# Fine Wine Price Prediction using Machine Learning

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### Introduction

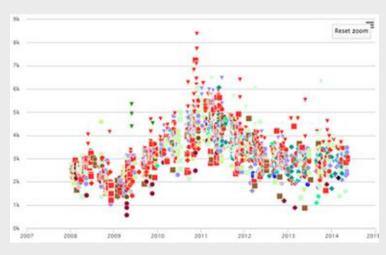
- 1. Background to Fine Wine market
- 2. Invinio
- 3. Research

## Background

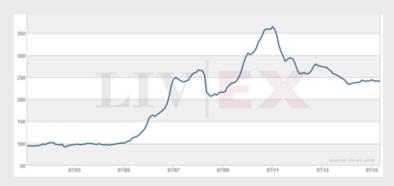
- Many people invest in Treasure assets like classic cars and fine wines.
- Fine wines are inefficiently priced.

 They are more prone to price bubbles than other markets.

#### 1982 Lafite Rothschild



Liv-ex 100



#### Invinio



- Invinio was set up to address these issues:
  - Intelligently price fine wines
  - Offer portfolio diversification
  - Facilitate trading
- Desire was to bring tools from hedge fund domain into this world.
  - Be completely numbers focused be proud that we know nothing relating to the product from a consumptive perspective.
- Make it easier for outsiders to access and start investing in fine wine market.
  - Ended up with a "Nutmeg for Fine Wines".

#### Research

- Love using ML in new domains (e.g. medicine, supply chain, financial services).
- Great opportunity to use ML on the wine price data set.
- My question is always:
  - o Is complexity justified by improved performance?
  - o Michelle did great job publication and media attention
  - Research had significant input from Prof. John Shawe-Taylor

## **Explanation vs Prediction**

#### Explanation

- Coming up with a hypothesis and testing
- Simple linear regression
- Common in wine literature

#### Prediction

- Focus purely on predictive accuracy
- "Black box" algorithms like neural nets
- Common in finance
- Idea: to use predictive machine learning algorithms on wine prices

#### Questions:

- How accurately we can predict the price of wine tomorrow based on today?
- o Is the increase in accuracy worth the effort?



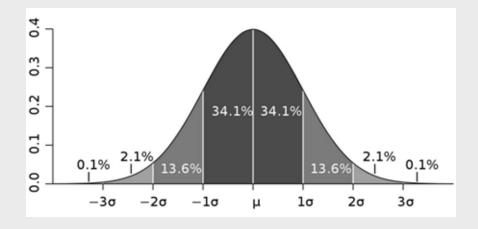
#### What is a Gaussian?

#### Normal distribution with mean and variance

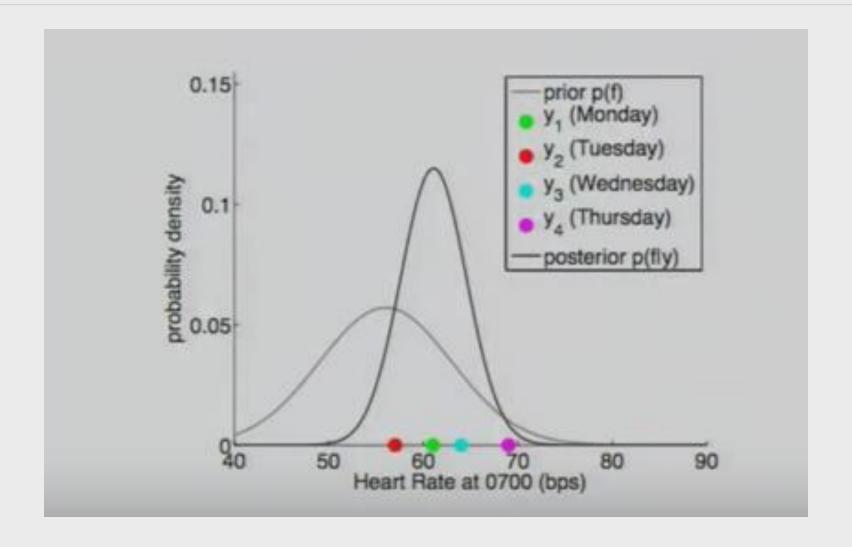
o Gives the probability of an event within an interval happening

#### Bayesian Inference

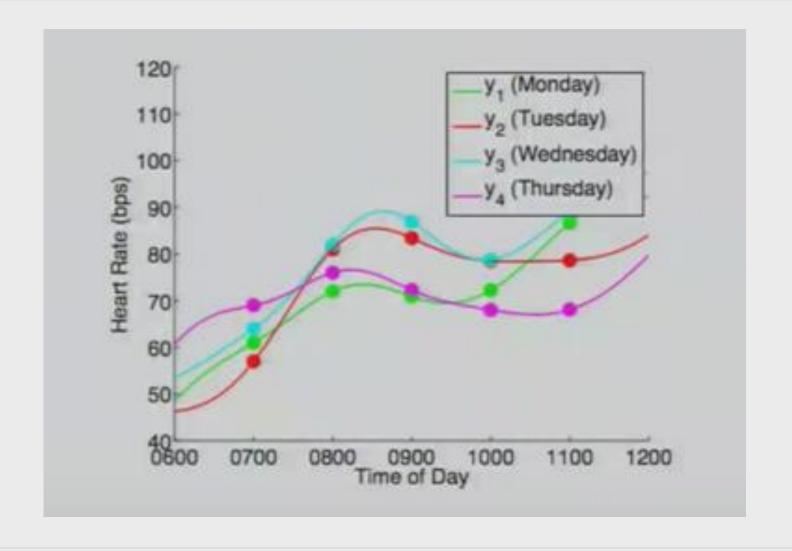
- o Prior: belief about a quantity before taking into account evidence
- Evidence: observed data about the quantity
- Posterior: updated belief in the quantity after considering evidence



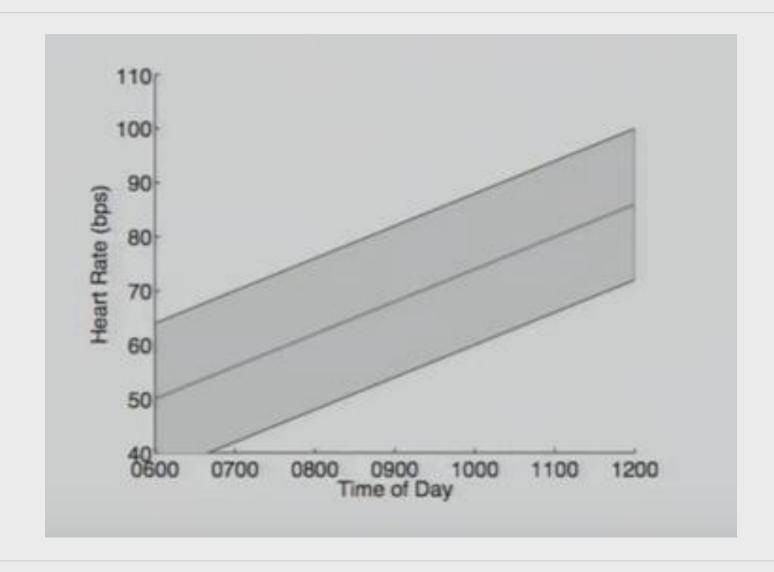
## Bayesian Inference



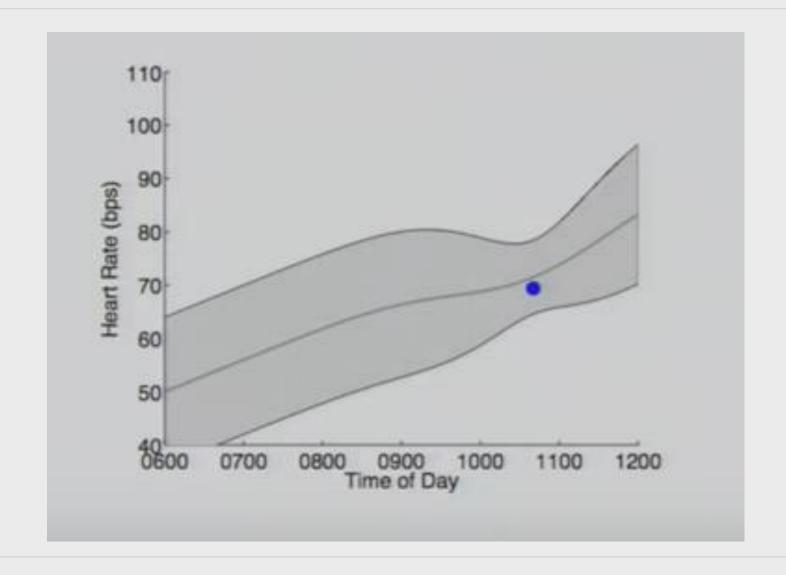
#### What if we are interested in a function over time?



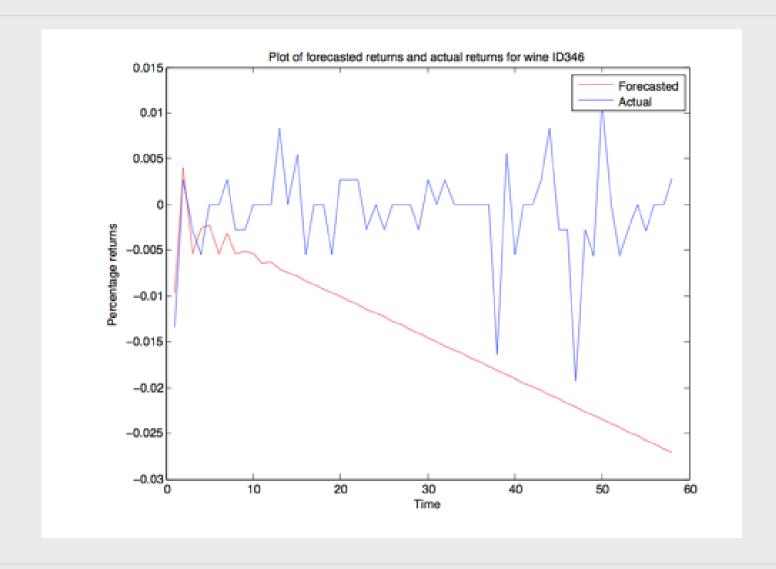
## Bayesian Inference (again)



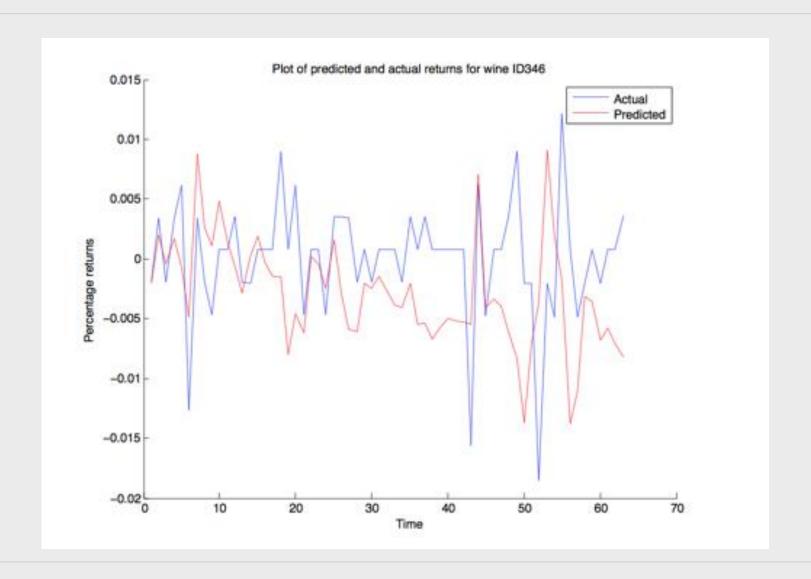
## Bayesian Inference (again)



## Simple time series forecasting



## Gaussian Process Regression





#### Results

- Liv-Ex 100 wines
- Trained on data from August 2013 March 2014
- Tested on data from Apr 2014 July 2014
- 16% decrease in mean squared error
- 50% increase in accuracy in determining positive/negative returns for the next time step





## Explanation vs. Prediction (again)

- Higher predictive power with Gaussian Processes
- Explanatory too!
  - o Weights in linear regression explains the observed data
  - o Gaussian Process regression is essentially Bayesian linear regression with a prior on the weights
  - o Gaussian Process regression lets us specify our belief on the relationship between predictors and response



#### Is all this worth it?



- Significantly improved predictive performance over simpler models
- o Plots: learning more about the nature of wine prices
- Using Gaussian Processes we get both an explanatory & predictive model

## Thank you for listening!

Any questions?