

**Table 1. Summary of the NEFSC ecosystem models with notations of salient features.**

Model Class	ESAM MRMs- Ecology		Main Purpose of Model	Major Strengths	Major Weaknesses	Stochastic or deterministic outputs	Analytical or Statistical	Primary model intent- simulation, estimation, scenario testing, pattern detection, etc.	Static or Dynamic	Predictions or Forecasts (Y or N)	(Tactical, Strategic, Both, Heuristic, N/A) output use wrt Mgt Advice	Major Assumptions	Reference/s	Emphazized in this Review (Y or N)
<b>Model</b>	S-R	Augment SS Assessment Models with Ecological Interactions	Enhanced ecological realism; common outputs as SS models	Precision vs Accuracy Uncertainty Debate in SS Assessment Context; misses other processes and dynamics	Both	Both	Estimation	Both	Y	Tactical	A stock-recruitment relationship exists and can be fitted; predation or cannibalism is distinct from depensation/compensation	Tyrrell et al. 2011, Lucey & Alade unpubl. data	N	
<b>Model</b>	SS Prod	Augment SS Assessment Models with Ecological Interactions	Enhanced ecological realism; common outputs as SS models	Precision vs Accuracy Uncertainty Debate in SS Assessment Context; misses other processes and dynamics	Both	Both	Estimation	Both	Y	Tactical	Usual for production models, esp. some equilibrium assumptions; consumption is equal to or greater than fisheries removals, has a distinct pattern/trend than fisheries removals, helps scale magnitude of pop estimates	Tyrrell et al. 2011, Link & Idoine 2009, Moustahfid et al. 2009a, NEFSC 2007a, Overholtz and Link 2007, Overholtz et al. 1999	N	
<b>Model</b>	Age Structured	Augment SS Assessment Models with Ecological Interactions	Enhanced ecological realism; common outputs as SS models	Precision vs Accuracy Uncertainty Debate in SS Assessment Context; misses other processes and dynamics	Both	Both	Estimation	Both	Y	Tactical	Usual for age/stage structured models; consumption-at-age is estimable; consumption is equal to or greater than fisheries removals, has a distinct pattern/trend than fisheries removals, helps scale magnitude of pop estimates	Tyrrell et al. 2011, Overholtz et al. 2008, Moustahfid et al. 2009b, NEFSC 2011, NEFSC 2010	N	
<b>Model</b>	Ecological Footprints	Context for SS Assessment Models with Ecological Interactions- Usually consumption	Enhanced ecological realism, based on copious FH data	Mainly just contextual	Deterministic	Statistical	Estimation	Both	N	Heuristic	Consumption estimates are representative of the population; parameters are reasonable, esp. Beta about 0.11 based on experiments and sensitivity analyses	Tyrrell et al. 2007, NEFSC 20007b, Overholtz and Link 2007, NEFSC 2010b, DFO 2010	N	

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ESAM MRMs- Environmental													
<b>Model</b>	S-R	Augment SS Assessment Models with Environmental Considerations	Enhanced environmental realism; common outputs as SS models	Env-Fish relationships tend to decouple; Precision vs Accuracy Uncertainty Debate in SS Assessment Context; misses other processes and dynamics	Both	Both	Estimation	Both	Y	Tactical	A stock-recruitment relationship exists and can be fitted; environmental drivers are distinct from depensation/compensation; environmental covariates imply an understood mechanism	Hare et al. 2010, Keyl & Wolff 2008	N
<b>Model</b>	SS Prod	Augment SS Assessment Models with Environmental Considerations	Enhanced environmental realism; common outputs as SS models	Env-Fish relationships tend to decouple; Precision vs Accuracy Uncertainty Debate in SS Assessment Context; misses other processes and dynamics	Both	Both	Estimation	Both	Y	Tactical	Usual for production models, esp. some equilibrium assumptions; environmental covariates imply an understood mechanism; environmental covariates provide distinct pattern/trend in addition to fisheries	Keyl & Wolff 2008	N
<b>Model</b>	Age Structured	Augment SS Assessment Models with Environmental Considerations	Enhanced environmental realism; common outputs as SS models	Env-Fish relationships tend to decouple; Precision vs Accuracy Uncertainty Debate in SS Assessment Context; misses other processes and dynamics	Both	Both	Estimation	Both	Y	Tactical	Usual for age/stage structured models; environmental covariates imply an understood mechanism; environmental covariates provide distinct pattern/trend in addition to fisheries	Keyl & Wolff 2008	N

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Multispecies MRMs	Model	MS PROD	Simulate BRPs for multiple SS that include various interactions- Ecological	Enhanced ecological realism; common outputs as SS models	Precision vs Accuracy Uncertainty Debate in SS Assessment Context	Both	Analytical	Simulation	Dynamic	Y	Both	Prey switching is always possible: no feedback on predator abundance by prey; group and system carrying capacities can be exceeded given unrealistic parameterization; functional form of spp interactions is loosely Lotka-volterra and fishing is linear	Gamble and Link 2009	Y
	Model	MSYPR	Estimate BRPs for multiple SS that include various interactions- Technical	Enhanced ecological realism; common outputs as SS models	Precision vs Accuracy Uncertainty Debate in SS Assessment Context; can miss other processes and dynamics	Deterministic	Analytical	Estimation	Dynamic	N	Both	Stock-recruitment relationships exist and can be estimated; technical interactions are consistent across gears	Murawski 1984	N
	Model	MVTS-Gompertz	Nonlinear, Non parameteric estimate BRPs for multiple SS that include various interactions- Ecological	Enhanced ecological realism; minimal statistical assumptions; common outputs as SS models	Precision vs Accuracy Uncertainty Debate in SS Assessment Context; can miss other processes and dynamics	Stochastic	Statistical	Estimation	Dynamic	Y	Tactical	Intrinsically linear dynamics	Fogarty and Liu in prep.	Y
	Model	MS SPMW	Estimate BRPs for multiple SS that include various interactions- Ecological	Enhanced ecological realism; common outputs as SS models	Precision vs Accuracy Uncertainty Debate in SS Assessment Context; can miss other processes and dynamics	Both	Analytical	Estimation	Dynamic	Y	Tactical	Usual for production models, esp. some equilibrium assumptions; ecological interactions have a distinct pattern/trend than fisheries removals, helps scale magnitude of pop estimates	Link et al. 2010, Mueter & Bohyaboy & Bundy et al. unpubl. data	Y
	Model	MSVPA-X	Estimate BRPs for multiple SS that include various interactions- Ecological	Enhanced ecological realism; common outputs as SS models	Precision vs Accuracy Uncertainty Debate in SS Assessment Context; can miss other processes and dynamics	Both	Statistical	Estimation	Dynamic	Y	Both	Usual for age/stage structured models; feeding sub-model assumes functional form of Type II or III functional response; selectivity of prey is primarily size-based, but with some type preference; mortalities are separable across fleets and predators; consumption helps scale magnitude of pop estimates	Tyrrell et al. 2008, Garrison et al. 2010, NEFSC 2006, Garrison and Link 2004, White et al. 2003	N

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Model Class	Aggregate Production	Main Purpose of Model	Major Strengths	Major Weaknesses	Stochastic or deterministic outputs	Analytical or Statistical	Primary model intent- simulation, estimation, scenario testing, pattern detection, etc.	Static or Dynamic	Predictions or Forecasts (Y or N) (Tactical)	Strategic, Both, Heuristic, N/A) output use wrt Mgt Advice	Major Assumptions	Reference/s	Emphasized in this Review (Y or N)
<b>Model</b>	AggPROD v of MS PROD	Simulate BRPs for Aggregate Groupings	Aggregate Properties Conserved, Simple Data Needs, Scenario Testing; common outputs as SS models	Amalgamating across spp may obfuscate life history factors	Both	Analytical	Simulation	Dynamic	Y	Tactical	Prey switching is always possible; no feedback on predator abundance by prey; aggregate properties of groups do not overly amalgamate spp information	Gamble and Link 2009, Link 2003	Y
<b>Model</b>	ASP-SPMW	Estimate BRPs for Aggregate Groupings	Aggregate Properties Conserved, Simple Data Needs, Scenario Testing; common outputs as SS models	Amalgamating across spp may obfuscate life history factors	Both	Statistical	Estimation	Static	N	Tactical	Usual for production models, esp. some equilibrium assumptions; covariates help scale magnitude of pop estimates; aggregate properties of groups do not overly amalgamate spp information	Mueter and Megrey 2006, Link et al. 2010	Y
<b>Model</b>	ASP-SPMW-Dynamic	Estimate BRPs for Aggregate Groupings	Aggregate Properties Conserved, Simple Data Needs, Scenario Testing; common outputs as SS models	Amalgamating across spp may obfuscate life history factors	Both	Analytical	Estimation	Dynamic	Y	Tactical	Covariates help scale magnitude of pop estimates; aggregate properties of groups do not overly amalgamate spp information; no equilibrium assumptions	Bohaby et al. unpubl. data, Link et al. 2010	Y
<b>Model</b>	Agg v of ASPIC	Estimate BRPs for Aggregate Groupings	Aggregate Properties Conserved, Simple Data Needs, Scenario Testing; common outputs as SS models	Amalgamating across spp may obfuscate life history factors	Both	Analytical	Estimation	Dynamic	Y	Tactical	Usual for production models, esp. some equilibrium assumptions; covariates help scale magnitude of pop estimates; aggregate properties of groups do not overly amalgamate spp information	Overholtz et al. 2008	Y
<b>Model</b>	Agg Mod - Overholtz/SAS	Estimate BRPs for Aggregate Groupings	Aggregate Properties Conserved, Simple Data Needs, Scenario Testing; common outputs as SS models	Amalgamating across spp may obfuscate life history factors	Both	Statistical	Estimation	Dynamic	N	Tactical	Usual for production models, esp. some equilibrium assumptions; covariates help scale magnitude of pop estimates; aggregate properties of groups do not overly amalgamate spp information	Overholtz et al. unpubl. Data	Y
<b>Model</b>	Agg Testing of MS PROD	Simulate BRPs for Aggregate Groupings	Aggregate Properties Conserved, Scenario Testing; common outputs as SS models	Amalgamating across spp may obfuscate life history factors	Stochastic	Analytical	Simulation	Dynamic	Y	Heuristic	Underlying dynamics of MS-PROD are able to be amalgamated into different groups; some equilibrium assumptions; mainly minimal as a simulator	Gaichas, Fogarty et al. unpubl. data	N

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Model Class	Energy Transfers (TL transfer, food web, network, etc.)	Main Purpose of Model	Major Strengths	Major Weaknesses	Stochastic or deterministic outputs	Analytical or Statistical	Primary model intent- simulation, estimation, scenario testing, pattern detection, etc.	Static or Dynamic	Predictions or Forecasts (Y or N)	(Tactical, Strategic, Both, Heuristic, N/A) output use wrt Mgt Advice	Major Assumptions	Reference/s	Emphazized in this Review (Y or N)
<b>Model</b>	Linear Production Potential	Estimate Fishery Production Potential	Aggregate Properties Conserved	Amalgamating across spp may obfuscate life history factors	Deterministic	Statistical	Pattern Detection	Static	N/A	Heuristic	Static transfer efficiencies across TLs; amalgamated properties within TL; catches distributed precisely across TL; PP estimable and partionable among new and recycled	Fogarty et al. 2008, NEFSC 2008	Y
<b>Model</b>	Stochastic Production Potential	Estimate Fishery Production Potential	Aggregate Properties Conserved	Amalgamating across spp may obfuscate life history factors	Stochastic	Statistical	Estimation	Static	Y	Tactical	Variable transfer efficiencies across TLs; amalgamated properties within TL; catches distributed precisely across TL; PP estimable and partionable among new and recycled	Fogarty et al. unpubl. data, Fogarty et al. 2008	Y
<b>Model</b>	Ecopath	Estimate Fishery Production Potential; Network Structure	User Friendly, good balancing tools	Too user friendly, hard to tell when balancing complete	Deterministic	Analytical	Estimation	Static	N/A	Strategic	Mass balance constraint; equilibrium assumption; vital rates germane as amalgamated across groups; transfers among groups reasonable estimatable/verifiable; balancing based off of multiple criteria here, not just EE	Link et al. 2006, 2008a, 2008b, 2009, Gaichas et al. 2009, Link 2010, Byron et al. in press; Walters et al. 1997; Christensen and Pauly 1992	Y
<b>Model</b>	Econetwrk	Estimate Fishery Production Potential; Network Structure	User Friendly, good balancing tools	Too user friendly, hard to tell when balancing complete	Deterministic	Analytical	Estimation	Static	N/A	Strategic	Mass balance constraint; equilibrium assumption; vital rates germane as amalgamated across groups; transfers among groups reasonable estimatable/verifiable; balancing based off of multiple criteria here, not just R/B	Link et al. 2006, 2008a, 2008b, 2009; Ulanowicz 2004	Y
<b>Model</b>	GOMAGG	Estimate Fishery Production Potential; Network Structure	Flexible/adaptable to myriad scenarios; based upon tuned networks	Not user friendly, assumptions on donor control not widely used in fisheries	Deterministic	Analytical	Simulation	Dynamic	Y	Strategic	Mass balance constraint; donor-controlled dynamics; vital rates germane as amalgamated across groups; transfers among groups reasonable estimatable/verifiable	Overholtz & Link 2009	N
<b>Model</b>	Topological Webs	Explore food web structure	Presence or absence needed; based on FH copious data	Not widely used nor applicable as DSS in fisheries	Deterministic	Statistical	Pattern Detection	Static	N/A	N/A	Linkages among groups detectable; network structure representative	Link et al. 2005, Link 2002, Link 1999	N

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<b>Model Class</b>	Full System												
<b>Model</b>	Atlantis	Simulate E2E full marine ecosystem	Flexible/adaptable to myriad scenarios; able to handle multiple processes, factors, and functional forms; highly modular; very inclusive	Not user friendly; very onerous to parameterize, initialize, and calibrate	Both	Analytical	Simulation, Scenario Testing	Dynamic	Y	Strategic	An entire ecosystem can modeled concurrently; myriad processes have multiple functional forms and these can be aptly chosen; many for each specific process being modeled	Link et al. 2010, Fulton et al. in press, Link et al. in press, Fulton et al. 2004	Y
<b>Model Class</b>	Misc												
<b>Model</b>	AAC	Estimate alpha ijs (spp interaction terms)	Addresses a commonly needed but hard to estimate parameter	Requires inputs that may not be extant	N/A	N/A	Calculate Parameters	Static	N/A	N/A	Growth rate, clearance rate/handling time, assimilation efficiency & consumption rate all have requisite info available and are estimable; assumes a Type II functional response of feeding, with variable forms available but harder to calculate	Gamble & Link unpubl. data	N
<b>Model</b>	Donut Selectivity Model	Estimate diet composition	Estimates DC based on 1st principles; surprisingly accurate; simple to use	Only gives selectivity if no prey field relative abundance available	N/A	N/A	Calculate Parameters	Static	N/A	N/A	Just provides preference in absence of ambient prey field (relative proportion); underlying framework based off of Hollings components of predation	Link 2004, Link & Keen 1999	N
<b>Model</b>	PSA	Determine susceptibility of biota; DSS	Straightforward ranking	Overly simplistic rankings or categories can lead towards scores tending towards central estimates	Deterministic	N/A	Estimation	Static	N/A	Strategic	Risk, susceptibility, and productivity can be deconstructed into salient, component features; there is enough contrast among spp to delineate levels of risk; the ranking categories are sufficient to detect risk	Patrick et al. 2010, 2009	N
<b>Model</b>	LeMans	Simulate size (length) structure of food web	Explores dynamics based on size structure	Missing other dynamics	Deterministic	Analytical	Simulation	Dynamic	Y	Heuristic	Dynamics of a fish community are driven largely by size of spp; else minimal as a simulating tool	Hall et al. 2006	N
<b>Model</b>	Size Spectra	Estimate size structure of food web	Based on copious data; Explores size structure	Missing other dynamics	Deterministic	Statistical	Pattern Detection	Static	N/A	Heuristic	Decay of abundance with size is due to known mechanisms; change in rate/slope represents perturbations; slope is robust in aquatic ecosystems	Methratta & Link 2006, Link 2005	N

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<b>Model Class</b>	Misc												
<b>Model</b>	CCA/CanCorr	MV Statistics to explore relationships among response & explanatory variables	Based on copious data	Usual MV statistical assumptions	Deterministic	Statistical	Pattern Detection	Static	Y	N/A	MV normality and linearity; minimal collinearity and redundancy; statistically significant canonical relationships not necessarily causal	Link et al. 2002, Link et al. 2009a	N
<b>Model</b>	DFA/MAFA	MV TS Statistics to explore relationships among response & explanatory variables	Based on copious data	Usual MV statistical assumptions	Deterministic	Statistical	Pattern Detection	Static	Y	N/A	MV normality; minimal collinearity and redundancy; significant canonical time series combined represent common responses, but not necessarily causal from covariates	Nye et al. 2009, Shackell et al. unpubl. data, Link et al. 2009b	N
<b>Model</b>	PCA/MDS	MV Statistics to explore relationships among variables	Based on copious data	Usual MV statistical assumptions	Deterministic	Statistical	Pattern Detection	Static	N	N/A	MV normality and linearity for PCA, more robust for non-parametric MDS; minimal collinearity and redundancy; fairly robust methods to detect patterns	Link et al. 2002, Methratta & Link 2006	N