Gaussian Process Active Learning

Gaussian Process Regression

n observations: $\mathbf{y} = \{y_1, \ldots, y_n\}$

covariance:

$$K = \begin{bmatrix} k(x_1, x_1) & k(x_1, x_2) & \cdots & k(x_1, x_n) \\ k(x_2, x_1) & k(x_2, x_2) & \cdots & k(x_2, x_n) \\ \vdots & \vdots & \ddots & \vdots \\ k(x_n, x_1) & k(x_n, x_2) & \cdots & k(x_n, x_n) \end{bmatrix}$$

For χ^* , $K_* = [k(x_*, x_1) \ k(x_*, x_2) \ \cdots \ k(x_*, x_n)]$ $K_{**} = k(x_*, x_*)$ How to get \mathcal{Y}_* ?

假设:数据服从均值为0的联合高斯分布。

$$\begin{bmatrix} \mathbf{y} \\ y_* \end{bmatrix} \sim \mathcal{N} \left(\mathbf{0}, \begin{bmatrix} K & K_*^{\mathrm{T}} \\ K_* & K_{**} \end{bmatrix} \right)$$
$$\mathbf{y}_* | \mathbf{y} \sim \mathcal{N} (K_* K^{-1} \mathbf{y}, K_{**} - K_* K^{-1} K_*^{\mathrm{T}})$$



Gaussian Process Classification



1. Finding the probability $p(f_*|\mathbf{f})$ is similar to GPR, i.e.

$$p(f_*|\mathbf{f}) = \mathcal{N}(K_*K^{-1}\hat{\mathbf{f}}, K_{**} - K_*(K')^{-1}K_*^{\mathrm{T}})$$

2. Squash f_* to find the probability of class membership.

$$\pi(f) = \operatorname{prob}(y = 1|f)$$

Bayesian Optimization





Acquisition Function



Active Learning for GP

$$\begin{aligned} \mathcal{Q}_{\mu_*}(\mathcal{U}) &= \operatorname{argmin}_{\widehat{\boldsymbol{x}}^{(i)} \in \mathcal{U}} |\mu_*(\widehat{\boldsymbol{x}}^{(i)})| & \text{Exploitation} \\ \mathcal{Q}_{\sigma_*^2}(\mathcal{U}) &= \operatorname{argmax}_{\widehat{\boldsymbol{x}}^{(i)} \in \mathcal{U}} \sigma_*^2(\widehat{\boldsymbol{x}}^{(i)}) & \text{Exploration} \\ \mathcal{Q}_{\operatorname{unc}}(\mathcal{U}) &= \operatorname{argmin}_{\widehat{\boldsymbol{x}}^{(i)} \in \mathcal{U}} \frac{|\mu_*(\widehat{\boldsymbol{x}}^{(i)})|}{\sqrt{\sigma_*^2(\widehat{\boldsymbol{x}}^{(i)})}} & \text{Trade-off} \end{aligned}$$

[Kapoor et.al, ICCV, 2007]

PU Learning

- 有限的正类样本
- 负类种类多样
- 查询到负类的处理方式
- 模型的更新

The distribution of the three groups of examples changes over a period of time as more and more examples get labeled

Discard negative:

- Label instances which are located near the surface of the SVDD sphere
- Identify instances in the unlabeled pool that fit the target class best, and label these instances first
- First pick the instances which the initial model considers to match the learned class the *least*

Keep negative:

- 转化为二分类的监督学习或半监督学习

GP in One-Class Classification



$$\mu_* = \mathbf{k}_*^T (\mathbf{K} + \sigma_n^2 \mathbf{I})^{-1} \mathbf{1}$$

The mean decreases with distance from the training data.



[Kemmler et.al, Pattern Recognition, 2013]

GPAL for PU



- Query Strategy: trade-off
- Model Update: 如何利用未标记数据