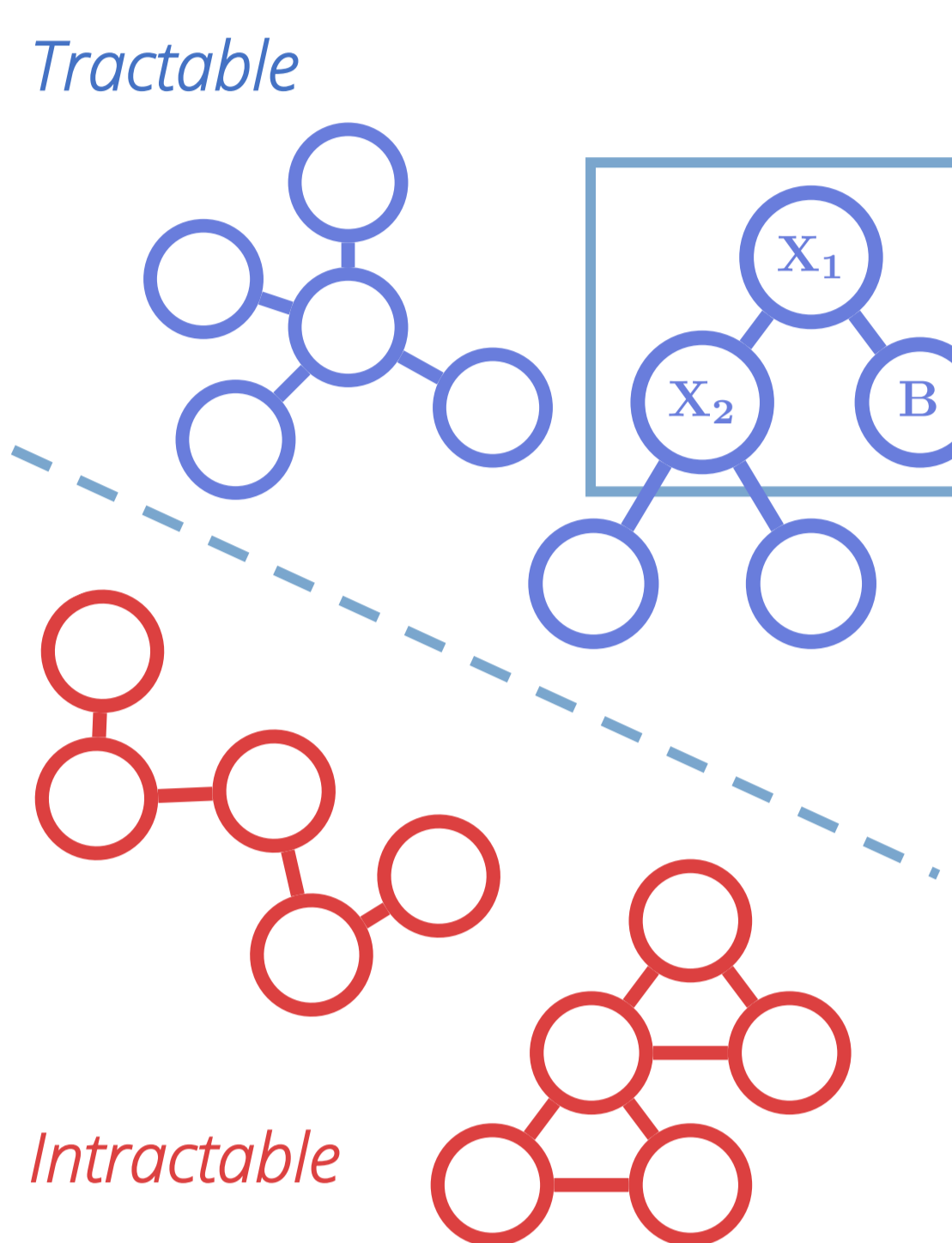


Hybrid Probabilistic Inference with Logical Constraints: Tractability and Message Passing



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Hardness : Treewidth, Diameter



Formula

$$0 \leq X_1 \leq 2$$

$$\wedge 0 \leq X_2 \leq 2$$

$$\wedge [X_1 + X_2 \leq 2]_{\ell_1}$$

$$\wedge [B \vee (1 \leq X_1)]_{\ell_2}$$

$$\wedge [B \vee (1 \leq X_1)]_{\ell_3}$$

Per-Literal Weight

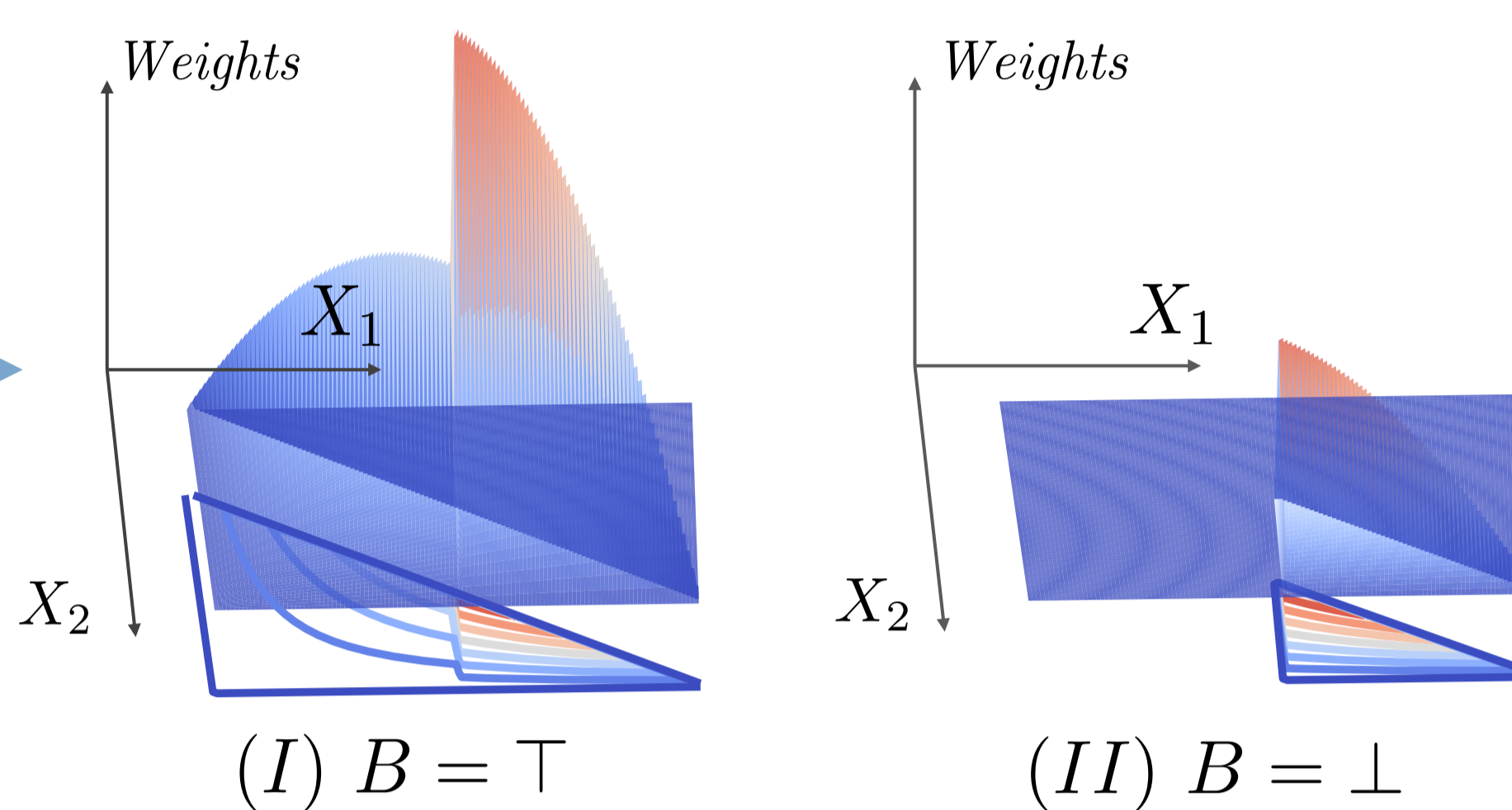
$$w_{\ell_1}(x_1, x_2) = x_1 \cdot x_2$$

$$w_{\ell_2}(B) = 3$$

$$w_{\ell_3}(x_1) = 2$$

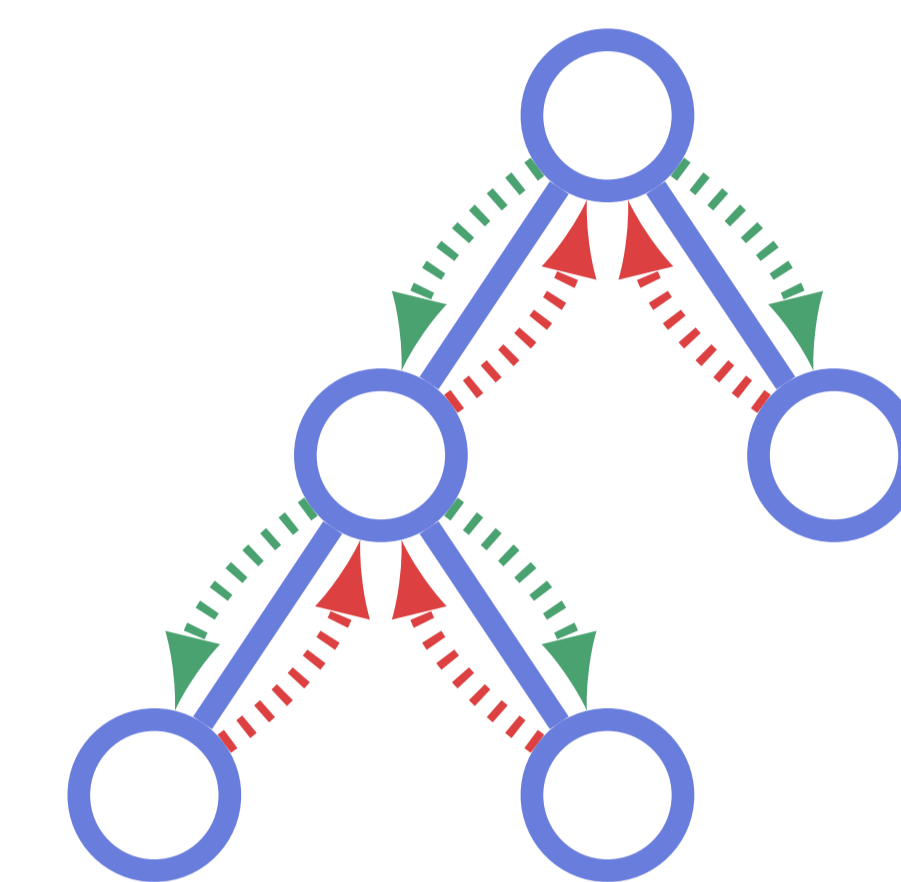
Logical Constraints : SMT(LRA)

Regions of weighted satisfying assignments



Probabilistic Inference : WMI

$$Pr(B) = \frac{WMI(I)}{WMI(I) + WMI(II)}$$



Up-/Down-ward Messages
in piecewise polynomial

WMI & MI

Weighted Model Integration (WMI) is a framework for hybrid probabilistic inference with logical constraints [1].

$$WMI(\Delta, w; \mathbf{X}, \mathbf{B}) \triangleq \sum_{\mathbf{b} \in \mathbb{B}^{|\mathbf{B}|}} \int_{\Delta(\mathbf{x}, \mathbf{b})} w(\mathbf{x}, \mathbf{b}) d\mathbf{x}$$

i.e. integration over the weighted assignments to hybrid variables, that satisfies a given logic constraint.

Logical constraints are Satisfiability Modulo Theories (SMT) formulas, as combinations of *Boolean* literals & *linear real arithmetic* (LRA) literals in Conjunctive Normal Form (CNF).

Primal graphs for SMT(LRA) formula: nodes \Leftrightarrow variables, edges \Leftrightarrow clauses.

Per-literal weights assign weight if the literal is SAT; otherwise assign one. Together they define a **joint weight**.

Answer queries

$$Pr(q) = \frac{WMI(\Delta \wedge q, w; \mathbf{X}, \mathbf{B})}{WMI(\Delta, w; \mathbf{X}, \mathbf{B})}$$

WMI-to-MI reduction [2]

- Hybrid WMI to real WMI by encoding Boolean literals into LRA literals.
- Real WMI to real MI by encoding polynomial per-literal weights into LRA constraints.

Contributions

- First we give *necessary* conditions for tractable MI problems by hardness proofs.
- Further we give *sufficient* conditions for tractability by proposing an efficient message passing scheme MP-MI for MI.

Hardness

What MI problems are intractable?

TLDR: Treewidth > 1 or Diameter $O(n)$.

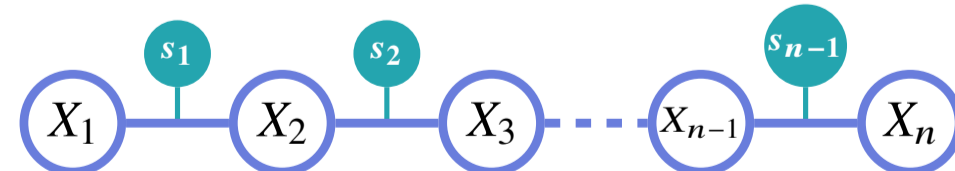
Thm. 1 Computing $MI(\Delta)$ of an SMT(LRA) formula Δ whose primal graph is a tree with diameter $O(n)$ is #P-hard, with $n = \#$ variables.

Thm. 2 Computing $MI(\Delta)$ of an SMT(LRA) formula Δ whose primal graph has treewidth two and diameter $O(\log n)$ is #P-hard, with $n = \#$ variables.

Proof sketch: Both are proved by reducing Subset Sum Problem (SSP) to MI problems:

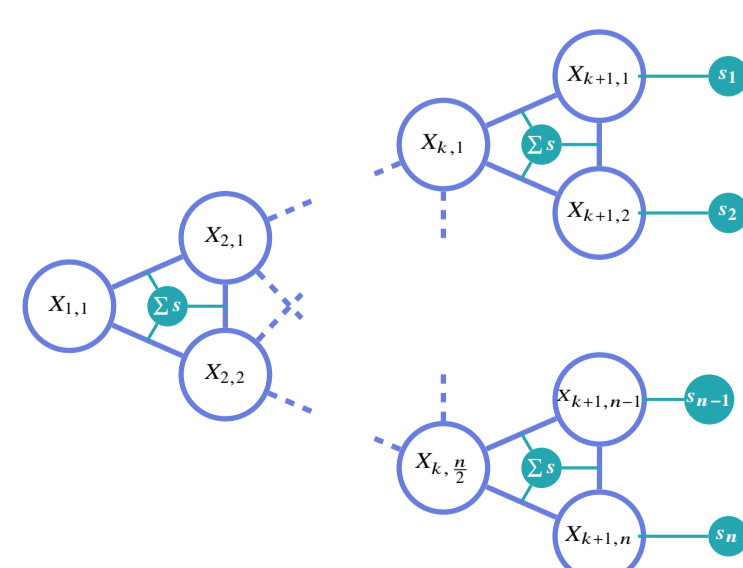
- For Thm.1, we construct an SMT(LRA) formula with primal graph as a path, s.t.

$$n^n MI(\Delta \wedge (l < X_n < u)) = \# \text{subsets}$$



- For Thm. 2, we construct an SMT(LRA) formula with primal graph having treewidth two and diameter $O(\log n)$, s.t.

$$(2n)^{2n-1} MI(\Delta \wedge (l < X_{1,1} < u)) = \# \text{subsets}$$



MP-MI

Exact message passing scheme for MI, with messages & beliefs in piecewise polynomials.

Messages

$$m_{i \rightarrow j}(x_j) = \int_{\mathbb{R}} \mathbb{1}[x_i, x_j \models \Delta_{i,j}] \mathbb{1}[x_i \models \Delta_i] \times \prod_{c \in \text{neigh}(i) \setminus \{j\}} m_{c \rightarrow i}(x_i) dx_i$$

Beliefs

$$b_i^+(x_i) = \prod_{c \in \text{ch}(i)} m_{c \rightarrow i}(x_i), \quad b_i(x_i) = b_i^-(x_i) = \prod_{c \in \text{neigh}(i)} m_{c \rightarrow i}(x_i)$$

Algorithm 1 MP-MI(Δ) – Message Passing Model Integration

```

1:  $V_{up} \leftarrow$  sort nodes in  $\mathcal{G}_\Delta$ , children before parents
2: for each  $X_i \in V_{up}$  do send-message( $X_i, X_{\text{parent}(X_i)}$ ) end
3:  $V_{down} \leftarrow$  sort nodes in  $\mathcal{G}_\Delta$ , parents before children
4: for each  $X_i \in V_{down}$  do
5:   for each  $X_c \in \text{ch}(X_i)$  do send-message( $X_i, X_c$ ) end
6: return  $\{b_i\}_{i: X_i \in \mathcal{G}_\Delta}$ 
send-message( $X_i, X_j$ )
1:  $b_i \leftarrow$  compute-beliefs
2:  $P \leftarrow$  critical-points( $b_i, \Delta_i, \Delta_{i,j}$ ),  $I \leftarrow$  intervals-from-points( $P$ )
3: for interval  $[l, u] \in I$  consistent with formula  $\Delta_i \wedge \Delta_{i,j}$  do
4:    $\langle l_s, u_s, f \rangle \leftarrow$  symbolic-bounds( $b_i, [l, u], \Delta_{i,j}$ )
5:    $f' \leftarrow \int_{l_s}^{u_s} f(x_i) dx_i$ ,  $m_{i \rightarrow j} \leftarrow m_{i \rightarrow j} \cup (l, u, f')$ 
6: return  $m_{i \rightarrow j}$ 

```

When the message passing scheme is done, the MI can be computed by

$$MI(\Delta) = \int_{\mathbb{R}} \mathbb{1}[x_i \models \Delta_i] \cdot b_i(x_i) dx_i.$$

And also amortize queries:

Node queries $MI(\Delta \wedge q) = \int_{\mathbb{R}} \mathbb{1}[x_i \models q] \mathbb{1}[x_i \models \Delta_i] b_i(x_i) dx_i.$

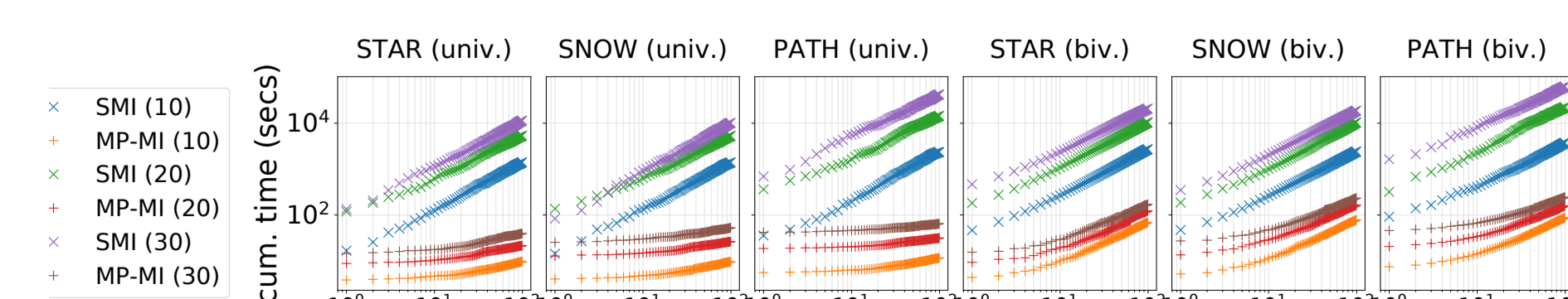
Edge queries

$$m_{j \rightarrow i}^*(x_i) = \int_{\mathbb{R}} b_j(x_j) / m_{i \rightarrow j}(x_j) \times \mathbb{1}[x_i, x_j \models \Delta_{i,j} \wedge q] \mathbb{1}[x_j \models \Delta_j] dx_j$$

Moment queries $\mathbb{E}[X_i^k] = \frac{1}{MI(\Delta)} \int_{\mathbb{R}} \mathbb{1}[x_i \models \Delta_i] \times x_i^k b_i(x_i) dx_i$

Experiments

Time for answer multiple queries to SMT(LRA) formulas with tree primal graphs of diameters $O(1)$, $O(\log n)$ & $O(n)$ respectively.



[Paper available here](#)

KR2ML @ NeurIPS 19,
Vancouver, Canada

References: [1] Vaishak Belle, Andrea Passerini, and Guy Van den Broeck. Probabilistic inference in hybrid domains by weighted model integration. IJCAI, 2015.
 [2] Zhe Zeng and Guy Van den Broeck. Efficient search-based weighted model integration. UAI, 2019.

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