

Forecasting mood in bipolar disorder

P.J. Moore¹ M. Little² P. McSharry³

¹Oxford Centre for Industrial and Applied Mathematics
University of Oxford

²MIT Media Lab, Cambridge, Massachusetts

³Smith School of Enterprise and the Environment
University of Oxford

October 17, 2011

Outline

Forecasting
mood in bipolar
disorder

Outline

Data
characteristics

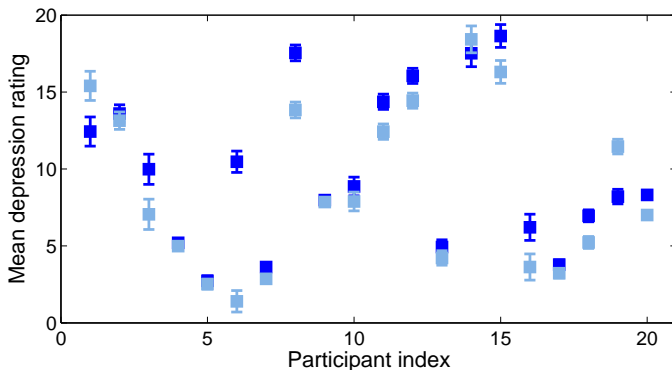
Forecasting
Depression

Proposals

Summary

- Some qualities of the depression data - nonstationarity and a wide variation in roughness
- Forecasting depression using simple exponential smoothing and Gaussian process regression
- Effect of gender on forecastability
- A proposal for modeling with hidden Markov models

Non Stationarity of Time Series



Dark blue and light blue markers represent the sample mean of the first and second halves of the depression time series for 20 patients. The error bars represent the standard error assuming that the numbers are Gaussian distributed and uncorrelated.

Forecasting mood in bipolar disorder

Outline

Data characteristics

Non Stationarity of Time Series

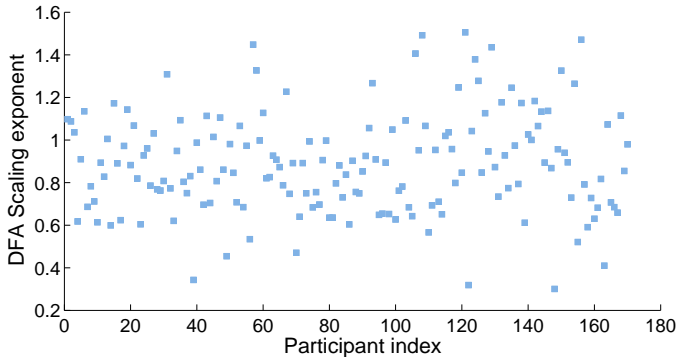
Roughness of time series

Forecasting Depression

Proposals

Summary

Roughness of time series



The scaling exponent provides a measure of the roughness of the time series. It can be seen that the data set is heterogeneous.

Forecasting Depression - method

Forecasting mood in bipolar disorder

Outline

Data characteristics

Forecasting Depression

Forecasting Depression

Exponential Smoothing

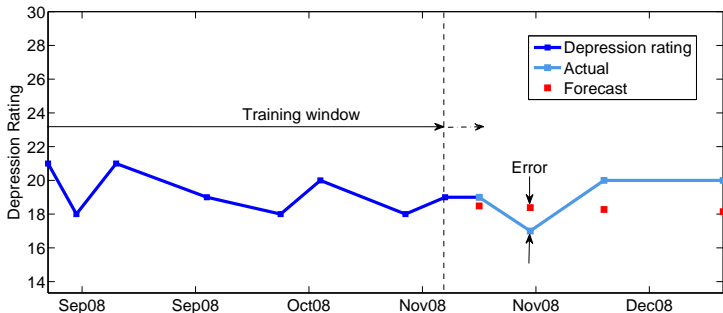
Gaussian Process Regression

Forecast Results

Gender Differences

Proposals

Summary



For each time series, a window is extended stepwise from a minimum offset to a maximum length. At each point a next step forecast is made. This process is repeated for all patients, and the root mean square error taken over all forecasts.

Forecasting method 1 - exponential smoothing

Forecasting
mood in bipolar
disorder

Outline

Data
characteristics

Forecasting
Depression

Forecasting
Depression

**Exponential
Smoothing**

Gaussian
Process
Regression

Forecast Results
Gender
Differences

Proposals

Summary

- Self esteem has been found to be modelled using simple exponential smoothing $\hat{y}_{t+1} = \hat{y}_t + \alpha(y_t - \hat{y}_t)$ (Fortes, *Dynamics of Self-Esteem and Physical Self, Quality and Quantity, 2004*).
- They suggest that two processes underlie the dynamics of self-esteem: *preservation* which tends to restore the previous value after a disturbance and adaptation, which tends to inflect the series in the direction of the perturbation
- The disturbance (or shock) can be interpreted as the outcome of stressors or positive experiences. The value of α might then provide a measure of the preservation/adaptation spectrum.

Forecasting method 2 - Gaussian process regression

Forecasting
mood in bipolar
disorder

Outline

Data
characteristics

Forecasting
Depression

Forecasting
Depression
Exponential
Smoothing

**Gaussian
Process
Regression**

Forecast Results
Gender
Differences

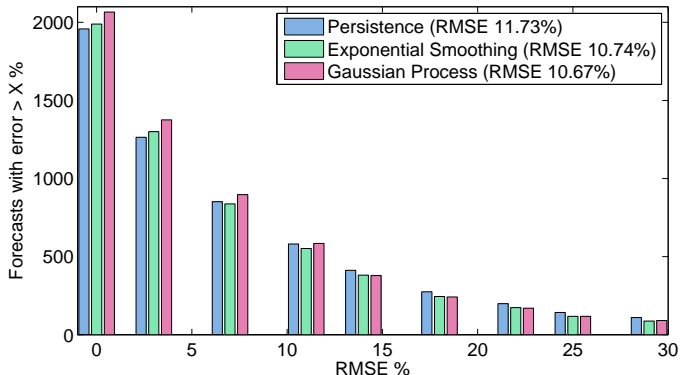
Proposals

Summary

- For a Gaussian Process the joint distribution between any finite number of these variables is Gaussian. The process is specified by a mean function and a covariance function.
- We postulate a certain form of parameterised covariance function and estimate the parameters from the data.
- The expected value and variance of the function can then be found for any input.

Forecasting Results

Cumulative error distribution



The figure shows the cumulative error distribution for 2440 forecasts. The X axis is the forecast error normalised by the maximum of the rating scale. There is little to choose between the methods, except that the nontrivial methods reduce higher errors slightly compared with persistence forecasting.

Forecasting mood in bipolar disorder

Outline

Data characteristics

Forecasting Depression

Forecasting Depression

Exponential Smoothing

Gaussian Process Regression

Forecast Results

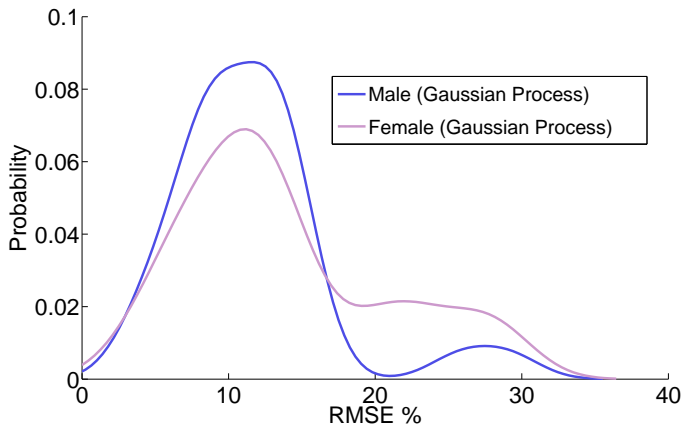
Gender Differences

Proposals

Summary

Gender Differences

Forecast error distributions for male and female patients



Kernel density estimation for distribution of forecast errors for two sets of 30 patients, one all male and the other all female ($p = 0.02$). The x-axis is the normalised RMS forecast error for a participant over a test set.

Forecasting mood in bipolar disorder

Outline

Data characteristics

Forecasting Depression

Forecasting Depression

Exponential Smoothing

Gaussian Process Regression

Forecast Results

Gender Differences

Proposals

Summary

Hidden Markov modelling

Forecasting mood in bipolar disorder

Outline

Data characteristics

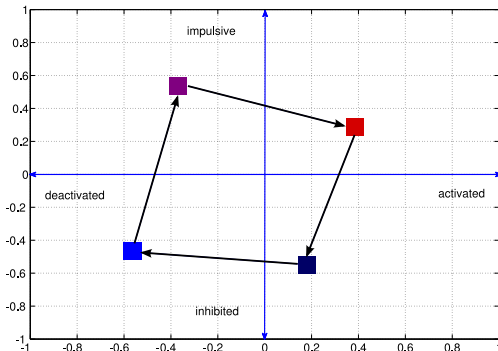
Forecasting Depression

Proposals

Hidden Markov modelling

Pattern recognition techniques

Summary



A state space model of mood in bipolar disorder. Activation is measured along the x-axis and impulsivity is measured along the y-axis. The rectangles are representative points corresponding to mood episodes, and the path represents a mood trajectory. The proposal is to model each mood episode with a hidden Markov model which generates observation vectors.

Pattern recognition techniques

Forecasting
mood in bipolar
disorder

Outline

Data
characteristics

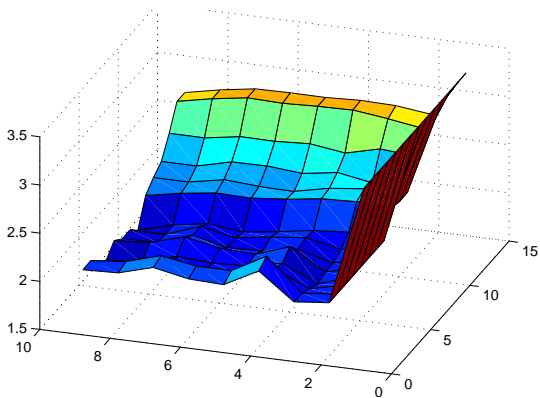
Forecasting
Depression

Proposals

Hidden Markov
modelling

**Pattern
recognition
techniques**

Summary



Currently working on an algorithm to build a library of mood patterns and to use these for forecasting. Results are so far little better than persistence forecasting, but there is potential for improvement.

Summary

Forecasting
mood in bipolar
disorder

Outline

Data
characteristics

Forecasting
Depression

Proposals

Summary

- The mood dataset is heterogeneous with respect to stationarity and roughness.
- Exponential smoothing and Gaussian process forecasting offer little improvement over persistence forecasts for depression.
- A difference was found in the forecast error between sets of male and female patients.
- Proposals for work using hidden Markov models and pattern recognition.