



# Seasonal Inhomogeneous Arrival Process Search and Evaluation

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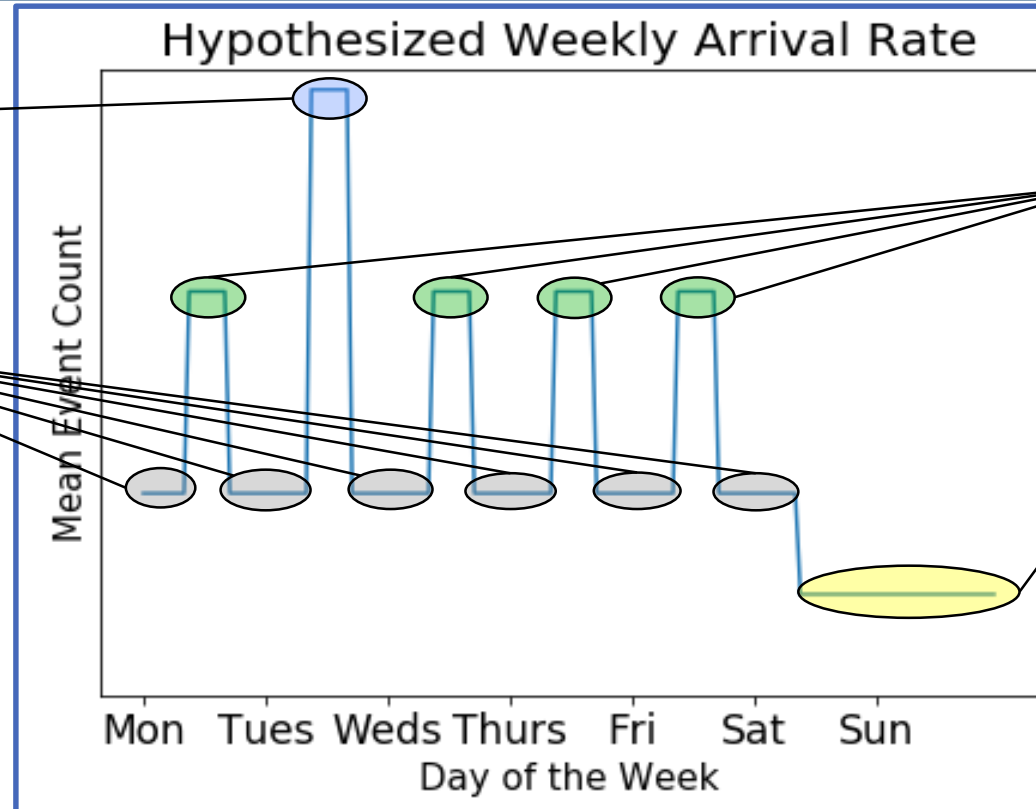
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# Model Hypothesis: Cybersecurity Domain

Higher expected arrival rate on "Patch Tuesdays"

Rates overnight on weekdays lower than during the day but higher than over the weekend



Most attacks expected to occur during work hours – weekdays may be part of the same "regime"

Fewer attacks during weekends

## Key Features

- Nonhomogeneous piecewise model
- Nonconsecutive intervals
- Repetitive
- Sparse data



# Model Vocabulary



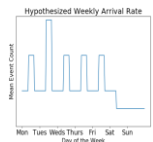
Breakpoints



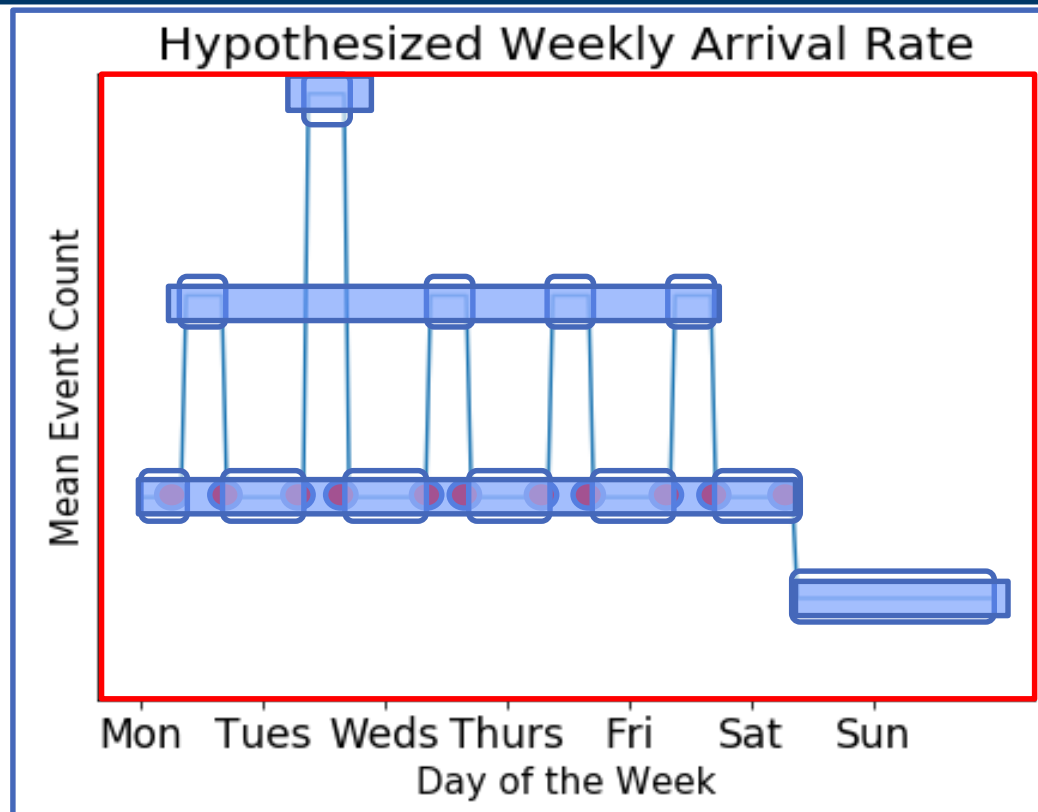
Intervals



Regimes



Interval Division



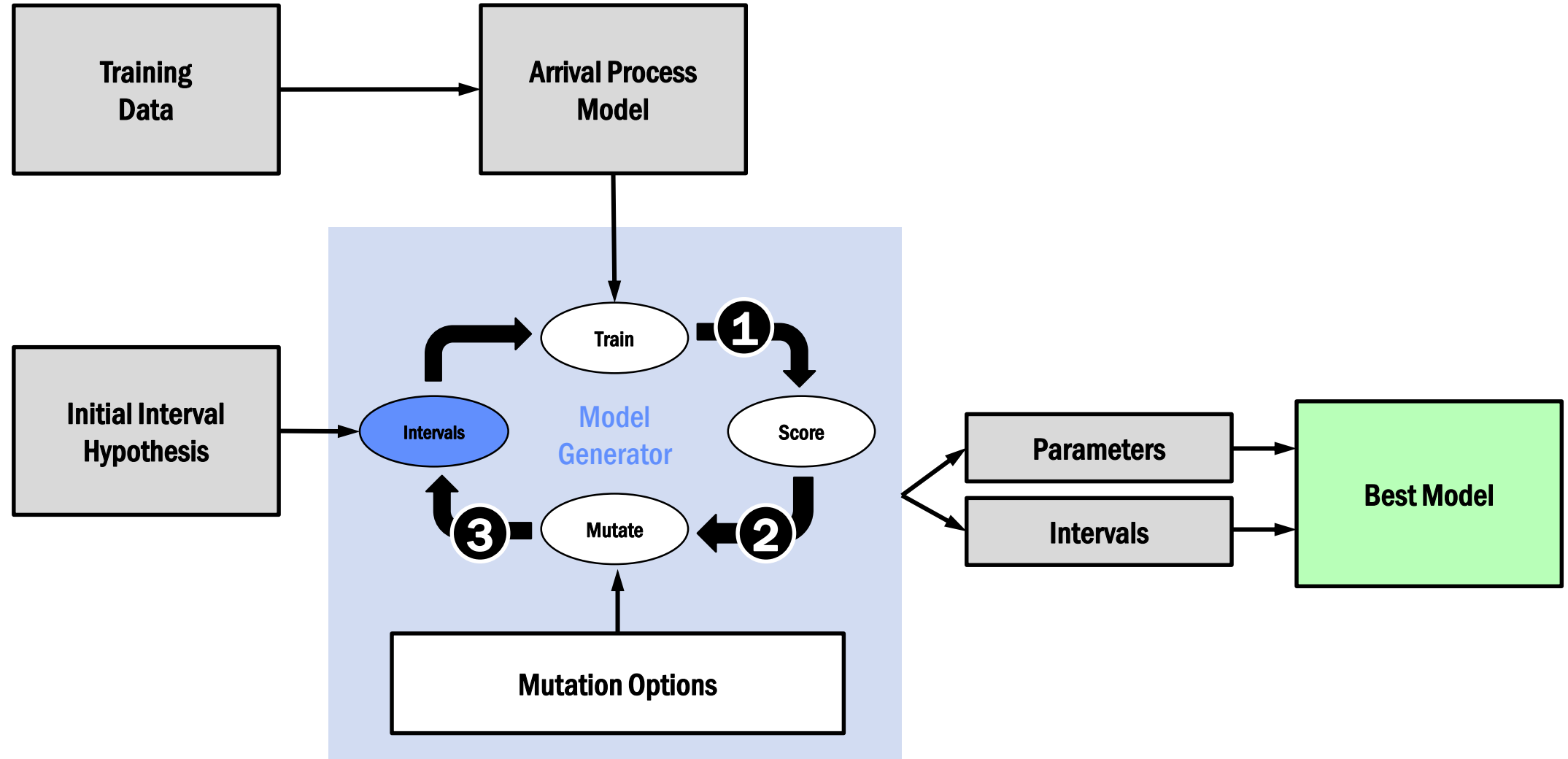
## Questions

- Which model type fits best?
- Where are the breakpoints?
- How should intervals be assigned to regimes?
- What are the correct parameters for each regime?



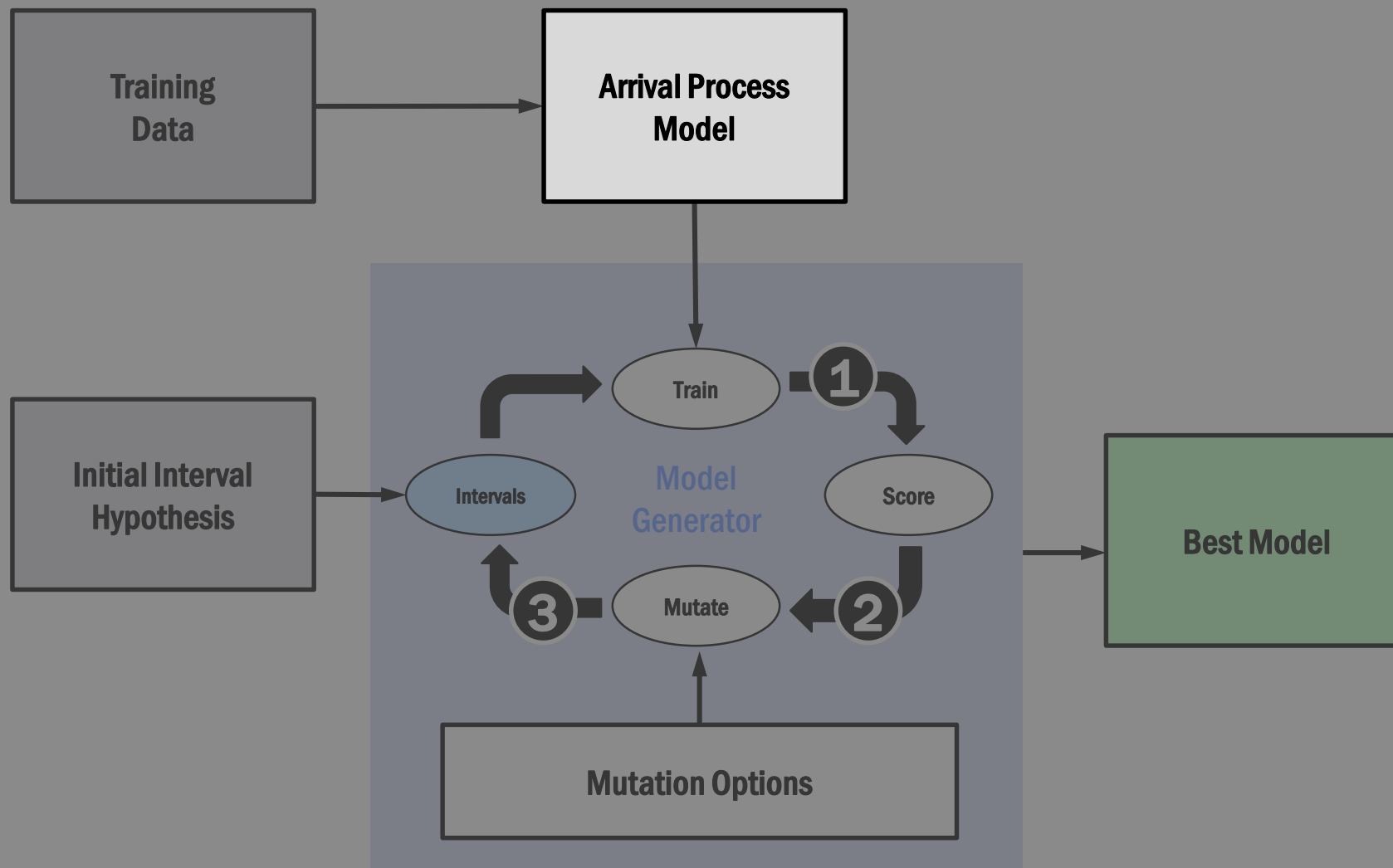
# SINAPSE Algorithm

Seasonal Inhomogeneous Arrival Process Search and Evaluation





# SINAPSE Algorithm

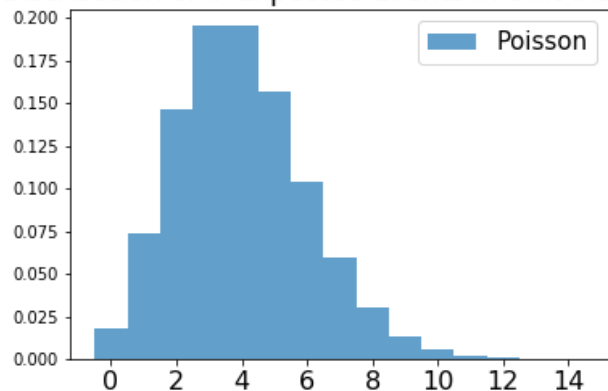




# Arrival Process Models

## Poisson

Distribution of # Expected Events in Unit Interval

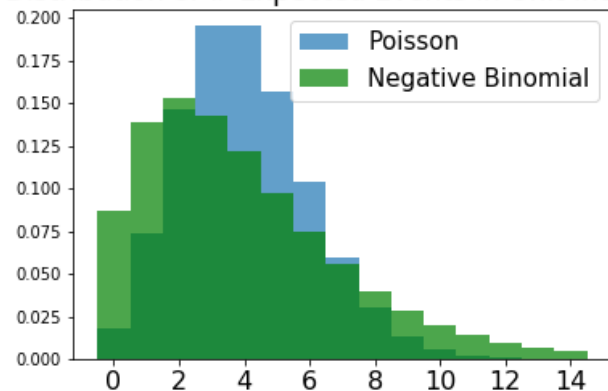


$$P(N = n; \lambda) = \frac{\lambda^n e^{-\lambda}}{n!}$$

- Models occurrence of events over time
- Time between events is independent, exponentially distributed

## Negative Binomial

Distribution of # Expected Events in Unit Interval

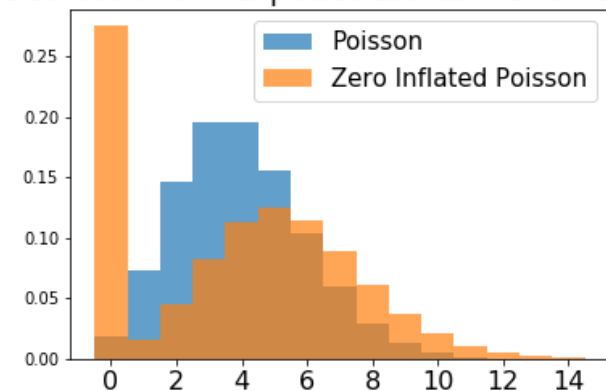


$$P(N = n; r, p) = \binom{n+r-1}{n} (1-p)^r p^n$$

- Variance may be greater than the mean
- More parameters, greater risk of overfitting

## Zero-Inflated Poisson

Distribution of # Expected Events in Unit Interval

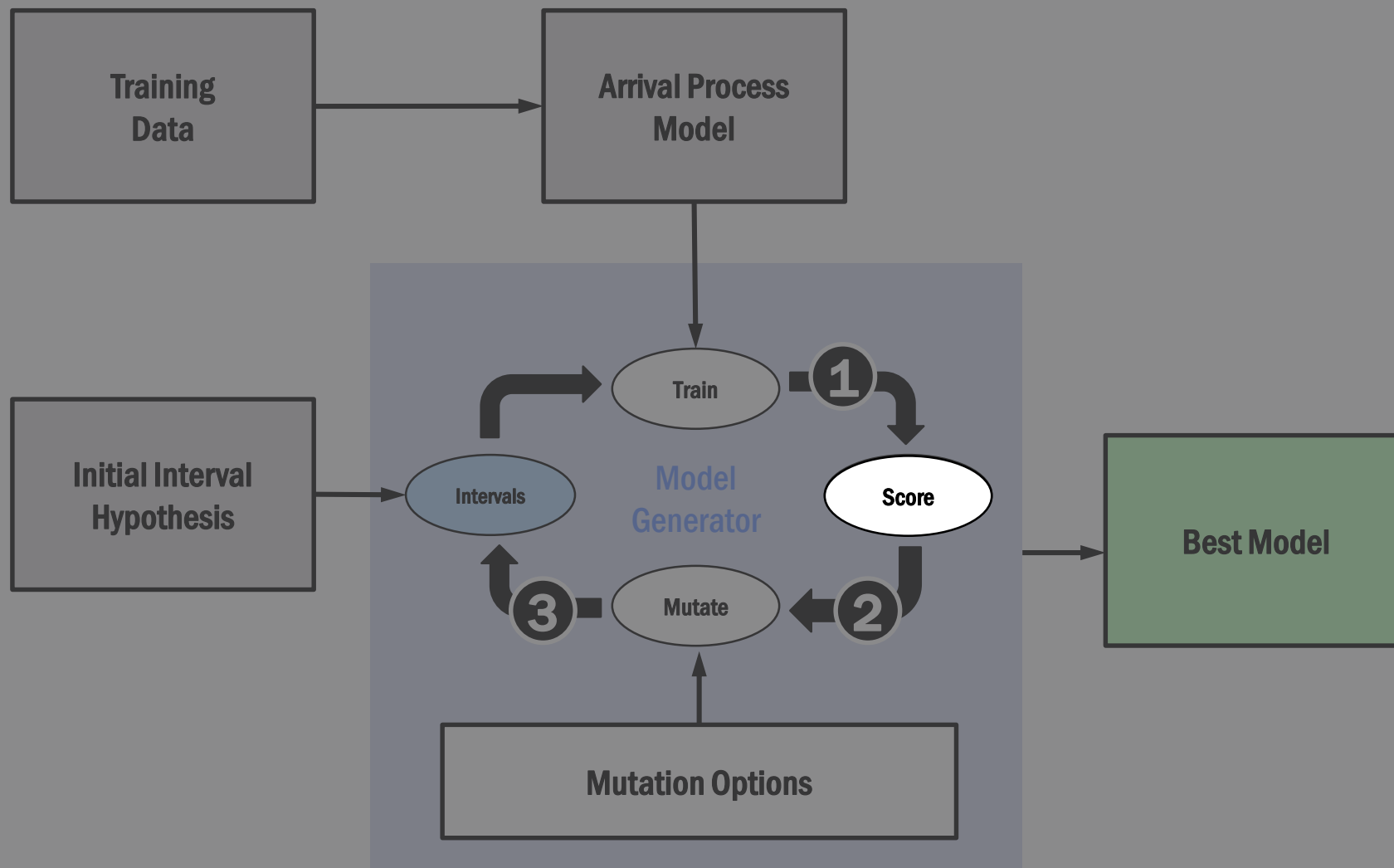


$$P(N = n; \lambda, \pi) = \begin{cases} \pi + (1 - \pi)e^{-\lambda} & n = 0 \\ (1 - \pi) \frac{\lambda^n e^{-\lambda}}{n!} & \text{otherwise} \end{cases}$$

- Poisson with excess zeros from outside process



# SINAPSE Algorithm





# Measure of Fit

$$AICc = 2P - 2 \log L(\theta|X) + \frac{2P(P+1)}{(n-P-1)}$$



Formula for Penalty Parameter P

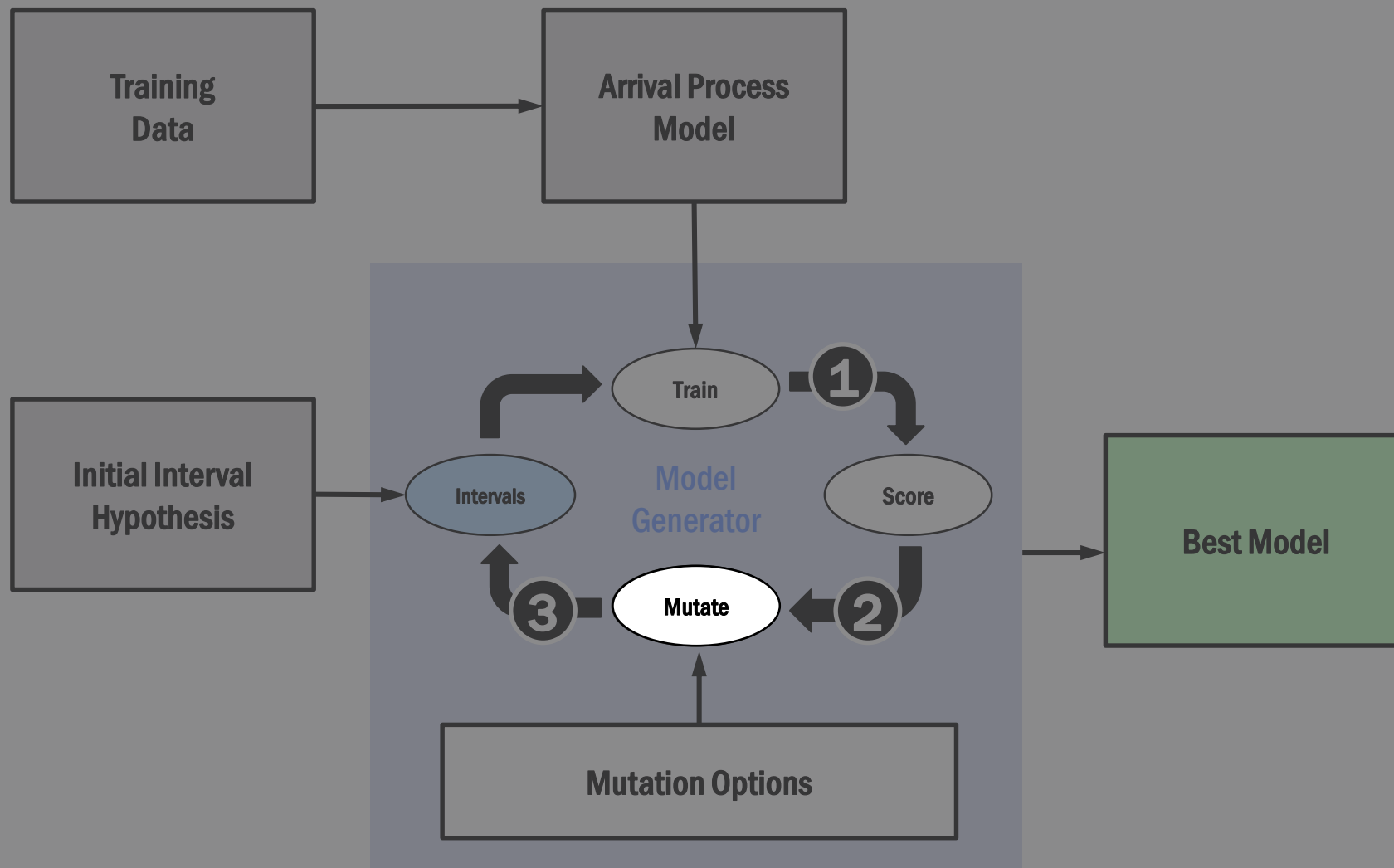
$$P = \# \text{ Intervals} + (\# \text{ Regimes}) * r$$

Arrival Process Model	Parameter Count per Regime (r)
Poisson	1
Negative Binomial	2
Zero Inflated Poisson	2



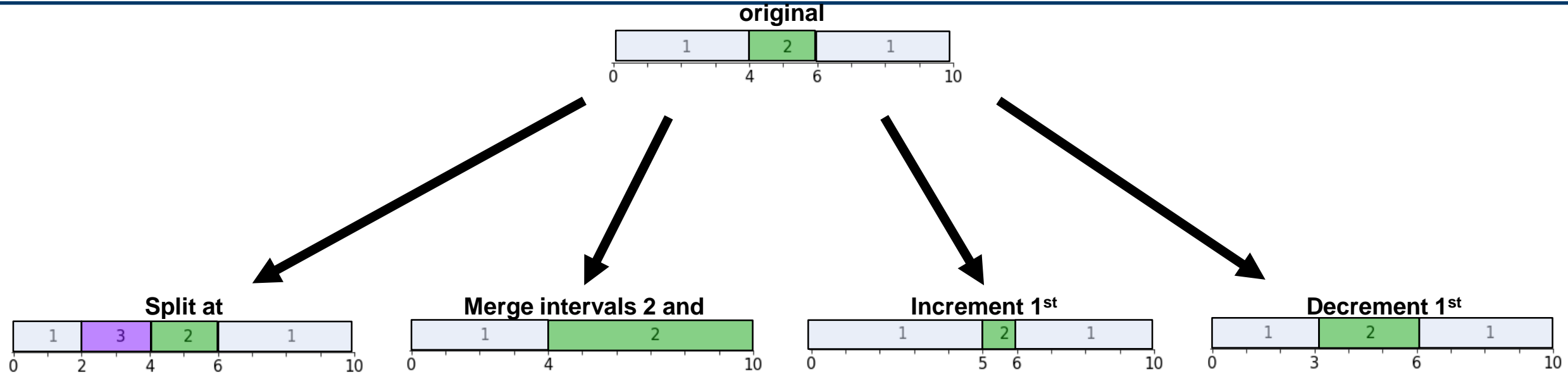


# SINAPSE Algorithm





# Mutation Options



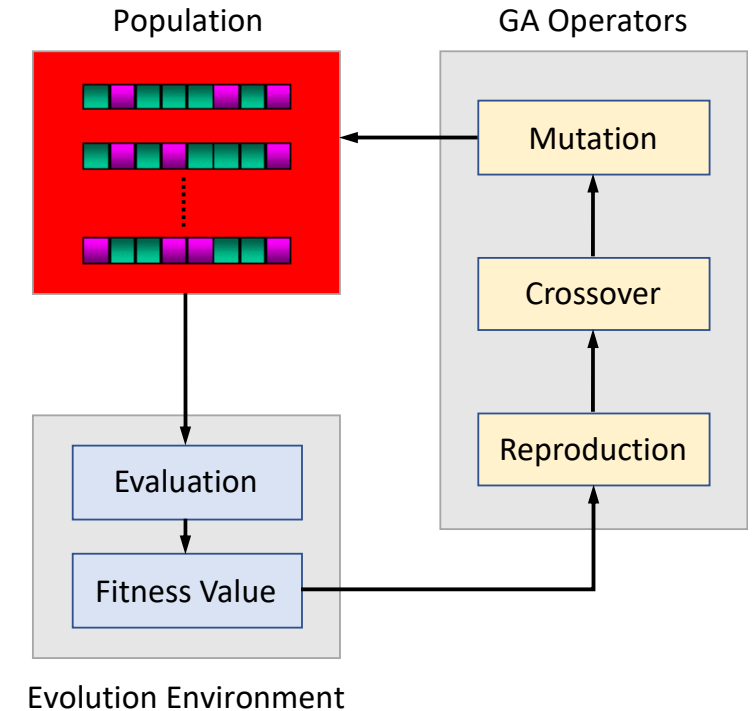
## Interval Mutations

1. Randomly split an interval by adding a new breakpoint, creating a new regime
2. Randomly merge two intervals
3. Randomly increment the location of a breakpoint
4. Randomly decrement the location of a breakpoint



# Genetic Algorithm

- **Optimization algorithm**
  - Minimize AICc
- **Randomly generate a population**
  - N intervals
- **Choose the best few intervals to create the next generation with**
  - Reproduction
  - Crossover
  - Mutation
- **Converge after specified number of generations**

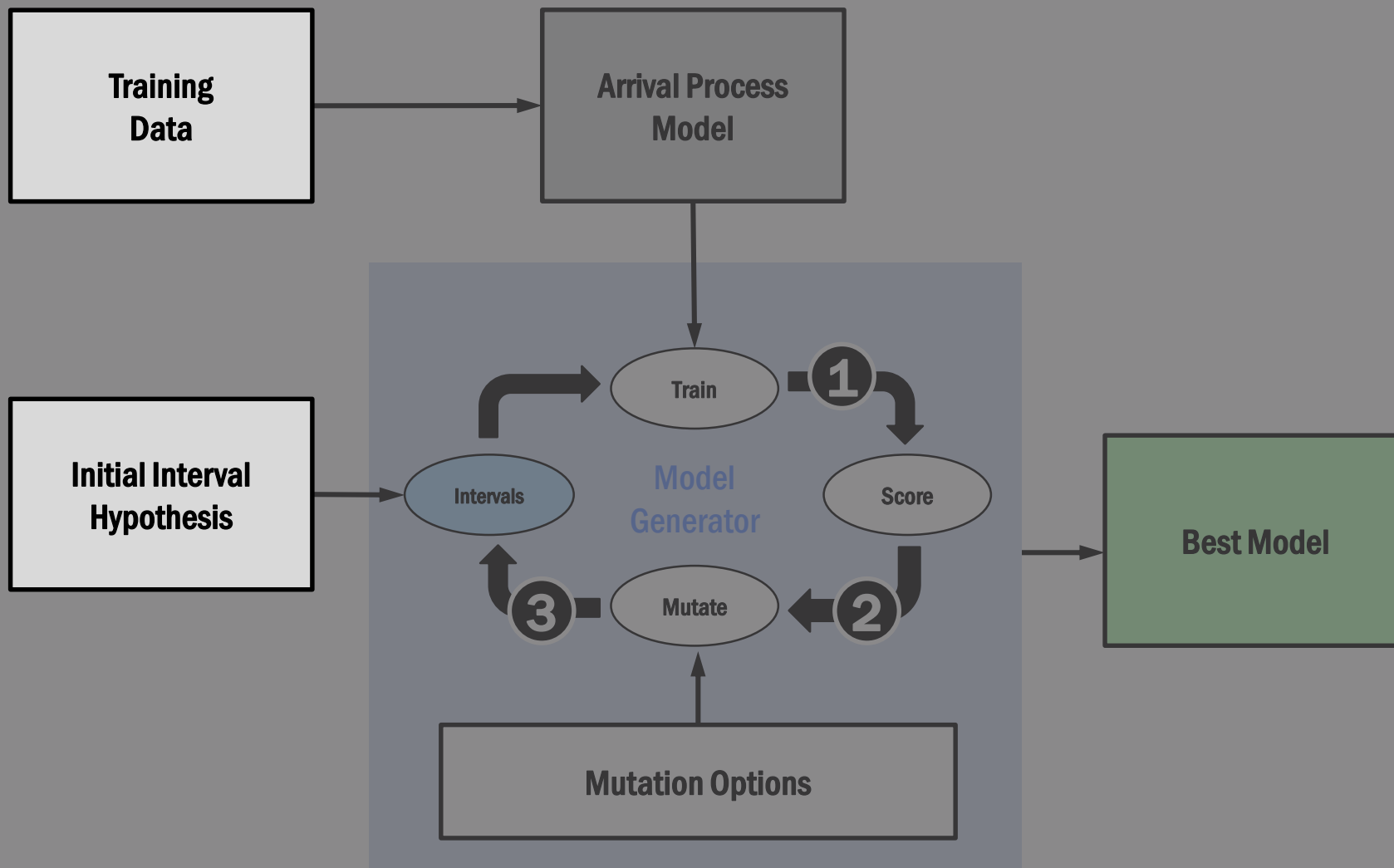


Genetic Algorithm Evolution Flow

Image from AL-Madi, Nagham & Khader, Ahamad Tajudin. (2008). De Jong's sphere model test for a Social-Based Genetic Algorithm (SBGA). IJCSNS. 8.

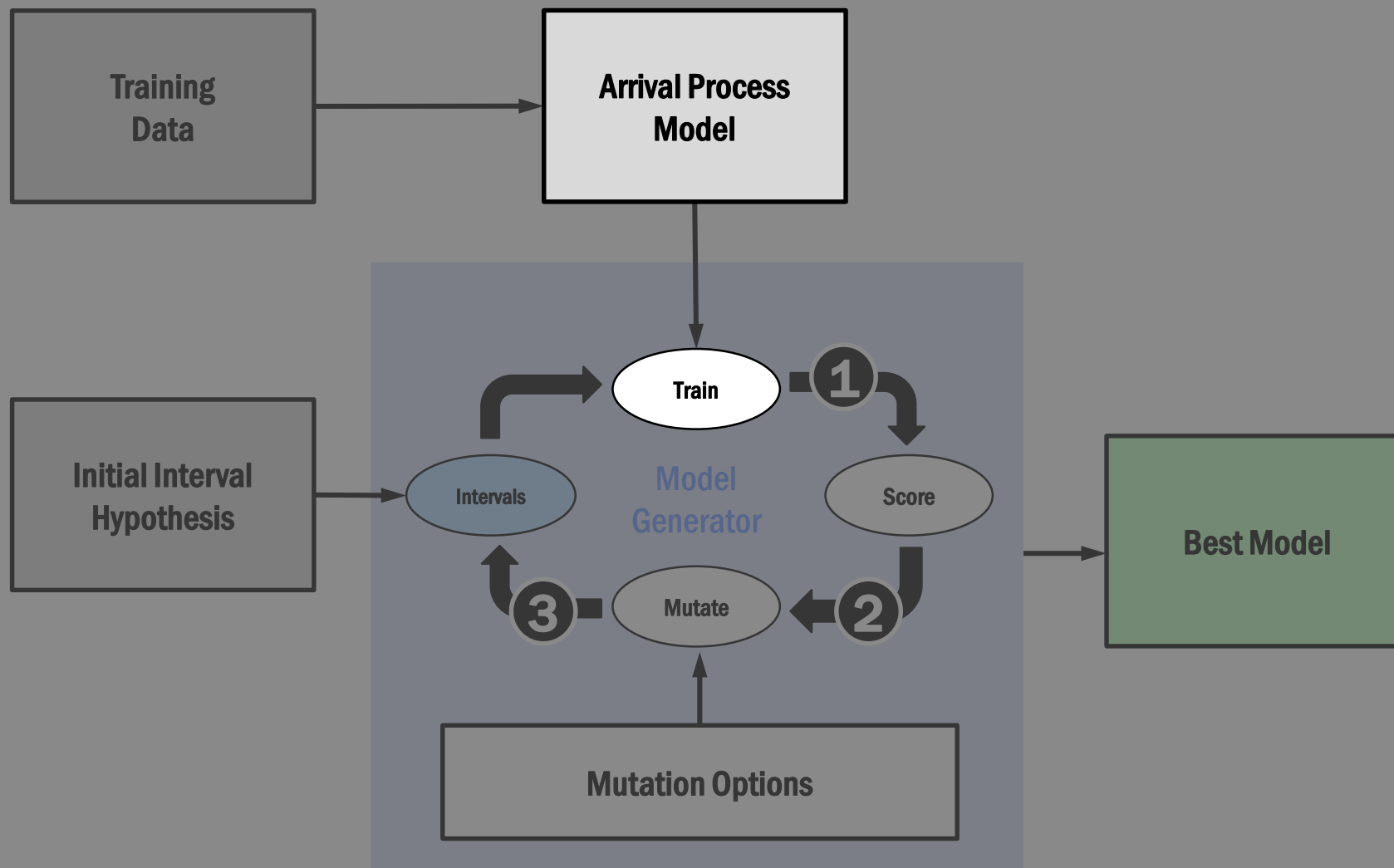


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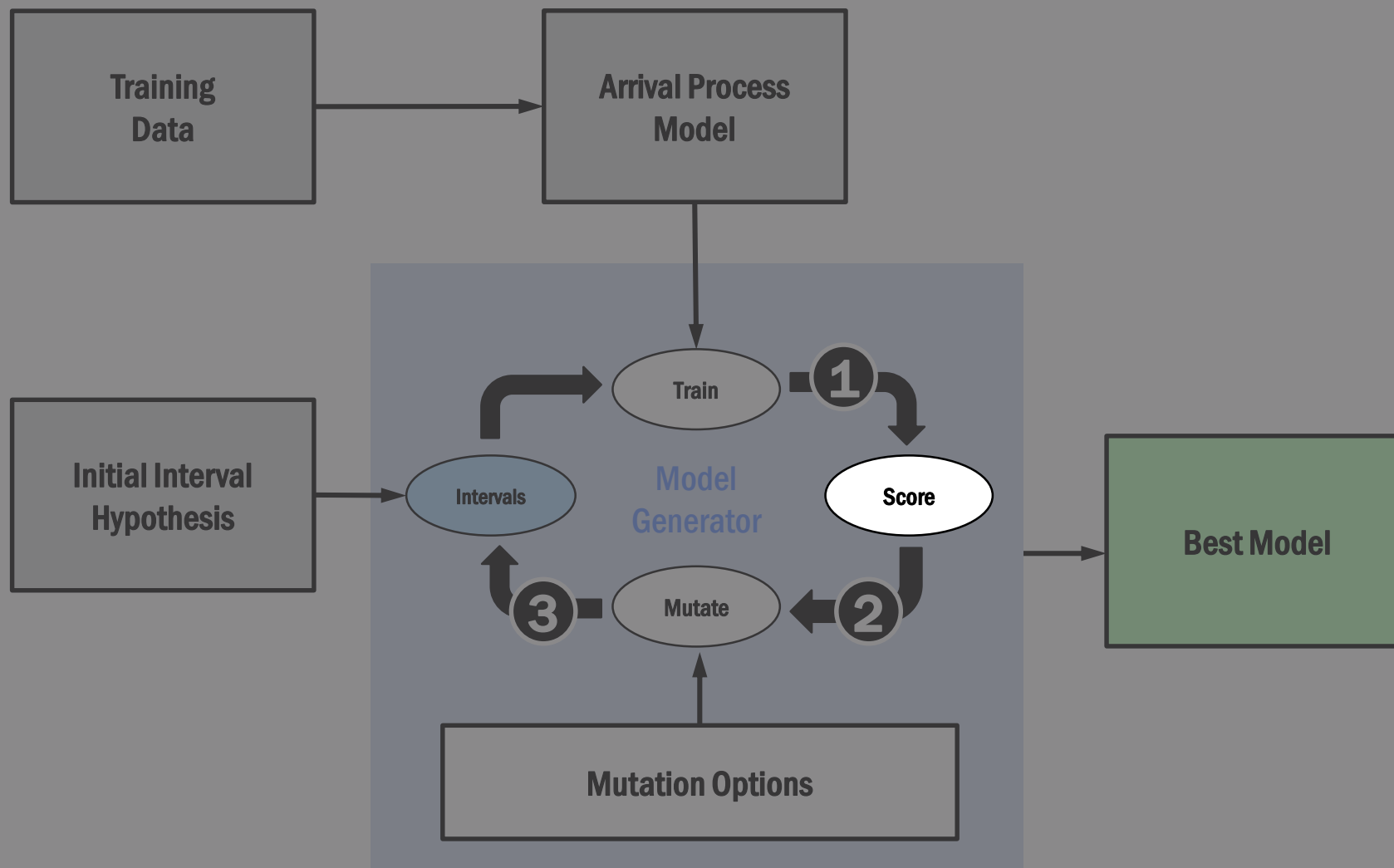


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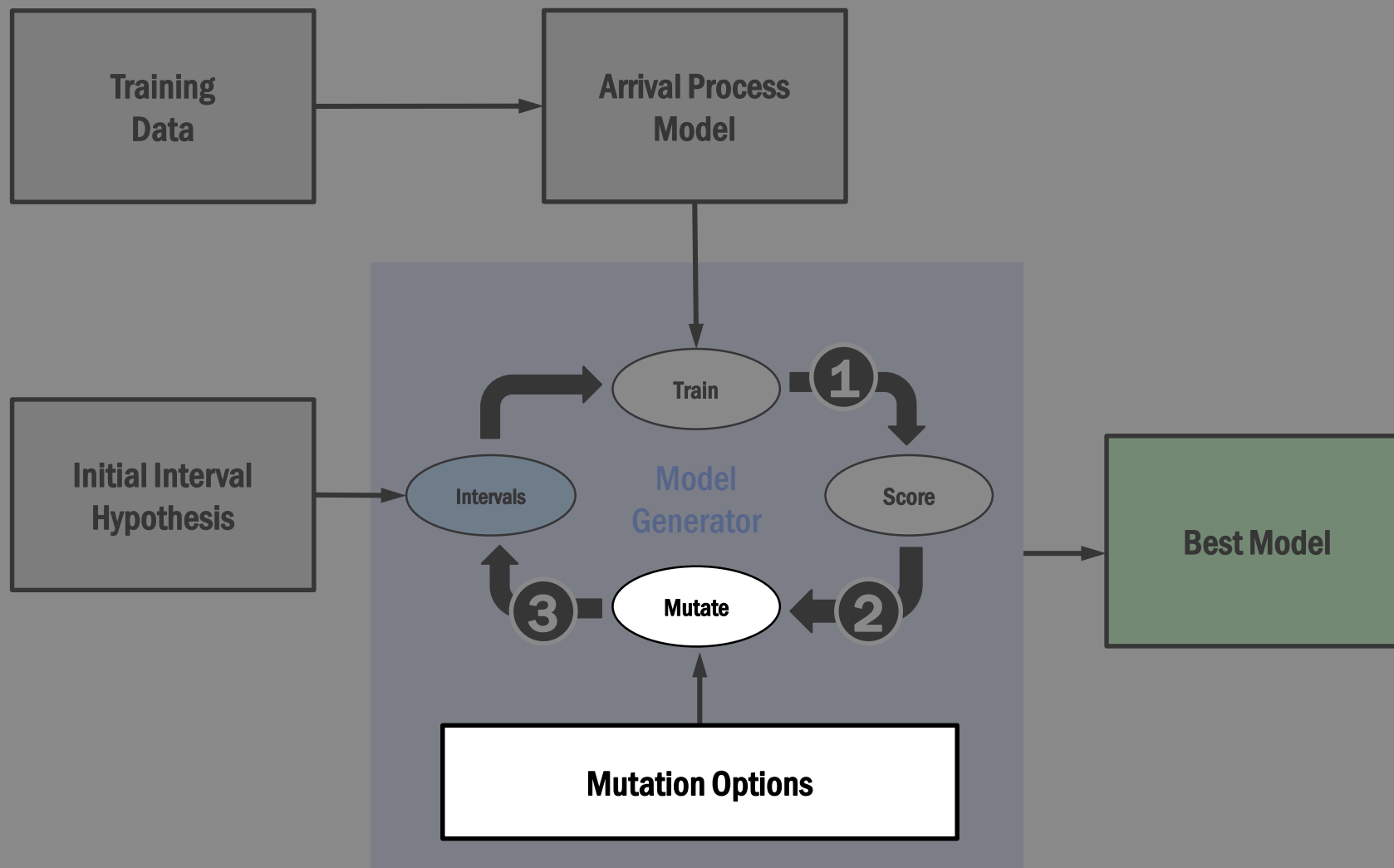


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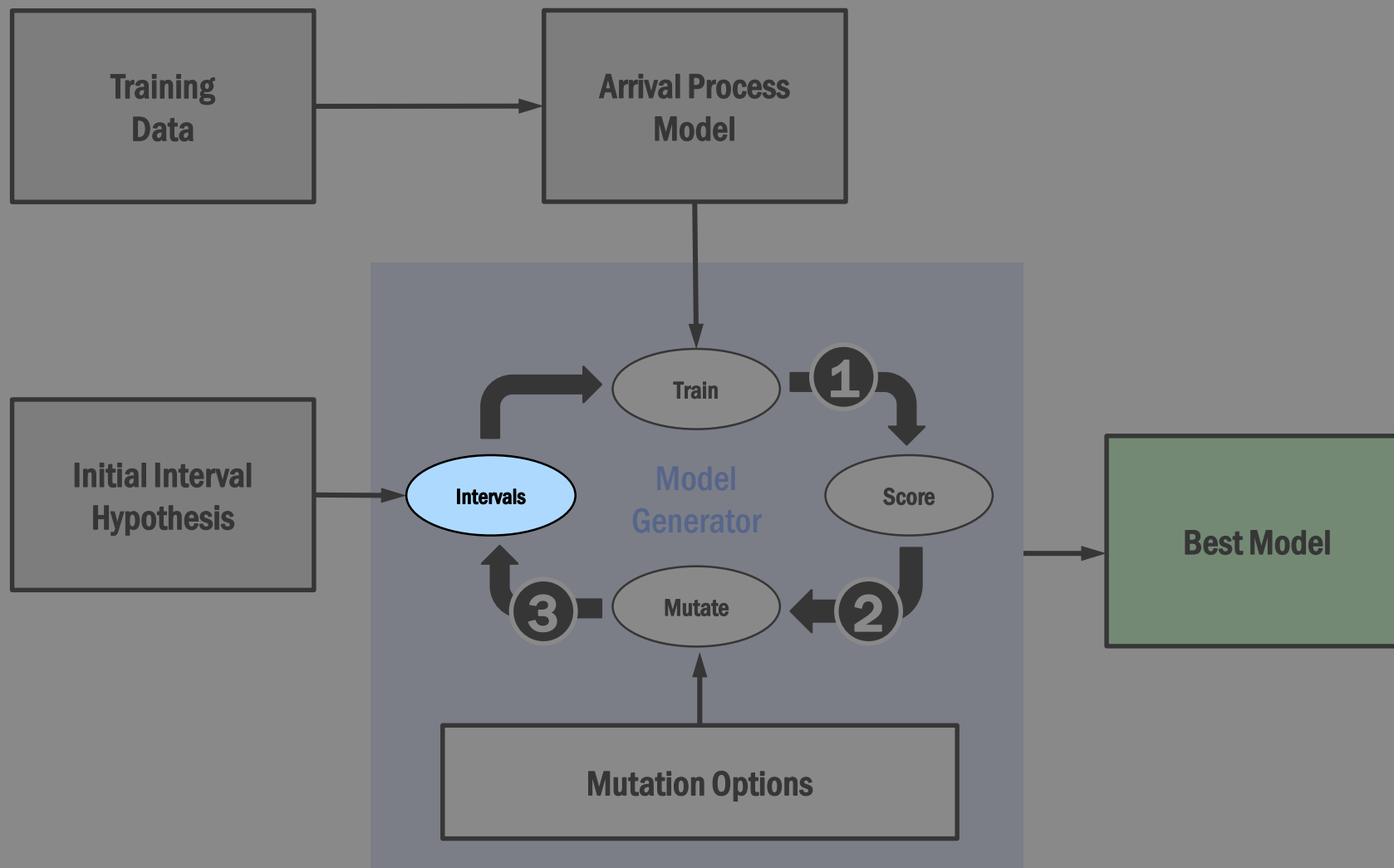


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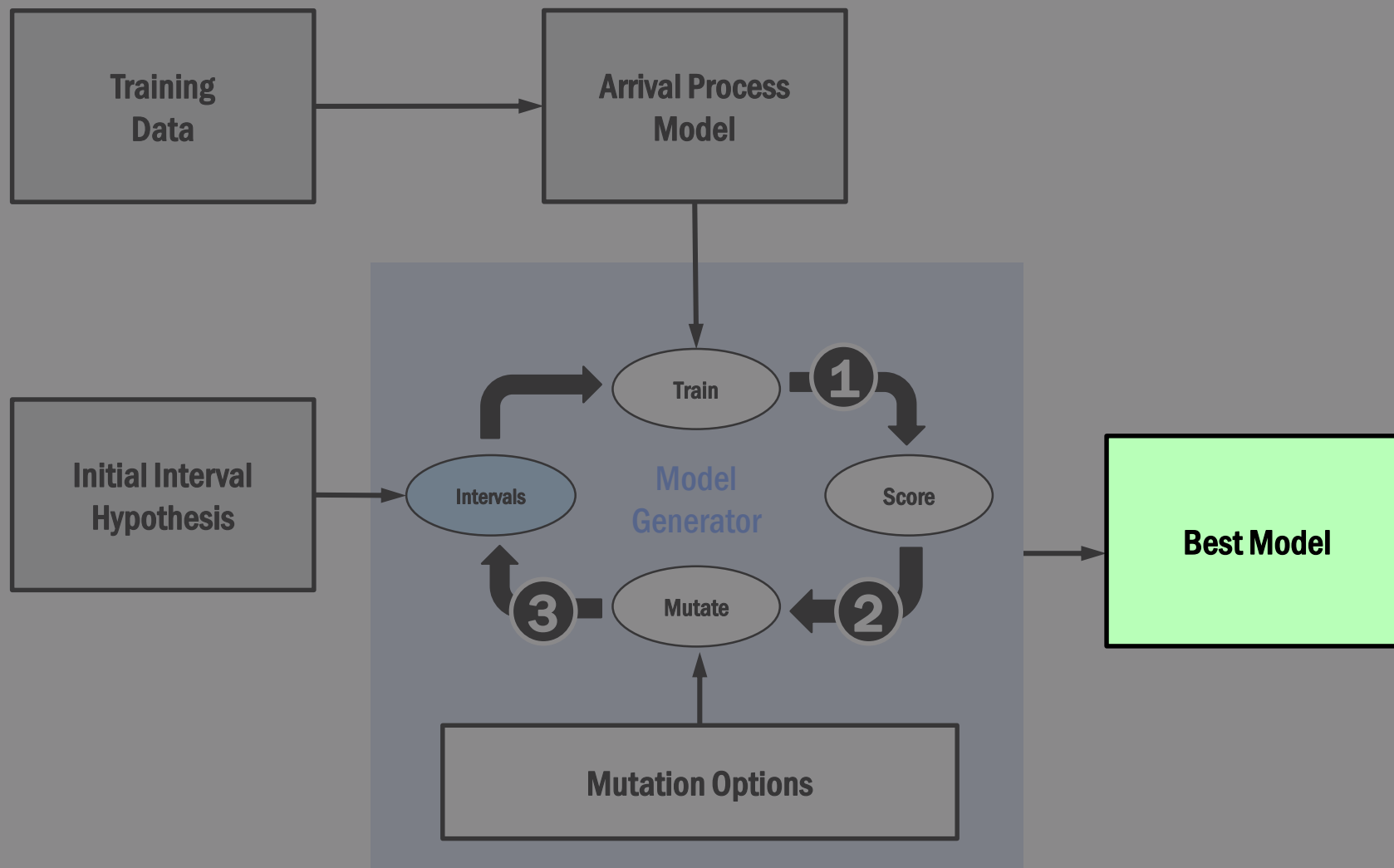
# SINAPSE Algorithm







# SINAPSE Algorithm





# Summary

- **SINAPSE algorithm fits seasonal time series with:**
  - Minimal parameter setting
  - Nonconsecutive intervals
  - Simple prediction
- **When the data has:**
  - Few samples
  - Sharp breakpoints
  - Fixed period seasonality
  - Many observations per period
- **It can be used for:**
  - Modeling seasonality
  - Predicting seasonal time series
  - Anomaly detection
- **See our poster and paper for more on applications and results**