# **GP** Basics

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Computational Intelligence in Economics & Finance Monday, August 16, 2004 Summer Workshop

> Sponsored by Taiwan's National Science Counsel & AI-Econ Research Center

**Focus: Selection of Run Parameters** Population vs. generation size Mutation vs. crossover Choosing fitness function Training vs. forecast fitness Predictability of dependent variable Financial market Applications Stock returns Stock prices Exchange rate

### **Population vs. Generation Size**

Basically there is no agreement in the literature.

- This is a research opportunity that will gain immediate attention.
- Suggested Strategy:
  - Select 5-10 series with known or measurable levels of complexity.
  - Run 100 searches twice to obtain best equation for each. Use 1000/100 for population/generation sizes once and 100/1000 for population/generation sizes another.
  - Select the fittest 10 from each run and compute the mean fitness for all selected outcomes.
  - Test the hypothesis that mean differences between those means is zero.

#### Population vs. Generation Size (Cont'd)

- Select i series, for i = 1, ..., n.
- Obtain lowest j MSEs, for j = 1, ..., k.
- Compute mean MSE for the i<sup>th</sup> series:

$$\overline{\mathsf{MSE}}_i = \frac{1}{k} \left( \sum_{j=1}^k \mathsf{MSE}_j \right)$$

Test H<sub>o</sub>:  $\overline{MSE}_{d} = \frac{1}{n} \sum_{i=1}^{n} (\overline{MSE}_{100/1000} - \overline{MSE}_{100/1000})_{i} = 0$ The test statistic is:

$$t = \frac{MSE_d}{\frac{s_d}{\sqrt{n}}}$$

#### Mutation vs. Crossover

- GP researchers favor crossover while EP researchers favor mutation.
- Strategy to test the efficacy of either may depend on the relative complexity of a series. I would guess that more complex series may demand higher % of mutation.
- This is another area of research that demands immediate attention and can be conducted in a manner similar to the one described for determining population and generation sizes.

# **Choosing Fitness Function**

There are several fitness functions to choose from.
MSE
MAPE
Conditional MSE

$$P\left(\sum_{t=(T+1)}^{(T+K+L)} (x_{t} - \hat{x}_{t})^{2} | \min_{t=1}^{T} (x_{t} - \hat{x}_{t})^{2} \right) \geq P\left(\sum_{t=(T+1)}^{(T+K+L)} (x_{t} - \hat{x}_{t})^{2} | \min_{t=T+1}^{(T+K)} (x_{t} - \hat{x}_{t})^{2} \& \text{reasonable} \sum_{t=1}^{T} (x_{t} - \hat{x}_{t})^{2} \right)$$

Conditional MAPE

$$P\left(\sum_{t=(T+1)}^{(T+K+L)} (x_t - \hat{x}_t)^2 | \min_{t=1}^T (x_t - \hat{x}_t)^2 \right) \ge P\left(\sum_{t=(T+1)}^{(T+K+L)} (x_t - \hat{x}_t)^2 | \min_{t=T+1}^{(T+K)} | x_t - \hat{x}_t | \& reasonable \sum_{t=1}^T (x_t - \hat{x}_t)^2 \right)$$

## Training vs. Forecast Fitness

- Selection of fitness function depends on the nature of the values of the dependent variable and the objective.
  - Selecting a function for cyclical data with large spreads between highs and lows can be a challenge. MSE will provide accurate predictions of larger values. MAPE will provide more accurate predictions of smaller values.
  - Conditional functions are useful when recent dynamics are believed to be more dominant than earlier dynamics.

Conditional				Uncondit	ional
	MAPE	MSE		MAPE	MSE
R 38 U	19.5	583.7	R 1 U	19.7	530.5
R 22 U	21.6	541.2	R 35 U	24.6	948.9
R 37 U	21.8	608.4	R 91 U	26.4	1082.8
R 42 U	23.8	565.6	R 79 U	28.2	1324.0
R 62 U	23.8	628.5	R 53 U	28.3	1890.7
R 8 U	24.0	605.5	R 26 U	29.5	1207.4
R 1 U	24.1	567.7	R 22 U	32.1	1003.1
R 44 U	24.6	701.9	R 54 U	33.7	1375.7
R 60U	25.5	653.1	R 83 U	34.9	1152.8
R 31 U	35.0	1987.7	R 13 U	36.8	1218.2

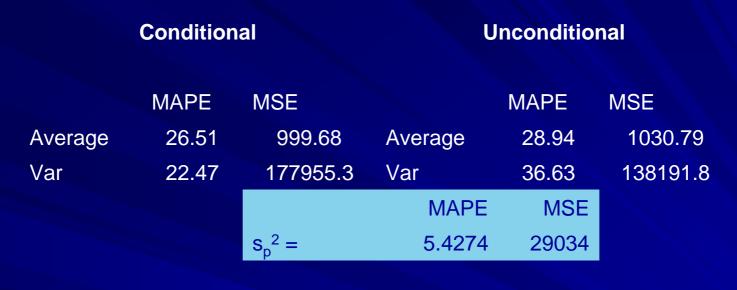
R 47 C	22.6	623.8	R 8 U	22.7	515.8
R 55 C	29.1	1553.2	R 88 U	25.4	608.5
R 70 C	29.9	932.7	R 15 U	28.3	784.8
R 41 C	31.1	985.1	R 2 U	32.0	922.2
R 15 C	31.2	1070.7	R 76 U	32.3	863.7
R 96 C	31.7	1111.2	R 24 U	34.7	979.4
R 50 C	33.1	1848.7	R 16 U	35.2	1098.6
R 66 C	35.0	1793.2	R 1 U	41.1	1428.6
R 63 C	35.7	1104.5	R 58 U	41.2	1741.1
R 44 C	36.6	1244.2	R 78 U	46.9	1872.0

R 94 C	21.2	614.0	R 50 U	20.9	549.0
R 33 C	21.6	652.3	R 52 U	21.0	565.3
R 7 C	22.0	1289.9	R 34 U	21.1	605.0
R 20 C	22.2	675.3	R 68 U	21.2	616.0
R 45 C	23.3	1003.4	R 47 U	23.9	816.2
R 74 C	24.2	1090.9	R 53 U	24.1	735.3
R 8 C	26.4	1667.0	R 60 U	24.9	1187.4
R 22 C	26.9	977.3	R 20 U	28.7	989.0
R 56 C	28.8	1177.2	R 90 U	29.3	1017.9
R 80 C	33.9	2169.9	R 10 U	36.2	2049.2

R 12 C	20.9	891.0	R 96 U	24.5	814.9
R 52 C	21.1	749.7	R 10 U	24.5	909.5
R 77 C	21.5	644.7	R 31 U	27.1	873.5
R 51 C	23.2	676.1	R 39 U	27.3	1122.7
R 8 C	23.9	741.6	R 17 U	28.6	992.2
R 85 C	25.6	1058.3	R 6 U	28.9	1172.4
R 23 C	29.7	1344.1	R 78 U	29.5	1235.9
R 11 C	30.5	1322.9	R 48 U	30.5	1143.8
R 87 C	30.8	1616.6	R 74 U	30.8	1525.3
R 33 C	33.0	1661.8	R 71 U	32.2	1365.9

R 41 C	20.2	631.4	R 34 U	22.7	539.4
R 22 C	20.3	637.1	R 71 U	22.9	653.6
R 85 C	21.9	610.9	R 66 U	22.9	695.4
R 14 C	24.5	608.1	R 30 U	23.2	653.8
R 82 C	25.0	708.9	R 84 U	24.5	649.4
R 13 C	26.3	878.4	R 69 U	25.8	804.8
R 86 C	28.4	944.2	R 53 U	28.7	1207.2
R 26 C	28.8	1024.6	R 39 U	31.0	933.9
R 42 C	29.4	1025.0	R 16 U	34.9	1152.8
R 94 C	30.1	1180.9	R 57 U	41.0	1414.1

#### **Comparative Analysis**



Ho: Ave. MAPE for Conditional = Ave. MAPE for Unconditional

MAPE	t =	-2.33	(72%)
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Ho: Ave. MSE for Conditional = Ave. MSE for Unconditional

MSE	t =	-0.408	(44%)
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#### Recommendation

- Use the conditional MSE fitness.
- Select the fittest 10.
- Plot actual, fitted, ex post forecast, and ex ante forecast.
- Discard of any illogical forecast.

Select that equation that seems to produce the best *ex post* forecast with the most logical *ex ante* forecast.

#### **Predictability of Dependent Variable**

The predictability η-metric measures percentage predictable information in a variable Y<sub>t</sub>. The η-metric applies to time series in general.

$$\eta = \max\left\{0, \frac{1}{n}\sum_{i=1}^{n} \left(1 - \frac{SSE_{\gamma}}{SSE_{S}}\right)_{n}\right\}$$

 $(S = Y_t \text{ with order of its observations randomly shuffled.})$ 

The metric can be used to conduct one of two hypothesis tests:

■ Ho: Y<sub>t</sub> is GP-predictable,

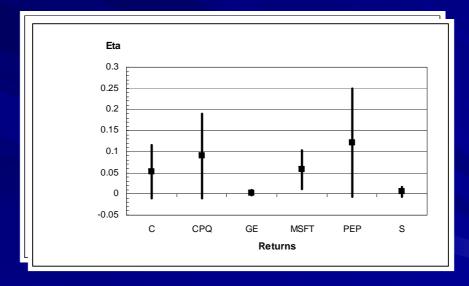
Ha:  $Y_t$  is not GP-predictable.

Ho: Y1<sub>t</sub> and Y2<sub>t</sub> are equally GP-predictable, Ha: Y1<sub>t</sub> and Y2<sub>t</sub> are not equally GP-predictable.

# **Application: Stock Returns**

Predictabi	lity of Stock					
	<b>30-m</b>	inute:	1-mi	nute:	PC	Rs:
Stocks	R²	η (%)	R²	η (%)	R²	η (%)
ВА	0.07	0.00	0.12	0.00	0.32	38.59
GE	0.03	0.00	0.10	1.07	0.59	72.44
GM	0.49	0.00	0.13	10.76	0.54	58.32
IBM	0.46	0.00	0.21	0.00	0.28	30.75
S	0.01	2.15	0.11	6.32	0.42	70.70
Т	0.08	0.00	0.45	50.10	0.59	41.48
WMT	0.07	13.74	0.30	0.00	0.83	91.46
XON	0.02	7.01	0.11	0.00	0.56	47.67

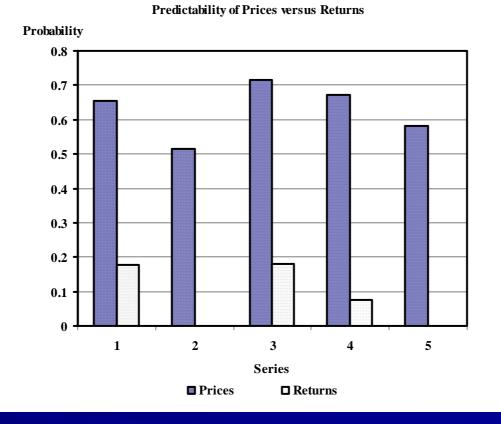
## Price Levels vs. Returns

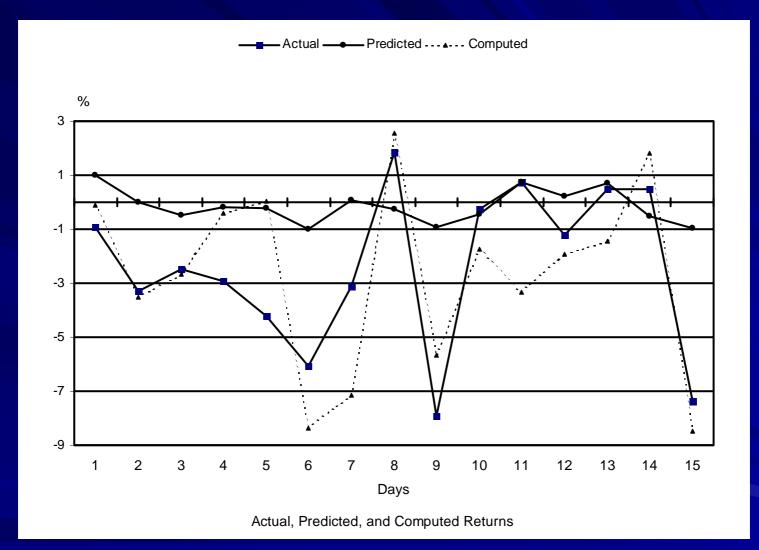


# **Application: Stock Prices**

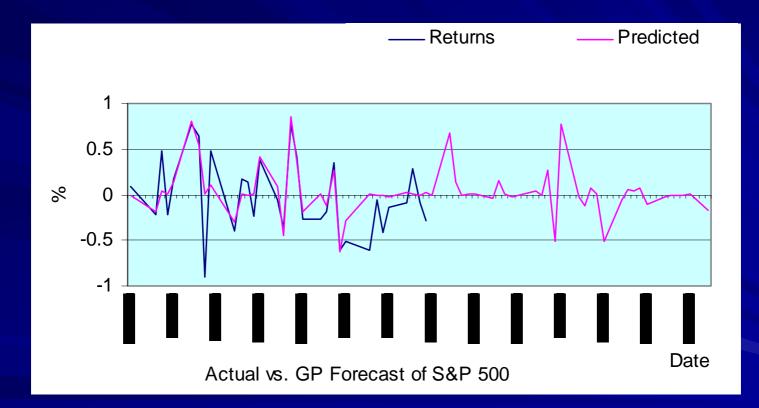
Testing mean difference between predicted and computed predicted prices

			MSE		
	n	t	Predicted	Computed	
С	100	-2.400	2.327	2.873	
CPQ	100	-2.895	0.831	1.405	
GE	100	-4.341	3.162	4.745	
MSFT	100	-4.097	1.636	2.158	
PEP	100	-2.611	0.987	1.788	
S	100	-2.695	1.910	2.707	
All	600	-6.374	1.809	2.613	

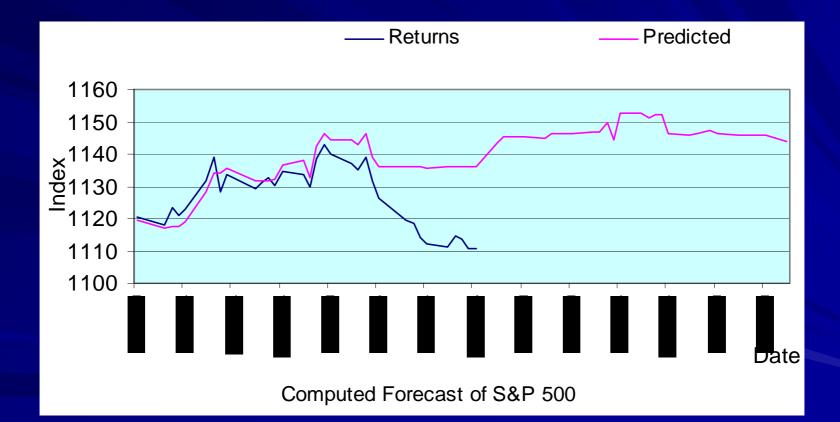




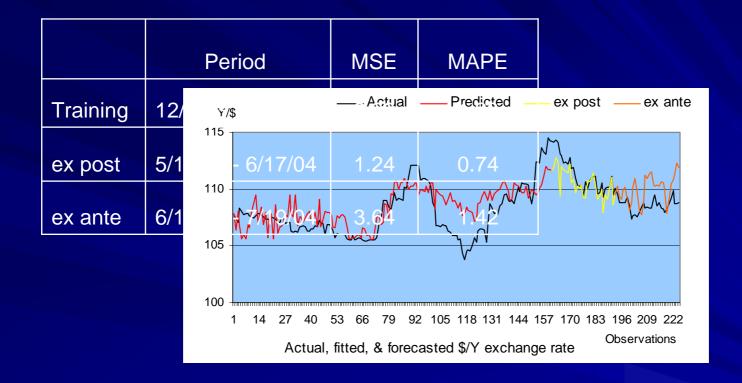
# S&P 500



## S&P 500 Index Forecast



#### Application: Yen/\$ Exchange Rate Forecast



# Summary & Conclusion

- Selecting between larger population and larger number of generations is unclear. It may depend on software written as well as complexity of series investigated.
- Favoring mutation over crossover seems logical although more research is needed.
- MSE seems to be the most logical fitness function to use. Using conditional MSE may help obtain better forecasts, but this also remains unclear and more research is needed to confirm its superiority.
- Using minimum lag length (MLL) models provides more useful for decision making forecasts. They may improve accuracy of forecasting of nonlinear systems.
- It may be easier to predict natural nonlinear phenomena than it is to predict outcomes of multi-agent decision where the agents are humans.

# Next Topic

#### Wavelets