

***“Improvement of wave height forecast
in deep and intermediate waters
with the use of stochastic methods”***

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Presentation Structure

- 1. Introduction**
- 2. Methodology**
- 3. Applications**
- 4. Conclusions**



Introduction



Introduction

OBJECTIVE: Improvement of the forecast power of numerical wave models using real time measurements

Improvement of wave predictions can be achieved by three different approaches:

- **Data assimilation techniques** (require re-run and space coverage)
- **Stochastic tools**
- **Neural networks techniques**

✚ **Only a few stochastic applications concerning real time improvement of the wave forecast**

✚ **Caires & Sterl, 2004: non parametric correction of ERA-40 reanalysis data, using a learning dataset**

In this work a stochastic technique is applied to measured data of wave height for improving forecasts of the wave model

Methodology



Bivariate Linear Regression models

$$\hat{Y}_{t+k\Delta t} = a_k X_{t+k\Delta t} + \beta_k Y_t + \gamma_k$$

\hat{Y} : Wave height estimate

Y : Wave height measurement

X : Wave height model forecast

- $k=0,1,2 \rightarrow \langle \text{model } 0,1,2 \rangle$
- *Trivariate linear model* $\rightarrow \langle \text{model } 3 \rangle$

Bivariate Non-Linear Regression models

$$Y^{\lambda}_{t+\Delta t} = \alpha_1 X^{\lambda}_{t+\Delta t} + \beta_1 Y^{\lambda}_t + \gamma_1, \quad \text{if } Y_t \leq c,$$

$$Y^{\lambda}_{t+\Delta t} = \alpha_1 X^{\lambda}_{t+\Delta t} + \beta_2 Y^{\lambda}_t + \gamma_2, \quad \text{if } Y_t \geq c$$

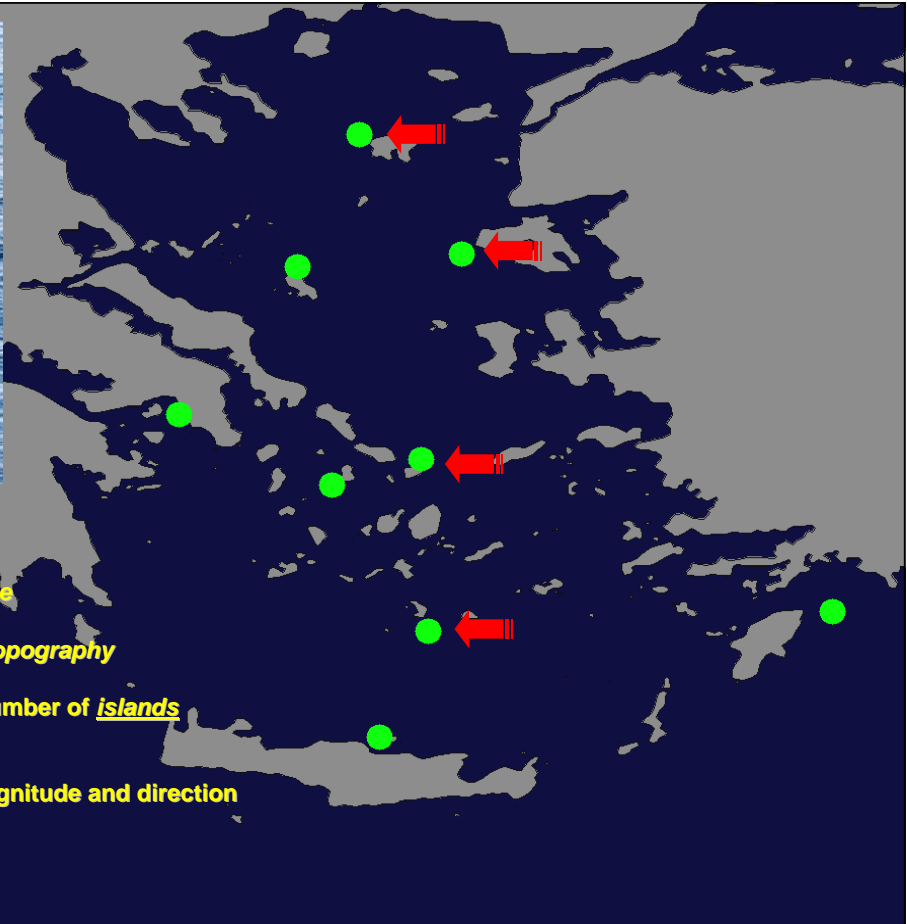
$$\gamma_2 = \gamma_1 + (\beta_1 - \beta_2) c^{\lambda} \quad (\text{continuity } c)$$

$\langle \text{model } 4 \rangle$

Applications



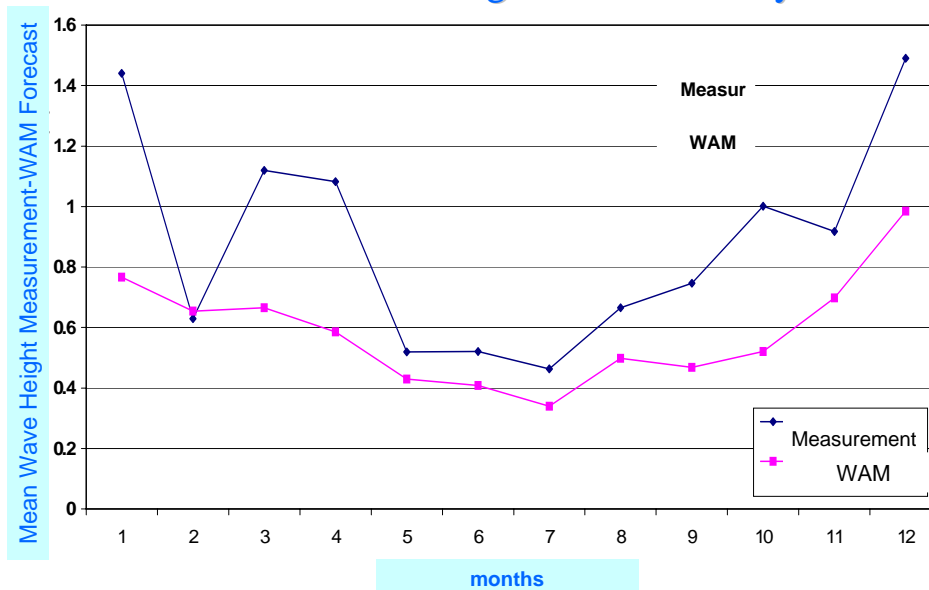
First Application: The Aegean Sea The Stations Used



- ✓very irregular coastline
- ✓complicated bottom topography
- ✓presence of a large number of *islands* scattered over the area
- ✓abrupt changes in magnitude and direction of the wind field

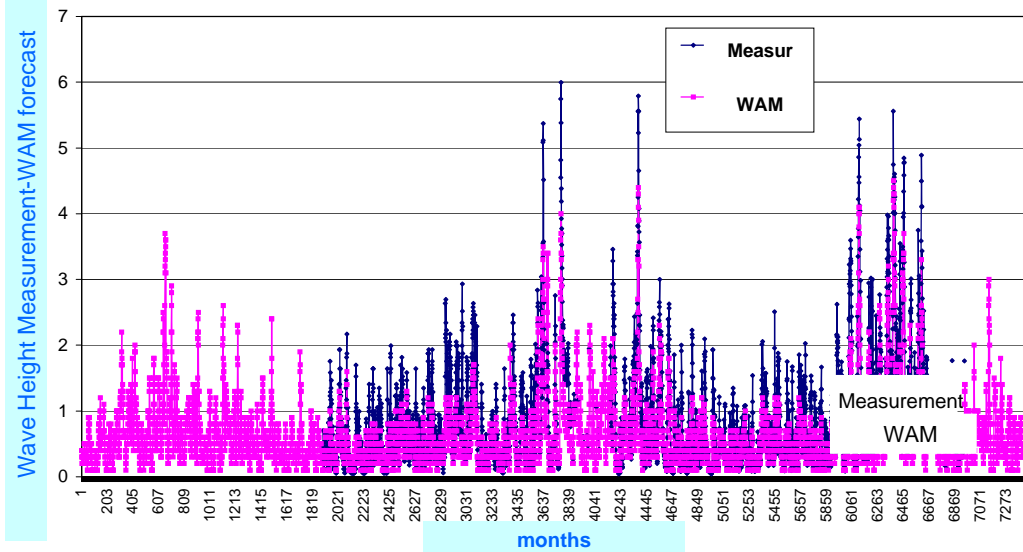
WAM Underestimation

Measurements through POSEIDON System



WAM Underestimation

Measurements through POSEIDON System

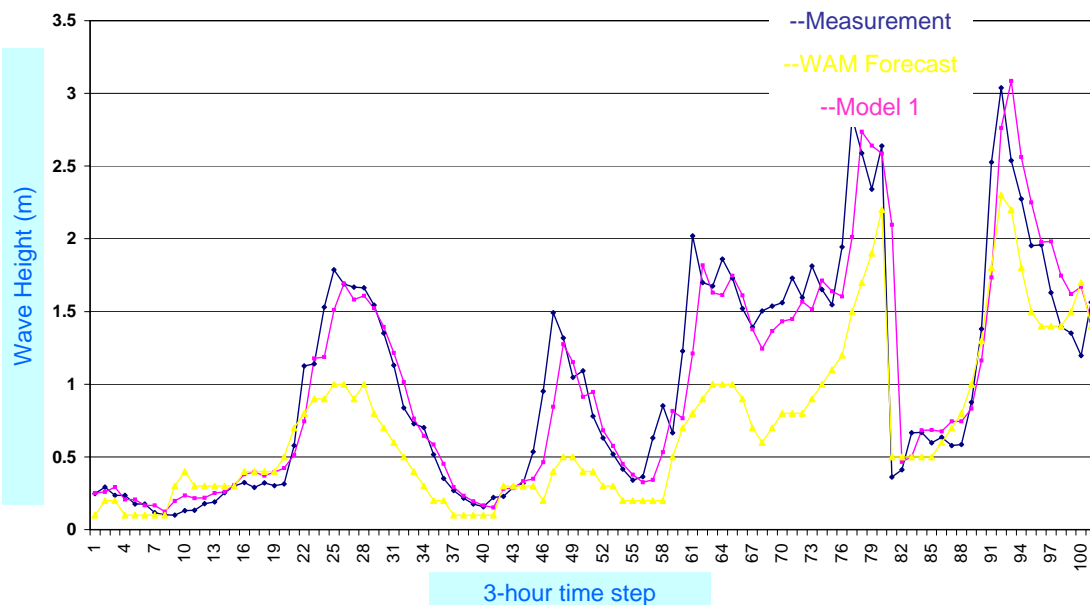


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ATHOS STATION

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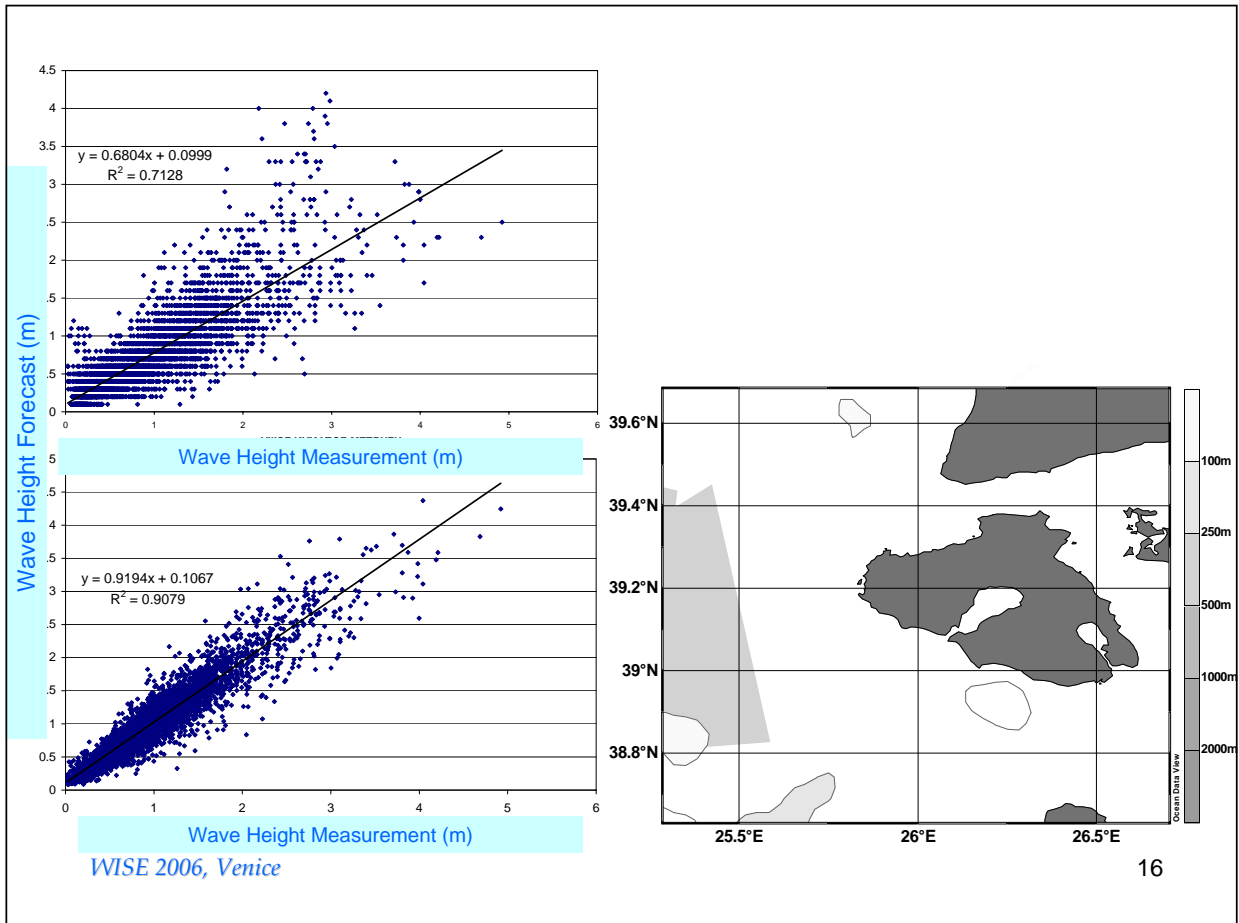
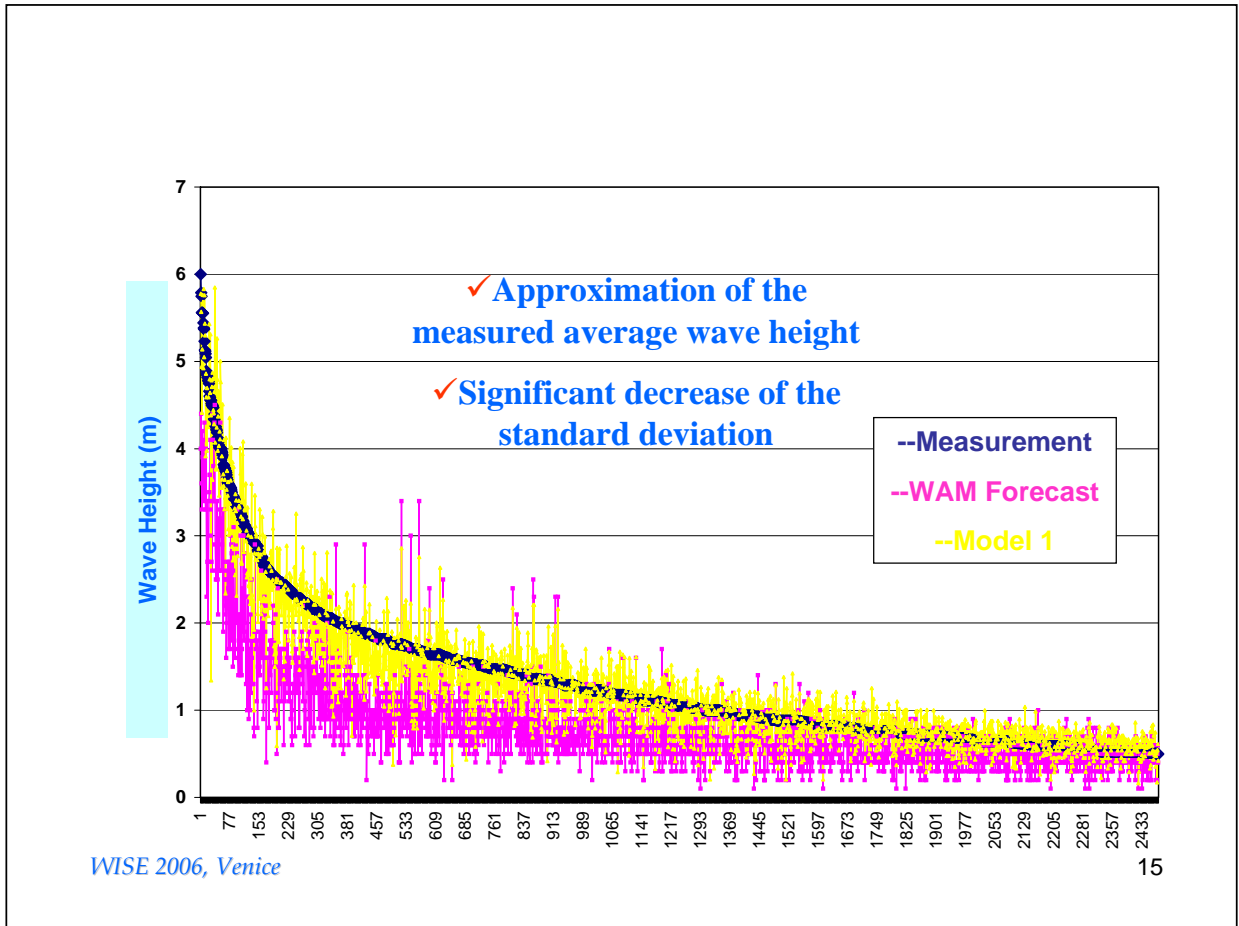
Model 1

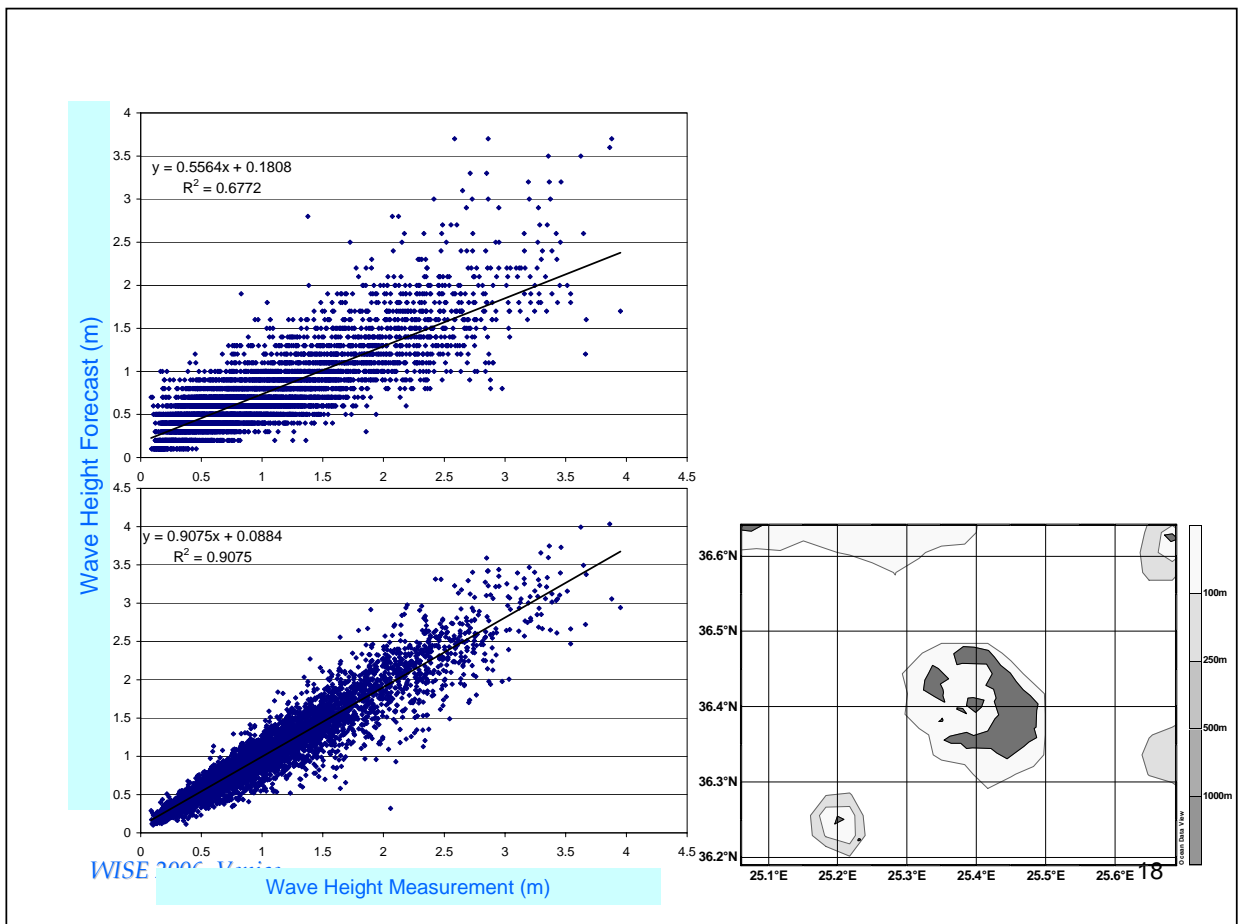
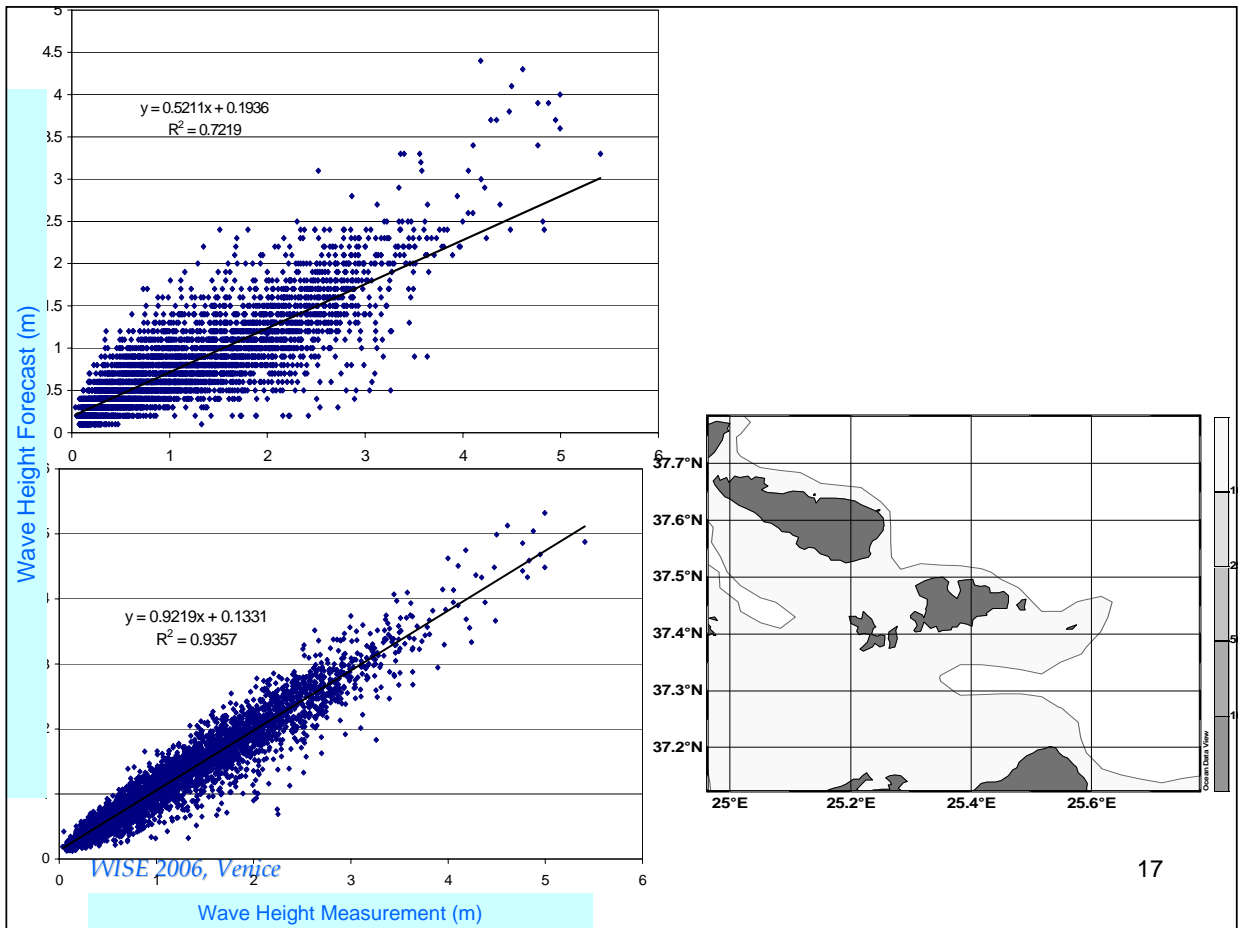


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Summary Table

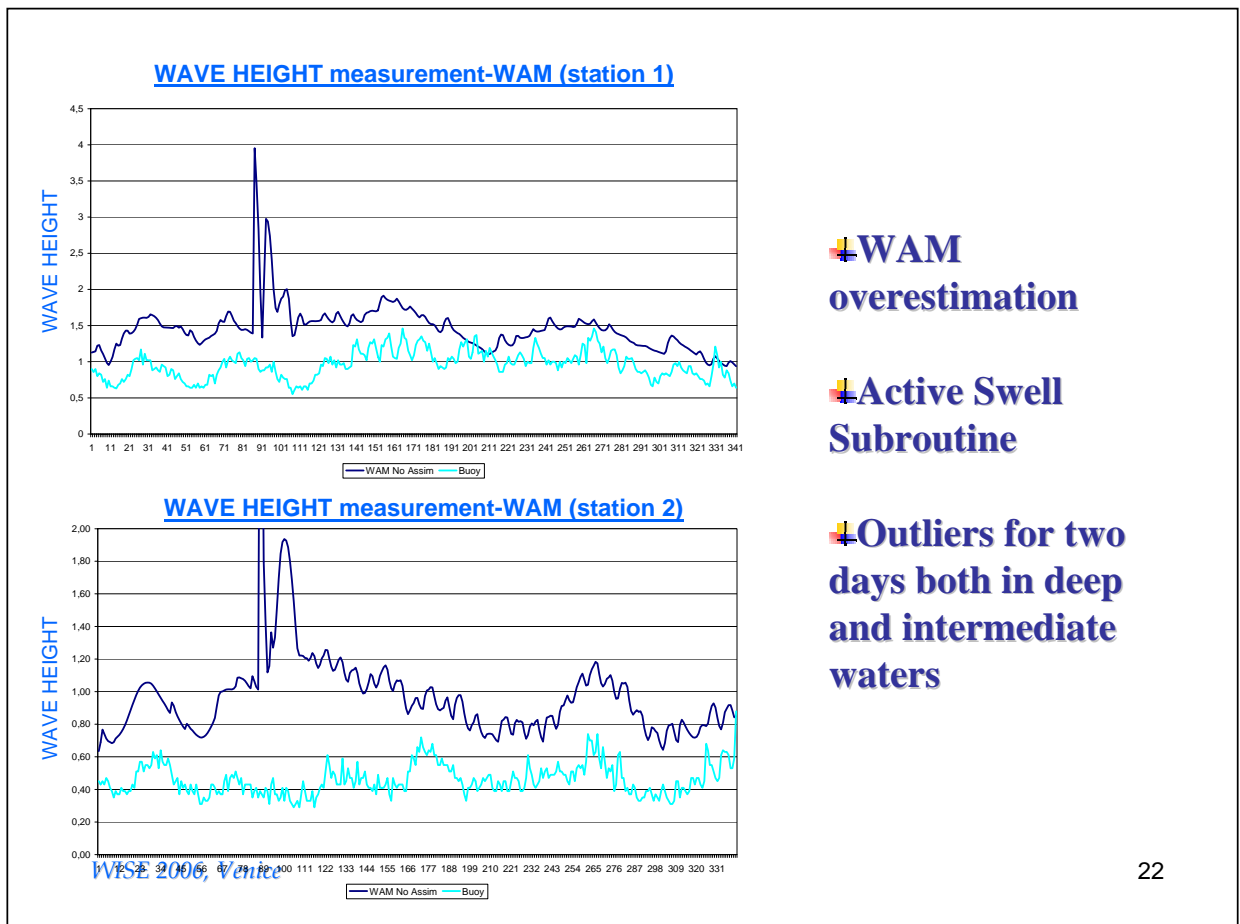
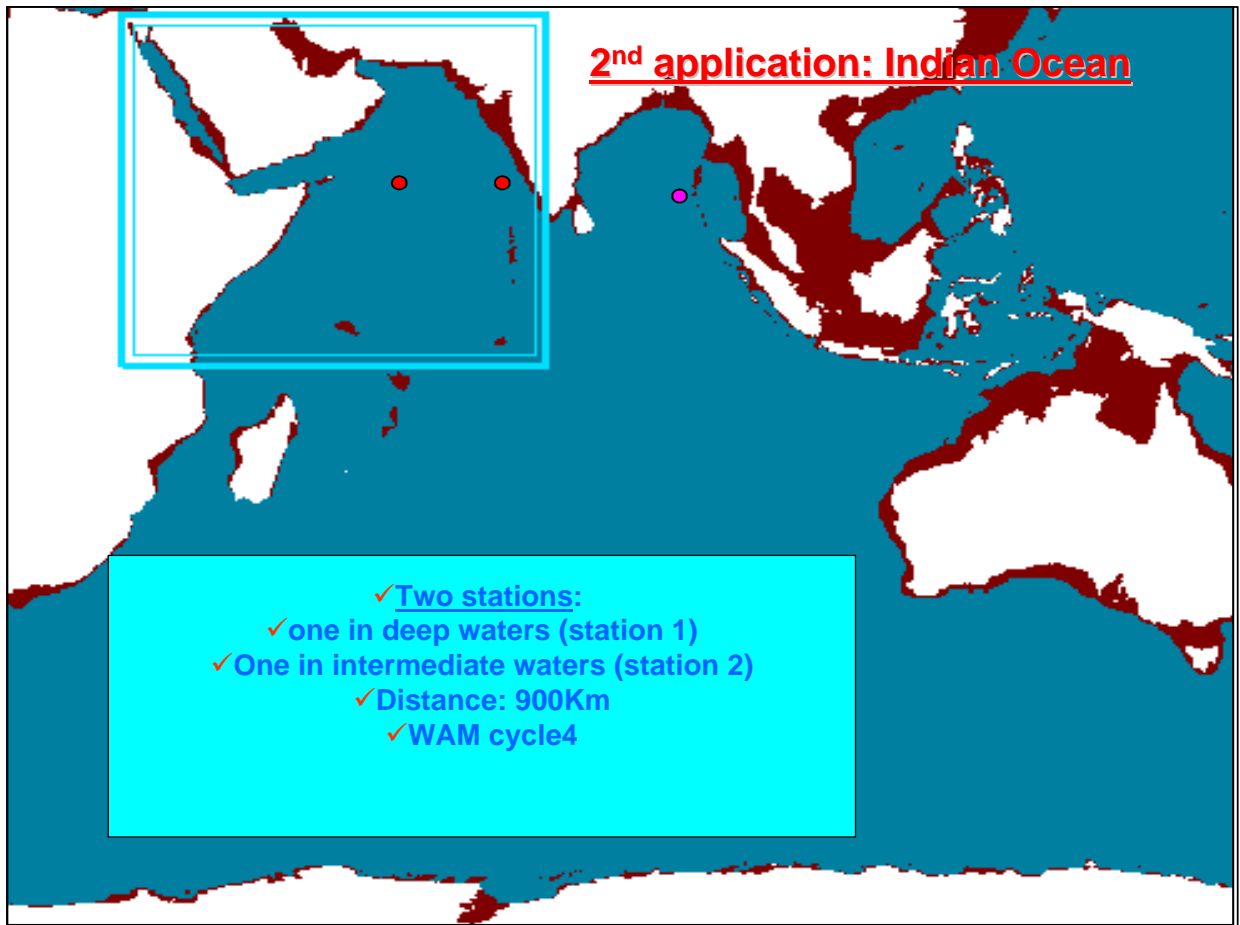
	R² Athos	R² Lesvos	R² Mykonos	R² Santorini
WAM MODEL	0.781	0.713	0.722	0.676
MODEL 1	0.920	0.892	0.927	0.906
MODEL 2	0.865	0.813	0.860	0.820
MODEL 3	0.924	0.895	0.931	0.911
MODEL 4a	0.929	0.897	0.936	0.909
MODEL 4b	0.929	0.898	0.935	0.908

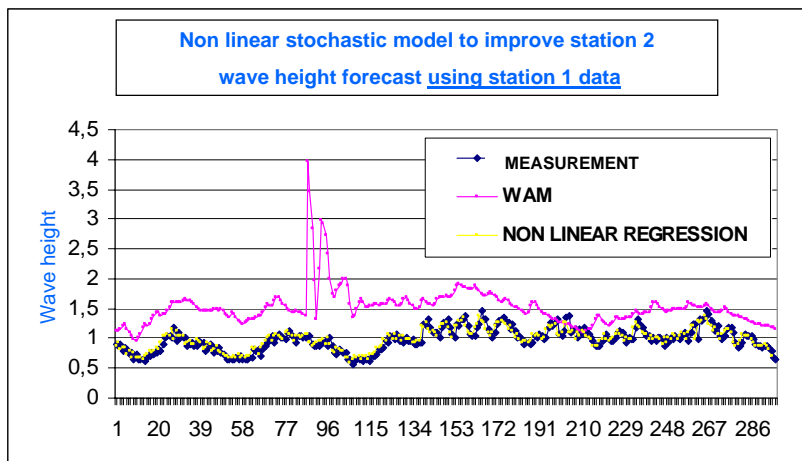
Total improvement obtained:

- ✚ For all four stations**
- ✚ For all four models tested**
- ✚ For all periods of implementation**
- ✚ For high and low values of the wave height**
- ✚ No outliers**

✓ Increase of the forecast time step decreases the predictive accuracy of our technique

✓ Our technique's R² equals WAM's for forecasts horizon more than 3 days





- ✚ **Remarkable improvement**
- ✚ **Coefficient of determination ($R^2=0.9$)**
- ✚ **Approximation of the measured average wave height**
- ✚ **Absorbing the outliers**
- ✚ **K time step selection**

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Conclusions



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- # All models improve significantly the forecast in applications tested
- # The parameters (weights) in all models indicate the same performance in all stations and in all periods of implementation
- # The relation between measurements and predictions seems to be the same (error source, systematic error)
- # If we have/measure the wave height for a small period of 3-4 months irrespectively of the season, we can:
 1. Estimate the variable weights and then
 2. Significantly improve in real time the forecast of the wave model without re-running it and within an area of considerable extent containing the point of measurement
- # Stochastic models are a useful tool for wave forecast. Future research is needed