





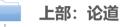
### 既是世间法、自当有分别

艾新波 / 2018 • 北京



#### 课程体系







- 第2章 所谓学习、归类而已
- 第3章 格言联璧话学习
- 第4章 源于数学、归于工程
- 中部:执具
  - 第5章 工欲善其事必先利其器
  - 第6章 基础编程
  - 第7章 数据对象

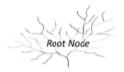


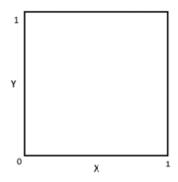




- 第10章 观数以形
- 第11章 相随相伴、谓之关联
  - 🗐 第12章 既是世间法、自当有分别
  - 第13章 方以类聚、物以群分
  - 第14章 庐山烟雨浙江潮

#### 不纯度与不确定性的减少

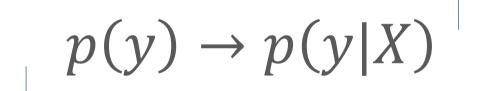




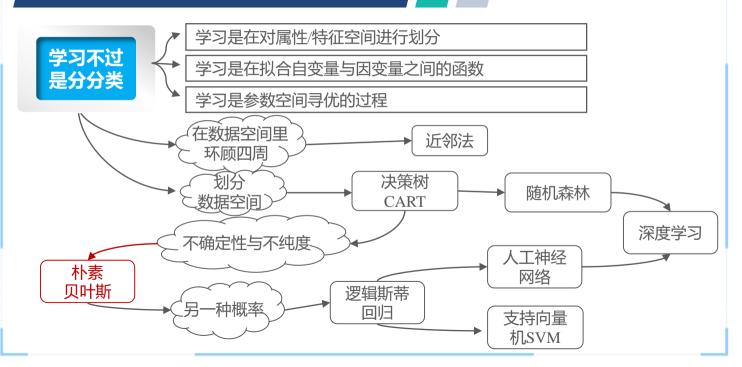
For more tutorials: annalyzin.wordpress.com

https://algobeans.com/

#### 另一种不确定性的减少: 有了更多的证据



#### 算法模型



#### 贝叶斯公式用于分类

#### 贝叶斯公式:

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

#### 换一种形式:

$$P($$
类别 $\mathbf{1}|$ 特征组合) = 
$$\frac{P($$
特征组合|类别 $\mathbf{1})P($ 类别 $\mathbf{1})}{P($ 特征组合)}

$$P($$
类别**2**|特征组合) = 
$$\frac{P($$
特征组合|类别**2**) $P($ 类别**2**)}{P(特征组合)}



#### 朴素贝叶斯

#### 贝叶斯公式:

$$p(C|F_1,\cdots,F_n) = \frac{p(C)p(F_1,\cdots,F_n|C)}{p(F_1,\cdots,F_n)}$$

#### 分母不发挥作用:

$$p(C|F_1,\cdots,F_n) \propto p(C)p(F_1|C)p(F_2|C,F_1)\cdots p(F_n|C,F_1,\cdots,F_{n-1})$$

#### 朴素一点——特征之间相互独立:

$$p(F_i|C,F_j) = p(F_i|C)$$

#### 朴素贝叶斯

#### 于是有:

$$p(C|F_1,\cdots,F_n) \propto p(C)p(F_1|C)p(F_2|C)\cdots p(F_n|C) = p(C)\prod_{i=1}^n p(F_i|C)$$

#### 用于分类:

$$classify(f_1, f_2, \dots, f_n) = \underset{c}{argmaxp}(C = c) \prod_{i=1}^{n} p(F_i = f_i | C = c)$$

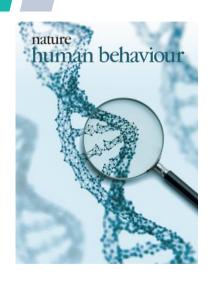
#### 朴素贝叶斯

Article

#### Machine learning of neural representations of suicide and emotion concepts identifies suicidal youth

Marcel Adam Just <sup>™</sup>, Lisa Pan, Vladimir L. Cherkassky, Dana L. McMakin, Christine Cha, Matthew K. Nock & David Brent

The clinical assessment of suicidal risk would be substantially complemented by a biologically based measure that assesses alterations in the neural representations of concepts related to death and life in people who engage in suicidal ideation. This study used machine-learning algorithms (Gaussian Naive Bayes) to identify such individuals (17 suicidal ideators versus 17 controls) with high (91%) accuracy, based on their altered functional magnetic resonance imaging neural signatures of death-related and life-related concepts. The most discriminating concepts were 'death', 'cruelty', 'trouble', 'carefree', 'good' and 'praise'. A similar classification accurately (94%) discriminated nine suicidal ideators who had made a suicide attempt from eight who had not. Moreover, a major facet of the concept alterations was the evoked emotion, whose neural signature served as an alternative basis for



Marcel Adam Just, et al. Machine learning of neural representations of suicide and emotion concepts identifies suicidal youth. Nature Human Behaviour (2017). doi:10.1038/s41562-017-0234-y

```
library (e1071)
imodel <- naiveBayes(wlfk~.,</pre>
                      data = cjb[train set idx, ])
predicted train <- predict(imodel,</pre>
                             newdata = cjb[train set idx,],
                             type = "class")
Metrics::ce(cjb$wlfk[train set idx], predicted train)
#> [1] 0.2920518
```

```
sp <- Sys.time() #记录开始时间
cat("\n[Start at:", as.character(sp))
for (i in 1:length(kfolds)) {
  curr fold <- kfolds[[i]] #当前这一折
  train set <- cjb[-curr fold,] #训练集
  test set <- cjb[curr fold,] #测试集
  imodel kfold <- naiveBayes(wlfk~.,</pre>
                             data = train set)
  predicted train <- predict(imodel kfold,</pre>
                              train set, type = "class")
```

```
imetrics("naiveBayes", "Train",
           predicted train, train set$wlfk)
  predicted test <- predict(imodel kfold,</pre>
                             test set, type = "class")
  imetrics("naiveBayes", "Test",
           predicted test, test set$wlfk)
ep <- Sys.time()
cat("\tFinised at:", as.character(ep), "]\n")
cat("[Time Ellapsed:\t",
    difftime(ep, sp, units = "secs"), " seconds1\n")
```

```
#> 41 naiveBayes Train 0.7040230
                                 0.2959770
#> 42 naiveBayes Test 0.6794872 0.3205128
#> 43 naiveBayes Train 0.7155172 0.2844828
#> 44 naiveBayes Test 0.7179487 0.2820513
#> 45 naiveBayes Train 0.7068966 0.2931034
#> 46 naiveBayes Test 0.7307692 0.2692308
#> 47 naiveBayes Train 0.7183908 0.2816092
#> 48 naiveBayes Test 0.7051282 0.2948718
. . . . . .
#> 59 naiveBayes Train 0.6944046 0.3055954
#> 60 naiveBayes Test 0.7142857 0.2857143
```

#### 究竟是一种什么关系

既然所有规律都是关系那么,请问:

## 朴素贝叶斯 究竟是什么关系

得到规律的表现形式是什么



# 謝謝聆听 Thank you

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课程 网址: https://github.com/byaxb/RDataAnalytics



