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一个比较长的 中文题目

摘要

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A long English title

Abstract

The body of your English Abstract.

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第一章 介绍

1.1 参考文献示例

文章 FullyConvolutionalNetwork^[1].

多篇文章引用示例^[1-4]。

1.2 图像插入示例

图像位置在本项目的 Img 文件夹下, 可以点上传后同本地上传, 建议按照章节命名编号, 便于保存.

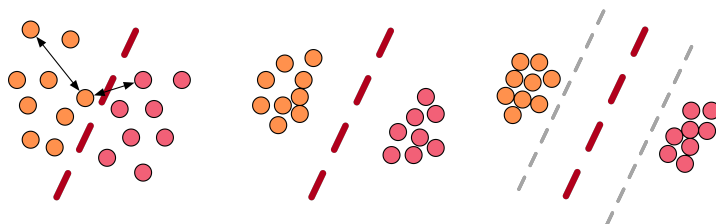


图 1.1 度量学习和分类问题的区别

如图 1.1

1.3 多张图像插入示例

多张图像可以分别标记 Label, 方便后文引用.

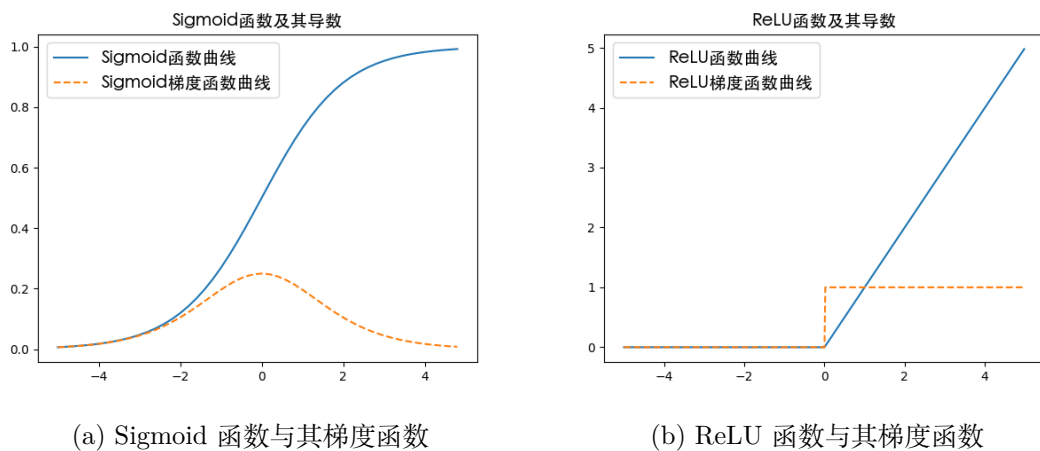


图 1.2 常用的损失函数

如图 1.2b.

1.4 其他环境测试

定理 1.1 这是一个定理.

定义 1.1 这是一个定义.

例 1.1 这是一个例子.

证明 1 这是一个证明.

解 1 这是一个解.

初号字.
小初号字.
一号字.
二号字.

小二号字.

三号字.

四号字.

小四号字.

五号字.

小五号字.

六号字.

七号字.

公式

$$\frac{\partial \mathcal{L}}{\partial x_i} = \frac{\frac{\partial \mathcal{L}}{\partial \tilde{x}_i} - \tilde{x}_i \sum_j \frac{\partial \mathcal{L}}{\partial \tilde{x}_j}}{\|x\|_2}. \quad (1.1)$$

表格示例

表 1.1 部分模型在城市景观数据集上的表现

序号	模型框架	特征提取器	Acc	mAcc	mIoU
1	FPN	ResNet-50	95.81%	81.86%	74.51%
2	FPN	ResNet-101	96.01%	83.39%	75.80%
3	PointRend	ResNet-50	95.95%	84.05%	76.47%
4	PointRend	ResNet-101	96.23%	85.70%	78.30%
5	DeeplabV3+	ResNet-50	96.44%	86.96%	80.08%
6	DeeplabV3+	ResNet-101	96.57%	87.86%	80.97%

第二章 相关工作

第三章 解决方法

第四章 讨论

第五章 总结与展望

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