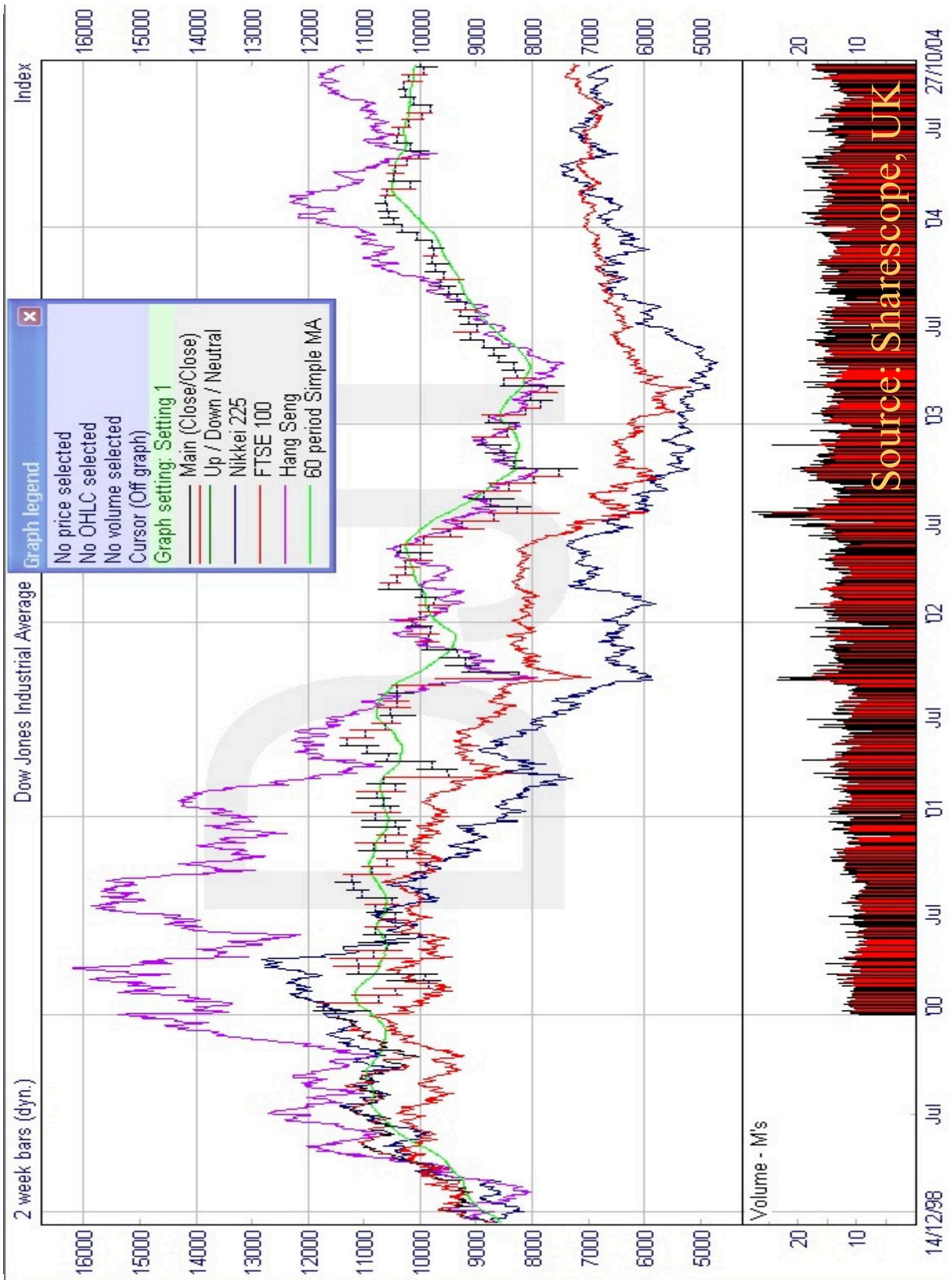


EDDIE: A Genetic Programming Tool for Financial Forecasting

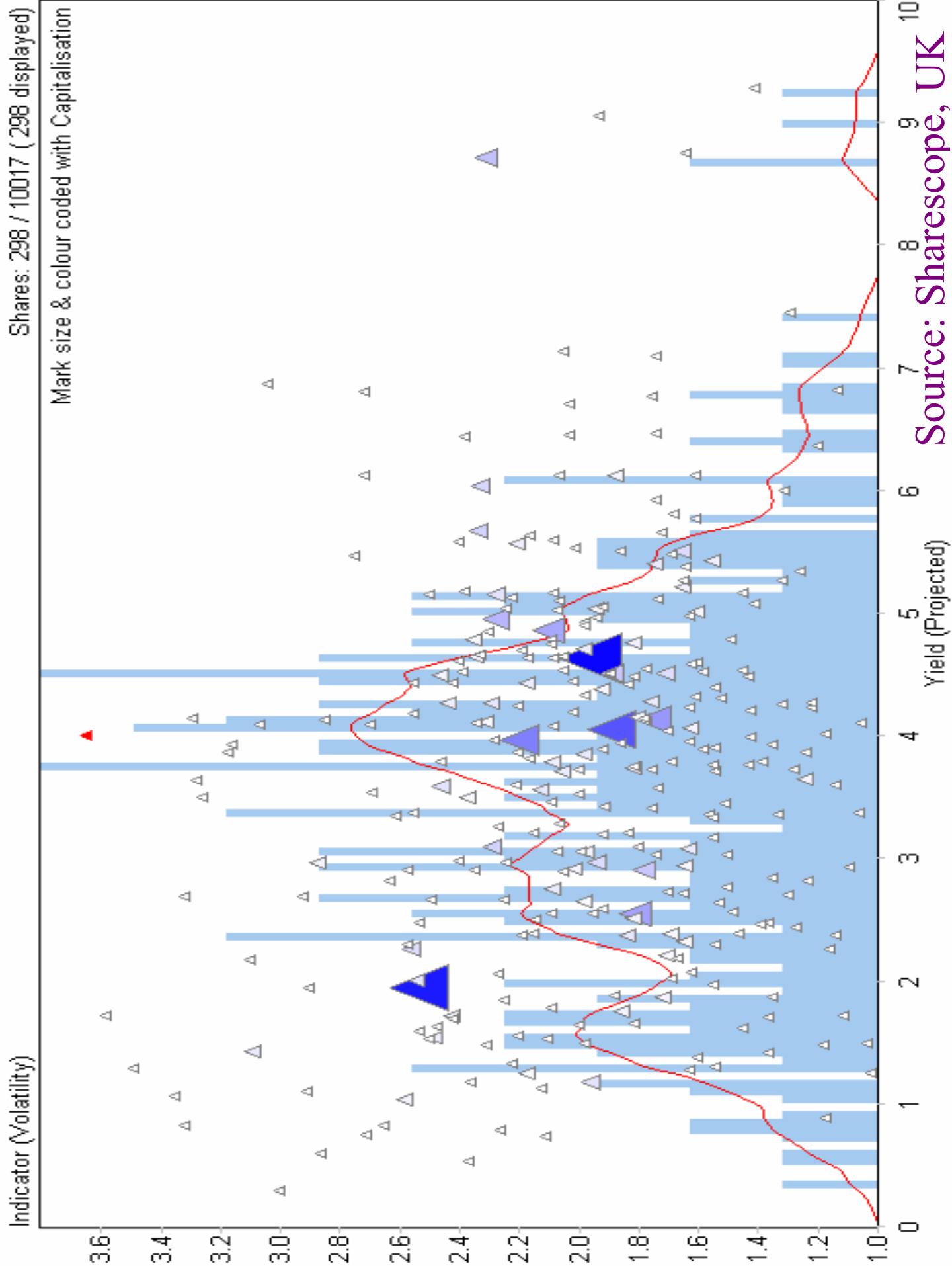
Edward Tsang

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EDDIE = Evolutionary Dynamic Data Investment Evaluator



Indicator (Volatility)



EDDIE Research Agenda

- ◆ Suppose your expert tells you that
 - Price-earning ratio
 - 12 or 50-days moving average
 - Interest rate
 -
- are relevant to future price of FTSE-100
- ◆ How would you actually use them to forecast?
- ◆ Goal: add value to your expert knowledge

Efficient Market Hypothesis (EMH)

- ◆ Financial assets (e.g. shares) pricing:
 - All available information is discounted
- ◆ If EMH holds, forecasting is impossible
 - Random walk theory
- ◆ Assumptions:
 - Efficient markets
 - Perfect information flow
 - Rational traders

Does the EMH Hold?

- ◆ It holds for the long term
- ◆ “*Fat Tail*” observation:
 - big changes today often followed by big changes
(either + or -) tomorrow
- ◆ How fast can one adjust asset prices given a new piece of information?
 - Faster machines certainly help
 - So should faster algorithms

EDDIE / FGP Overview

EDDIE = Evolutionary Dynamic Data Investment Evaluator

FGP = Financial Genetic Programming



- ♦ EDDIE / FGP learns from past history
 - Using Genetic Programming
- ♦ It generates decision trees
 - Which allows it to *explain* its recommendations
- ♦ Used learned rules to answer questions such as:
 - Will prices rise by 4% within the next 21 days?
- ♦ It works with *domain experts* on
 - what *features* are relevant?
 - are the rules generated *reasonable*?

Working with Experts

- ◆ EDDIE is not designed to *replace experts*
 - It is designed to work *with* experts
- ◆ GP is only a tool
 - it needs expert input to be effective
- ◆ Experts channel knowledge into EDDIE:
 - by suggesting what factors are relevant
 - by evaluation of the rules generated
- ◆ EDDIE adds value expert input

Expert Knowledge in EDDIE

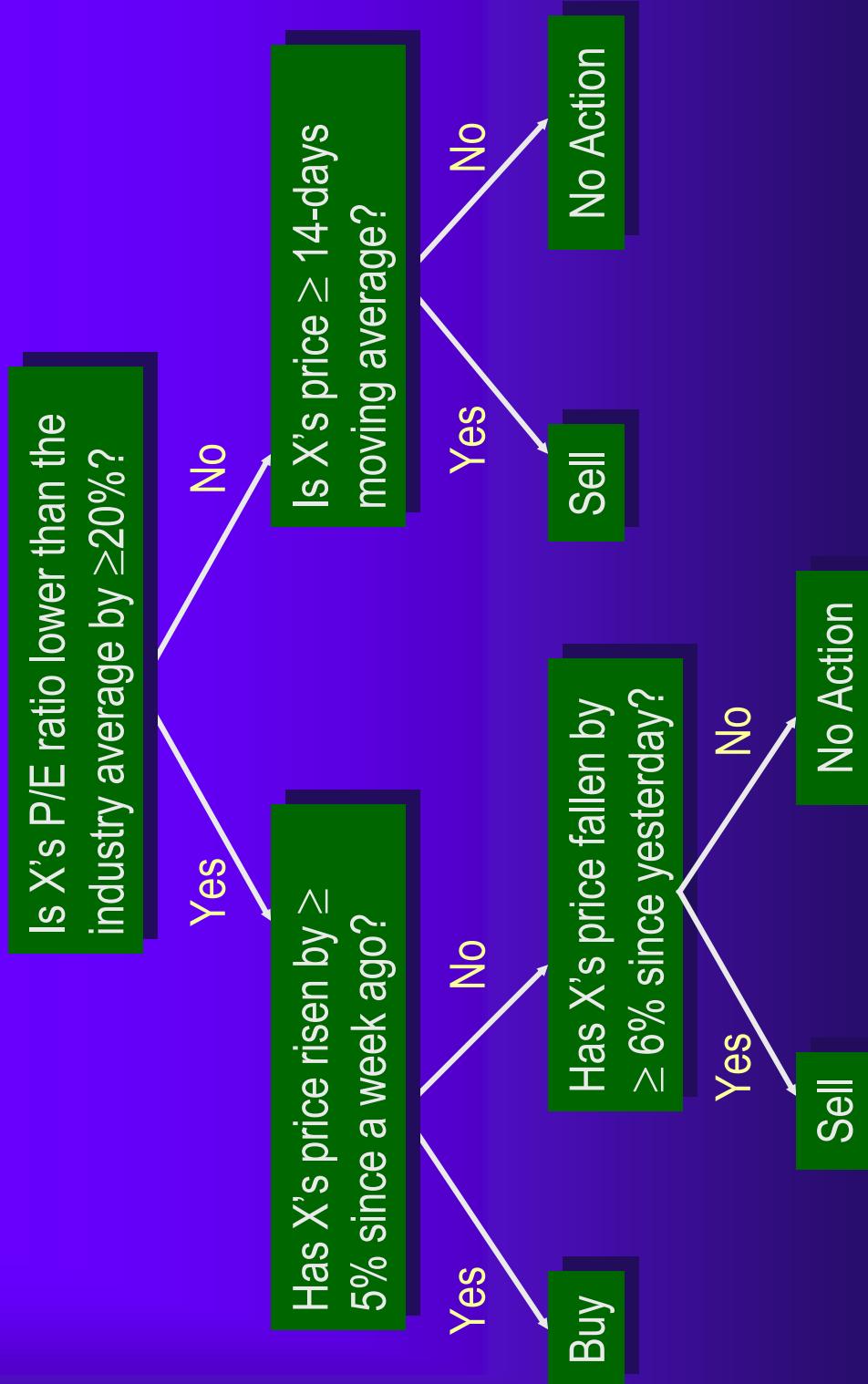


- ◆ Effective channeling of expert knowledge into EDDIE is the key to success

Technical Details

Inside EDDIE / FGP

An Example Decision Tree



Syntax of GDTs in EDDIE-2

```
<Tree> ::= "if-then-else" <Condition> <Tree> <Tree> | Decision  
<Condition> ::= <Condition> "And" <Condition> |  
          <Condition> "Or" <Condition> |  
          "Not" <Condition> |  
Variable <RelationOperation> Threshold  
<RelationOperation> ::= ">" | "<" | "="
```

Variable is an indicator / feature

Decision is an integer, “Positive” or “Negative” implemented

Threshold is a real number

- ◆ Richer language \Rightarrow larger search space

A taste of user input

Given More input: Expert adds:

Define target:

Daily closing	50 days m.a.	Volat- ility	↑4% in 21 days?
90	80	50	0	1	1
99	82	52	0	1	1
87	83	53	1	1	1
82	82	51	1	1	1
.....

EDDIE adds value to user input

- ◆ User inputs *indicators*
 - e.g. moving average, volatility, predictions
 - ◆ EDDIE makes *selectors*
 - e.g. ‘50 days moving average > 89.76 ’
 - ◆ EDDIE combines selectors into *trees*
 - by discovering interactions between selectors
- Finding thresholds (e.g. 89.76) and interactions by human experts is laborious

Research Methodology

- ◆ Concentrate on predicting:
 $G = \text{"will prices go up/down by } r\% \text{ within the next } n \text{ days?"}$
- ◆ To evaluate EDDIE, choose r and n such that 50% of the days achieve G
 - Performance against random decisions
 - Also compared against ID3 / C4.5
- ◆ Measure prediction accuracy
 - Return on investment also used for reference

Testing of EDDIE

- ◆ S&P 500 Index, 1963 to 1974
- ◆ Dow Jones Industrial Average Index
- ◆ Combining expert predictions on Heng Seng Index
- ◆ Shares: IBM, HSBC, BAA, BHP, ANZ, 1991 to 2000
- ◆ UK handicap horse races 1993

EDDIE on S&P 500 daily close

- ◆ Trained: 2/4/63 to 2/7/70 (1800 days)
 - ◆ Tested: 6/7/70 to 25/1/74 (900 days)
 - ◆ Target: “rise of 4% within 63 days”
 - ◆ Input: textbook technical indicators
 - e.g. n days moving averages/ min/ max prices
 - ◆ Result: 54% accuracy, 43% annual return
- Reference: *Software, Practice & Experience*, 28(10) 1998

Performance Measures

		Reality	
		Predictions	
		Negative	Positive
		True Negative	False Positive
		False Negative	True Positive

- ◆ Rate of correctness = $(TN + TP) \div \text{Total}$
- ◆ Rate of failure = $FP \div (FP + TP) = 1 - precision$
- ◆ Rate of missing chances = $FN \div (FN + TP) = 1 - recall$

EDDIE on IBM 1991-1997

- ◆ 60% of recommendations correct
 - where opportunities occur in 45% of the days

IBM		Training Period (1991.10.30-1995.12.27)		
RC	61.4%	0	1	Opport.: 42.3%
RMC	89.1%	654	10	664
RF	15.9%	434	53	487
		1088	63	1151
IBM		Over Test Period (1995.12.28-1997.03.05)		
RC	56.5%	0	1	Opport.: 45.0%
RMC	90.0%	104	6	110
RF	40.0%	81	9	90
AR	210.0%	185	15	8% 200

FGP2 on HSBC 1996-2000

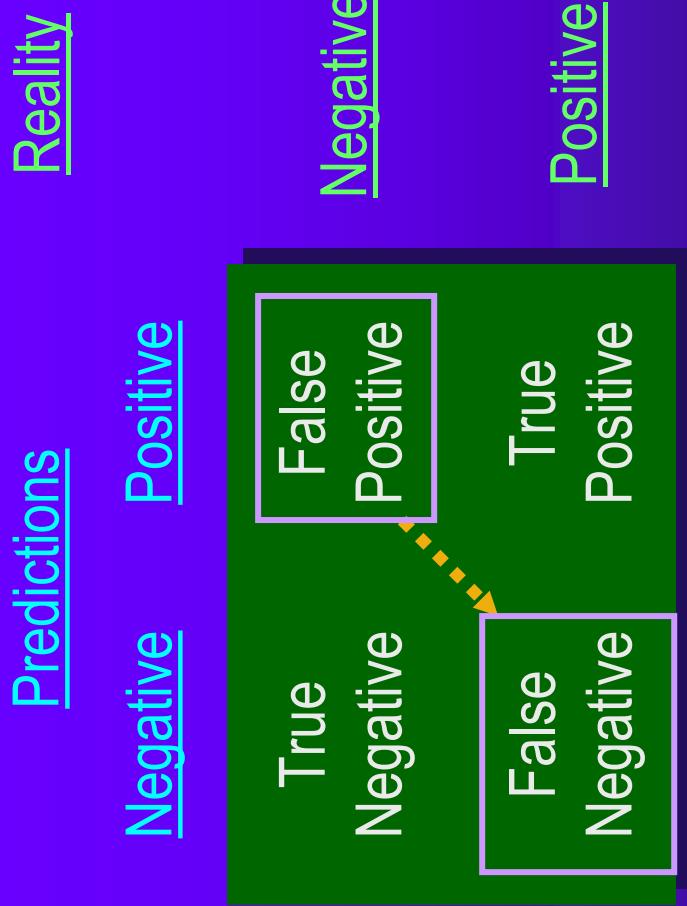
- No recommendations made

HSBA	Training	12/03/96 to 28/5/1999	Opport.:	52.1%
RC	54.8%	0	1	
RMC	83.8%	389	13	402
RF	15.5%	366	71	437
		755	84	839
HSBA	Testing	31/5/1999 to 03/03/00	Opport.:	41.0%
RC	59.0%	0	1	
RMC	100.0%	118	0	118
RF	N.A.	82	0	82
		200	0	200

Improving Precision

Reducing Rate of Failure

Reducing Rate of Failure (RF)



- ♦ $RF = FP \div (FP + TP)$
 - ♦ $RMC = FN \div (FN + TP)$
 - ♦ Reduce RF at the cost of RMC
- $$= 1 - precision$$
- $$= 1 - recall$$

FGP: Constrained Fitness

- ◆ Constraints can help guiding the search

$$\text{Fitness} = w_{rc} \times RC' - w_{mc} \times RMC - w_{rf} \times RF$$

- ◆ $RC' = \begin{cases} RC & \text{if } P+ \in [\text{Min}, \text{Max}] \\ 0 & \text{otherwise} \end{cases}$

Positive

Negative

True	False
False	True

Positive

Negative

False

True

Positive

Negative

Reducing RF

- ◆ Desirable to reduce *Rate of Failure*
 - Missing opportunities may be more acceptable than losing money
- ◆ Our approach:
 - Augment fitness with constraints
 - Tighter constraints means lower RF
 - Even if lower RF \Rightarrow more missed chances
- ◆ Our goal:
 - Allow one to tune RF to one's preference
 - without affecting overall Rate of Correctness

EGP-2 On IBM 1991-1997

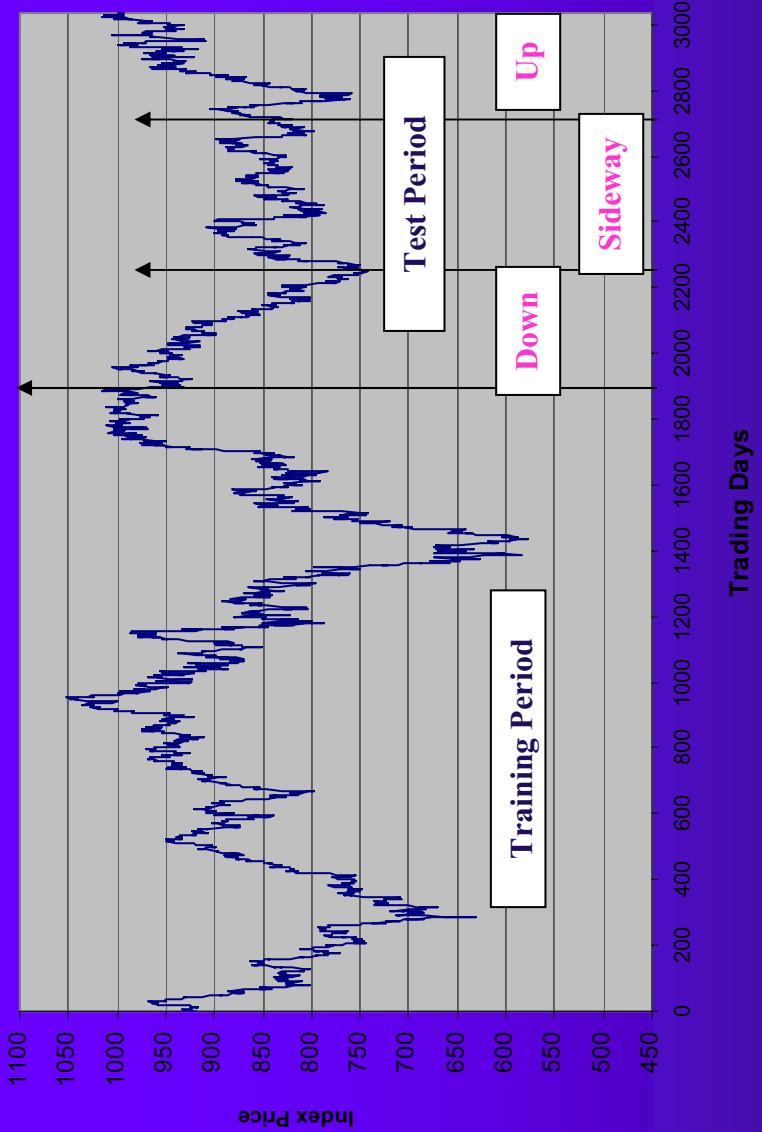
- With low rate of failure specified
- Results are more reliable

General fitness function:	
IBM	Over Test Period (1995.12.28-1997.03.05)
RC	56.5%
RMC	90.0%
RF	40.0% (circled)
AR	210.0%

Constrained fitness function:	
IBM	Test Period (1995.12.28-1997.03.05)
RC	59.0%
RMC	87.8%
RF	21.4% (circled)
AR	511.0%

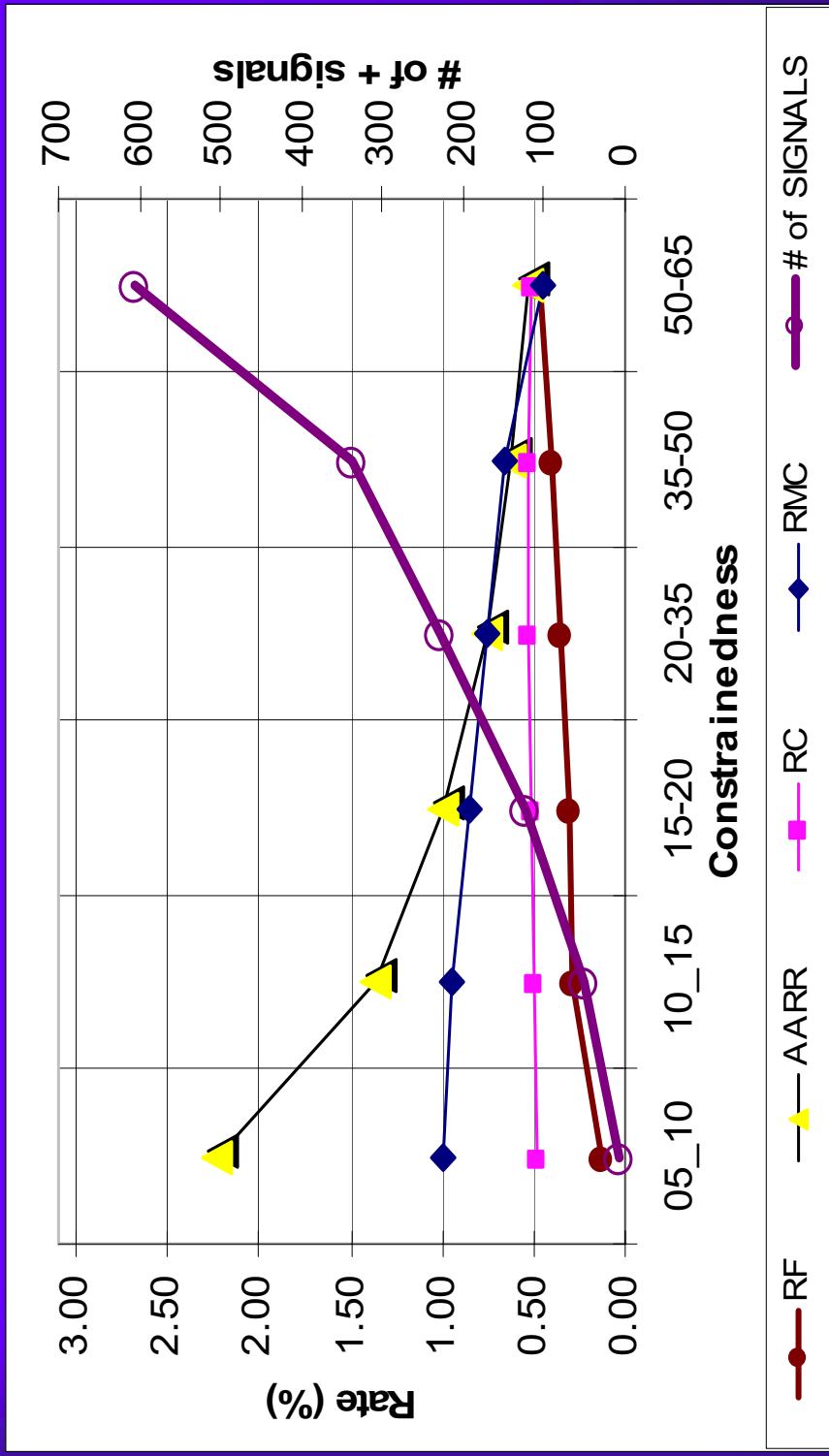
FGP-2 On DJIA Data

DJIA Index Closing Prices



- Training: 1,900 days 07/04/1969 to 11/10/1976
- Testing: 1,135 days 12/10/1976 to 09/04/1981
- Target: “rise of 4% within 63 days”

Effect of constraints on FGP-2



- ◆ Observation: RMC can be traded for RF without significantly affecting RC

Our FGP Experience

- ◆ Patterns exist
 - Would they repeat themselves in the future?
(EMH debated for decades)
- ◆ EDDIE has found patterns
 - Not in every series
(we don't need to invest in every index / share)
- ◆ EDDIE extending user's capability
 - and give its user an edge over investors of the same caliber

High Frequency Data: Example of an Order Book

	Price	Volume	Orders
Seller 4	3.86	2,000	1
Seller 3	3.85	10,000	5
Seller 2	3.84	5,000	1
Seller 1	3.83	1,000	1
Buyer 1	3.82	6,000	3
Buyer 2	3.81	8,000	3
Buyer 3	3.80	5,000	1
Buyer 4	3.79	17,000	3

EDDIE in Arbitrage, Historical Note



1995: EDDIE

(Evolutionary Dynamic Data Investment Evaluator)

James Butler Edward Tsang



1996: FGP

(Financial Genetic Programming)

Jin Li



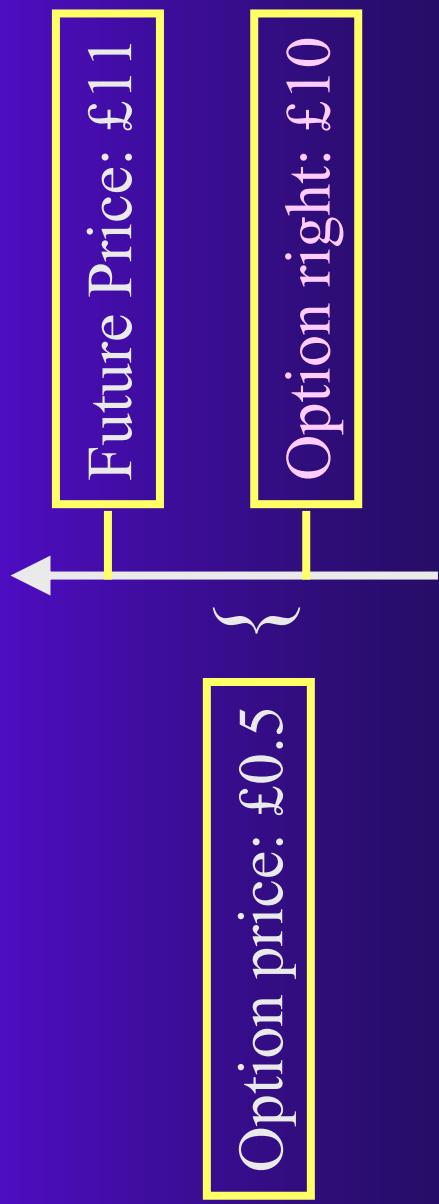
2000:
FGP+Arbitrage

Arbitrage Research

Sheri Markose Hakan Er

Arbitrage Opportunities

- ◆ Futures are obligations to buy or sell at certain prices
- ◆ Options are rights to buy at a certain price
- ◆ If they are not aligned, one can make risk-free profits
 - Such opportunities should not exist
 - But they do in London



EDDIE in Arbitrage

Typical arbitrage result

- ◆ Arbitrage opportunities found
 - They shouldn't exist?
 - They exist for 12-45 seconds
- ◆ EDDIE to predict arbitrage
 - 15 minutes in advance
 - Find clear opportunities only
- ◆ Human expertise needed
 - 9 months data preprocessing
 - Over 10 data set revisions

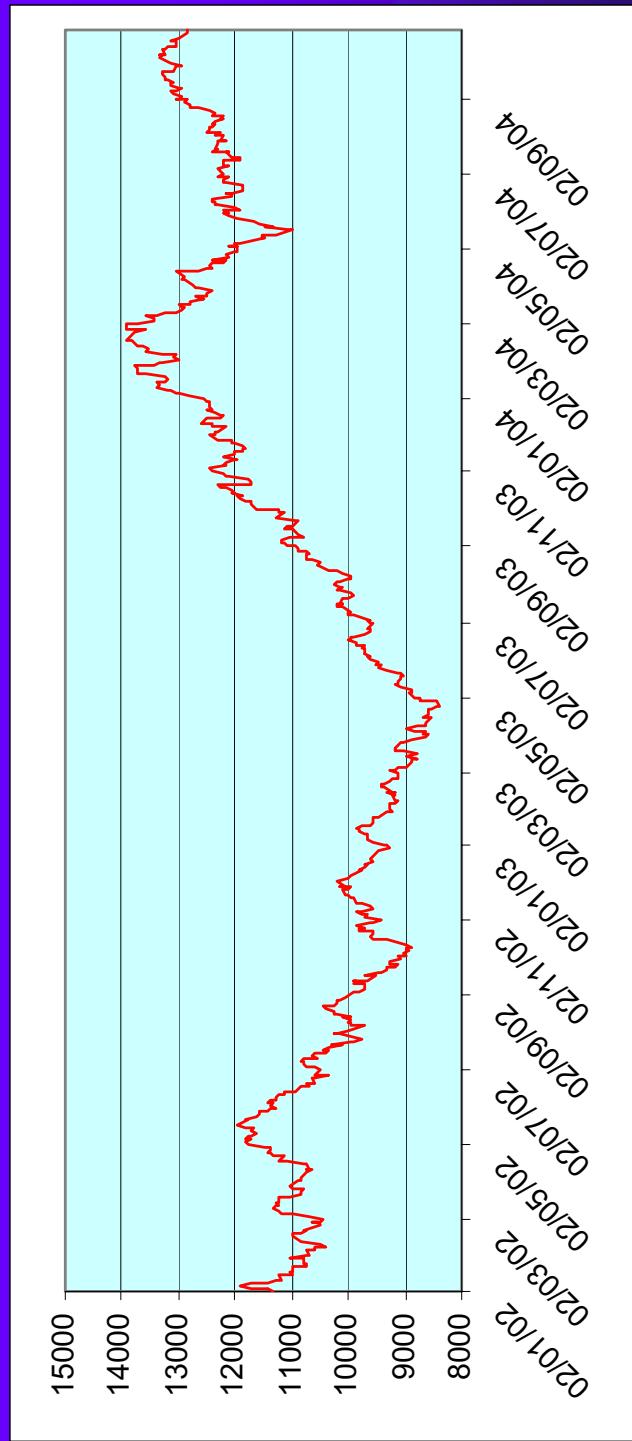
Positive

Negative

4,900	0	96	4
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Profitable arbitrage opportunities are rare;
Can't afford to miss too many

EDDIE Artificial Market



EDDIE as Intelligent Agent

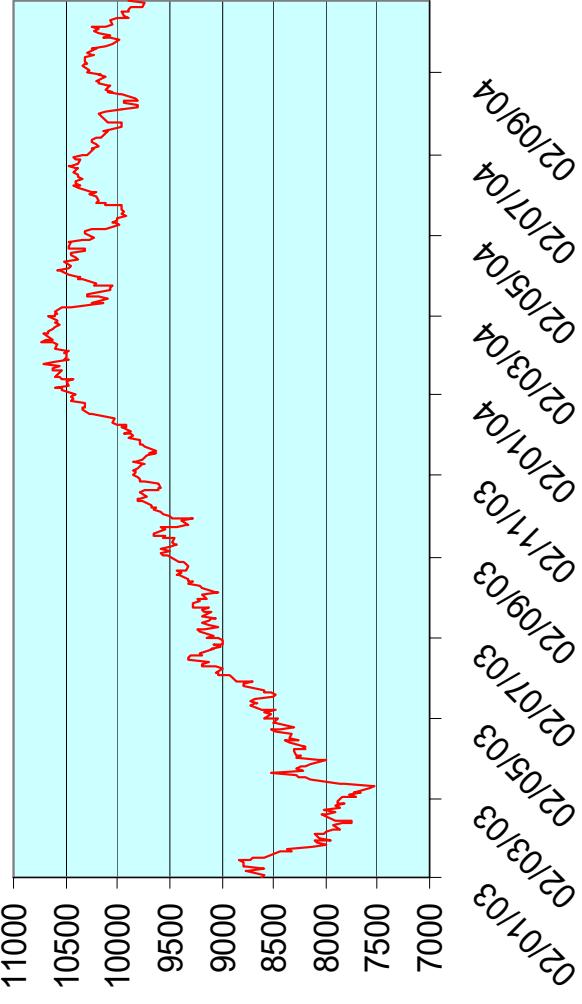
- ◆ Let artificial agents for a market

- Technical traders (EDDIE)
 - Fundamental traders (Economists)
 - Noise traders
- ◆ How would the prices look like?
 - ◆ Under what conditions will they produce real market stylus?



Serafin Martinez
EDDIE Red Q

Dow Jones Industrial Index



The Red Queen Effect

- ◆ Power Law wealth distribution

- The weakest must re-train themselves

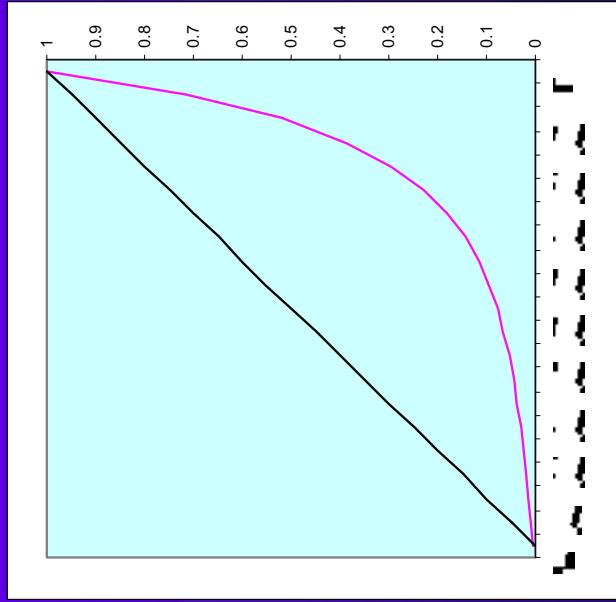
- ◆ Red queen effect

- You have to run as fast as you can to stay in the same place

- ◆ EDDIE is used for re-training



Sheri Markose Serafin Martinez
Red Queen EDDIE Red Q



EDDIE in Business

From research to practice:
Surfing one step ahead of each wave

What can EDDIE do for you?

- ◆ If it changes 50-50 chances to 55-45
 - in your favour
 - you must be better off in the long run...
- ◆ It helps you to beat other investors of the same calibre
 - It provides an extra expert opinion
 - If all your experts give you the same opinion, you have better chance to succeed
- ◆ It works day and night, you can't...

EDDIE/FGGP is no magic

- ◆ A tool is useful when...
 - it can do something good, and
 - we know how to use it, and
 - we know its limitations
- ◆ EDDIE / FGP is such a tool
 - No expert input, no useful forecast
(It only adds value to expert input)
 - It can only find patterns that exist
(No point asking it to predict the lottery)

Don't expect to see...

- ◆ Miracles –
 - we can't predict the unpredictable!
- ◆ Prediction of everything
 - May not find patterns for the future
 - E.g. patterns found in IBM/BAA, but not HSBC
 - So no positive actions recommended
- ◆ Fancy interface
 - At the moment, we concentrate to make EDDIE *predict accurately*

Current Research

- ◆ EDDIE for Arbitrage
 - Spot, option and future prices don't always synchronize
 - Hence one can make risk-free return?
- ◆ EDDIE for Forecasting
 - When to sell? How to combine signals?
 - What is the return in reality?
- ◆ GP for modelling volatility
 - coefficients fitting for GARCH-like functions
 - Discovering new functions forms?
- ◆ GP for market understanding
 - Learning agents form artificial market

The Computational Finance Research Team



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Early Tools



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FGP



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29 November 2004

Biliana Alexandrova-
Kabadjova
Small world

Alma Garcia
Forecasting

Questions, Discussion