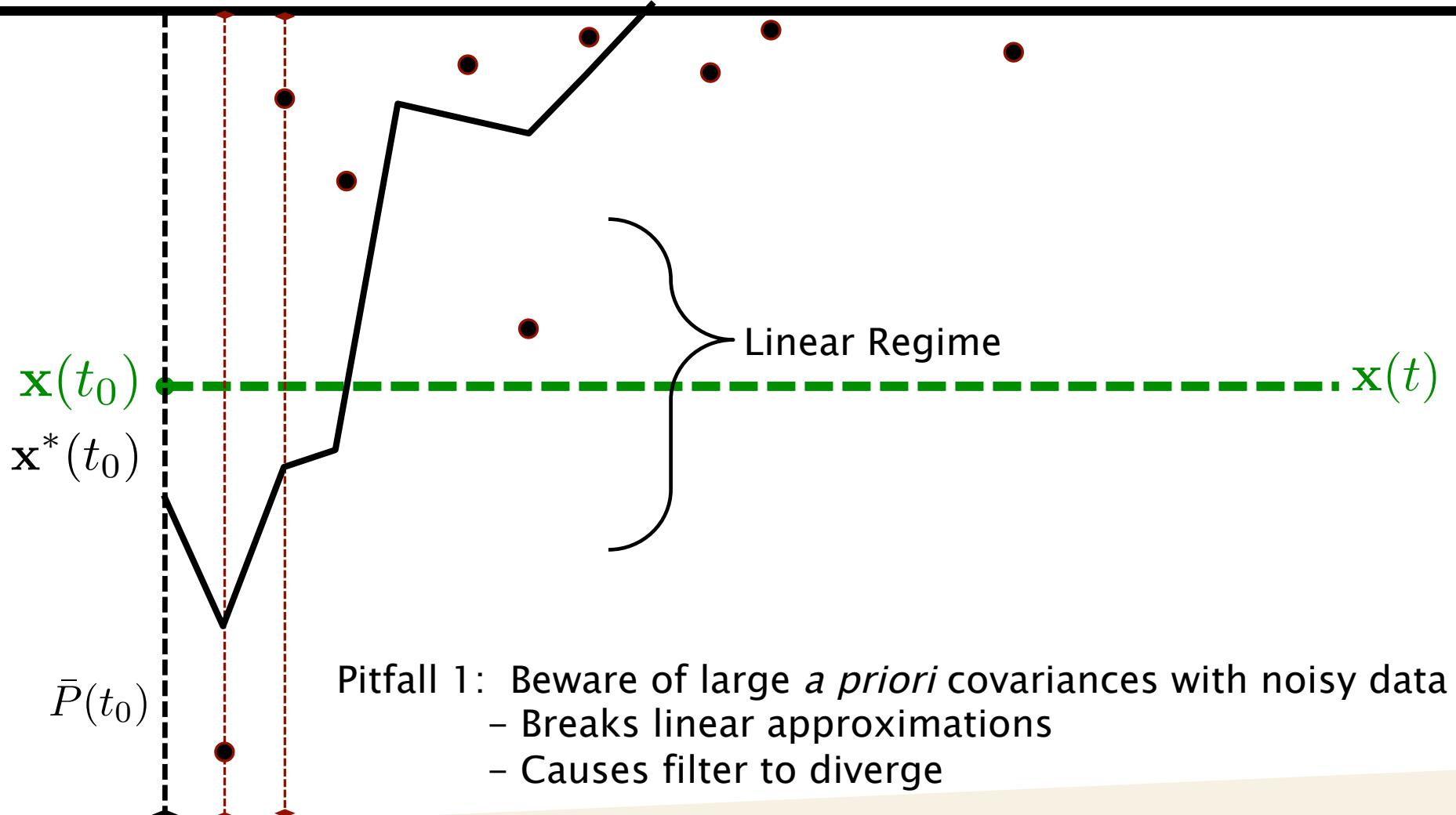


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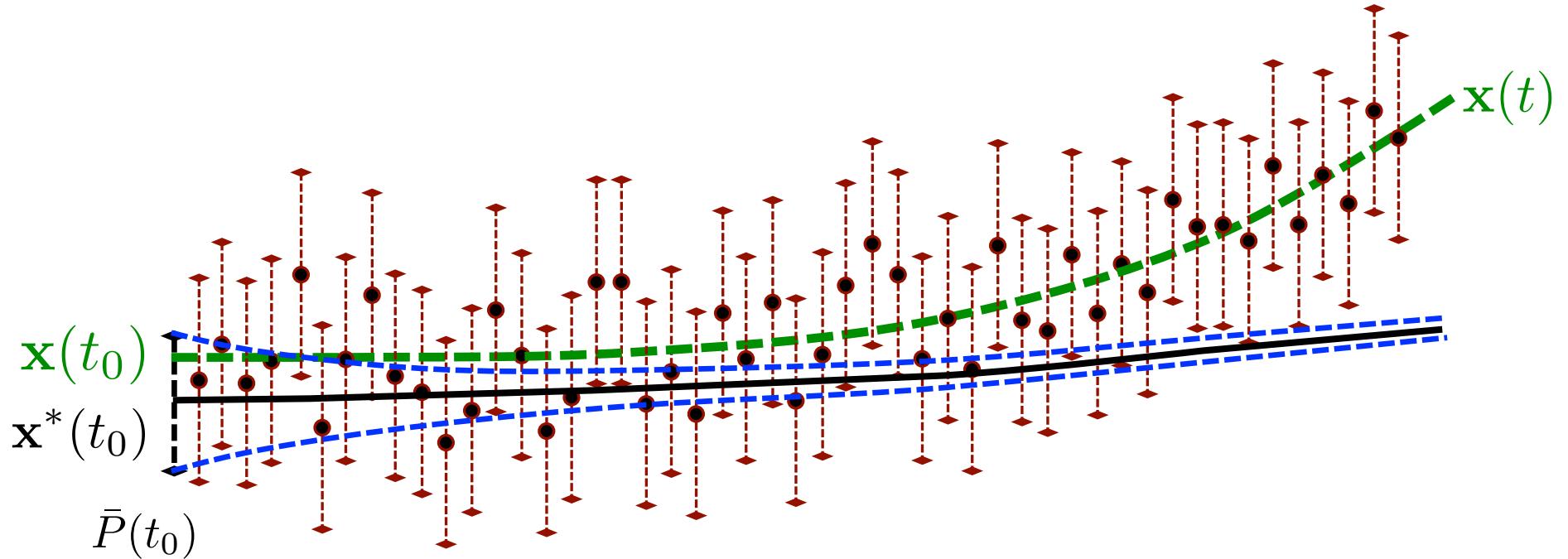
## ► EKF



# Stat OD Conceptualization



# Stat OD Conceptualization



Pitfall 2: Beware of collapsing covariance

- Prevents new data from influencing solution
- More prevalent for longer time-spans



# The End

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- ▶ Final Project
- ▶ Homework 5 is not graded...neither is the test.
- ▶ Homework 6 due Today
- ▶ Homework 7 due next week (Tuesday!)
  - It's okay to use Matlab to compute partials and to output them. But verify them.
- ▶ Concept Quizzes to resume Monday!
- ▶ Guest lecturer next week 10/25

