

Comisión Interamericana del Atún Tropical  
Inter-American Tropical Tuna Commission



IATTC plan for model weighting in the yellowfin and bigeye tuna stock assessment

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2nd Workshop on improving the risk analysis for the tropical tunas in the eastern Pacific Ocean: model weighting for integrated stock assessments (by videoconference)

# Outline

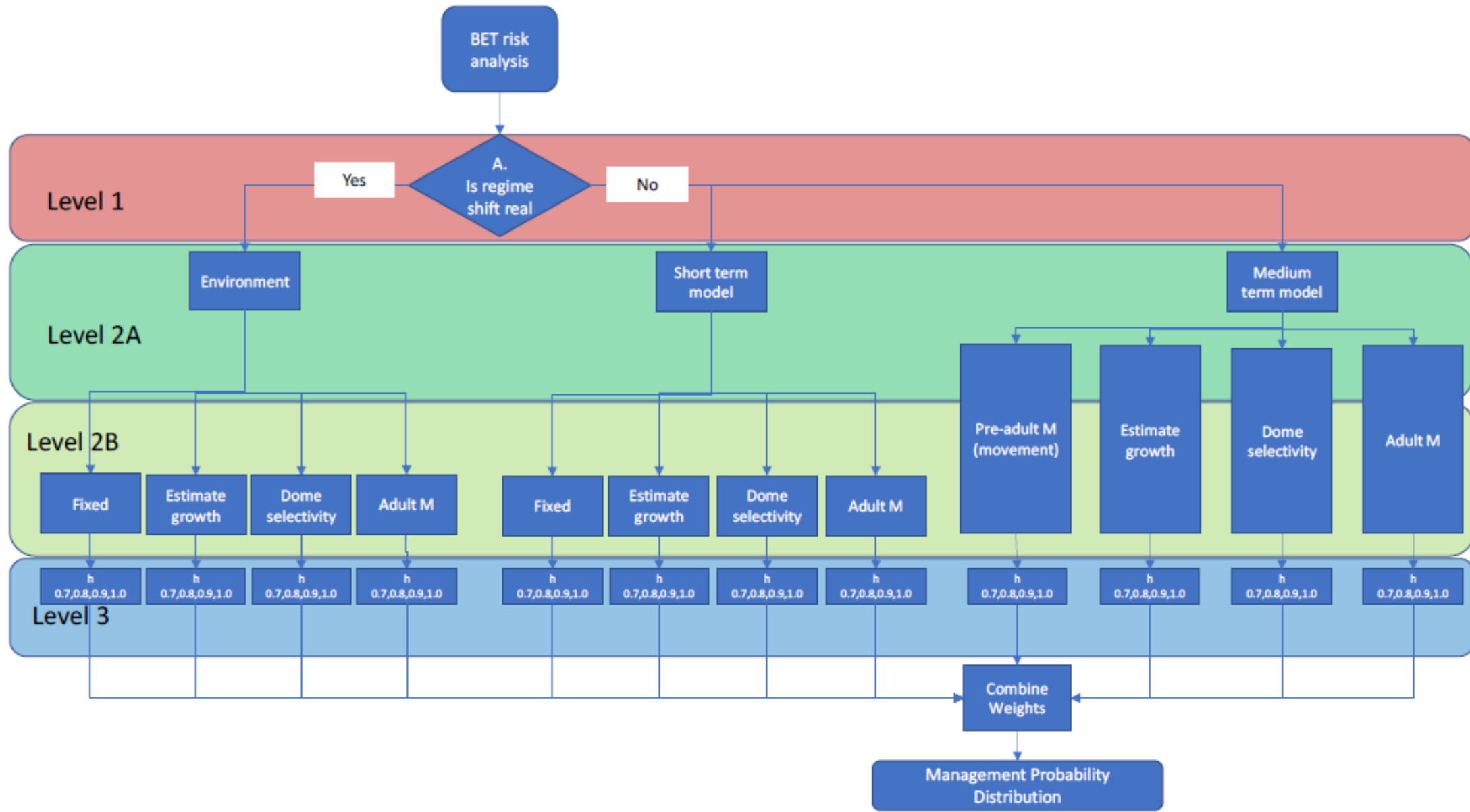
- Recap 2020 risk analysis approach
- Proposal for updated risk analysis approach

# The 2020 risk analysis approach

Described in Maunder et al. 2020 (SAC-11- INF-F):

- 1. Identify alternative hypotheses ('states of nature') about the population dynamics of the stock that address the main issues in the assessments**
- 2. Implement stock assessment models representing alternative hypotheses**
- 3. Assign relative weights to each hypothesis (model)**
- 4. Compute combined probability distributions for management quantities using model relative weights**

# The 2020 flow chart for bigeye tuna



# The 2020 method for assigning model weights

- Level 1 (regime shift hypothesis) is weighted independently solely on experts opinion
- Level 2 is weighted based on several criteria:
  - Expert opinion
  - Convergence
  - Fit to data
  - Plausible parameter estimates
  - Plausible model results
  - Model diagnostics
  - Recruitment shift metric
  - Empirical selectivity vs. estimated selectivity
- Level 3 (steepness hypothesis) is weighted independently solely on experts opinion

# The 2020 method for assigning model weights

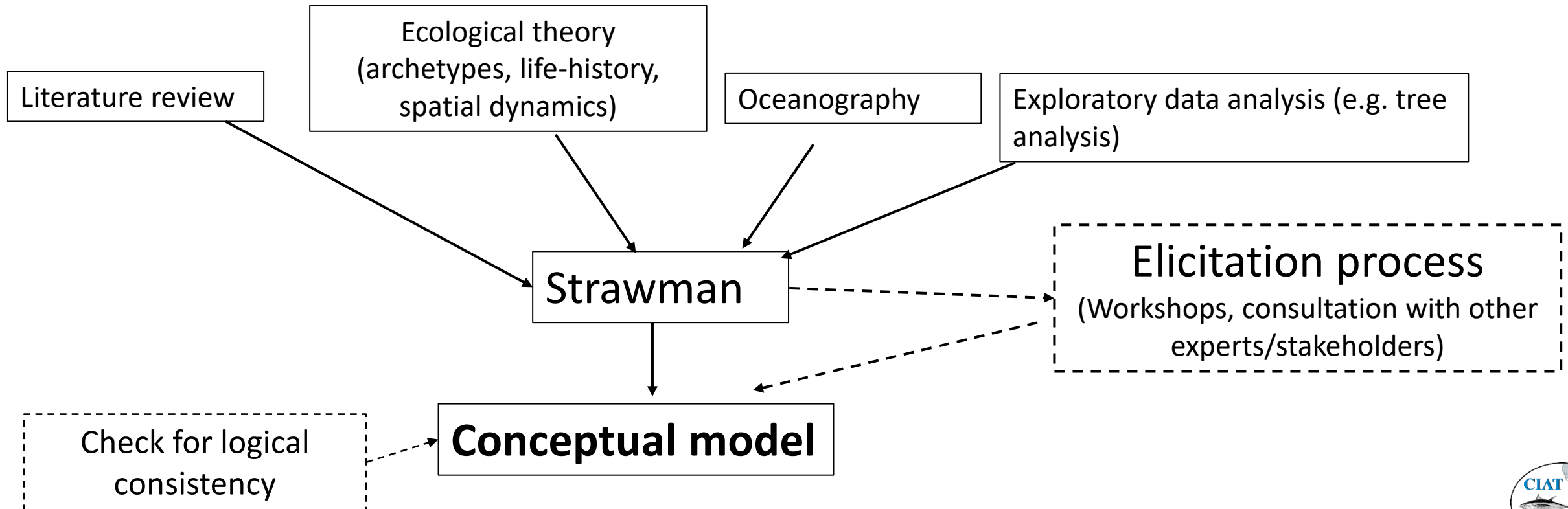
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# Proposal for new approach

1. Develop a **conceptual model** for the fisheries system:

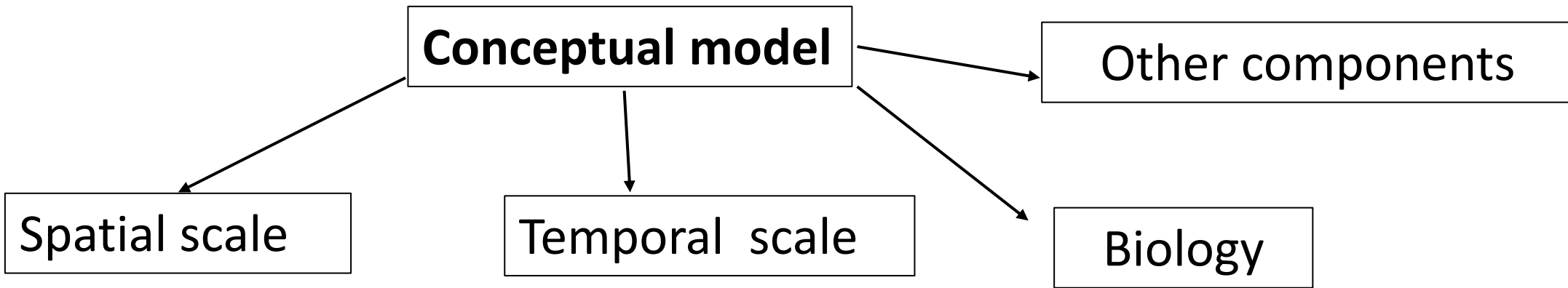
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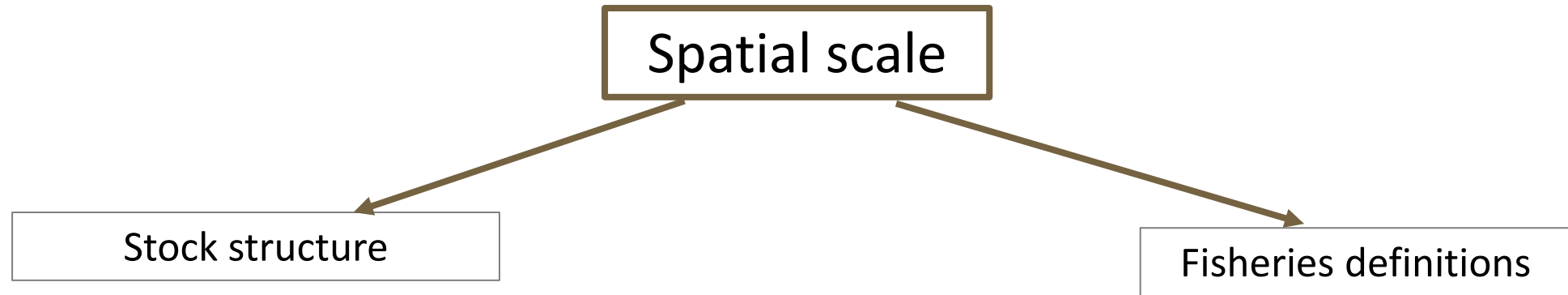




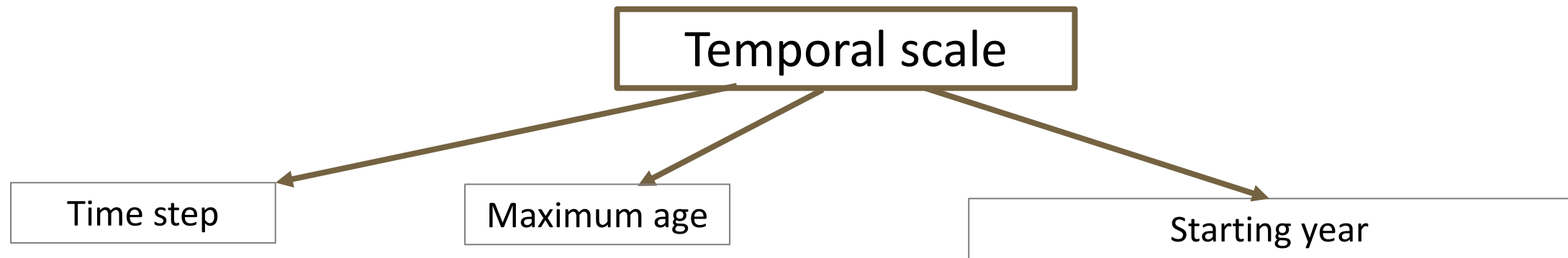
# Develop a conceptual model



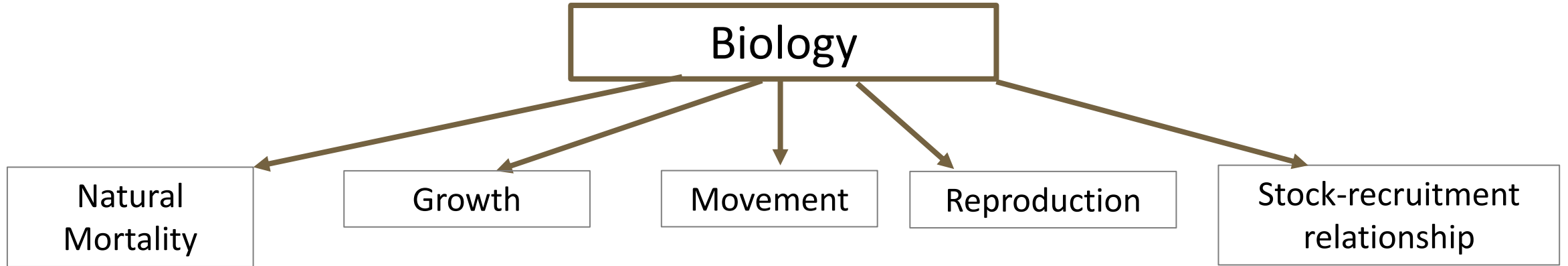
# Develop a conceptual model



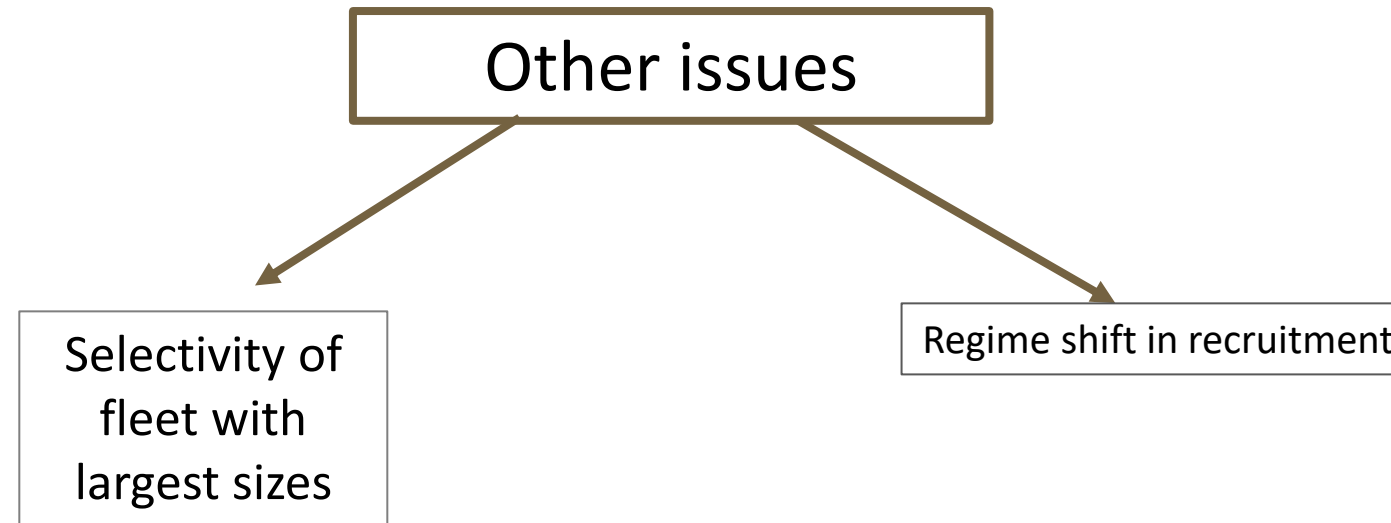
# Develop a conceptual model



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# Develop a conceptual model



# Proposal for new approach

1. Develop a **conceptual model** for the fisheries system:
  - Include justifications for assumptions/hypotheses
  - **Use an elicitation / consultation process if needed**
  - Organize hypotheses in a hierarchical way
  - Identify independent (orthogonal) uncertainty axes
    - Parameters that will have different fixed values (Level 3 hypotheses) tried for other hypotheses
  - Identify processes (parameters) with no information in the data used to fit the model
    - **Use knowledge from published simulation studies to understand whether a parameter is estimable (e.g. Lee et al on M and h)**
    - **Perform simulation to investigate estimability of the parameter on the fishery system under study**
    - Identify processes (parameters) that can be informed by external data:
      - Include prior knowledge as i) fixed value, ii) parameter range, iii) **prior distribution, iv) joint prior distribution for correlated processes (e.g. Natural mortality and growth)**
      - **Decide whether to include this info in the estimation or in the weighting**
  - Identify practical and impractical hypotheses (can they be implemented given the available data/tools?)
    - Impractical hypotheses will have weight of 0, and considered in the future when resource become available
  - **Check the conceptual model for logical consistency**

# Proposal for new approach

## 2. Implement stock assessment models representing alternative hypotheses

- Implement a base or “ancestral” model(s) from which other models will be developed:
  - Models can use different data or have completely different structure according to the hypothesis they represent
- Run a set of diagnostics to learn about the base /“ancestral” model(s) (**decide in advance what those diagnostics will be**)
  - Improve base or “ancestral” models/change conceptual model if issues arise that merit new hypotheses
- Decide on a candidate set of reference models (and **decide in advance which diagnostics to used, noting some diagnostics are for model understanding while others are for model validity \***)
  - A. Large set of models that hopefully covers all the models to improve diagnostics
  - B. Smaller set of curated models to represent specific hypotheses:
    - Use diagnostics to improve the models or discard them (how or how much to improve models?)
- Fit all reference models to data:
  - Data will need to have its own quality control rules to be able to be included in the model
- Arrive at a final list of models:
  - The final list will be more or less inclusive according to the diagnostics used

# Proposal for new approach

## 3. Assign relative weights to each hypothesis (model)

- **Decide in advance the weighting scheme based on the objectives of the assessment**, which may include in this order:
  - Use equal weight as default or as alternative
  - Decide prior weight of overarching hypotheses that cannot be evaluated by the application of models (e.g. stock structure)
  - For independent processes not informed by the data (e.g. steepness, movement rates):
    - use prior distributions (e.g. from meta-analyses, expert opinion or other) as weights , if they were not used in the estimation
  - **Use predictive ability (hindcast) to weight models:**
    - Ideally the predicting ability should be estimated on the quantity of interest
    - Quantities of interest in stock assessment are in general not directly observable
    - Observable quantities are used and should be chosen carefully (prefer those that are proxies for quantities of interest)
  - Weight nested hypotheses/models with conditional weights (sum to 1 within a branch)
- Final weights consider the prior weight of overarching hypotheses, independent process weights, weight based on hindcast and the position of the model in the hierarchy



# Proposal for new approach

## 4. Compute combined probability distributions for management quantities using model relative weights

- Use probability staking rather than model averaging to estimate the tails of the distribution
- To compute the final probability distribution by:
  - Normal approximations based on the estimate and standard error
  - Some standard errors are approximated
  - The resulting distribution is rescaled to obtain  $P(\text{Quantity} | \text{Model}=m)$ .
  - Works well when the data is very informative
  - Probability distribution may be asymmetrical
  - Evaluate appropriateness of the approximation using Posteriors derived from limited MCMC analyses

# List of model diagnostics (Pass/Fail can be used to improve models as well)

## Standard diagnostics

- **Pass/Fail** Model convergence: e.g. size of the gradient at the MLE, Hessian matrix, jitter analysis.
- **P/F** Check for parameters in the bounds
- **P/F run test** Evaluation of residuals: need to take into account the assumed distribution, use PIT residuals for composition data
- **Improve** Effective sample sizes and variances
- **Weighting** Cross validation and **hindcasting**
- **P/F** Bayesian model checking
- **P/F** Retrospective analysis: The interpretation is context dependent. When the data is more informative at the end of the time-series may result in pattern. If we have no retrospective pattern there is no learning from the data in the tails.

## Stock Assessment specific

- **P/F** R0 likelihood component profile
- **Improvement** Age-structured Production Model (ASPM)
- **P/F** Catch curve analysis
- **P/F** Empirical selectivity
- **Improvement** **Patterns of process error** (e.g. recruitment residuals, but other process error may need to be checked too such as time-varying selectivity)

## Plausibility

- **P/F** Parameter values (e.g. smaller F for highly exploited stocks)
- **P/F** Results
- **Not used** **Projection with current catches** (if the system cannot support current catches in the future, the production function may be implausible)

# Notes

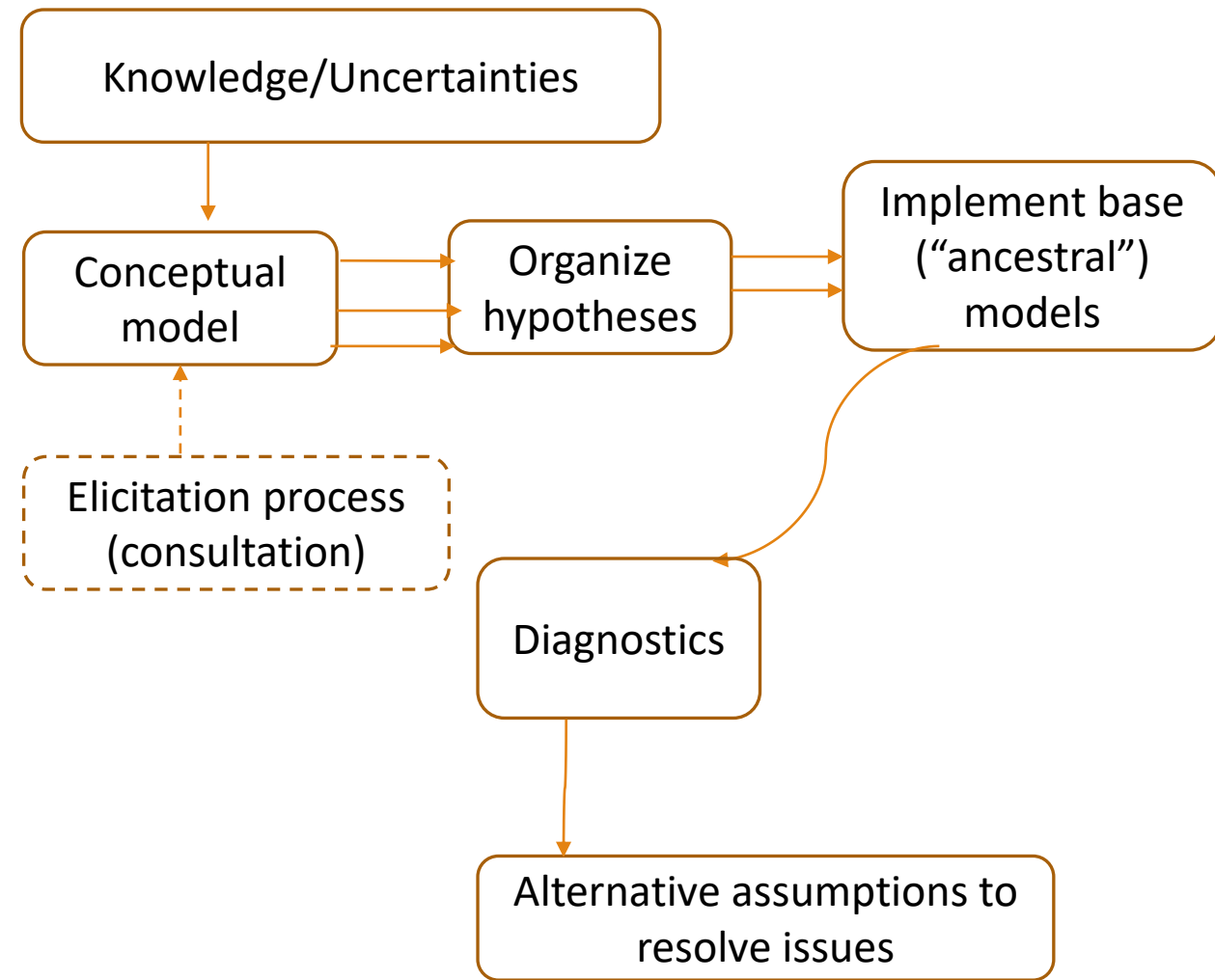
- Values for pass/fail or criteria for each diagnostics need to be worked out
- If the model fails one of the P/S diagnostics, it will be excluded
- Investigate emergent properties, what are the combination of parameters that cause the model to break

# Hindcasting

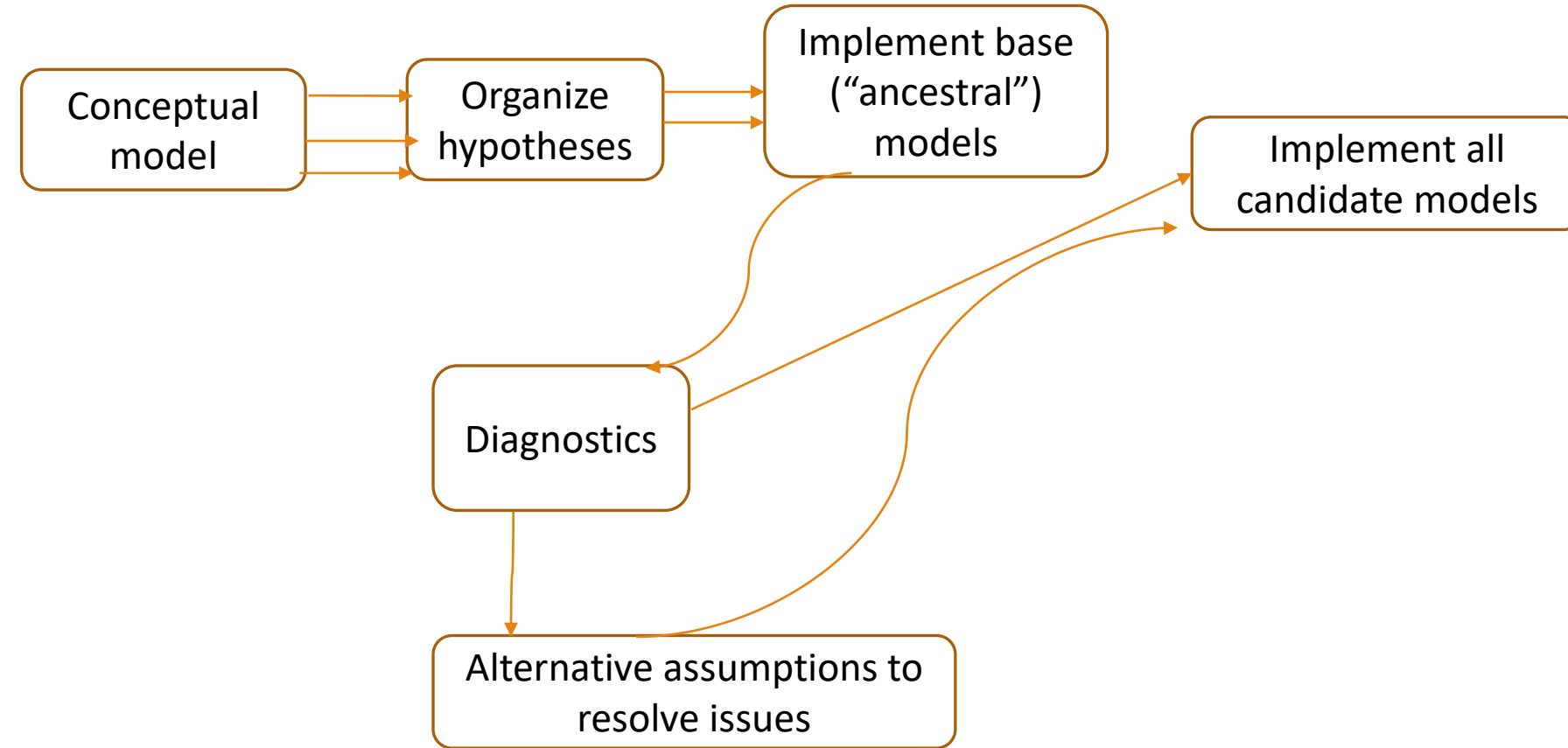
## Decisions to make (in advance):

- Need some simulation studies to make informed decisions
- What to predict:
  - index of abundance
  - average size/quantile in the survey/index (if available)
- For how long?
  - One set ahead in the time step of the model (quarterly)
- What data to exclude?
  - All the data in the year except the catches or effort
- How many peels?
  - Minimum of 10 quarters (2.5 years)

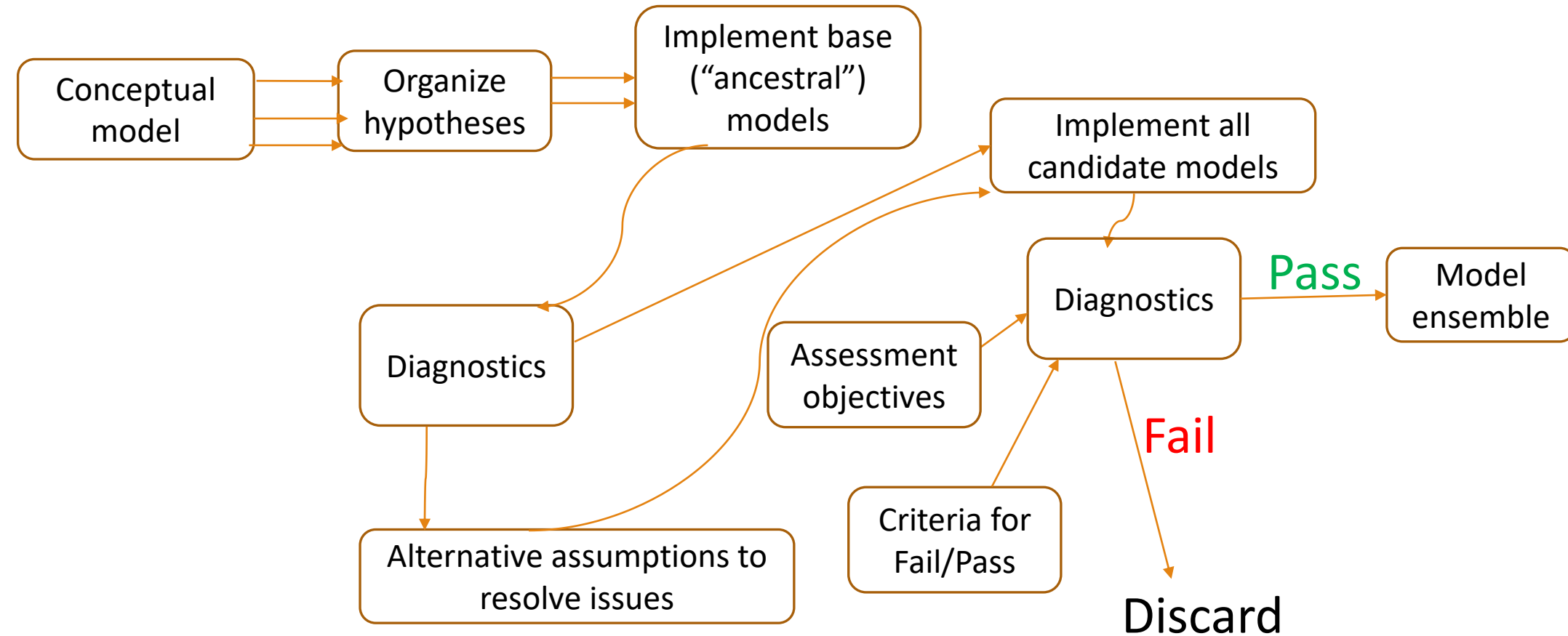
# Summary: Ensemble model approach



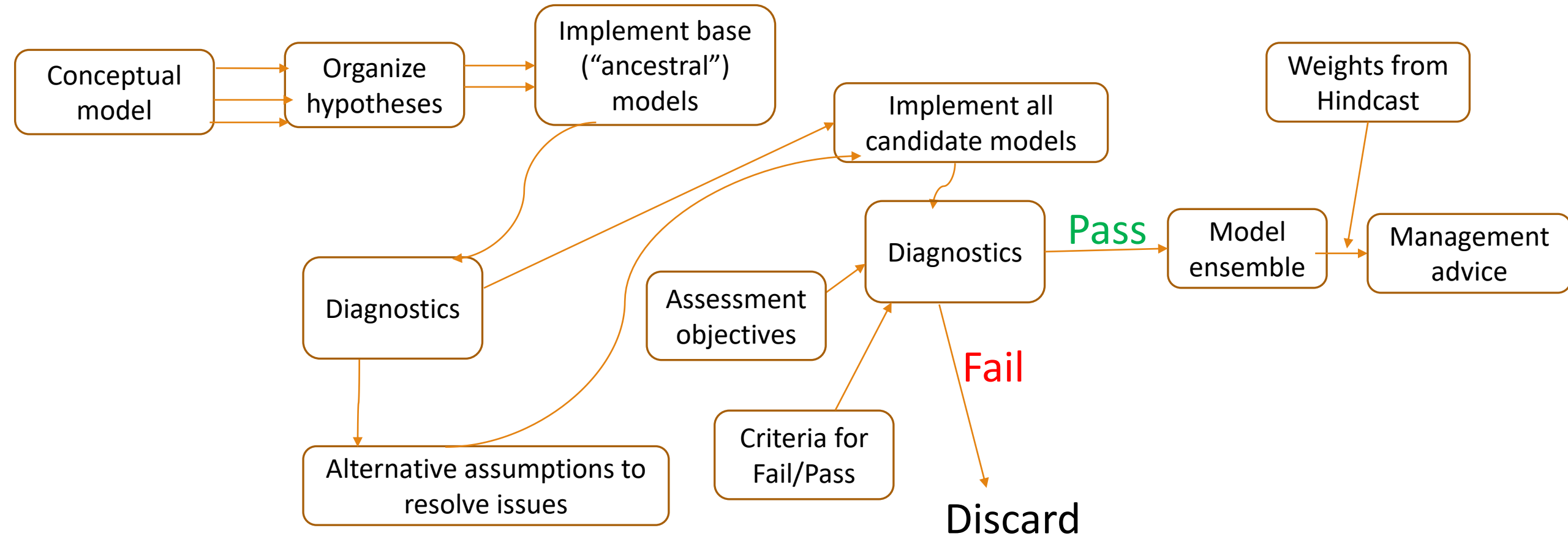
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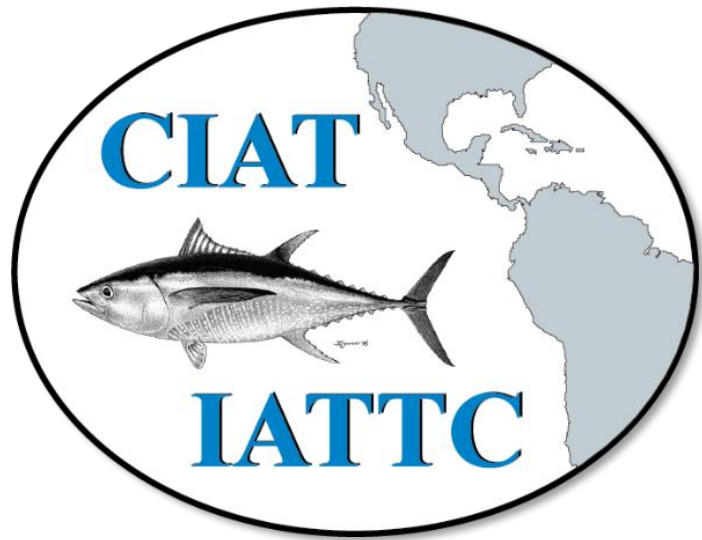
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Questions