



Methodology and Architecture for Using Oceanographic and Meteorological Ensembles to Improve Forecast Skill and Guidance in Support of Tactical Planning and Operations

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Outline

- **Motivation**
- **Background**
- **Systems**
- **Results**
- **Conclusions**
- **Future Work**



Motivation

- Decision makers and operators must make complex plans and decisions ad-hoc and often go into action on short notice
 - Must process a large amount of information quickly
 - Only able to use what is at hand and easily understood
- METOC ensembles can be an important source of information for the decision maker; however, they:
 - Provide ambiguous and redundant information
 - Are difficult to interpret and apply directly

The value that ensemble prediction brings to problem solving is easily overlooked if the information isn't in a form that makes it immediately understandable and actionable

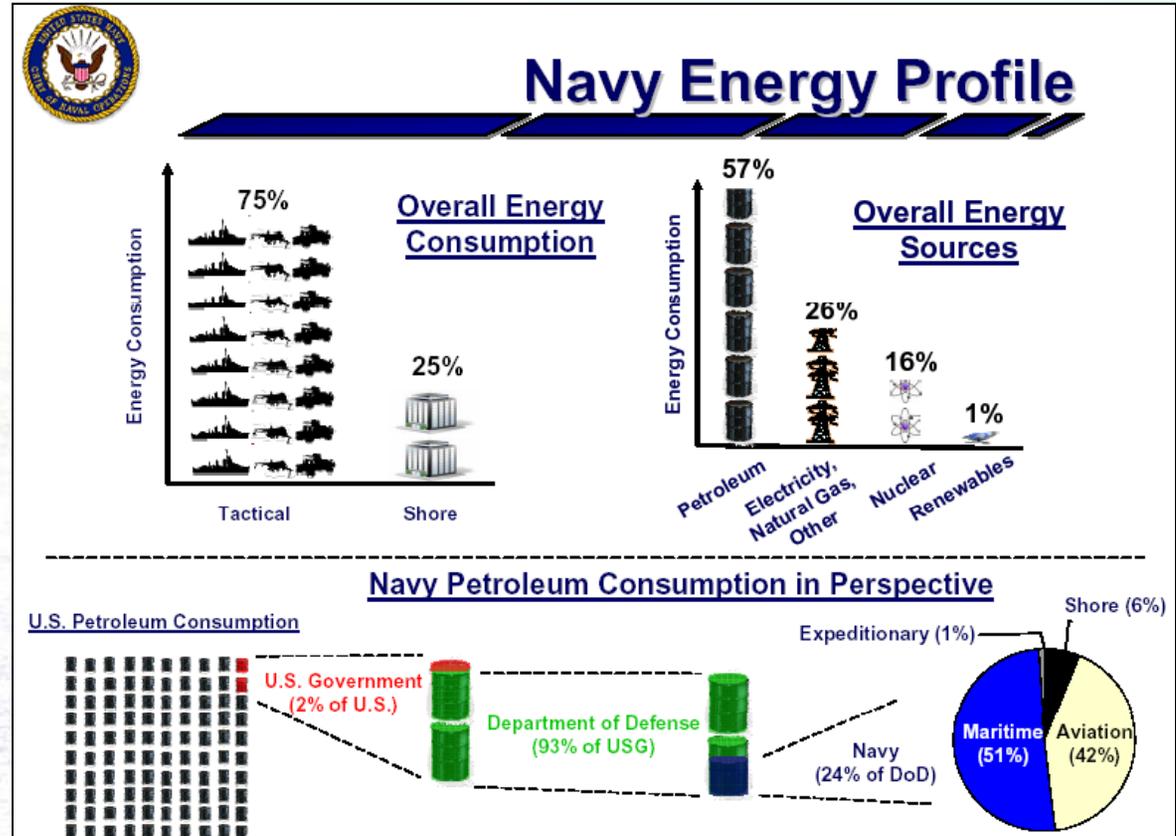
Goal: Derive a consensus forecast from ensemble data to improve an existing US Navy ship routing application



Need for Smart Ship Routing



- **FY2007 Defense Authorization Act Section 360** makes it Department of Defense (DOD) policy to improve the fuel efficiency of weapons platforms, consistent with mission requirements, and requires a report on DOD progress in implementing the policy



**Navy - FY08: ~35 M bbls
Transit Estimate ~11 M bbls**



Background



Results from previous work

- Simple bias correction worked well for most parameters to improve the ensemble mean
- At longer forecast times, gradients in the ensemble means are too weak
- Simple techniques for PDF spread adjustment did not add value
- For short duration forecasts (24 to 72h), training data sets of as little as 25 days are adequate for reducing model bias

Parameter	Tau	Best Ensemble Average Conditioning Method Compared to Analysis	Error Compared to Analysis		
			Ave Error		Improvement
			Analysis – Ensemble	Analysis – Ensemble Mean	
Sig Wave Hgt	06	Bias Correction	.222	.164	.058 M
	24	Bias Correction	.243	.174	.069 M
	48	Bias Correction	.313	.230	.083 M
SFC Wind Speed	06	Bias Correction	1.04	.64	.4 m/s
	24	Bias Correction	1.38	1.0	.38 m/s
	48	Bias Correction	1.68	1.25	.43 m/s
SFC Wind Direction	06	Raw Ensemble	30	23	7°
	24	Raw Ensemble	38	30	8°
	48	Raw Ensemble	47	38	9°



Consensus Forecasts for Smart Routing



Post Processed Ensemble Forecast Improvement			
Needed Parameters	24 Hr FCST	72 Hr FCST	144 Hr FCST
SWL_WAVE_DIR	BIAS_CORR	RAW	RAW
SWL_WAVE_HT	BIAS_CORR	BIAS_CORR	BIAS_CORR
SWL_WAVE_PER	BIAS_CORR	BIAS_CORR	BIAS_CORR
WIND_DIR	RAW	RAW	RAW
WIND_SPD	BIAS_CORR	BIAS_CORR	RAW
WIND_WAVE_DIR	BIAS_CORR	BIAS_CORR	BIAS_CORR
WIND_WAVE_HT	BIAS_CORR	BIAS_CORR	BIAS_CORR
WIND_WAVE_PER	BIAS_CORR	BIAS_CORR	RAW

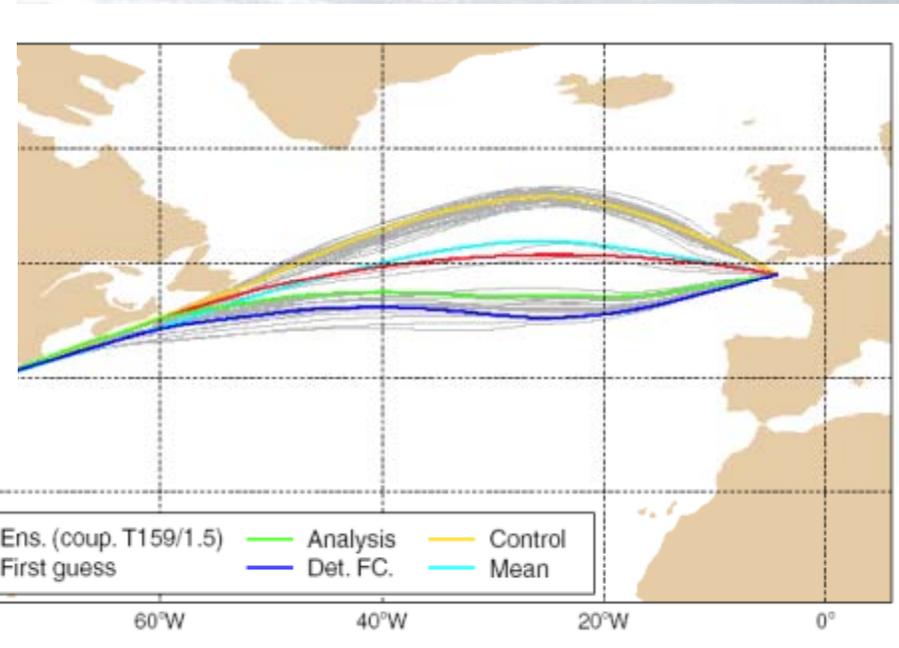
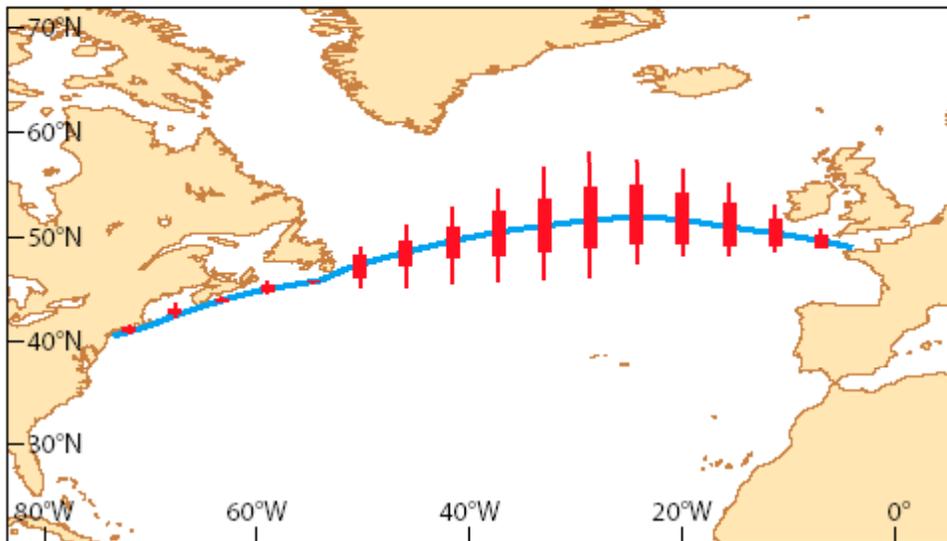
- **Developed Ensemble Forecast Application System (EFAS) software to provide ensemble post-processing of FNMOC operational fields**
- **Developed an interface for EFAS to Ocean Systems Inc. Ship Tracking and Routing System (STARS) algorithm used at FNMOC**
- **STARS also needs surface currents, SST, and climatology data**



STARS Ship Routing Engine



- Generates ship routes from voyage start/stop info, ship class, specified wind/sea limits (damage avoidance) for route and divert recommendations
- The Ship Route Engine finds the most efficient (least power) path, minimizing fuel burn, that also avoids land and bad weather
- Fuel burn along a specified route can also be computed



- Early approaches to supporting route decision making with ensemble information
- Which route to recommend?



EFAS Work Flow for Ship Routing



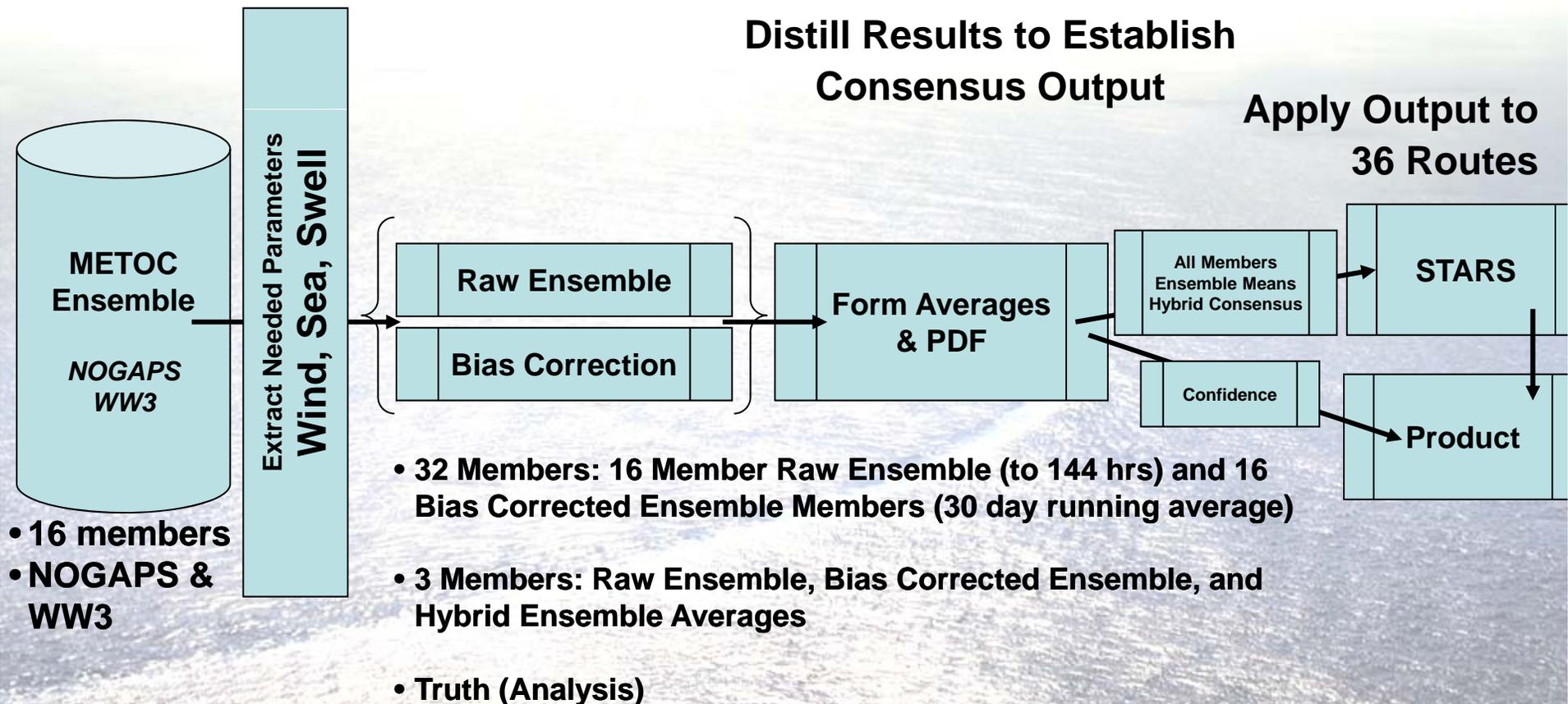
Ensemble Data

Determine Parameters Needed

Apply Calibration

Distill Results to Establish Consensus Output

Apply Output to 36 Routes





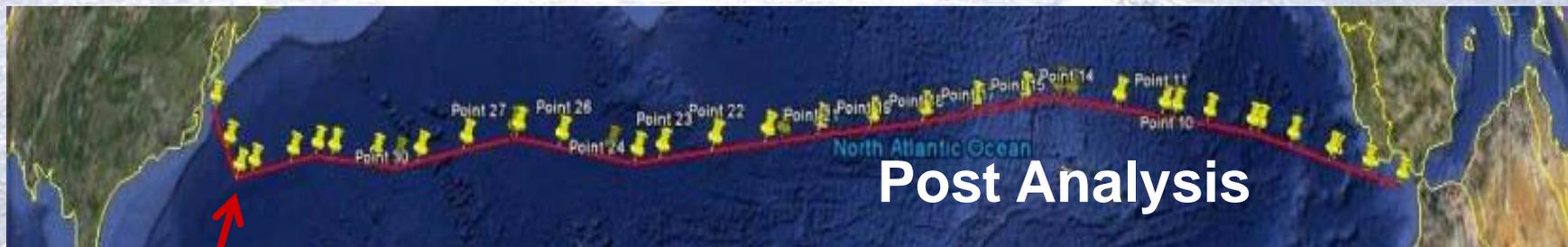
EFAS - STARS Route Recommendation



RAW Ensemble Horse Power
1532

BIAS CORR Horse Power
1530

HYBRID Horse Power
1530

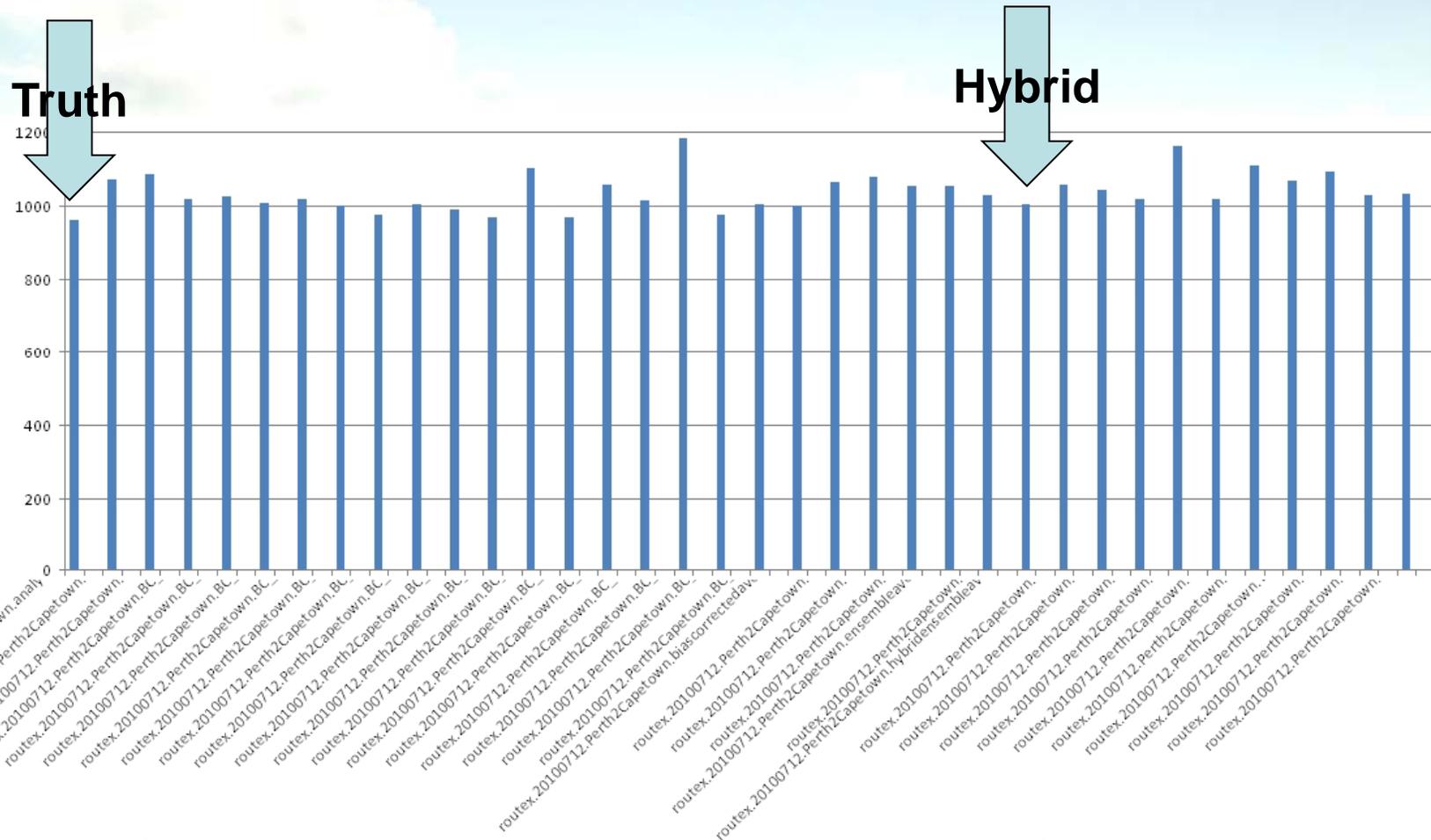


Impact of Climatology
Added distance/Horse Power

Post Analysis Route Horse Power
1570



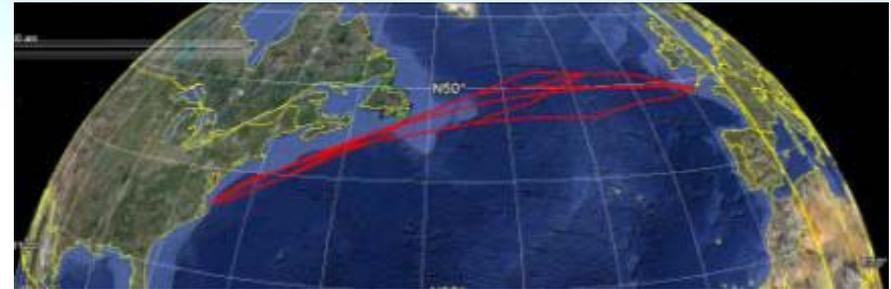
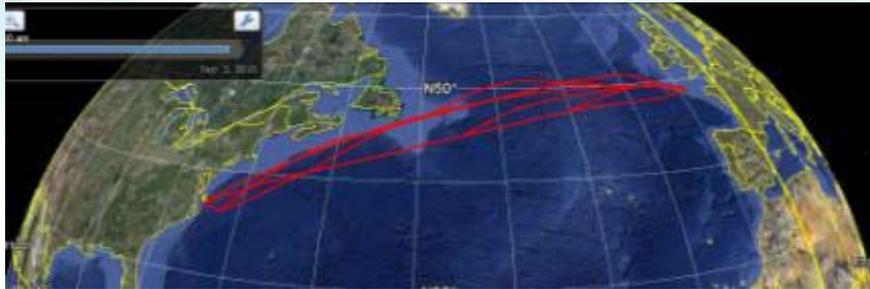
Horse Power Expended



- On average, the Hybrid route provides the HP expenditure closest to the analysis “truth”
- To score the routes, need to sail each route through the analysis “truth”



Smart Ship Routing using EFAS



Norfolk, VA to English Channel 28 Aug 2010

HP Average = 1130.2	Min HP = 1072.6
HP Truth = 1108.9	Max HP = 1305.8
Hybrid HP = 1136.8	Delta HP = 233.2
	% Variation = 20.6%
	Min Dist = 2943.5
Ave Distance = 2990.1	Max Dist = 3038.7
	Delta Dist = 95.2
	% Variation = 3%

Ensemble Sailing GC w/ no Optimizer

HP Average = 1089.8	Min HP = 1023.3
	Max HP = 1276.7
	Delta HP = 253.4
	% Variation = 23.3%

Distance = 2084

English Channel to Norfolk, VA 31 Aug 2010

HP Average = 1220.1	Min HP = 1105.1
HP Truth = 1252.5	Max HP = 1472.5
Hybrid HP = 1214.3	Delta HP = 367.4
	% Variation = 30.1%
	Min Dist = 2943.5
Ave Distance = 2969.6	Max Dist = 3002.0
	Delta Dist = 58.5
	% Variation = 2%

Ensemble Sailing GC w/ no Optimizer

HP Average = 1195.2	Min HP = 1100.6
	Max HP = 1372.9
	Delta HP = 272.3
	% Variation = 22.8%

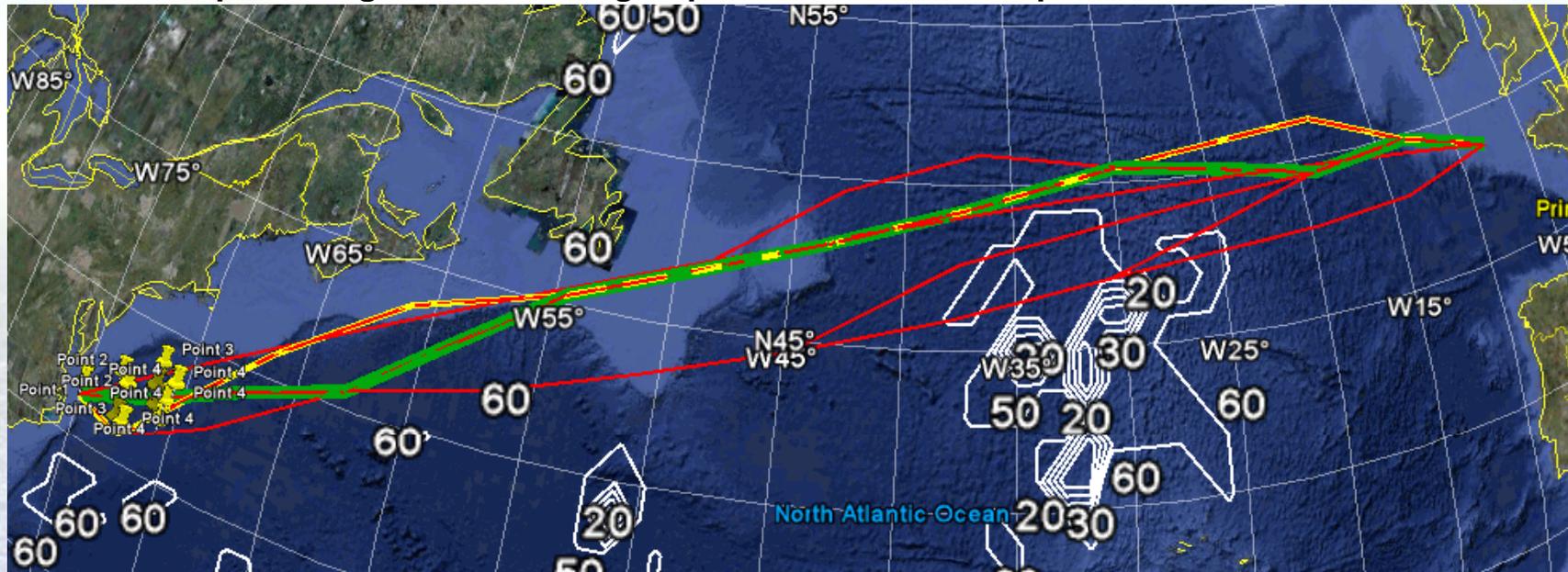
Distance = 2084



Norfolk, VA to English Channel Routes with Wind Confidence Chart



Confidence: percentage of ensemble grid points within a user-specified forecast error tolerance



HYBRID Ensemble

Arrival Time = 201009032123
Horse Power = 1136.8
Distance = 3018.6

Analysis FCST Route

Arrival Time = 201009032123
Horse Power = 1108.9
Distance = 2974.9



Value of Smart Ship Routing



1. EFAS – STARS may help reduce US Navy fuel expenditure

- In our limited tests (60 routes) the horse power expended varies between 8% to 30% across the ensembles
- Hybrid ensemble average shows improvement of up to 10% over a randomly selected ensemble member
- At \$70/bbl, each 1% improvement in routing saves about \$7.5 M/yr in fuel costs

2. Can the improved forecast be shown to avoid costs of damage from heavy weather, reducing yard time/repair & extending service life?

- ADM Harvey - Keeping ships in shape to make their 25-, or 30- or even 40-year planned operational life spans is fundamental to the Navy's three-decade fleet plan. And that plan is already aggressive: it will push many ships, including cruisers, destroyers and amphibious ships, as much as 10 or 15 years beyond their original design life.
- \$???? B can be avoided in Ship Repair
- \$???? B can be put into Ship Building



Summary



- Decision makers and operators require timely information and guidance for planning, improved safety and operational efficiency
- Ensembles can be represented into a deterministic-like Consensus Forecast that conveys:
 - A consensus value (presently we use an average of the post processed ensemble)
 - Confidence in the consensus within a user-defined tolerance of forecast error
- Regarding Post-Processing:
 - The model analysis is used for ensemble bias corrections and calibrations of the grid
 - In most cases, for our parameters, bias correction gives best results
 - 30 day training period sufficient to condition the ensemble
- The consensus forecast outperforms the deterministic forecast
- This provides a method to interface TDAs to METOC ensembles



Future Work

- **Compute the fuel usage for each route through the “truth” and evaluate the fuel penalties**
- **Develop a technique for deriving confidence in fuel usage calculations**
- **Extend our ensemble database from 144 to 240 hrs**
- **Introduce ocean currents and SST ensembles**
- **Evaluate mesoscale ensembles to improve near land transits**
- **Develop improved bias correction (testing Kalman filter) and spread adjustment methods**
- **Evaluate “Best Member” and “Most Likely Value” approaches**
- **Implement improved ship and power models used for optimization of the type being developed for Tempest (US Navy’s next-generation sea-keeping and dynamic stability software)**