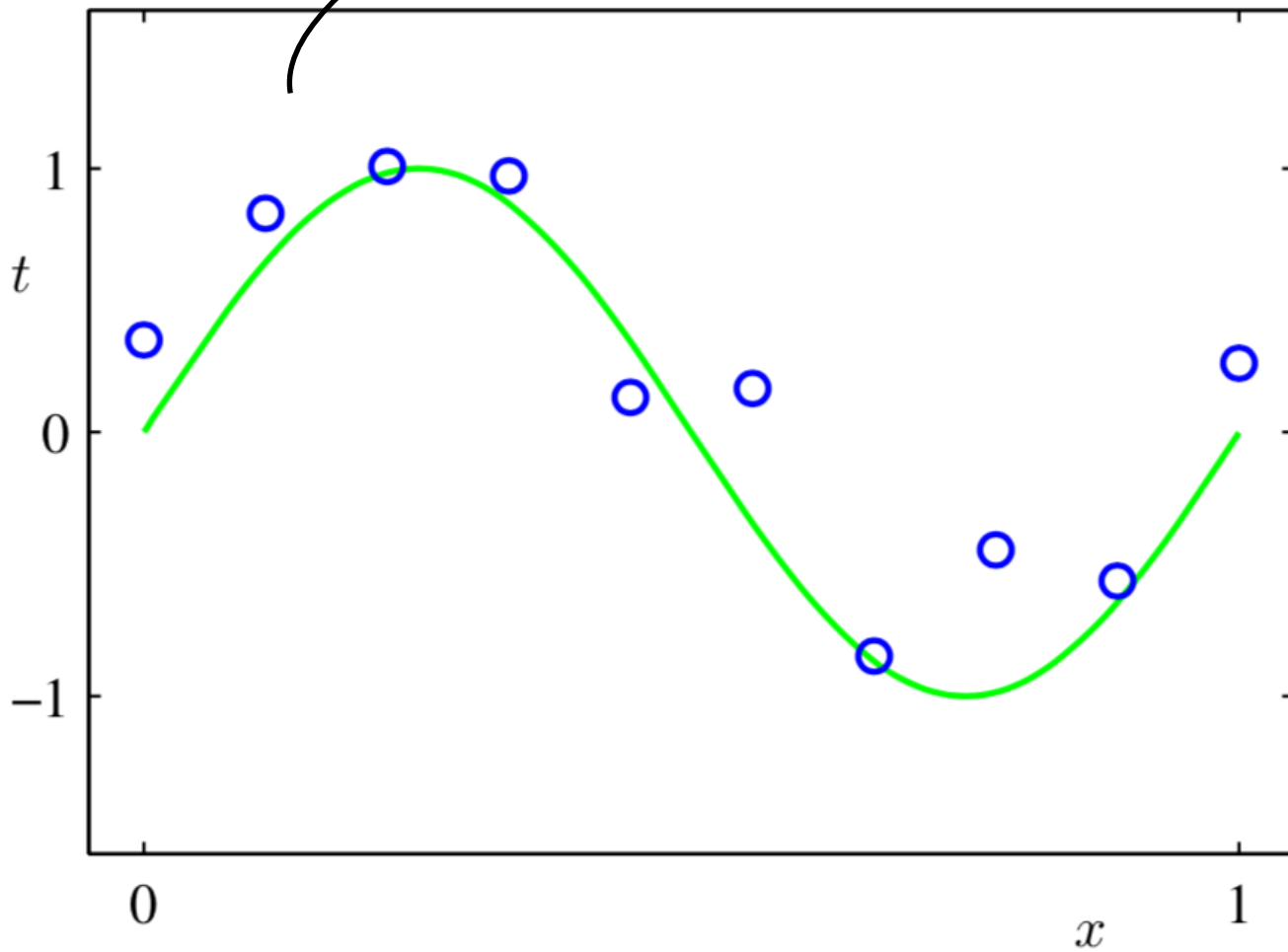


SAFEOPT & STAGEOPT

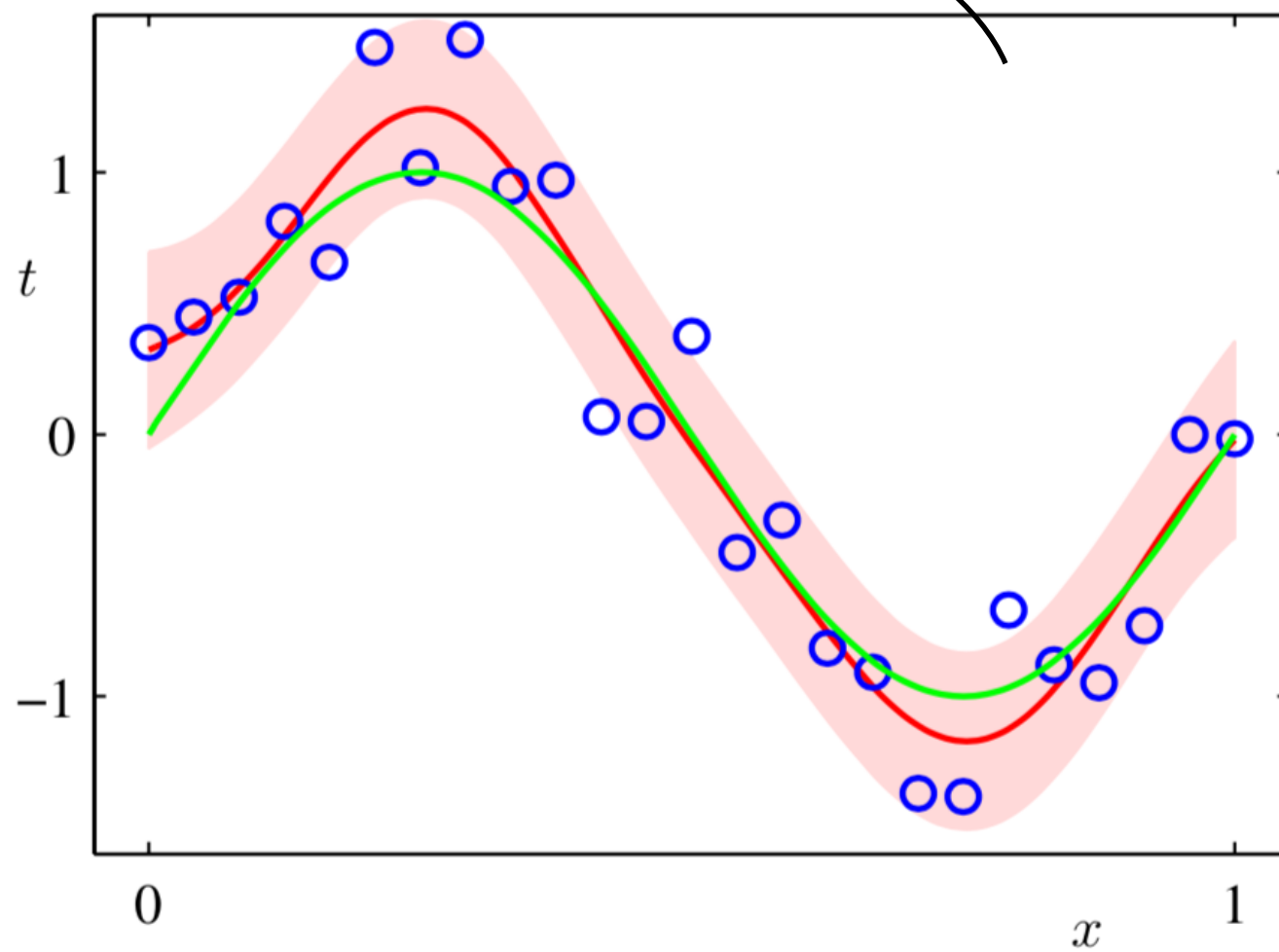
Daoran Jing
2019.07.31

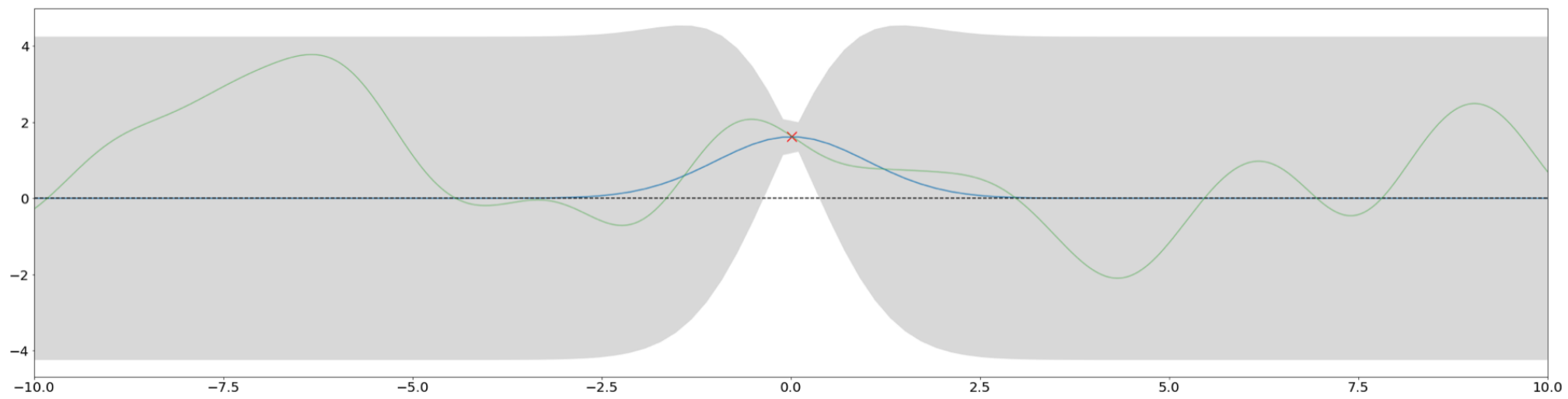
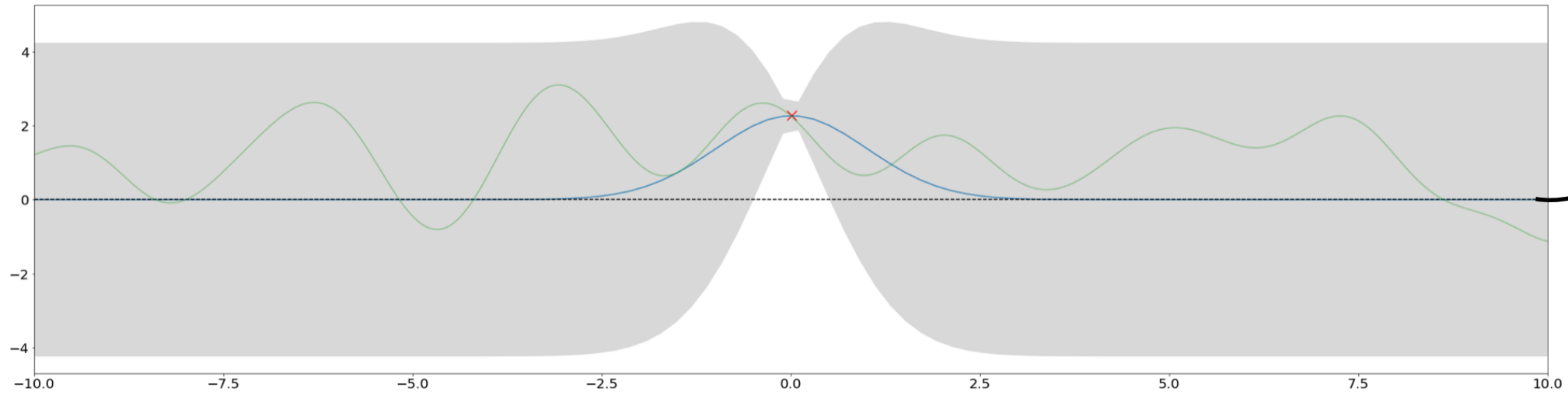
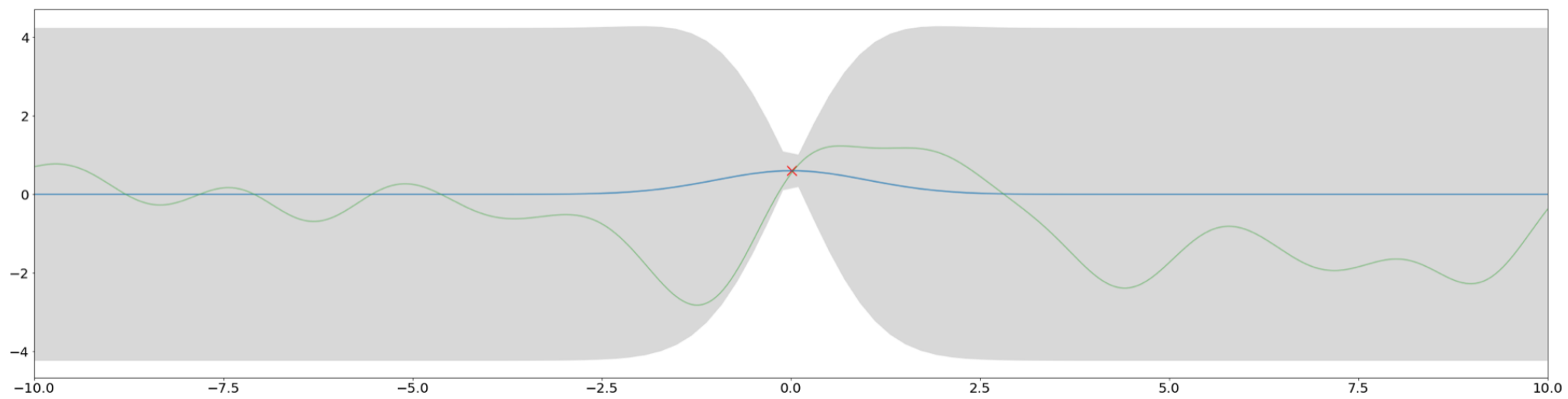
Fit theory & Bayesian methods

Only fit theory

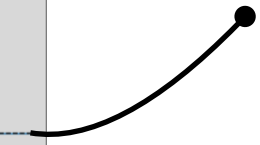


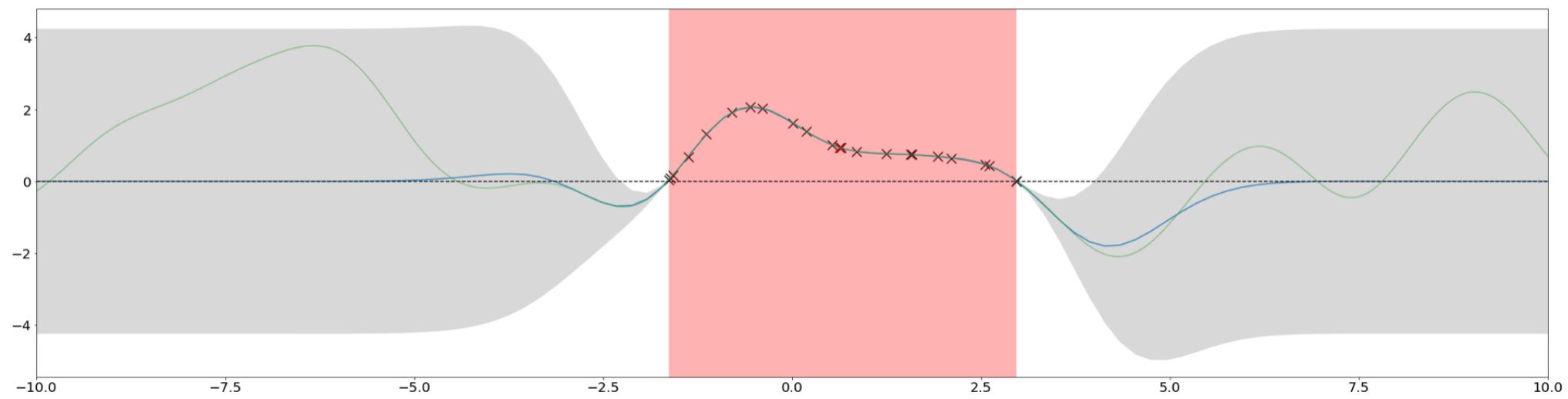
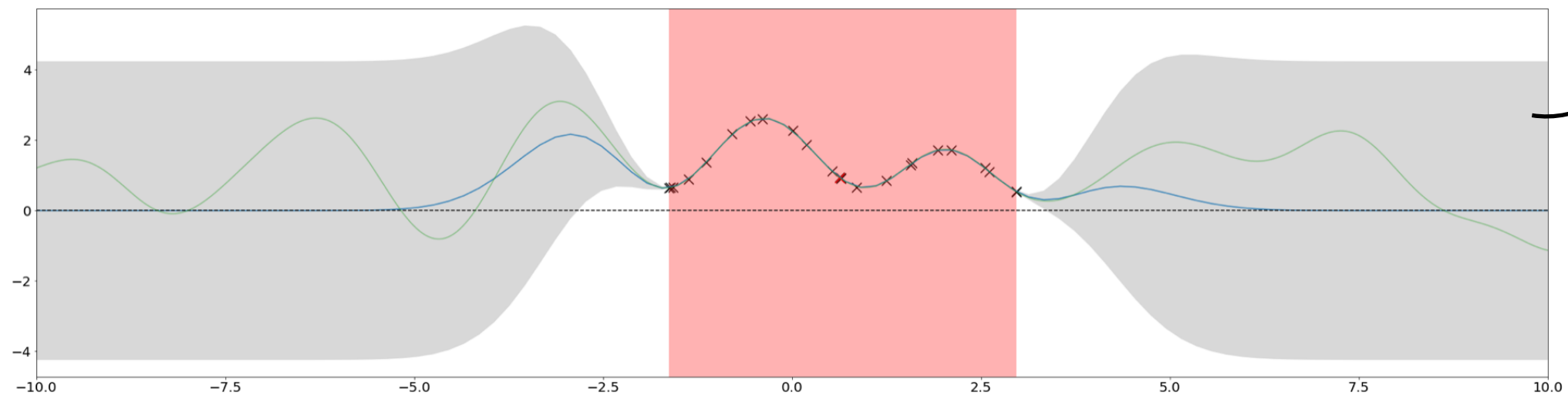
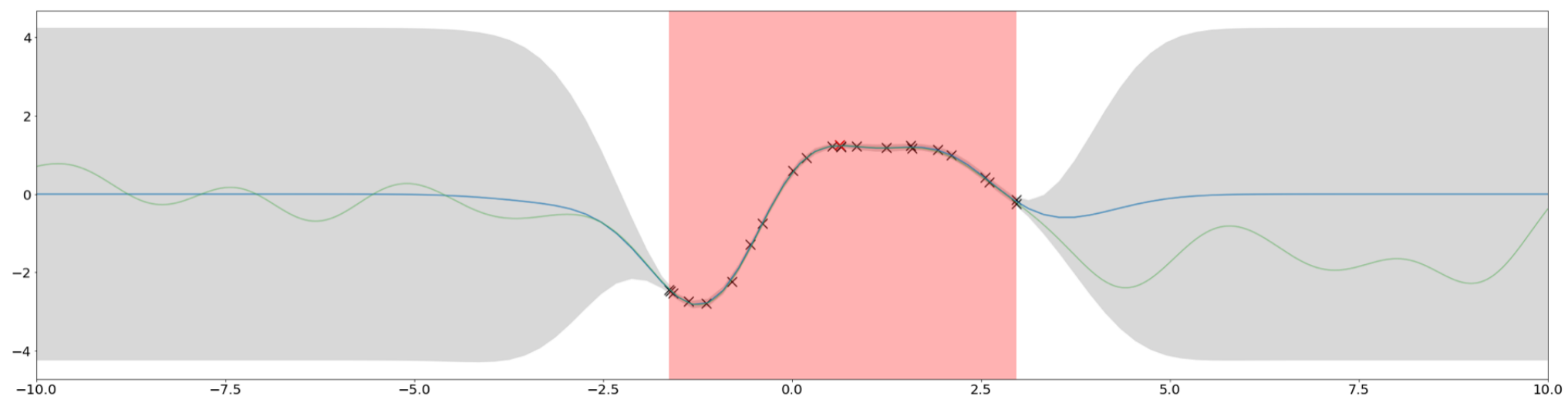
Using Bayesian method





Safety threshold





The result



• SafeOpt

- $Q_t(x) := [\mu_{t-1}(x) \pm \beta_t^{1/2} \sigma_{t-1}(x)]$

- $g_t(x) := |\{x' \in D/S_t \mid u_t(x) - Ld(x, x') \geq h\}|$

Algorithm 1 SAFE OPT

- 1: **Input:** sample set D ,
GP prior (μ_0, k, σ_0) ,
Lipschitz constant L ,
seed set S_0 ,
safety threshold h
 - 2: $C_0(\mathbf{x}) \leftarrow [h, \infty)$, for all $\mathbf{x} \in S_0$
 - 3: $C_0(\mathbf{x}) \leftarrow \mathbb{R}$, for all $\mathbf{x} \in D \setminus S_0$
 - 4: $Q_0(\mathbf{x}) \leftarrow \mathbb{R}$, for all $\mathbf{x} \in D$
 - 5: **for** $t = 1, \dots$ **do**
 - 6: $C_t(\mathbf{x}) \leftarrow C_{t-1}(\mathbf{x}) \cap Q_{t-1}(\mathbf{x})$
 - 7: $S_t \leftarrow \bigcup_{\mathbf{x} \in S_{t-1}} \{\mathbf{x}' \in D \mid \ell_t(\mathbf{x}) - Ld(\mathbf{x}, \mathbf{x}') \geq h\}$
 - 8: $G_t \leftarrow \{\mathbf{x} \in S_t \mid g_t(\mathbf{x}) > 0\}$
 - 9: $M_t \leftarrow \{\mathbf{x} \in S_t \mid u_t(\mathbf{x}) \geq \max_{\mathbf{x}' \in S_t} \ell_t(\mathbf{x}')\}$
 - 10: $\mathbf{x}_t \leftarrow \operatorname{argmax}_{\mathbf{x} \in G_t \cup M_t} (w_t(\mathbf{x}))$
 - 11: $y_t \leftarrow f(\mathbf{x}_t) + n_t$
 - 12: Compute $Q_t(\mathbf{x})$, for all $\mathbf{x} \in S_t$
 - 13: **end for**
-

• StageOpt

- Safe region expansion

- Optimization

Algorithm 1 STAGEOPT

```

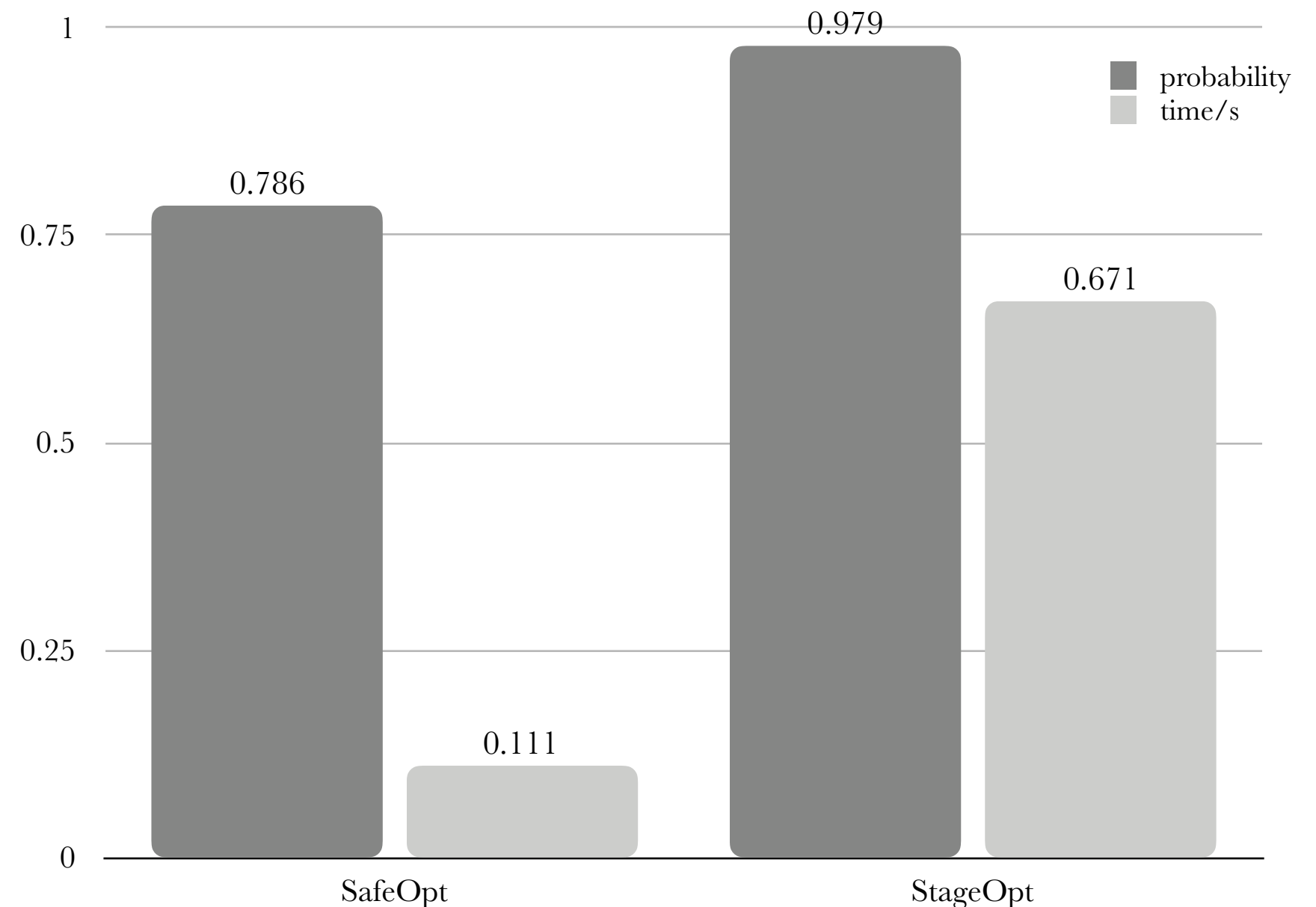
1: Input: sample set  $D, i \in \{1, \dots, n\}$ ,
           GP prior for utility function  $f$ ,
           GP priors for safety functions  $g_i$ ,
           Lipschitz constants  $L_i$  for  $g_i$ ,
           safe seed set  $S_0$ ,
           safety threshold  $h_i$ ,
           accuracies  $\epsilon$  (for expansion),  $\zeta$  (for optimization).
2:  $C_0^i(\mathbf{x}) \leftarrow [h_i, \infty)$ , for all  $\mathbf{x} \in S_0$ 
3:  $C_0^i(\mathbf{x}) \leftarrow \mathbb{R}$ , for all  $\mathbf{x} \in D \setminus S_0$ 
4:  $Q_0^i(\mathbf{x}) \leftarrow \mathbb{R}$ , for all  $\mathbf{x} \in D$ 
5:  $C_0^f(\mathbf{x}) \leftarrow \mathbb{R}$ , for all  $\mathbf{x} \in D$ 
6:  $Q_0^f(\mathbf{x}) \leftarrow \mathbb{R}$ , for all  $\mathbf{x} \in D$ 
7: for  $t = 1, \dots, T_0$  do
8:    $C_t^i(\mathbf{x}) \leftarrow C_{t-1}^i(\mathbf{x}) \cap Q_{t-1}^i(\mathbf{x})$ 
9:    $C_t^f(\mathbf{x}) \leftarrow C_{t-1}^f(\mathbf{x}) \cap Q_{t-1}^f(\mathbf{x})$ 
10:   $S_t \leftarrow \bigcap_i \bigcup_{\mathbf{x} \in S_{t-1}} \{\mathbf{x}' \in D \mid \ell_t^i(\mathbf{x}) - L_i d(\mathbf{x}, \mathbf{x}') \geq h_i\}$ 
11:   $G_t \leftarrow \{\mathbf{x} \in S_t \mid e_t(\mathbf{x}) > 0\}$ 
12:  if  $\forall i, \epsilon_t^i < \epsilon$  then
13:     $\mathbf{x}_t \leftarrow \operatorname{argmax}_{\mathbf{x} \in G_t, i \in \{1, \dots, n\}} w_t^i(\mathbf{x})$ 
14:  else
15:     $\mathbf{x}_t \leftarrow \operatorname{argmax}_{\mathbf{x} \in S_t} \mu_{t-1}^f(\mathbf{x}) + \beta_t \sigma_{t-1}^f(\mathbf{x})$ 
16:  end if
17:   $y_{f,t} \leftarrow f(\mathbf{x}_t) + n_{f,t}$ 
18:   $y_{i,t} \leftarrow g_i(\mathbf{x}_t) + n_{i,t}$ 
19:  Compute  $Q_{f,t}(\mathbf{x})$  and  $Q_{i,t}(\mathbf{x})$ , for all  $\mathbf{x} \in S_t$ 
20: end for
21: for  $t = T_0 + 1, \dots, T$  do
22:   $C_t^f(\mathbf{x}) \leftarrow C_{t-1}^f(\mathbf{x}) \cap Q_{t-1}^f(\mathbf{x})$ 
23:   $\mathbf{x}_t \leftarrow \operatorname{argmax}_{\mathbf{x} \in S_t} \mu_{t-1}^f(\mathbf{x}) + \beta_t \sigma_{t-1}^f(\mathbf{x})$ 
24:   $y_{f,t} \leftarrow f(\mathbf{x}_t) + n_{f,t}$ 
25:   $y_{i,t} \leftarrow g_i(\mathbf{x}_t) + n_{i,t}$ 
26:  Compute  $Q_{f,t}(\mathbf{x})$  and  $Q_{i,t}(\mathbf{x})$ , for all  $\mathbf{x} \in S_t$ 
27: end for

```

- **Probability of guaranteeing safety**

- Two safety function [f1, f2], with noise[1e-5, 1e-3]

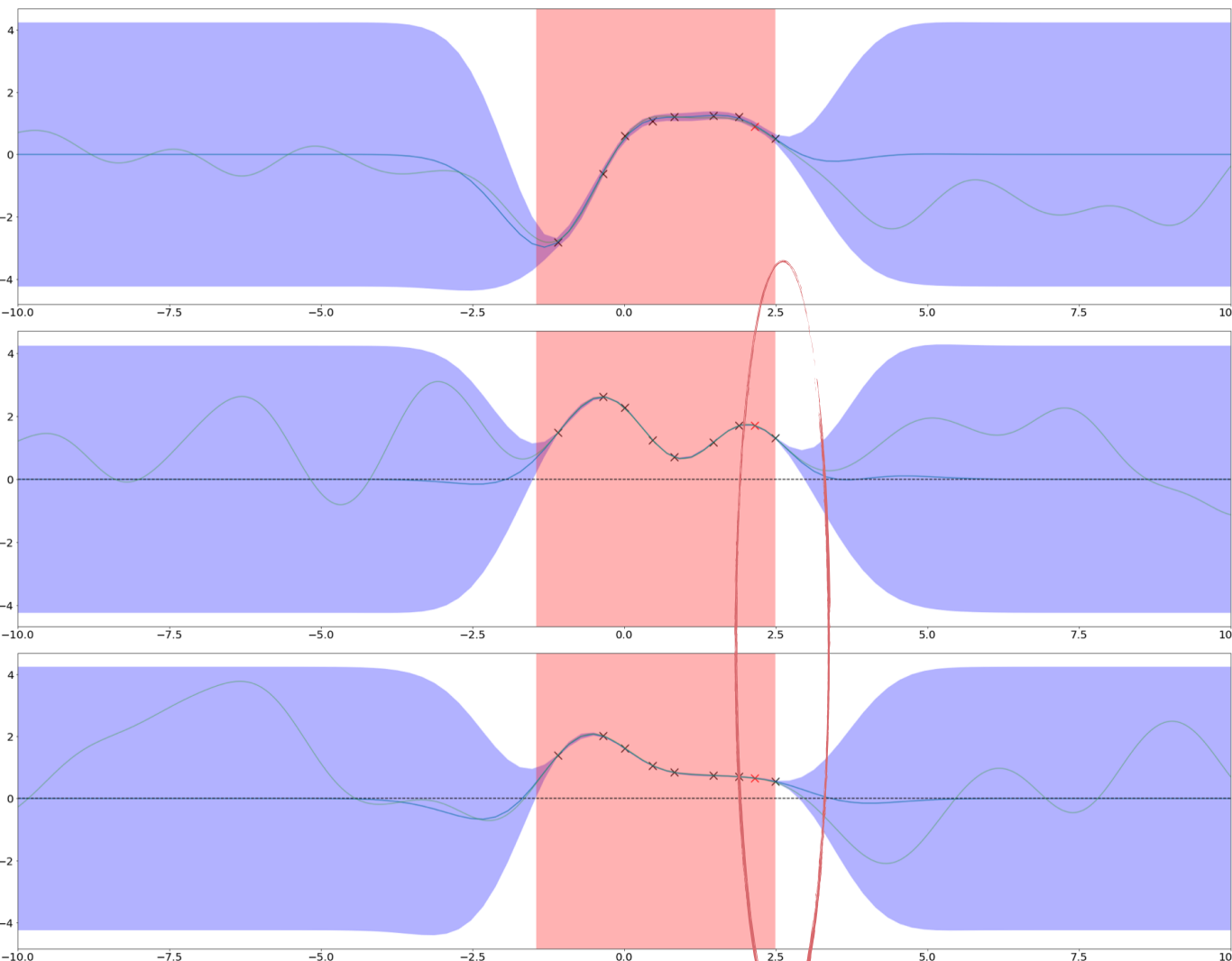
- repeat 1000 times and get results:



• Safe region result

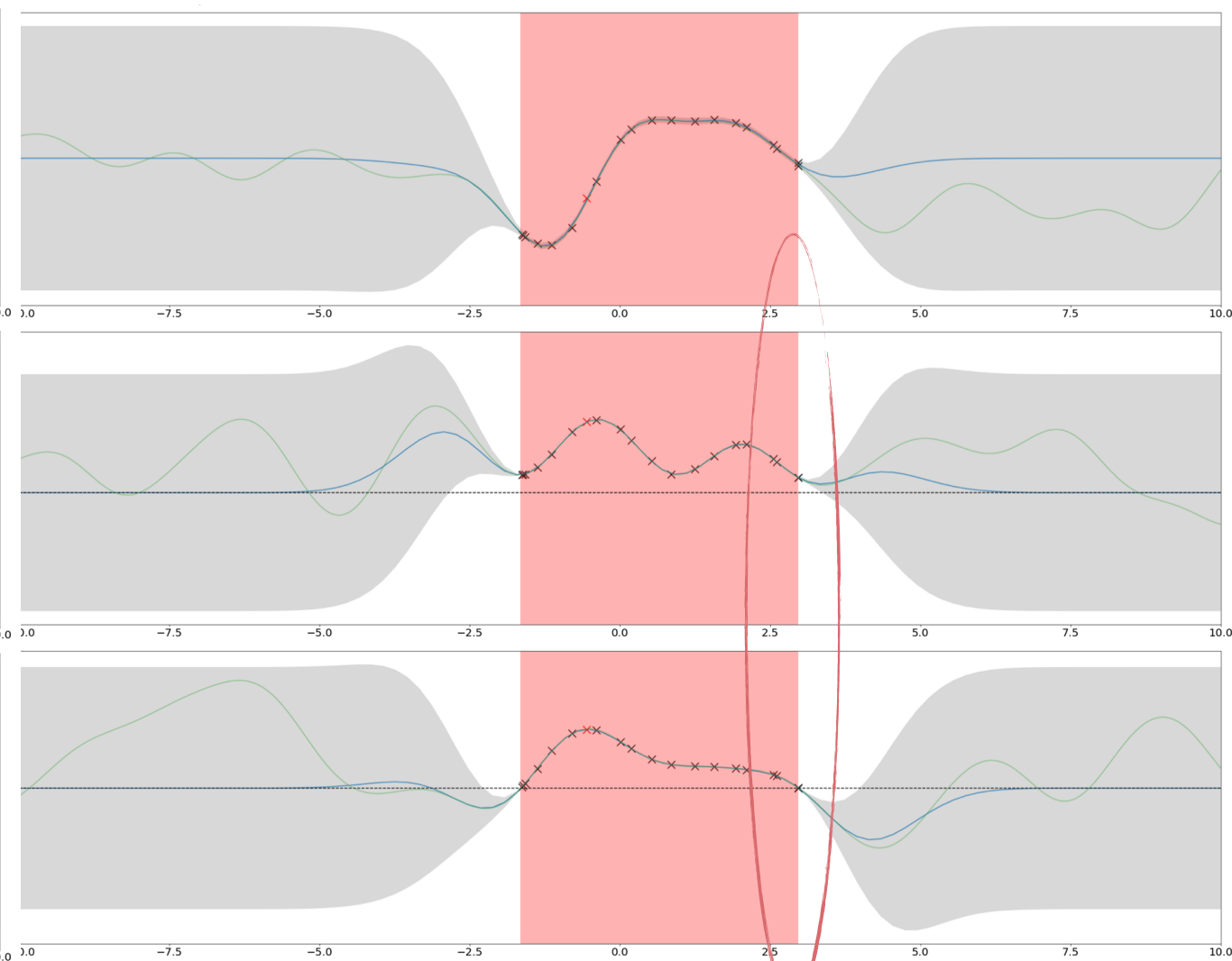
- Two safety function $[f1, f2]$, with noise $[1e-5, 1e-3]$, number of iterations = 20
- StageOpt has more probabilities to get larger safe region.

SafeOpt



$[-1.45145, 2.41241]$

StageOpt

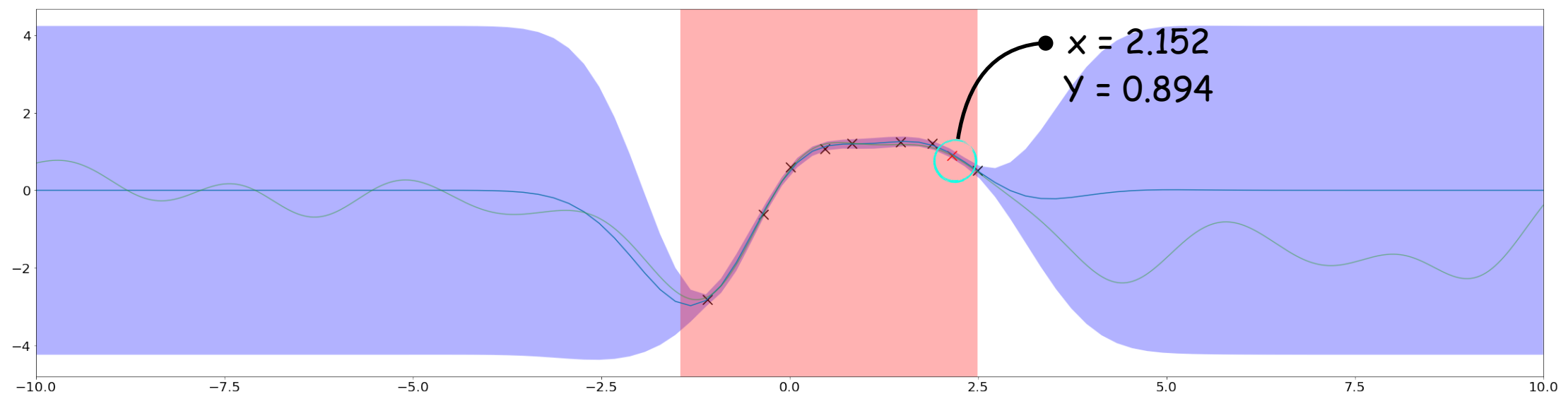


$[-1.65165, 2.97297]$

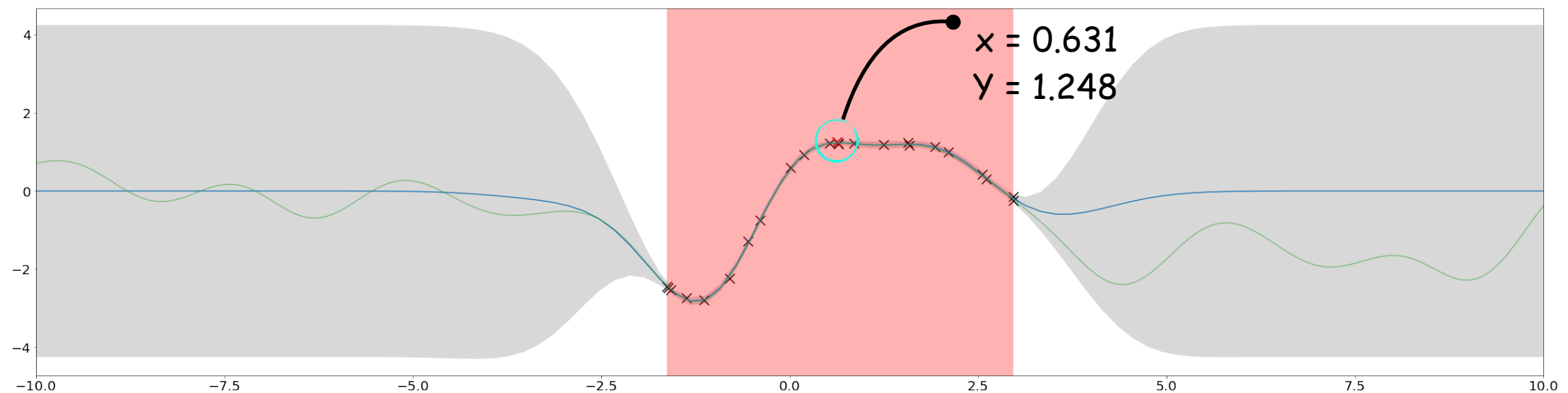
- **Optimization result**

- Two safety function $[f1, f2]$, with noise $[1e-5, 1e-3]$, number of iterations = 25
- StageOpt has more probabilities to get greater optimization result and more controllable

SafeOpt:



StageOpt:



- **Compare StageOpt and SafeOpt**

Strategy Name	Guarantee safety	Safe region	Optimization result	Controllabil ity	Numbers of iterations
StageOpt	Higher probability	Larger	Greater	Controllable	Smaller
SafeOpt	High probability	Smaller	Smaller	Uncontrollable	Greater

- **Future Work**

- How to deal with a smaller signal to noise ratio?
- How to apply these works to continuous intervals?
- If the utility function is changing?