# Infrared: a declarative tree decomposition-powered framework for bioinformatics 

Additional file 1

Hua-Ting Yao, Bertrand Marchand, Sarah J. Berkemer, Yann Ponty, Sebastian Will

Recall the following few definitions and notation from the main text. Given the cluster tree $(T, \chi, \phi)$ and a node $u$ of T. If $u$ has the parent $v$, then $\operatorname{diff}(u)=\chi(u) \backslash \chi(v)$ and $\operatorname{sep}(u)=\chi(v) \cap \chi(u)$; otherwise, if $u$ is the root, $\operatorname{diff}(u)=\chi(u)$ and $\operatorname{sep}(u)=\emptyset$. Moreover, we defined $T_{u}$ as the subtree under node $u$, $\chi\left(T_{u}\right)=\bigcup_{u^{\prime} \in T_{u}} \chi\left(u^{\prime}\right)$ and $\operatorname{diff}\left(T_{u}\right)=\chi\left(T_{u}\right) \backslash \operatorname{sep}(u)$. We use the following lemma.
Lemma S1. $\left\{\operatorname{diff}\left(T_{c}\right) \mid c\right.$ child of $\left.u\right\}$ is a partition of $\chi\left(T_{u}\right) \backslash \chi(u)=\operatorname{diff}\left(T_{u}\right) \backslash \operatorname{diff}(u)$.
Proof. Let $u$ be a node of the cluster tree $(T, \chi, \phi), c$ and $c^{\prime}$ be children of $u$, then $\operatorname{diff}\left(T_{c}\right) \cap \operatorname{diff}\left(T_{c^{\prime}}\right)=\emptyset$, since-due to tree decomposition property (3) -any common element must be in $\chi(u)$ in contradiction to the definition of diff on subtrees; moverover,

$$
\begin{aligned}
\bigcup\left\{\operatorname{diff}\left(T_{c}\right) \mid c \text { child of } u\right\} & =\bigcup_{c \text { child of } u} \chi(c) \backslash \operatorname{sep}(c) \\
& =\bigcup_{c \text { child of } u} \chi(c) \backslash \bigcup_{c \text { child of } u} \operatorname{sep}(c) \\
& =\bigcup_{c \text { child of } u} \chi(c) \backslash \chi(u)=\operatorname{diff}\left(T_{u}\right) \backslash \operatorname{diff}(u) .
\end{aligned}
$$

## 1 Correctness of optimization

Proof of Proposition 1. We are going to show the correctness by structural induction on the cluster tree. For the base case, let $u$ be a leaf of the cluster tree. Sinc the subtree $T_{u}$ is the node $u$ itself, Eq. 3 for a leaf $u$ is written as

$$
m_{u \rightarrow v}(\bar{x})=\max _{\substack{\bar{x}^{\prime} \in \mathcal{A}_{\text {diff }(u)} \\ \bar{x} \cup \bar{x}^{\prime} \text { is valid }}} \sum_{f \in \phi(u)} \alpha_{f} f\left(\bar{x} \cup \bar{x}^{\prime}\right) .
$$

Thus, Proposition 1 holds for leaves $u$, since they do not have children.
For the induction step, let $u$ be an internal node of the cluster tree and assume Proposition 1 holds for all children of $u$. Denote the children of $u$ as $c_{1} \ldots, c_{k}$. We directly apply the induction hypothesis to the r.h.s of the claim.

Our goal is to show equivalence to the r.h.s. of Eq. (3). By Lemma S1 set diff $\left(T_{u}\right) \backslash \operatorname{diff}(u)$ is disjointly decomposed into diff $\left(T_{c_{1}}\right) \ldots \operatorname{diff}\left(T_{c_{k}}\right)$; moreover, $\phi\left(T_{u}\right) \backslash \phi(u)$ is partitioned into $\phi\left(T_{c_{i}}\right), i=1 \ldots k$. It follows that for every fixed assignment $\bar{x} \cup \bar{x}^{\prime}$, the maximizations over $\bar{y}_{i} \in \operatorname{diff}\left(T_{c_{i}}\right)$ are independent for different $i$; as well the sum over the functions of bag $u$ is independent of the maximizations. Thus, we can move the maximizations over the assignments $\bar{y}_{i}$ in front of the summations.

$$
=\max _{\substack{\bar{x}^{\prime} \in \mathcal{A}_{\text {diff }(u)} \\
\bar{x} \cup \bar{x}^{\prime} \text { is valid }}} \max _{\substack{\bar{y}_{1} \in \mathcal{A}_{\text {diff }}\left(T_{\left.c_{1}\right)}\right) \\
\bar{x} \cup \bar{x}^{\prime} \cup \bar{y}_{1} \text { is valid }}} \max _{\substack{\bar{y}_{k} \in \mathcal{A}_{\text {diff( } \left.T_{\left.c_{k}\right)}\right)} \\
\bar{x} \cup \bar{x}^{\prime} \cup \bar{y}_{k} \text { is } \text { valid }}}\left[\begin{array}{l}
\sum_{f \in \phi(u)} \alpha_{f} f\left(\bar{x} \cup \bar{x}^{\prime}\right) \\
+\sum_{c_{i} \text { child of } u} \sum_{f \in \phi\left(T_{\left.c_{i}\right)}\right)} \alpha_{f} f\left(\bar{x} \cup \bar{x}^{\prime} \cup \bar{y}_{i}\right)
\end{array}\right]
$$

Then, we combine the maximizations over the single partial assignments $\bar{y}_{i}$; moreover we replace the sums with a single sum ranging over the combined assignments $\bar{y}$ of $\operatorname{diff}\left(T_{u}\right) \backslash \operatorname{diff}(u)$.

$$
=\max _{\substack{\bar{x}^{\prime} \in \mathcal{A} \\ \bar{x} \cup \bar{x}^{\prime} \cup \bar{y} \text { if }(u) \\ \bar{y} \in \mathcal{A}_{\text {diflid }}}} \max _{\substack{x} \bar{x}_{u}^{\prime} \cup \backslash \text { diff } \cup(u)}\left[\sum_{f \in \phi\left(T_{u}\right)} \alpha_{f} f\left(\bar{x} \cup \bar{x}^{\prime} \cup \bar{y}\right)\right]
$$

The last term equals r.h.s. of Eq. (3), showing the induction step.

## 2 Correctness of partition function computation

Proof of Proposition 2. Again, we will show the correctness by induction. Starting with the base case where $u$ is a leaf of the cluster tree, the subtree $T_{u}$ is the node $u$ and the child set of $u$ is empty. Eq 5 is then written as

$$
m_{u \rightarrow v}(\bar{x})=\sum_{\substack{\bar{x}^{\prime} \in \mathcal{A}_{\text {diff }(u)} \\ \bar{x} \cup \bar{x}^{\prime} \text { is valid }}} \prod_{f \in \phi(u)} \exp \left(\alpha_{f} f\left(\bar{x} \cup \bar{x}^{\prime}\right)\right)
$$

which shows the correctness for leaves.
For the induction step, let $u$ be an internal node of the cluster tree; we assume that the message $m_{c_{i} \rightarrow u}$ for all children $c_{i}, i \in\{1, \ldots, k\}$ of $u$ is correctly computed as in Proposition 2 . Following the scheme as in the previous proof, we first rewrite the r.h.s of the claim applying the induction hypothesis,

$$
\begin{aligned}
& \text { r.h.s. (6) }= \\
& \sum_{\substack{\bar{x}^{\prime} \in \mathcal{A}_{\text {diff }(u)} \\
\bar{x} \cup \bar{x}^{\prime} \text { is valid }}}\left[\begin{array}{l}
\prod_{f \in \phi(u)} \exp \left(\alpha_{f} f\left(\bar{x} \cup \bar{x}^{\prime}\right)\right) \\
\times \prod_{c_{i} \text { child of } u} \sum_{\substack{\left.\bar{y}_{i} \in \mathcal{A}_{\text {diff }} \\
\bar{x} \cup \bar{x}_{c_{i}}\right) \\
\bar{y}_{i} \text { is valid }}} \prod_{f \in \phi\left(T_{c_{i}}\right)} \exp \left(\alpha_{f} f\left(\bar{x} \cup \bar{x}^{\prime} \cup \bar{y}_{i}\right)\right)
\end{array}\right]
\end{aligned}
$$

As the set $\operatorname{diff}\left(T_{u}\right) \backslash \operatorname{diff}(u)$ is disjointly decomposed into $\operatorname{diff}\left(T_{c_{1}}\right) \ldots \operatorname{diff}\left(T_{c_{k}}\right)$, the second term in the bracket can be seen as the Cartesian product of $k$ sets. We can move the summations in front of the production,

For every fixed assignment $\bar{x} \cup \bar{x}^{\prime}$, the product over the functions of bag $u$ is independent of the summations. Thus, we can move the summations over the assignments $\bar{y}_{i}$ in front of the product,

$$
=\sum_{\substack{\bar{x}^{\prime} \in \mathcal{A}_{\text {diff }(u)} \\
\bar{x} \cup \bar{x}^{\prime} \text { is valid }}} \sum_{\substack{\bar{y}_{1} \in \mathcal{A}_{\text {diff }\left(T_{c_{1}}\right)} \\
\bar{x} \cup \bar{x}^{\prime} \cup \bar{y}_{1} \text { is valid }}} \ldots \sum_{\substack{\bar{y}_{k} \in \mathcal{A}_{\text {diff }\left(T_{c_{k}}\right)} \\
\bar{x} \cup \bar{x}^{\prime} \cup \bar{y}_{k} \text { is valid }}}\left[\begin{array}{l}
\prod_{f \in \phi(u)} \exp \left(\alpha_{f} f\left(\bar{x} \cup \bar{x}^{\prime}\right)\right) \\
\times \prod_{c_{i} \text { child of } u} \prod_{f \in \phi\left(T_{\left.c_{i}\right)}\right)} \exp \left(\alpha_{f} f\left(\bar{x} \cup \bar{x}^{\prime} \cup \bar{y}_{i}\right)\right)
\end{array}\right]
$$

Then, we combine the summations over the single partial assignments $\bar{y}_{i}$; moreover we replace the products with a single product ranging over the combined assignments $\bar{y}$ of $\operatorname{diff}\left(T_{u}\right) \backslash \operatorname{diff}(u)$ as $\phi\left(T_{u}\right) \backslash \phi(u)$ is partitioned into $\phi\left(T_{c_{i}}\right), i=1 \ldots k$.

$$
\left.=\sum_{\substack{\bar{x}^{\prime} \in \mathcal{A}_{\text {diff }(u)} \\ \bar{x} \cup \bar{x}^{\prime} \cup \bar{y} \text { is valid }}} \sum_{\substack{\bar{y} \in \mathcal{A}_{\text {diff }\left(T_{u}\right) \backslash \text { diff }(u)}}}\left[\prod_{f \in \phi\left(T_{u}\right)} \exp \left(\alpha_{f} f\left(\bar{x} \cup \bar{x}^{\prime} \cup \bar{y} \text { is valid }\right\} \bar{y}\right)\right)\right]
$$

The last term equals r.h.s. of Eq. (5), showing the induction step.

## 3 Correctness of sampling by stochastic traceback

Lemma S2. At edge $u \rightarrow v$, Alg. 3 choses a partial assignment $\bar{x} \in \mathcal{A}_{\operatorname{diff}(u)}$ with probability

$$
\mathbb{P}_{\mathrm{alg}}(\bar{x} \mid x)=\frac{w_{u \rightarrow v}(\bar{x} \mid x)}{m_{u \rightarrow v}(x)}
$$

where $x$ is the partial assignment determined in the previous iterations and weight $w_{u \rightarrow v}(\bar{x} \mid x)$ is defined as
$w_{u \rightarrow v}(\bar{x} \mid x)=\prod_{f \in \phi(u)} \exp \left(\alpha_{f} f(x \cup \bar{x})\right) \prod_{c \text { child of } u} m_{c \rightarrow u}(x \cup \bar{x})=\sum_{\substack{\bar{y} \in \mathcal{A}_{\text {diff }\left(T_{u}\right) \backslash \text { diff }}^{x \cup u)} \\ x \cup \bar{x} \cup \bar{y} \text { is valid }}} \prod_{f \in \phi\left(T_{u}\right)} \exp \left(\alpha_{f} f(x \cup \bar{x} \cup \bar{y})\right)$.

Proof. Let us order the partial assignments $\bar{x}_{i}$ in $\mathcal{A}_{\text {diff }(u)}$ with $i$ from 1 to $\left|\mathcal{A}_{\text {diff }(u)}\right|$, each is associated with weight $w_{u \rightarrow v}\left(\bar{x}_{i} \mid x\right)=\prod_{f \in \phi(u)} \exp \left(\alpha_{f} f\left(x \cup \bar{x}_{i}\right)\right) \prod_{c \text { child of } u} m_{c \rightarrow u}\left(x \cup \bar{x}_{i}\right)$. Note that the sum of all weights $\sum_{i=1}^{\left|\mathcal{A}_{\text {diff }}(u)\right|} w_{u \rightarrow v}\left(\bar{x}_{i} \mid x\right)$ equals to $m_{u \rightarrow v}(x)$.

For a selected value $t$, Alg. 3 enumerates assignments $\bar{x}_{i}$ starting from $i=1$. In every iteration, it subtracts the weight $w_{u \rightarrow v}\left(\bar{x}_{i} \mid x\right)$. It is terminated as soon as $t$ is negative, let this index be $j$, and assigns $\bar{x}_{j}$ to $\operatorname{diff}(u)$. In other words, $\bar{x}_{j}$ is assigned to $\operatorname{diff}(u)$ if

$$
\sum_{i=0}^{j} w_{u \rightarrow v}\left(\bar{x}_{i} \mid x\right) \leq t<\sum_{i=0}^{j+1} w_{u \rightarrow v}\left(\bar{x}_{i} \mid x\right)
$$

with $w_{u \rightarrow v}\left(\bar{x}_{0} \mid x\right)$ is defined as 0 for $i=0$. Since $t$ is randomly and uniformly selected from 0 to $m_{u \rightarrow v}(x)$, the probability that Alg. 3 chooses $\bar{x}$ is

$$
\mathbb{P}_{\mathrm{alg}}(\bar{x} \mid x)=\frac{w_{u \rightarrow v}(\bar{x} \mid x)}{m_{u \rightarrow v}(x)}
$$

Finally, the weight $w_{u \rightarrow v}(\bar{x} \mid x)=\prod_{f \in \phi(u)} \exp \left(\alpha_{f} f(x \cup \bar{x})\right) \prod_{c \text { child of } u} m_{c \rightarrow u}(x \cup \bar{x})$ can be seen as a special case of Eq. 6 by replacing the partial assignment set $\mathcal{A}_{\text {diff }(u)}$ with a singleton $\{\bar{x}\}$. Applying Proposition 2 gives

$$
w_{u \rightarrow v}(\bar{x} \mid x)=\sum_{\substack{\bar{y} \in \mathcal{A}_{\text {diff }}\left(T_{u}\right) \backslash \text { diff }(u) \\ x \cup \bar{x} \cup \bar{y} \text { is valid }}} \prod_{f \in \phi\left(T_{u}\right)} \exp \left(\alpha_{f} f(x \cup \bar{x} \cup \bar{y})\right) .
$$

Proof of Proposition 3. Let us assume that the non-root nodes of the cluster tree are labeled in preorder as $u_{1}, u_{2}, \ldots, u_{\ell}$. Alg. 3 assigns values in the order of $\operatorname{diff}\left(u_{1}\right), \ldots, \operatorname{diff}\left(u_{\ell}\right)$; it finally generates a total assignment since $\cup_{i=1}^{\ell} \operatorname{diff}\left(u_{i}\right)=\mathcal{X}$.

Let $x^{*}=\bar{x}_{1}^{*} \cup \cdots \cup \bar{x}_{\ell}^{*}$ be a total assignment where $\bar{x}_{i}^{*}$ is a partial assignment of $\mathcal{A}_{\text {diff }\left(u_{i}\right)}$ for $i$ from 1 to $\ell$. The probability of $x^{*}$ can be written as product of conditional probabilities

$$
\mathbb{P}\left(x^{*}\right)=\mathbb{P}\left(\bar{x}_{1}^{*}\right) \mathbb{P}\left(\bar{x}_{2}^{*} \mid \bar{x}_{1}^{*}\right) \cdots \mathbb{P}\left(\bar{x}_{\ell}^{*} \mid \cup_{k=1}^{\ell-1} \bar{x}_{k}^{*}\right)=\cup_{i=1}^{\ell} \mathbb{P}\left(\bar{x}_{i}^{*} \mid \cup_{k=1}^{i-1} \bar{x}_{k}^{*}\right)
$$

with the conditional Boltzmann probability

$$
\begin{align*}
& \sum_{\bar{x} \in \mathcal{A}_{\text {diff }\left(u_{i+1}\right) \cup \ldots \cup \operatorname{diff}\left(u_{\ell}\right)}} \prod_{f \in \phi_{\text {all }}} \exp \left(\alpha_{f} f\left(\cup_{k=1}^{i-1} \bar{x}_{k}^{*} \cup \bar{x}_{i}^{*} \cup \bar{x}\right)\right) \\
& \mathbb{P}\left(\bar{x}_{i}^{*} \mid \cup_{k=1}^{i-1} \bar{x}_{k}^{*}\right)=\frac{\bar{x}^{\bar{x} \text { is valid }}}{\sum_{\substack{\bar{x}^{\prime} \in \mathcal{A}_{\text {diff }\left(u_{i}\right)} \\
\bar{x}^{\prime} \text { is valid }}}^{\substack{\bar{x} \in \mathcal{A}_{\text {diff }}\left(u_{i+1}\right) \cup \ldots \text { Udiff }\left(u_{\ell}\right) \\
\bar{x} \text { is valid }}} \prod_{f \in \phi_{\text {all }}} \exp \left(\alpha_{f} f\left(\cup_{k=1}^{i-1} \bar{x}_{k}^{*} \cup \bar{x}^{\prime} \cup \bar{x}\right)\right)} \tag{S2}
\end{align*}
$$

where $\phi_{\text {all }}$ is the set of all network functions in the feature network. Note that for $i=1$, the union $\cup_{k=1}^{0} x_{k}^{*}$ is empty.

At the edge $u_{i} \rightarrow v$, where the variables $\operatorname{diff}\left(u_{1}\right), \ldots, \operatorname{diff}\left(u_{i-1}\right)$ have been assigned by $\bar{x}_{1}^{*}, \ldots, \bar{x}_{i-1}^{*}$, the probability that Alg. 3 chooses $\bar{x}_{i}^{*}$ for $\operatorname{diff}\left(u_{i}\right)$ is given by Lemma S2 as

$$
\begin{equation*}
\mathbb{P}_{\mathrm{alg}}\left(\bar{x}_{i}^{*} \mid \cup_{k=1}^{i-1} \bar{x}_{k}^{*}\right)=\frac{w_{u_{i} \rightarrow v}\left(\bar{x}_{i}^{*} \mid \cup_{k=1}^{i-1} \bar{x}_{k}^{*}\right)}{m_{u_{i} \rightarrow v}\left(\cup_{k=1}^{i-1} \bar{x}_{k}^{*}\right)} . \tag{S3}
\end{equation*}
$$

Our goal is to show that two conditional probabilities $\mathbb{P}\left(\bar{x}_{i}^{*} \mid \cup_{k=1}^{i-1} \bar{x}_{k}^{*}\right)$ of Eq. S2 and $\mathbb{P}_{\text {alg }}\left(\bar{x}_{i}^{*} \mid \cup_{k=1}^{i-1} \bar{x}_{k}^{*}\right)$ of Eq. S3 are actually the same.

We rewrite $\mathbb{P}_{\text {alg }}$ of Eq. S 3 using Eq. S 1 and the definition of conditional partition function as in Eq. 5

$$
\begin{aligned}
& \mathbb{P}_{\mathrm{alg}}\left(\bar{x}_{i}^{*} \mid \cup_{k=1}^{i-1} \bar{x}_{k}^{*}\right)
\end{aligned}
$$

Comparing with $\mathbb{P}\left(\bar{x}_{i}^{*} \mid \cup_{k=1}^{i-1} \bar{x}_{k}^{*}\right)$ of Eq. S2 only a subset of network functions $\phi\left(T_{u}\right)$ are taken into account. We are going to bring the rest into the expression. First, let us consider the functions assigned to the nodes that Alg. 3 has visited before edge $u_{i} \rightarrow v$, denoted by $\phi_{\text {visited }}:=\cup_{k=1}^{i-1} \phi\left(u_{k}\right)$. Multiplying nominator and denominator by partial partition function $\prod_{f \in \phi_{\text {visited }}} \exp \left(\alpha_{f} f\left(\cup_{k=1}^{i-1} \bar{x}_{k}^{*}\right)\right)$ restricted to $\phi_{\text {visited }}$ and the partial assignment $\cup_{k=1}^{i-1} \bar{x}_{k}^{*}$ gives

$$
=\frac{\left(\sum_{\substack{\bar{y} \in \mathcal{A}_{\text {diff }\left(T_{u_{i}}\right) \backslash \operatorname{diff}\left(u_{i}\right)} \\ x \cup \bar{x}_{i}^{*} \cup \bar{y} \text { is valid }}} \prod_{f \in \phi\left(T_{u_{i}}\right)} \exp \left(\alpha_{f} f\left(\cup_{k=1}^{i-1} \bar{x}_{k}^{*} \cup \bar{x}_{i}^{*} \cup \bar{y}\right)\right)\right)\left(\prod_{f \in \phi_{\text {visited }}} \exp \left(\alpha_{f} f\left(\cup_{k=1}^{i-1} \bar{x}_{k}^{*}\right)\right)\right)}{\left(\sum_{\substack{\bar{x}^{\prime} \in \mathcal{A}_{\text {diff }\left(u_{i}\right)} \\ \bar{x}^{\prime} \text { is valid }}} \sum_{\substack{\bar{y} \in \mathcal{A}_{\text {diff }\left(T_{u_{i}}\right) \backslash \operatorname{diff}\left(u_{i}\right)}^{x \cup \bar{x}^{\prime} \cup \bar{y} \text { is valid }}}} \prod_{f \in \phi\left(T_{\left.u_{i}\right)}\right)} \exp \left(\alpha_{f} f\left(\cup_{k=1}^{i-1} \bar{x}_{k}^{*} \cup \bar{x}^{\prime} \cup \bar{y}\right)\right)\right)\left(\prod_{f \in \phi_{\text {visited }}} \exp \left(\alpha_{f} f\left(\cup_{k=1}^{i-1} \bar{x}_{k}^{*}\right)\right)\right)}
$$

As $\operatorname{diff}\left(T_{u_{i}}\right)$ is disjoint to the dependency of any function of $\phi_{\text {visited }}$, we can further group the products

We consider now the rest of the network functions, denoted by $\phi_{\text {rest }}$. These functions are assigned to the cluster nodes, denoted by $\mathcal{U}_{\text {rest }}$, that have not been visited before the edge $u_{i} \rightarrow v$ or are not in the subtree of $u_{i}$. Note that $T_{u_{i}} \cup \mathcal{U}_{\text {rest }}=\left\{u_{i}, \ldots, u_{\ell}\right\}$. The partial partition function restricted to $\phi_{\text {rest }}$ for fixed partial assignment $\cup_{k=1}^{i-1} \bar{x}_{k}^{*}$ is

$$
\sum_{\substack{\bar{y}^{\prime} \in \mathcal{A}_{\cup_{u} \in \mathcal{U}_{\text {rest }}}^{\bar{y}^{\prime} \text { if valid }(u)}}} \prod_{f \in \phi_{\text {rest }}} \exp \left(\alpha_{f} f\left(\cup_{k=1}^{i-1} \bar{x}_{k}^{*} \cup \bar{y}^{\prime}\right)\right) .
$$

Multiplying again the partial partition function on nominator and denominator gives

$$
\begin{aligned}
& \left(\sum_{\substack{\bar{y} \in \mathcal{A}_{\text {diff }}\left(T_{u_{i}}\right) \backslash \text { diff }\left(u_{i}\right) \\
x \cup \bar{x}_{i}^{*} \cup \bar{y} \text { is } \\
\text { valid }}} \prod_{f \in \phi_{\text {visited }} \cup \phi\left(T_{u_{i}}\right)} \exp \left(\alpha_{f} f\left(\cup_{k=1}^{i-1} \bar{x}_{k}^{*} \cup \bar{x}_{i}^{*} \cup \bar{y}\right)\right)\right) \\
& \left(\begin{array}{c}
\sum_{\bar{y}^{\prime} \in \mathcal{A}_{\cup_{u} \in \mathcal{U}_{\text {rest }} \operatorname{diff}(u)}}^{\bar{y}^{\prime} \text { is valid }} \\
\left.\prod_{t \in \phi_{\text {rest }}} \exp \left(\alpha_{f} f\left(\cup_{k=1}^{i-1} \bar{x}_{k}^{*} \cup \bar{y}^{\prime}\right)\right)\right) \\
\end{array}\right. \\
& =\frac{\bar{y}^{\prime} \text { is valid }}{\left(\begin{array}{c}
\left.\sum_{\substack{\bar{x}^{\prime} \in \mathcal{A}_{\text {diff( }} \\
\bar{x}^{\prime} \text { is valid }}} \sum_{\substack{\bar{y} \in \mathcal{A}_{\text {diff }}\left(T_{u_{i}}\right) \backslash \operatorname{diff}\left(u_{i}\right) \\
x \cup \bar{x}^{\prime} \cup \bar{y} \text { is valid }}} \prod_{f \in \phi_{\text {visited }} \cup \phi\left(T_{u_{i}}\right)} \exp \left(\alpha_{f} f\left(\cup_{k=1}^{i-1} \bar{x}_{k}^{*} \cup \bar{x}^{\prime} \cup \bar{y}\right)\right)\right)
\end{array}\right)} \\
& \left(\sum_{\substack{\bar{y}^{\prime} \in \mathcal{A}_{\cup_{u} \in \mathcal{U}_{\text {rest }} \operatorname{diff}(u)} \\
\bar{y}^{\prime} \text { is valid }}} \prod_{f \in \phi_{\text {rest }}} \exp \left(\alpha_{f} f\left(\cup_{k=1}^{i-1} \bar{x}_{k}^{*} \cup \bar{y}^{\prime}\right)\right)\right)
\end{aligned}
$$

Because of the disjointness between $\operatorname{diff}\left(T_{u_{i}}\right) \backslash \operatorname{diff}\left(u_{i}\right)$ and $\cup_{u \in \mathcal{U}_{\text {rest }}} \operatorname{diff}(u)$, we can group the products and combine the partial assignments into one $\bar{x}=\bar{y} \cup \bar{y}^{\prime}$ as

$$
=\frac{\sum_{\bar{x} \in \mathcal{A}_{\text {diff }\left(u_{i+1}\right) \cup \ldots \text { diff }\left(u_{\ell}\right)}}^{\substack{\bar{x} \text { is valid }}} \prod_{f \in \phi_{\text {all }}} \exp \left(\alpha_{f} f\left(\cup_{k=1}^{i-1} \bar{x}_{k}^{*} \cup \bar{x}_{i}^{*} \cup \bar{x}\right)\right)}{\sum_{\substack{\bar{x}^{\prime} \in \mathcal{A}_{\text {diff (wi) }} \\ \bar{x}^{\prime} \text { is valid }}} \sum_{\left.\bar{x} \in \mathcal{A}_{\text {diff }}\left(u_{i+1}\right) \cup \ldots \text { udiff(u) }\right)} \prod_{f \in \phi_{\text {all }}} \exp \left(\alpha_{f} f\left(\cup_{k=1}^{i-1} \bar{x}_{k}^{*} \cup \bar{x}^{\prime} \cup \bar{x}\right)\right)} .
$$

Thus, $\mathbb{P}_{\text {alg }}\left(\bar{x}_{i}^{*} \mid \cup_{k=1}^{i-1} \bar{x}_{k}^{*}\right)=\mathbb{P}\left(\bar{x}_{i}^{*} \mid \cup_{k=1}^{i-1} \bar{x}_{k}^{*}\right)$.

