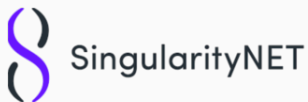


Partial Operator Induction with Beta Distribution

Nil Geisweiller

AGI-18

Prague



NOVAMENTE

Problem:

Combining Models from Different Contexts

Theory:

Solomonoff Operator Induction and Beta Distribution

Practice:

Inference Control Meta-Learning

Problem:

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Problem:

Combining Models from Different Contexts

Theory:

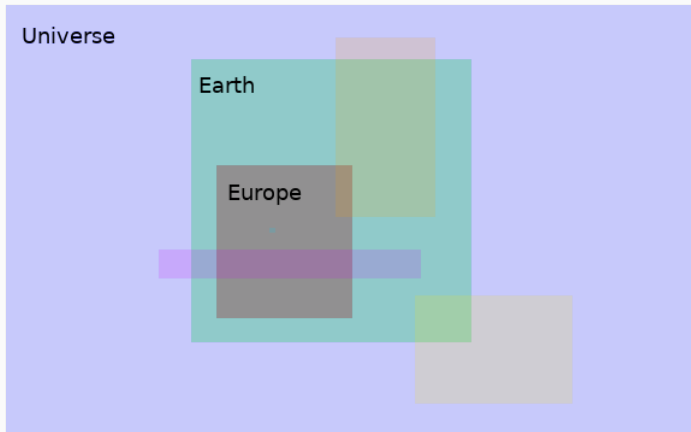
Solomonoff Operator Induction and Beta Distribution

Practice:

Inference Control Meta-Learning

Problem: Models from different contexts

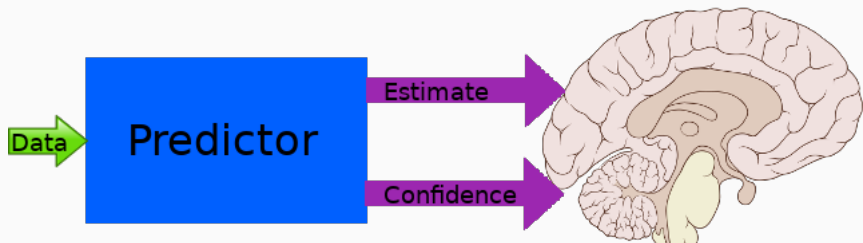
How to combine models obtained from different contexts?



Large Contexts → Underfit

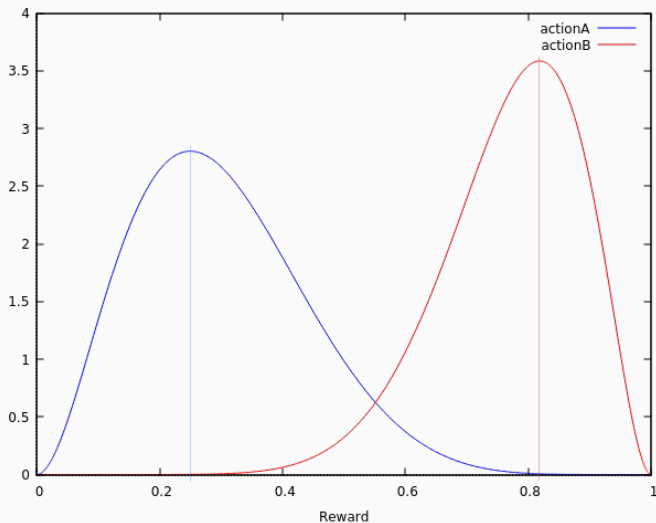
Small Contexts → Overfit

Problem: Preserve Uncertainty



Problem: Preserve Uncertainty

Exploration vs Exploitation (Thompson Sampling)



Problem: ImplicationLink

ImplicationLink <TV>

R

S

≡

Second Order

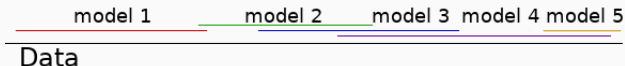
$P(S|R)$

Beta Distribution in disguise

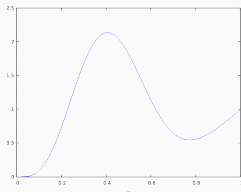
Solution

Bayesian Model Averaging / Solomonoff Operator Induction,
modified to:

1. Support partial models



2. Produce a probability distribution estimate, rather than probability estimate.



3. Specialize for Beta distributions

Problem:

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Solomonoff Operator Induction

Bayesian Model Averaging + Universal Distribution

Probability Estimate:

$$\hat{P}(A_{n+1}|Q_{n+1}) = \sum_j a_0^j \prod_{i=1}^{n+1} O^j(A_i|Q_i)$$

where:

- $Q_i = i^{th}$ question
- $A_i = i^{th}$ answer
- $O^j = j^{th}$ operator
- $a_0^j =$ prior of j^{th} operator

Specialization of Solomonoff Operator Induction

OpenCog implication link

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ImplicationLink <TV>  
  R  
  S
```

≡

Class of parameterized operators

$$O_p^j(A_i|Q_i) = \text{if } R^j(Q_i) \text{ then } \begin{cases} p, & \text{if } A_i = A_{n+1} \\ 1 - p, & \text{otherwise} \end{cases}$$

Beta Distribution

Probability Density Function:

$$pdf_{\alpha,\beta}(x) = \frac{x^{\alpha-1}(1-x)^{\beta-1}}{B(\alpha,\beta)}$$

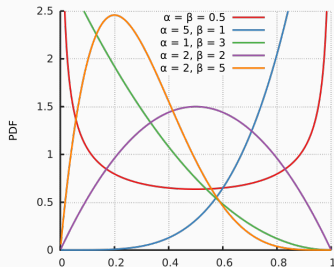
Beta Function:

$$B_x(\alpha,\beta) = \int_0^x p^{\alpha-1}(1-p)^{\beta-1} dp$$

$$B(\alpha,\beta) = B_1(\alpha,\beta)$$

Conjugate Prior:

$$pdf_{m+\alpha,n-m+\beta}(x) \propto x^m(1-x)^{n-m} pdf_{\alpha,\beta}(x)$$



Artificial Completion

$$O_p^j (A_i|Q_i) = \text{if } R^j(Q_i) \text{ then } \begin{cases} p, & \text{if } A_i = A_{n+1} \\ 1 - p, & \text{otherwise} \end{cases}$$

Data

Artificial Completion

$$O_{p,C}^j(A_i|Q_i) = \begin{cases} \text{if } R^j(Q_i) \text{ then } \begin{cases} p, & \text{if } A_i = A_{n+1} \\ 1 - p, & \text{otherwise} \end{cases} \\ \text{else } C(A_i|Q_i) \end{cases}$$

Data

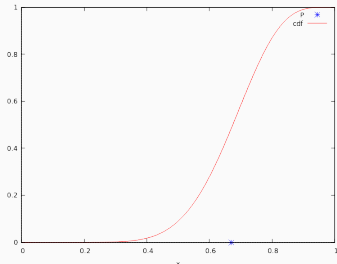
Second Order Solomonoff Operator Induction

Probability Estimate:

$$\hat{P}(A_{n+1}|Q_{n+1}) = \sum_j a_0^j \prod_{i=1}^{n+1} O^j(A_i|Q_i)$$

Probability Distribution Estimate:

$$\hat{cdf}_{(A_{n+1}|Q_{n+1})}(x) = \sum_{O^j(A_{n+1}|Q_{n+1}) \leq x} a_0^j \prod_{i=1}^n O^j(A_i|Q_i)$$



Combing Solomonoff Operator Induction and Beta Distributions

$$\hat{cdf}_{(A_{n+1}|Q_{n+1})}(x) \propto \sum_j a_0^j r^j B_x(m^j + \alpha, n^j - m^j + \beta) B(m^j + \alpha, n^j - m^j + \beta)$$

where

- n^j = number of observations explained by j^{th} model
- m^j = number of true observations explained by j^{th} model
- r^j = likelihood of the unexplained data

$r^j = ???$

Combing Solomonoff Operator Induction and Beta Distributions

$$\hat{cdf}_{(A_{n+1}|Q_{n+1})}(x) \propto \sum_j a_0^j r^j B_x(m^j + \alpha, n^j - m^j + \beta) B(m^j + \alpha, n^j - m^j + \beta)$$

where

- n^j = number of observations explained by j^{th} model
- m^j = number of true observations explained by j^{th} model
- r^j = likelihood of the unexplained data

$$r^j = ??? \approx 2^{-v^{(1-c)}}$$

- $v = n - n^j$ = number of unexplained observations
- c = compressibility parameter
 - $c = 1 \rightarrow$ explains remaining data
 - $c = 0 \rightarrow$ can't explain remaining data

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Learn how to reason efficiently

Inference Control Meta-learning

Learn how to reason efficiently

Methodology:

1. Solve sequence of problems (via reasoning)

Learn how to reason efficiently

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2. Store inference traces

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Methodology:

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4. Build control rules

Implication <TV>

And

<inference-pattern>

<rule>

<good-inference>

Learn how to reason efficiently

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<inference-pattern>

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5. **Combine control rules to guide future reasoning**

Combine Control Rules

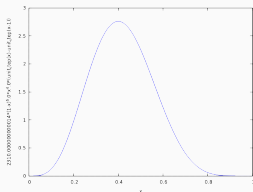
Implication <TV1>

And

<inference-pattern-1>

deduction-rule

<good-inference>



$$c = 1$$

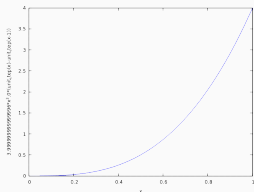
Implication <TV2>

And

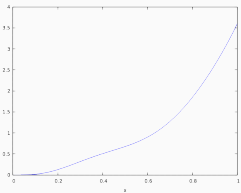
<inference-pattern-2>

deduction-rule

<good-inference>



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Combine Control Rules

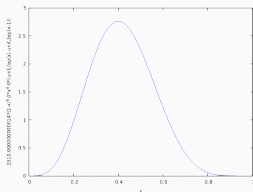
Implication <TV1>

And

<inference-pattern-1>

deduction-rule

<good-inference>



$$c = 0.5$$

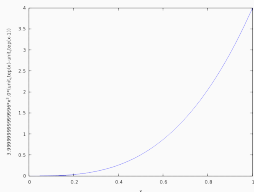
Implication <TV2>

And

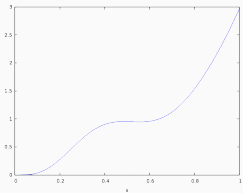
<inference-pattern-2>

deduction-rule

<good-inference>



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Combine Control Rules

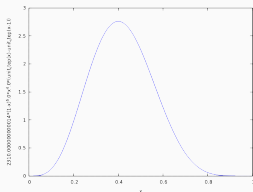
Implication <TV1>

And

<inference-pattern-1>

deduction-rule

<good-inference>



$$c = 0.1$$

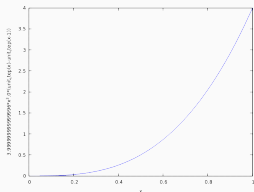
Implication <TV2>

And

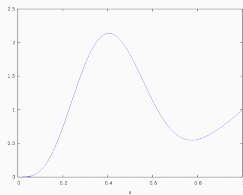
<inference-pattern-2>

deduction-rule

<good-inference>



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Conclusion

Contribution:

- Second Order Solomonoff Operator Induction
- Specialized for Beta Distribution
- Attempt to Deal with Partial Models

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Thank you!