Assignment Project Exam Help

https://powcodendoming

Dr Karim Seghouane

Add William Mathematics & Statistics Statistics William Powcoder

Outline

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 \S{ii} . Introduction & overview

§iii. Basic concepts WeChat powcoder

Admin

- Baldwin Spencer Theatre or via zoom
- Practical lab Jiadong Mao
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 - ► f2f. G2/Tues. 9:00-10:00, G3/Tues. 14:15-15:15 in PAR-Peter Hall-G70 (Wilson Laboratory) G4/Fri. 10:00-11:00 in PAR-Peter Hall-G69 (Thompson Lab)
 - A 0 0: GV Web 15:17-20:15, B O V V 10:15 10:16:15 G6/Thur. 16:15-17:15, G7/Wed J. 16:15-17:15 and G8/Thur. 12:00-13:00
- Consultation time
 - Wedn. 08:00-09:00
 - Frid. 08:00-09:00



Admin - Assessment

- Interesting assignments due early, mid and late semester

 - ► Each written assignment is worth 15%
- * Add We Chat powcoder

Admin - LMS

- All relevant teaching material will be posted on LMS (including the supplementary and the additional material)
- Due to the time limits during the lectures, you will need to go over some mathematical details & deepen your knowledge outside of the lectures time.
- ► MadenmWneschenattep@Wscoder
- ► Discussion board?

Admin - References

Assignment Project Exam Help * Elements of Statistical Learning by Hastie Trevor,

- Tibshirani Robert & Friedman Jerome (2009).
- An Introduction to Statistical Learning by James Gareth Witten Daniela, Hastie Trevor & Tibshirani Robert (2013).
- ► Introducing Monte Carlo Methods with R by R.P. Christian & G. Casella (2010).
- Christian & G. Casella (2010).

 Computational Statistics by t. Privers & Co. Cecing (2005).
- Academic articles, links to blogs & videos on LMS.

Admin - Communication

Assignments will be addressed during office hours and if time permits.

- It is expected that questions regarding these matters will be asked quing consultation four shauring the laboratories.
- ► There will be no consultation hours during non-teaching periods.
- ► IMIGEOUT WITE GUIN MODO WE COULD St two weeks of the examination period.
- I will provide an extra consultation time on week 12.
- Students should plan ahead with any queries regarding assignments or course material.

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Lecture schedule (provisional)

- **▶ Data mining** (7 weeks):
 - linear model selection and regularisation; kernel and local regression; basis (National of pipe regression) general additive modes (GAM); classification and regression trees; bagging, random forests and boosting; support vector machines (SVM); component analysis and deep learning.
- ► Compliation V statistics (a fveeter) WCOCET
 - ► EM algorithm; Bayes computing, Monte Carlo methods; and bootstrap methods.

Question to the class...

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How would you explain in a few sendences to a general audient type. // powcoder. com

- What a statistician / data scientist does?
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Some answers

- ▶ What a statistician / data scientist does?
- Freat a model/(hox) to understand the releathinseep
- make sense of data, extract important patterns and trends, we can call this learning from data
- trends, we can call this learning from data

 where the man proof a learning from data

 trends, we can call this learning from data

 to the control of the le
- Predict/decide or describe/understand

The power of (statistical) model



Once upon a time ...

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Image source: http://freshlearners.blogspot.com.au/2015/07/most-recent-communications-technology.html

12/25

Introduction

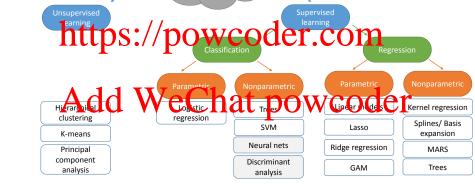
Today

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Image source: http://psmit.com/about.html

Taxonomy of covered methods



Regression vs. Clasification

- hetersion / he invis tenedid the numerical quitcome of a subject based on its features.
- Clasification The aim is to predict the class belonging of a subject based on its features. Add WeChat powcoder

Parametric vs. Nonparametric

- hatamatic approach who have an explicit assumption about the functional form. (Restriction on the shape)
- Nonparametric approach does not assume any functional form for the underlying model structure.

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Parametric models

Nonparametric models

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Advantages S Signer to The Continue nterper

- Fast to fit (CPU time)
- Requires to fit a small number of parameters, therefore requires

Disadvantages

- Highly constrained to the specified function form.
- It is hard to a atol the mathematical form of the model to the shapevin DGP?
- Can fit well relatively simple data structures (complex models -> danger of over-fitting).

Advantages C Advan

- Weak(er) assumptions about the underlying function.
- Outperforms in prediction context,

Disadvantages

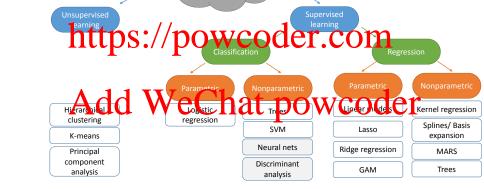
- Slow(er) to fit (CPU time).
- Requires large(r) data sets

Harder to interpret
Higher cangur of over-litting

Supervised vs. Unsupervised

- Supervised learning In case our data set contains the response (outcome) measurements the fitted model relates the different features to the response.
- Unsupervised learning In case our data set contains only information about the different features of the subjects, the fitter into el ains de learnest le une tribute into en un subjects in the features.

Taxonomy of covered methods



How models are fitted

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 We minimize some criterion (e.g. squared error with some
- We minimize some criterion (e.g. squared error with some additional penalty)

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Trade-offs

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- Intitity oredicts in two controls
 Bias Variance trade-off.
- - ightharpoonup variance how much \hat{f} changes with different data set.

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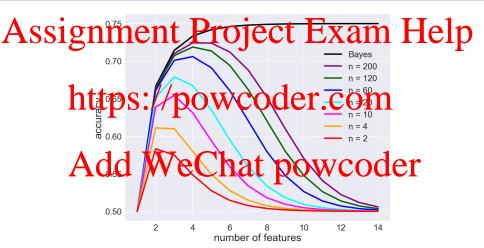
Complexity

Assignments to the light of the last of the supervised learning methods.

- Notion of curse of dimensionality: for a fixed sample size, the expected cossification ever will have by intreasing the number of features, but eventually will decrease. This is a consequence of the large size of high-dimensional spaces, which leadure over espendingly large training, sample sizes.
- Scissors effect: the expected error typically decreases as sample size increases, and more complex classification rules achieve smaller error for large sample sizes; however, simpler classification rules can perform better under small sample sizes, by virtue of needing less data.

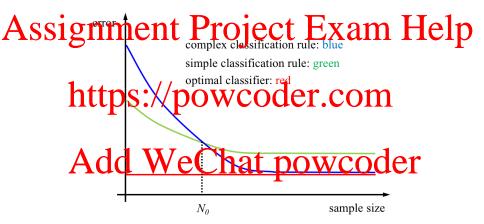
Introduction

Complexity



Expected accuracy in a discrete classification problem for various training sample sizes as a function of the number of predictors.

Complexity



Expected error as a function of sample size for two classification rules. There is a problem-dependent critical sample size N0, under which one should use the simpler classification rule.

