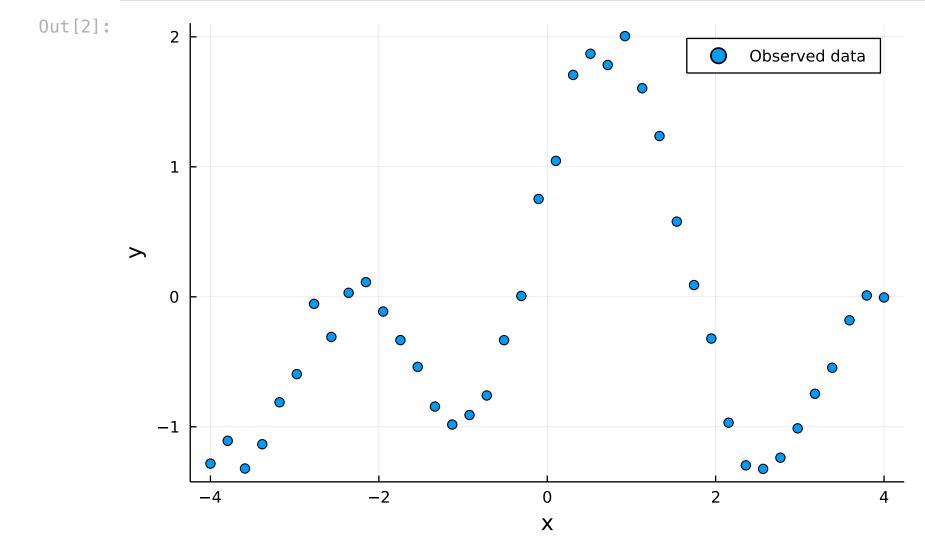
「ガウス過程と機械学習」

P.81 ガウス過程回帰

```
In [1]:
    using LinearAlgebra
    using Plots
```

観測データの生成

```
In [2]: N = 40
x_{train} = range(-4, 4, length=N)
f(x) = sin(2x) + cos(x-\pi/5) + 0.1 * randn()
y_{train} = f.(x_{train})
scatter(x_{train}, y_{train}, xlab="x", ylab="y", label="Observed data")
```



パラメータの設定

0.14108273887509437

```
In [3]:  \theta = \text{rand}(3) \\ \theta_1, \ \theta_2, \ \theta_3 = \theta 
Out[3]:  3 \text{-element Vector{Float64}:} \\ 0.9312686010495987 \\ 0.7877613526010105
```

ガウスカーネル関数を定義

```
In [4]: k(x1, x2, \theta_1, \theta_2, \theta_3) = \theta_1 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot ^2 \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(x1' \cdot - x2) \cdot / \theta_2) + \theta_3 \cdot \exp(-(
```

 $n_{\text{upt}}[A]$ k (generic function with 1 method)

ガウス過程回帰を計算

```
Out[5]:
```

In []: