Appendices

A DETAILED ANALYSES

A.1 Derivation of Expectation of Ψ for $\sigma' = Id \circ \sigma$

We use Indicator random variables $X_{i,j}$ to determine whether the join of $i \in R$, $j \in S$ is made.

$$X_{i,j} = \begin{cases} 0 & i, j \text{ not joined in } \Psi \\ 1 & i, j \text{ joined in } \Psi \end{cases}$$

The tuples are probed as long as they are stored.

$$E[X_{i,j}] = I[i.k = j.k] \cdot p(p_R q_s) = \frac{\varepsilon_r \varepsilon_s}{p} I[i.k = j.k]$$

The complete derivation is given by:

$$E[|\Psi|] = \sum E[X_{i,j}] = \sum_{i,k=i,k} E[X_{i,j}] = \frac{\varepsilon_R \varepsilon_S}{p} \gamma_{1,1} = \frac{\varepsilon_R \varepsilon_S}{p} J$$

A.2 Proof of Variance of \hat{I} for $\sigma' = Id \circ \sigma$

We still use the notations in Appendix A.1. Obviously,

$$\operatorname{Var}[\hat{J}] = \frac{p^2}{(\varepsilon_R \varepsilon_S)^2} \operatorname{Var}[|\Psi|]$$

And we can obtain the result by calculating $Var[|\Psi|]$.

$$Var[|\Psi|] = \sum Cov(X_{r,s}, X_{r',s'})$$

= $\sum E[X_{r,s}, X_{r',s'}] - E[X_{r,s}]E[X_{r',s'}]$

There is a case where there are two tuples a, b with the same key in R, S respectively. a and b are joined if and only if a is sampled by σ_R and b is sampled by σ_S .

$$\begin{split} &E[X_{r,s},X_{r',s'}](r\neq r',s\neq s',r.k\neq r'.k)\\ &=E[X_{r,s}]E[X_{r',s'}]\\ &E[X_{r,s},X_{r',s'}](r\neq r',s\neq s',r.k=r'.k)\\ &=pq_R^2q_S^2I[r.k=s.k=r'.k=s'.k]\\ &E[X_{r,s},X_{r,s'}](s\neq s')\\ &=pq_Rq_S^2I[r.k=s.k=s'.k]\\ &E[X_{r,s},X_{r',s}](r\neq r')\\ &=pq_R^2q_S^2I[r.k=s.k=r'.k]\\ &E[X_{r,s},X_{r',s}]=E[X_{r,s}] \end{split}$$

So the final variance is given by:

$$\begin{split} & \operatorname{Var}[\hat{J}] \\ &= \frac{p^2}{(\varepsilon_R \varepsilon_S)^2} \operatorname{Var}[|\Psi|] \\ &= \frac{p^2}{(\varepsilon_R \varepsilon_S)^2} \bigg((\gamma_{2,2} - \gamma_{2,1} - \gamma_{1,2} + \gamma_{1,1}) (p q_R^2 q_S^2 - p^2 p_R^2 q_s^2) \\ &+ \gamma_{1,1} (p p_R q_S - p^2 p_R^2 q_S^2) \\ &+ (\gamma_{2,1} - \gamma_{1,1}) (p q_R^2 q_S - p^2 p_R^2 q_S^2) \bigg) \\ &= \frac{1 - p}{p} \gamma_{2,2} + \frac{p - \epsilon_S}{p \epsilon_S} \gamma_{2,1} + \frac{p - \epsilon_R}{p \epsilon_R} \gamma_{1,2} + \frac{(p - \epsilon_S)(p - \epsilon_R)}{p \epsilon_S \epsilon_R} \gamma_{1,1} \end{split}$$

A.3 Derivation of Expectation of Ψ for $\sigma' = Id$

There is equal chance for the two arrival order of *i*, *j*.

$$E[X_{i,j}] = \frac{p(q_R + q_S)}{2} I[i.k = j.k] = \frac{\varepsilon_R + \varepsilon_S}{2} I[i.k = j.k]$$

In particular, it is important to note that according to Definition 2, stream joins need to take into account timestamps, we cannot assume that a join can be generated as long as one of i, j is sampled and use the principle of inclusion-exclusion to get the following result

$$E[X_{i,j}] = p(q_R + q_S - q_R q_S))I[i.k = j.k]$$

The complete derivation is given by:

$$E[|\Psi|] = \sum E[X_{i,j}] = \sum_{i,k=i,k} E[X_{i,j}] = \frac{\varepsilon_R + \varepsilon_S}{2} \gamma_{1,1} = \frac{\varepsilon_R + \varepsilon_S}{2} J$$

A.4 Proof of Variance of \hat{J} for $\sigma' = Id$

We still use the notations in Appendix A.3. Obviously,

$$Var[\hat{J}] = 4Var[|\Psi|]/(\varepsilon_R + \varepsilon_S)^2$$

and we can obtain the result by calculating $\text{Var}[|\Psi|]$.

$$Var[|\Psi|] = \sum Cov(X_{r,s}, X_{r',s'})$$

= \sum E[X_{r,s}, X_{r',s'}] - E[X_{r,s}]E[X_{r',s'}]

There is a case where there are two tuples a, b with the same key in R, S respectively. When a is sampled by σ_R and b is not sampled by σ_S , if a comes before b then a join is generated otherwise not. So, we should discuss about $E[X_{r,s}, X_{r',s'}]$ with taking timestamps into account, which means the permutation of the arrival.

$$E[X_{r,s}, X_{r',s'}](r \neq r', s \neq s', r.k \neq r'.k)$$

$$= E[X_{r,s}]E[X_{r',s'}]$$

$$E[X_{r,s}, X_{r',s'}](r \neq r', s \neq s', r.k = r'.k)$$

$$= \frac{1}{4}p(q_R + q_S)^2I[r.k = s.k = r'.k = s'.k]$$

$$E[X_{r,s}, X_{r,s'}](s \neq s')$$

$$= \frac{1}{3}p(q_R + q_Sq_R + q_S^2)I[r.k = s.k = s'.k]$$

$$E[X_{r,s}, X_{r',s}](r \neq r')$$

$$= \frac{1}{3}p(q_R^2 + q_Sq_R + q_S)I[r.k = s.k = r'.k]$$

$$E[X_{r,s}, X_{r,s}] = E[X_{r,s}]$$

So the final variance is given by:

$$\begin{aligned} & \text{Var}[\hat{J}] \\ &= & 4 \text{Var}[|\Psi|] / (\varepsilon_R + \varepsilon_S)^2 \\ &= & \frac{4}{(\varepsilon_R + \varepsilon_S)^2} \\ & \left((\gamma_{2,2} - \gamma_{2,1} - \gamma_{1,2} + \gamma_{1,1}) \right. \\ & \left. \times \left(\frac{1}{4} p (q_R + q_S)^2 - \frac{1}{4} p^2 (q_R + q_S)^2 \right) \right. \\ & \left. + (\gamma_{1,2} - \gamma_{1,1}) \left(\frac{1}{3} p (q_R + q_S q_R + q_S^2) - \frac{1}{4} p^2 (q_R + q_S)^2 \right) \right. \end{aligned}$$

$$+(\gamma_{2,1} - \gamma_{1,1}) \left(\frac{1}{3} p(q_R^2 + q_S q_R + q_S) - \frac{1}{4} p^2 (q_R + q_S)^2 \right)$$

$$+\gamma_{1,1} \left(\frac{1}{2} p(q_R + q_S) - \frac{1}{4} p^2 (q_R + q_S)^2 \right)$$

$$= \left(\frac{1-p}{p} \right) \gamma_{2,2} + \frac{(\varepsilon_S - 3\varepsilon_R)(\varepsilon_S + \varepsilon_R) + 4p\varepsilon_R}{3p(\varepsilon_R + \varepsilon_S)^2} \gamma_{2,1}$$

$$+ \frac{(\varepsilon_R - 3\varepsilon_S)(\varepsilon_R + \varepsilon_S) + 4p\varepsilon_S}{3p(\varepsilon_R + \varepsilon_S)^2} \gamma_{1,2} + \frac{-\varepsilon_R - \varepsilon_S - 2p}{3p(\varepsilon_R + \varepsilon_S)} \gamma_{1,1}$$

A.5 Derivation of Expectation of Ψ for

$$\sigma = RS, \sigma' = Id \circ \sigma$$

For tuple t we introduce a virtual property o representing the order in which the tuple reaches the server in the two streams R, S, taking unique values from 1 to n (i.e., |R|+|S|). It is virtual because it can be indirectly deduced from the timestamp. For given two orders, the expectation of an indicator is as follows:

$$E[X_{r,s}|r.o,s.o] = \begin{cases} 1 & \max(r.o,s.o) \leq \alpha \\ \frac{(\max(r.o,s.o)-2)}{(\max(r.o,s.o))} & \max(r.o,s.o) > \alpha \end{cases}$$
With acual probability of according order the indicators.

With equal probability of arrival order, the indicator for any two tuples has the following expectation:

$$\begin{split} E[X_{r,s}] &= \frac{I[r.k = s.k]}{(n)(n-1)} \sum_{r.o,s.o} E[X_{r,s}|r.o,s.o] \\ &= \frac{I[r.k = s.k]}{(n)(n-1)} \left(\sum_{i=2}^{\alpha} 2(i-1) + \sum_{i=\alpha+1}^{n} 2(i-1) \frac{\alpha(\alpha-1)}{i(i-1)} \right) \\ &= \frac{I[r.k = s.k]\alpha(\alpha-1)(1 + 2H_n - 2H_\alpha)}{(n)(n-1)} \end{split}$$

Thus, the estimation is given by:

$$E[\Psi] = \frac{\alpha(\alpha - 1)(1 + 2H_n - 2H_\alpha)}{(n)(n-1)}J$$

A.6 Variance of \hat{J} for $\sigma = RS$, $\sigma' = Id \circ \sigma$

For the sake of simplicity, we only consider the following expectation to produce join results when selected, without using a logical judgment function to indicate whether the key is the same. By taking full advantage of the property of reservoir sampling to sample a combination with equal probability at any moment after filling the reservoir, we first discuss the intermediate results with order effects.

$$\begin{split} E[X_{r,s}, X_{r',s} | r.o > r'.o > s.o] \\ &= \left\{ \begin{array}{ll} 1 & r.o \leq \alpha \\ \frac{\alpha(\alpha-1)}{r.o(r.o-1)} & r.o > \alpha, r'.o \leq \alpha \\ \frac{\alpha(\alpha-1)^2}{r.o(r.o-1)(r'.o-1)} & r'.o > \alpha \end{array} \right. \end{split}$$

$$\begin{split} E[X_{r,s},X_{r',s}\big|r.o>s.o>r'.o] \\ = \left\{ \begin{array}{ll} 1 & r.o \leq \alpha \\ \frac{\alpha(\alpha-1)}{r.o(r.o-1)} & r.o>\alpha,s.o \leq \alpha \\ \frac{\alpha(\alpha-1)^2}{r.o(r.o-1)(s.o-1)} & s.o>\alpha \end{array} \right. \end{split}$$

$$\begin{split} E[X_{r,s}, X_{r',s} | s.o > r.o > r'.o] \\ &= \left\{ \begin{array}{ll} 1 & s.o \leq \alpha \\ \frac{\binom{s.o-3}{\alpha-3}}{\binom{s.o}{\alpha}} & s.o > \alpha \end{array} \right. \end{split}$$

 $E[X_{r,s}, X_{r',s'}|r.o > r'.o > s.o > s'.o]$

$$= \begin{cases} 1 & r.o \leq \alpha \\ \frac{\alpha(\alpha-1)}{r.o(r.o-1)} & r.o > \alpha, s.o \leq \alpha \\ \frac{\alpha^2(\alpha-1)^2}{r.o(r.o-1)s.o(s.o-1)} & s.o > \alpha \end{cases}$$

$$\begin{split} E[X_{r,s}, X_{r',s'} | r.o > s.o > r'.o > s'.o] \\ &= \left\{ \begin{array}{ll} 1 & r.o \leq \alpha \\ \frac{\alpha(\alpha-1)}{r.o(r.o-1)} & r.o > \alpha, s.o \leq \alpha \\ \frac{\alpha(\alpha-1)^2(\alpha-2)}{r.o(r.o-1)^2(\alpha-2)} & s.o > \alpha \end{array} \right. \end{split}$$

$$\begin{split} E[X_{r,s}, X_{r',s'} | r.o > s.o > s'.o > r'.o] \\ &= \left\{ \begin{array}{ll} 1 & r.o \leq \alpha \\ \frac{\alpha(\alpha - 1)}{r.o(r.o - 1)} & r.o > \alpha, s.o \leq \alpha \\ \frac{\alpha(\alpha - 1)^2(\alpha - 2)}{r.o(r.o - 1)(s.o - 1)(s.o - 2)} & s.o > \alpha \end{array} \right. \end{split}$$

With equal probability of arrival order, the indicator for tuples' join has the following expectation:

$$\begin{split} &E[X_{r,s}]\\ &= \frac{I[r.k = s.k]}{(|R|+|S|)(|R|+|S|-1)} \sum_{r.o,s.o} E[X_{r,s}|r.o,s.o]\\ &= \frac{I[r.k = s.k]}{(|R|+|S|)(|R|+|S|-1)} \left(\sum_{i=2}^{\alpha} 2(i-1) + \sum_{i=\alpha+1}^{|R|+|S|} 2(i-1) \frac{\alpha(\alpha-1)}{i(i-1)}\right)\\ &= \frac{I[r.k = s.k]\alpha(\alpha-1)(1 + 2H_{|R|+|S|} - 2H_{\alpha})}{(|R|+|S|)(|R|+|S|-1)} \end{split}$$

There is a case where there are two tuples a, b with the same key in R, S respectively. If a comes before b, a and b are joined if and only if a is sampled by σ_R , b is sampled by σ_S , and a is still in the reservoir when b arrives. So, to calculate the variance of \hat{J} , we should discuss about $E[X_{r,s}, X_{r',s'}]$ with taking timestamps into account, which means the permutation of the arrival.

In case that the matches occur on one tuple in S and two different tuples in R.

$$\begin{split} &E[X_{r,s},X_{r',s}](r\neq r')\\ &=\frac{I[r.k=s.k=r'.k]}{(n)(n-1)(n-2)}\sum_{r.o,r'.o,s.o}E[X_{r,s},X_{r',s}|r.o,r'.o,s.o]\\ &=\frac{I[r.k=s.k=r'.k]}{(n)(n-1)(n-2)}\sum_{r.o,r'.o,s.o}2(E[X_{r,s},X_{r',s}|r.o>r'.o>s.o]+\\ &E[X_{r,s},X_{r',s}|r.o>s.o>r'.o]+E[X_{r,s},X_{r',s}|s.o>r.o>r'.o])\\ &=\frac{I[r.k=s.k=r'.k]}{(n)(n-1)(n-2)}\sum_{r.o,r'.o,s.o}2(2E[X_{r,s},X_{r',s}|r.o>r'.o>s.o]+\\ &E[X_{r,s},X_{r',s}|s.o>r.o>r'.o])\\ &=\frac{I[r.k=s.k=r'.k]}{(n)(n-1)(n-2)}((\sum_{i=3}^{\alpha}2(i-1)(i-2)+\sum_{i=\alpha+1}^{n}2\alpha(\alpha-1)\frac{\alpha(\alpha-1)}{i(i-1)}) \end{split}$$

$$\begin{split} &+\sum_{i=\alpha+2}^{n}\sum_{j=\alpha+1}^{i-1}4(j-1)\frac{\alpha(\alpha-1)^{2}}{i(i-1)(j-1)})+(\sum_{i=3}^{\alpha}(i-1)(i-2)+\\ &-\sum_{i=\alpha+1}^{n}(i-1)(i-2)\frac{\alpha(\alpha-1)(\alpha-2)}{i(i-1)(i-2)}))\\ &=\frac{I[r.k=s.k=r'.k]}{(n)(n-1)(n-2)}((\frac{2\alpha(\alpha-1)(\alpha-2)}{3}+\frac{2\alpha(\alpha-1)^{2}(n-\alpha)}{n}+\\ &-4\alpha(\alpha-1)^{2}\left(\frac{\alpha}{n}+H_{n}-H_{\alpha}-1\right))+(\frac{\alpha(\alpha-1)(\alpha-2)}{3}\\ &+\alpha(\alpha-1)(\alpha-2)(H_{n}-H_{\alpha})))\\ &=\frac{I[r.k=s.k=r'.k]}{(n)(n-1)(n-2)}(\alpha(\alpha-1)(\alpha-2)(1+H_{n}-H_{\alpha})+2\alpha(\alpha-1)^{2}\\ &\left(2H_{n}-2H_{\alpha}+\frac{\alpha}{n}-1\right)) \end{split}$$

In case that the matches occur on two different tuples in S and two different tuples in R.

$$\begin{split} &E[X_{r,s},X_{r',s'}](r\neq r',s\neq s')\\ &=\frac{I[r.k=r'.k,s.k=s'.k]}{(n)(n-1)(n-2)(n-3)}\sum_{r.o,r'.o,s.o,s'.o}^{N} 8(E[X_{r,s},X_{r',s'}|r.o>s.o\\ &>s'.o>r'.o]+E[X_{r,s},X_{r',s'}|r.o>s.o>r'.o>s'.o]+E[X_{r,s},X_{r',s'}|r.o>s.o>r'.o>s'.o]+E[X_{r,s},X_{r',s'}|r.o>s.o>r'.o>s'.o]+E[X_{r,s},X_{r',s'}|r.o>r'.o>s.o>s'.o])\\ &=\frac{I[r.k=r'.k,s.k=s'.k]}{(n)(n-1)(n-2)(n-3)}\sum_{r.o,r'.o,s.o,s'.o}^{N} 8(E[X_{r,s},X_{r',s'}|r.o>r'.o>r'.o>s.o>s'.o])\\ &=\frac{I[r.k=r'.k,s.k=s'.k]}{(n)(n-1)(n-2)(n-3)}\sum_{r.o,r'.o,s.o,s'.o}^{N} 8(((n)(n-1)(n-2)(n-3)))\\ &=\frac{I[r.k=r'.k,s.k=s'.k]}{(n)(n-1)(n-2)(n-3)}\sum_{i=\alpha+1}^{n}\sum_{j=\alpha+1}^{i-2}(i-j-1)(j-1)\frac{\alpha^2(\alpha-1)^2}{i(i-1)j(j-1)}))\\ &+2(\frac{\alpha(\alpha-1)(\alpha-2)(\alpha-3)}{24}+\sum_{i=\alpha+3}^{n}\sum_{j=\alpha+1}^{i-2}(i-j-1)(j-1)\frac{\alpha^2(\alpha-1)^2}{i(i-1)j(j-1)}))\\ &+\sum_{i=\alpha+2}^{n}\sum_{j=\alpha+1}^{i-1}\frac{(j-1)(j-2)}{2}\frac{\alpha(\alpha-1)^2(\alpha-2)}{i(i-1)(j-1)(j-2)})))\\ &=\frac{I[r.k=r'.k,s.k=s'.k]}{(n)(n-1)(n-2)(n-3)}(\alpha(\alpha-1)(\alpha-2)(\alpha-3)+\\ &+\alpha^2(\alpha-1)^2(\alpha-2)\sum_{i=\alpha+1}^{n}\frac{i-j+1}{i(i-1)}+8\alpha(\alpha-1)^2(\alpha-2)\sum_{i=\alpha+1}^{n}\frac{i-1-\alpha}{i(i-1)}+\\ &8\alpha^2(\alpha-1)^2\sum_{i=\alpha+3}^{n}\sum_{j=\alpha+1}^{i-2}\frac{i-j+1}{i(i-1)j}+8\alpha(\alpha-1)^2(\alpha-2)\sum_{i=\alpha+2}^{n}\frac{i-1-\alpha}{i(i-1)}+\\ &\frac{\sum_{i=\alpha+1}^{i-1}\frac{1}{i(i-1)}}{i(i-1)(n-2)(n-3)}(\alpha(\alpha-1)(\alpha-2)(\alpha-3)+\\ &+\alpha^2(\alpha-1)^2(\alpha-2)(\frac{1}{\alpha}-\frac{1}{n})+\\ &8\alpha(\alpha-1)^2(\alpha-2)(\frac{1}{\alpha}-\frac{1}{n})+\\ &8\alpha(\alpha-1)^2(\alpha-2)(H_n-H_{\alpha+1}-\frac{\alpha}{\alpha+1}+\frac{\alpha}{n})+\\ \end{cases}$$

$$4\alpha^{2}(\alpha-1)^{2}(H_{n}^{2}-H_{n}^{(2)}-H_{\alpha+2}^{2}+H_{\alpha+2}^{(2)}-2H_{n}H_{\alpha} +2H_{\alpha}H_{\alpha+2}-H_{n}+H_{\alpha}-\frac{2}{\alpha+2}-\frac{\alpha}{n}+1+\frac{2}{\alpha+1}))$$

In case that the matches occur on one tuple in S and one tuple in R.

$$E[X_{r,s}, X_{r,s}] = E[X_{r,s}]$$

On the basis of considering the probability, combining all the above cases, the final variance of $|\Psi|$ is given by:

 $Var[|\Psi|]$

$$\begin{split} &= \sum Cov(X_{r,s},X_{r',s'}) \\ &= \sum E[X_{r,s},X_{r',s'}] - E[X_{r,s}]E[X_{r',s'}] \\ &= \frac{\gamma_{1,1}}{n(n-1)}\alpha(\alpha-1)(1+2H_n-2H_\alpha) \\ &+ \frac{(\gamma_{1,2}-\gamma_{1,1})+(\gamma_{2,1}-\gamma_{1,1})}{n(n-1)(n-2)}(\alpha(\alpha-1)(\alpha-2) \\ &(1+H_n-H_\alpha)+2\alpha(\alpha-1)^2(2H_n-2H_\alpha+\frac{\alpha}{n}-1)) \\ &+ \frac{I[r.k=s.k,r'.k=s'.k]}{n(n-1)(n-2)(n-3)}(\alpha(\alpha-1)(\alpha-2)(\alpha-3) \\ &+ 4\alpha^2(\alpha-1)^2(\alpha-2)(\frac{1}{\alpha}-\frac{1}{n})+8\alpha(\alpha-1)^2(\alpha-2) \\ &(H_n-H_{\alpha+1}-\frac{\alpha}{\alpha+1}+\frac{\alpha}{n}))+4\alpha^2(\alpha-1)^2 \\ &(H_n^2-H_n^2)-H_{\alpha+2}^2+H_{\alpha+2}^{(2)}-2H_nH_\alpha+2H_\alpha H_{\alpha+2} \\ &-H_n+H\alpha-\frac{2}{\alpha+2}-\frac{\alpha}{n}+1+\frac{2}{\alpha+1})) \\ &-\frac{I[r.k=s.k]I[r'.k=s'.k]}{n^2(n-1)^2}(1+2H_n-2H_\alpha)+\frac{\alpha(\alpha-1)}{n(n-1)(n-2)} \\ &((-10\alpha+8)H_n+(10\alpha-8)H_\alpha+2\alpha-2-\frac{4\alpha(\alpha-1)}{n}) \\ &+\frac{1}{n(n-1)(n-2)(n-3)}(\alpha(\alpha-1)(\alpha-2)(\alpha-3) \\ &+4\alpha^2(\alpha-1)^2(\alpha-2)(\frac{1}{\alpha}-\frac{1}{n})+8\alpha(\alpha-1)^2(\alpha-2) \\ &(H_n-H_{\alpha+1}-\frac{\alpha}{\alpha+1}+\frac{\alpha}{n}))+\frac{\alpha^2(\alpha-1)^2}{(n-1)^2(n-2)(n-3)} \\ &(-4H_\alpha^2-4H_{\alpha+2}^2-4H_n^{(2)}+4H_{\alpha+2}^{(2)}-8H_n+8H_\alpha+3 \\ &+8H_\alpha H_{\alpha+2}-\frac{4\alpha}{n}-\frac{8}{\alpha+2}+\frac{8}{\alpha+1}) \\ &+\frac{\alpha^2(\alpha-1)^2}{n(n-1)^2(n-2)(n-3)}(16H_n^2+20H_\alpha^2+4H_n^{(2)}+4H_{\alpha+2}^{(2)}-8H_n+8H_\alpha+3 \\ &+8H_\alpha H_{\alpha+2}-4H_{\alpha+2}^{(2)}-32H_nH_\alpha-8H_\alpha H_{\alpha+2}+24H_n \\ &-24H_\alpha+1+\frac{4\alpha}{n}+\frac{8}{\alpha+2}-\frac{8}{\alpha+1}) \\ &+\frac{6\alpha^2(\alpha-1)^2}{n^2(n-1)^2(n-2)(n-3)}(1+4H_n^2+4H_\alpha^2+4H_n \\ &-24H_\alpha+1+\frac{4\alpha}{n}+\frac{8}{\alpha+2}-\frac{8}{\alpha+1}) \\ &+\frac{6\alpha^2(\alpha-1)^2}{n^2(n-1)^2(n-2)(n-3)}(1+4H_n^2+4H_\alpha^2+4H_n \\ &-4H_\alpha-8H_nH_\alpha)) \\ &+\eta_{2,2}(\frac{1}{n(n-1)(n-2)(n-3)}(\alpha(\alpha-1)(\alpha-2)(\alpha-3) \\ &+\eta_{2,2}(\frac{1}{n(n-1)(n-2)(n-3)}(\alpha(\alpha-1)(\alpha-2)(\alpha-3)) \\ &+\eta_{2,2}(\frac{1}{n(n-1)(n-2)(n-3)}(\alpha(\alpha-1)(\alpha-2)(\alpha-3)) \\ &+\eta_{2,2}(\frac{1}{n(n-1)(n-2)(n-3)}(\alpha(\alpha-1)(\alpha-2)(\alpha-3) \\ &+\eta_{2,2}(\alpha-1)^2(\alpha-2)(\alpha-3) \\ &+\eta_{2,2}(\alpha-1)^2(\alpha-$$

$$\begin{split} &+4\alpha(\alpha-1)^2(\alpha-2)(1+\frac{\alpha}{n}+2H_n-2H_{\alpha+1}-\frac{2\alpha}{\alpha+1}))\\ &+\frac{\alpha^2(\alpha-1)^2}{(n-1)^2(n-2)(n-3)}(-4H_{\alpha}^2-4H_{\alpha+2}^2-4H_{n}^{(2)}+4H_{\alpha+2}^{(2)})\\ &-8H_n+8H_{\alpha}+3+8H_{\alpha}H_{\alpha+2}-\frac{4\alpha}{n}\\ &-\frac{8}{\alpha+2}+\frac{8}{\alpha+1})+\frac{\alpha^2(\alpha-1)^2}{n(n-1)^2(n-2)(n-3)}(16H_n^2+20H_{\alpha}^2\\ &+4H_n^{(2)}+4H_{\alpha+2}^2-4H_{\alpha+2}^{(2)}-32H_nH_{\alpha}-8H_{\alpha}H_{\alpha+2}+24H_n\\ &-24H_{\alpha}+1+\frac{4\alpha}{n}+\frac{8}{\alpha+2}-\frac{8}{\alpha+1})+\frac{6\alpha^2(\alpha-1)^2}{n^2(n-1)^2(n-2)(n-3)}\\ &(1+4H_n^2+4H_{\alpha}^2+4H_n-4H_{\alpha}-8H_nH_{\alpha}))\\ &+(\gamma_{1,2}+\gamma_{2,1})(\frac{1}{n(n-1)(n-2)}(\alpha(\alpha-1)(\alpha-2)(1+H_n-H_{\alpha})\\ &+2\alpha(\alpha-1)^2(2H_n-2H_{\alpha}+\frac{\alpha}{n}-1))\\ &-\frac{1}{n(n-1)(n-2)(n-3)}(\alpha(\alpha-1)(\alpha-2)(\alpha-3)+4\alpha(\alpha-1)^2\\ &(\alpha-2)(1+\frac{\alpha}{n}+2H_n-2H_{\alpha+1}-\frac{2\alpha}{\alpha+1}))+\frac{\alpha^2(\alpha-1)^2}{(n-1)^2(n-2)(n-3)}\\ &(-4H_{\alpha}^2-4H_{\alpha+2}^2-4H_n^{(2)}+4H_{\alpha+2}^{(2)}-8H_n+8H_{\alpha}+3+8H_{\alpha}H_{\alpha+2}\\ &-\frac{4\alpha}{n}-\frac{8}{\alpha+2}+\frac{8}{\alpha+1})+\frac{\alpha^2(\alpha-1)^2}{n(n-1)^2(n-2)(n-3)}(16H_n^2+20H_{\alpha}^2\\ &+4H_n^{(2)}+4H_{\alpha+2}^2-4H_{\alpha+2}^{(2)}-32H_nH_{\alpha}-8H_{\alpha}H_{\alpha+2}+24H_n\\ &-24H_{\alpha}+1+\frac{4\alpha}{n}+\frac{8}{\alpha+2}-\frac{8}{\alpha+1})+\frac{6\alpha^2(\alpha-1)^2}{n^2(n-1)^2(n-2)(n-3)}\\ &(1+4H_n^2+4H_{\alpha}^2+4H_n-4H_{\alpha}-8H_nH_{\alpha}))\\ &=\gamma_{1,1}\Theta(\frac{\alpha^2\log\frac{n}{\alpha}}{n^2})+\gamma_{2,2}\Theta(\frac{\alpha^4\log\frac{n}{\alpha}}{n^4})\\ &+(\gamma_{1,2}+\gamma_{2,1})\Theta(\frac{\alpha^3\log\frac{n}{\alpha}}{n^3}) \end{split}$$

The final variance of \hat{J} is given by:

$$\begin{split} Var[\hat{J}] &= \frac{n^2(n-1)^2}{\alpha^2(\alpha-1)^2(1+2H_n-2H_\alpha)^2} Var[\Psi] \\ &= \gamma_{1,1}(\frac{n^2}{\alpha^2\log\frac{n}{\alpha}}) + \gamma_{2,2}(\frac{1}{\log\frac{n}{\alpha}}) + (\gamma_{1,2}+\gamma_{2,1})\Theta(\frac{n}{\alpha\log\frac{n}{\alpha}}) \end{split}$$

A.7 Derivation of Expectation of Ψ for

$$\sigma = RS, \sigma' = Id$$

For given two orders, the expectation of an indicator is as follows:

$$E[X_{r,s}|r.o,s.o] = \begin{cases} 1 & \max(r.o,s.o) \leq \alpha \\ \frac{\alpha}{\max(r.o,s.o)} & \max(r.o,s.o) > \alpha \end{cases}$$
With equal probability of arrival order the indicate

With equal probability of arrival order, the indicator for tuples' join has the following expectation:

$$E[X_{r,s}] = 2\frac{I[r.k = s.k]}{n(n-1)} (\frac{\alpha(\alpha-1)}{2} + \sum_{i=\alpha+1}^{n} (i+1)\frac{\alpha}{i})$$
$$= \frac{I[r.k = s.k]}{n(n-1)} (2\alpha n - 2\alpha H_{\alpha} - \alpha(\alpha+1) + 2\alpha H_{\alpha})$$

Thus, the estimation is given by:

$$E[\Psi] = \frac{2\alpha n - 2\alpha H_{\alpha} - \alpha(\alpha+1) + 2\alpha H_{\alpha}}{n(n-1)} J$$

A.8 Variance of \hat{J} for $\sigma = RS$, $\sigma' = Id$

Just like the condition of $\sigma = \sigma' = RS$, we also only consider the following expectation to produce join results when selected, without using a logical judgment function to indicate whether the key is the same. By taking full advantage of the property of reservoir sampling to sample a combination with equal probability at any moment after filling the reservoir, we first discuss the intermediate results with order effects.

$$E[X_{r,s}, X_{r',s} | r.o > r'.o > s.o]$$

$$= \begin{cases} 1 & r.o \leq \alpha \\ \frac{\alpha}{r.o} & r.o > \alpha, r'.o \leq \alpha \\ \frac{\alpha}{r.o} & r'.o > \alpha \end{cases}$$

$$E[X_{r,s}, X_{r',s} | s.o > r.o > r'.o]$$

$$= \begin{cases} 1 & s.o \leq \alpha \\ \frac{\alpha(\alpha-1)}{s.o(s.o-1)} & s.o > \alpha \end{cases}$$

$$E[X_{r,s}, X_{r',s'} | r.o > s.o > r'.o > s'.o]$$

$$= \begin{cases} 1 & r.o \leq \alpha \\ \frac{\alpha}{r.o} & r.o > \alpha, r'.o \leq \alpha \\ \frac{\alpha^2}{r.or'.o} & r'.o > \alpha \end{cases}$$

$$E[X_{r,s}, X_{r',s'} | r.o > r'.o > s.o > s'.o]$$

$$= \begin{cases} 1 & r.o \leq \alpha \\ \frac{\alpha^2}{r.or'.o} & r'.o > \alpha, r'.o \leq \alpha \\ \frac{\alpha(\alpha-1)}{r.o(r'.o-1)} & r'.o > \alpha \end{cases}$$

With equal probability of arrival order, the indicator for tuples' join has the following expectation:

$$E[X_{r,s}] = 2\frac{I[r.k = s.k]}{n(n-1)} (\frac{\alpha(\alpha-1)}{2} + \sum_{i=\alpha+1}^{n} (i+1)\frac{\alpha}{i})$$
$$= \frac{I[r.k = s.k]}{n(n-1)} (2\alpha n - 2\alpha H_n - \alpha(\alpha+1) + 2\alpha H_\alpha)$$

There is a case where there are two tuples a, b with the same key in R, S respectively. If a comes before b, a and b are joined if and only if a is sampled by σ_R and a is still in the reservoir when b arrives. So, to calculate the variance of \hat{J} , we should discuss about $E[X_{r,s}, X_{r',s'}]$ with taking timestamps into account, which means the permutation of the arrival.

In case that the matches occur on one tuple in S and two different tuples in R.

$$E[X_{r,s}X_{r',s}])(r \neq r')$$

$$= \frac{I[r.k = r'.k = s.k]}{n(n-1)(n-2)} \sum_{r.o,r'.o,s.o} 2(EX_{[r,s}X_{r',s}|r.o > r'.o > s.o]$$

$$+EX_{[r,s}X_{r',s}|r.o > s.o > r'.o] + EX_{[r,s}X_{r',s}|s.o > r.o > r'.o])$$

$$\begin{split} &=\frac{I[r.k=r'.k=s.k]}{n(n-1)(n-2)}\sum_{r.o,r'.o,s.o}(4EX_{[r,s}X_{r',s}|r.o>r'.o>s.o]\\ &+2EX_{[r,s}X_{r',s}|s.o>r.o>r'.o])\\ &=\frac{I[r.k=r'.k=s.k]}{n(n-1)(n-2)}(4(\sum_{i=3}^{\alpha}\frac{(i-1)(i-2)}{2}+\sum_{i=\alpha+1}^{n}\frac{\alpha(\alpha-1)}{2}\cdot\frac{\alpha}{i}\\ &+\sum_{i=\alpha+2}^{n}\sum_{j=\alpha+1}^{i-1}(j-1)\frac{\alpha}{i})+2(\sum_{i=3}^{n}\frac{(i-1)(i-2)}{2}\\ &+\sum_{i=\alpha+1}^{n}(i-1)(i-2)\frac{\alpha(\alpha-1)}{i(j-1)}))\\ &=\frac{I[r.k=r'.k=s.k]}{n(n-1)(n-2)}(\alpha n^2+\alpha(2\alpha-7)n+(2\alpha^3-7\alpha^2+5\alpha+2)H_n\\ &-2\alpha(\alpha-1)(\alpha-2)H_{\alpha}+(\alpha-2)(\alpha+1)H_{\alpha+1}-2\alpha(\alpha^2-\alpha-3)) \end{split}$$

In case that the matches occur on two different tuples in S and two different tuples in R.

$$\begin{split} &E[X_{r,s}X_{r',s'}](r\neq r,s\neq s')\\ &=\frac{I[r.k=s.k,r'.k=s'.k]}{n(n-1)(n-2)(n-3)}(8E[X_{r,s}X_{r',s'}|r.o>s.o>r'.o>s'.o]\\ &+16E[X_{r,s}X_{r',s'}|r.o>r'.o>s.o>s'.o])\\ &=\frac{I[r.k=s.k,r'.k=s'.k]}{n(n-1)(n-2)(n-3)}(8(\frac{\alpha(\alpha-1)(\alpha-2)(\alpha-3)}{24}\\ &+\sum_{i=\alpha+1}^{n}(\frac{\alpha(\alpha-1)(\alpha-2)}{6}+\frac{(i-\alpha-1)\alpha(\alpha-1)}{2})\frac{\alpha}{i}\\ &+\sum_{i=\alpha+3}^{n}\sum_{j=\alpha+1}^{i-2}(i-j-1)(j-1)\frac{\alpha(\alpha-1)}{i(j-1)})\\ &+16(\frac{\alpha(\alpha-1)(\alpha-2)(\alpha-3)}{24}+\sum_{i=\alpha+1}^{n}\frac{\alpha(\alpha-1)(\alpha-2)}{6}\cdot\frac{\alpha}{i}\\ &+\sum_{i=\alpha+2}^{n}\sum_{j=\alpha+1}^{i-1}\frac{(j-1)(j-2)}{2}\cdot\frac{\alpha(\alpha-1)}{i(j-1)}))\\ &=\frac{I[r.k=s.k,r'.k=s'.k]}{n(n-1)(n-2)(n-3)}(4\alpha(\alpha-1)n^2-4\alpha(\alpha-1)(\alpha+1)n\\ &+\alpha(\alpha-1)(\alpha^2+3\alpha-14)-12\alpha(\alpha-1)(\alpha-2)H_n+12\alpha^2(\alpha-1)\\ &H_{\alpha}-4\alpha(\alpha-1)^2(\alpha-2)H_{\alpha+2}+4\alpha(\alpha-1)(\alpha+1)(\alpha-4)H_{\alpha+1}) \end{split}$$

On the basis of considering the probability, combining all the above cases, the final variance of $|\Psi|$ is given by:

$$\begin{split} &Var[|\Psi|] \\ &= \sum Cov(X_{r,s}, X_{r',s'}) \\ &= \sum E[X_{r,s}, X_{r',s'}] - E[X_{r,s}]E[X_{r',s'}] \\ &= \frac{\gamma_{1,1}}{n(n-1)}(2\alpha n - 2\alpha H_n - \alpha(\alpha+1) + 2\alpha H_\alpha) \\ &+ \frac{\gamma_{1,2} + \gamma_{2,1}}{n(n-1)(n-2)}(\alpha n^2 + \alpha(2\alpha-7)n + (2\alpha^3 - 7\alpha^2 + 5\alpha + 2)H_n \\ &- 2\alpha(\alpha-1)(\alpha-2)H_\alpha + (\alpha-2)(\alpha+1)H_{\alpha+1} - 2\alpha(\alpha^2 - \alpha - 3)) \\ &+ \frac{\gamma_{1,1} + \gamma_{2,2} - \gamma_{1,2} - \gamma_{2,1}}{n(n-1)(n-2)(n-3)}(4\alpha(\alpha-1)n^2 - 4\alpha(\alpha-1)(\alpha+1)n \\ &+ \alpha(\alpha-1)(\alpha^2 + 3\alpha - 14) - 12\alpha(\alpha-1)(\alpha-2)H_n + 12\alpha^2(\alpha-1) \\ &H_\alpha - 4\alpha(\alpha-1)^2(\alpha-2)H_{\alpha+2} + 4\alpha(\alpha-1)(\alpha+1)(\alpha-4)H_{\alpha+1}) \end{split}$$

$$\begin{split} &-\frac{\gamma_{1,1}+\gamma_{2,2}-\gamma_{1,2}-\gamma_{2,1}}{n(n-1)(n-2)(n-3)}(2\alpha n-2\alpha H_n-\alpha(\alpha+1)+2\alpha H_\alpha)^2\\ &=\gamma_{1,1}(\frac{-4\alpha}{n(n-1)}+\frac{\alpha(-1-\alpha+2H_\alpha)}{n(n-1)}-\frac{2\alpha H_n}{n(n-1)}\\ &-\frac{2\alpha(2\alpha-7)}{(n-1)(n-2)}-\frac{2}{n(n-1)(n-2)}(-2\alpha(\alpha^2-\alpha-3)\\ &+(2\alpha^3-7\alpha^2+5\alpha+2)H_n-2\alpha(\alpha-1)(\alpha-2)H_\alpha+(\alpha-2)\\ &(\alpha+1)H_{\alpha+1})-\frac{4\alpha n^2-4\alpha(4\alpha+3)n}{(n-1)^2(n-2)(n-3)}+\frac{1}{(n-1)^2(n-2)(n-3)}\\ &(-2\alpha(8\alpha^2+19\alpha-5)+16\alpha^2(\alpha+2)H_\alpha+4\alpha(\alpha-1)(\alpha+1)(\alpha-4)\\ &H_{\alpha+1}-4\alpha(\alpha-1)^2(\alpha-2)H_{\alpha+2}-4\alpha^2H_\alpha^2)\\ &+\frac{1}{n(n-1)^2(n-2)(n-3)}(2\alpha(2\alpha^3+4\alpha^2+11\alpha-7)-8\alpha^2\\ &(4\alpha+1)H_\alpha+4\alpha(\alpha-1)^2(\alpha-1)H_{\alpha+2}-4\alpha(\alpha-1)(\alpha+1)\\ &(\alpha-4)H_{\alpha+1}+20\alpha^2H_{\alpha^2})-\frac{6}{n^2(n-1)^2(n-2)(n-3)}\\ &(a^2(\alpha+1)^2+4\alpha^2H_\alpha^2-4\alpha^2(\alpha+1)H_\alpha)\\ &+\frac{H_n}{n^2(n-1)^2(n-2)(n-3)}(8\alpha^2\alpha^3-8\alpha n^2(2\alpha^2-4\alpha+3-\alpha H_\alpha))\\ &+8\alpha n(4\alpha^2+4\alpha+3-5\alpha H_\alpha)-24\alpha^2(\alpha+1-2H_\alpha))-\frac{4\alpha^2H_n^2}{n^2(n-1)^2})\\ &+\frac{1}{n(n-1)(n-2)}(-2\alpha(\alpha^2-\alpha-3)\\ &+(2\alpha^3-7\alpha^2+5\alpha+2)H_n-2\alpha(\alpha-1)(\alpha-2)H_\alpha+(\alpha-2)\\ &(\alpha+1)H_{\alpha+1})+\frac{4\alpha n^2-4\alpha(4\alpha+3)n}{(n-1)^2(n-2)(n-3)}-\frac{1}{(n-1)^2(n-2)(n-3)}\\ &(-2\alpha(8\alpha^2+19\alpha-5)+16\alpha^2(\alpha+2)H_\alpha+4\alpha(\alpha-1)(\alpha+1)(\alpha-4)\\ &H_{\alpha+1}-4\alpha(\alpha-1)^2(\alpha-2)H_{\alpha+2}-4\alpha^2H_\alpha^2)\\ &-\frac{1}{n(n-1)^2(n-2)(n-3)}(2\alpha(2\alpha^3+4\alpha^2+11\alpha-7)-8\alpha^2\\ &(4\alpha+1)H_\alpha+4\alpha(\alpha-1)^2(\alpha-1)H_{\alpha+2}-4\alpha(\alpha-1)(\alpha+1)\\ &(\alpha-4)H_{\alpha+1}+20\alpha^2H_{\alpha^2}+\frac{6}{n^2(n-1)^2(n-2)(n-3)}\\ &(\alpha^2(\alpha+1)^2+4\alpha^2H_\alpha^2-4\alpha^2(\alpha+1)H_\alpha)\\ &-\frac{H_n}{n^2(n-1)^2(n-2)(n-3)}(8\alpha^2n^3-8\alpha n^2(2\alpha^2-4\alpha+3-\alpha H_\alpha)\\ &+\frac{4\alpha^2H_n^2}{n^2(n-1)^2(n-2)(n-3)}\\ &(\alpha^2(\alpha+1)^2+4\alpha^2H_\alpha^2-4\alpha^2(\alpha+1)H_\alpha)\\ &-\frac{H_n}{n^2(n-1)^2(n-2)(n-3)}(8\alpha^2n^3-8\alpha n^2(2\alpha^2-4\alpha+3-\alpha H_\alpha)\\ &+\frac{4\alpha^2H_n^2}{n^2(n-1)^2(n-2)(n-3)}\\ &(\alpha^2(\alpha+1)^2+4\alpha^2H_\alpha^2-4\alpha^2(\alpha+1)H_\alpha)\\ &-\frac{H_n}{n^2(n-1)^2(n-2)(n-3)}(8\alpha^2n^3-8\alpha n^2(2\alpha^2-4\alpha+3-\alpha H_\alpha)\\ &+\frac{4\alpha^2H_n^2}{n^2(n-1)^2(n-2)(n-3)}(8\alpha^2n^3-8\alpha n^2(2\alpha^2-4\alpha+3-\alpha H_\alpha)\\ &+\frac{4\alpha^2H_n^2}{n^2(n-1)^2(n-2)(n-3)}(8\alpha^2n^3-8\alpha n^2(2\alpha^2-4\alpha+3-\alpha H_\alpha)\\ &+\frac{4\alpha^2H_n^2}{n^2(n-1)^2(n-2)(n-3)}(8\alpha^2n^3-8\alpha n^2(2\alpha^2-4\alpha+3-\alpha H_\alpha)\\ &+\frac{4\alpha^2H_n^2}{n^2(n-1)^2(n-2)(n-3)}(8\alpha^2n^3-8\alpha n^2(2\alpha^2-4\alpha+3-\alpha H_\alpha)\\ &+\frac{4\alpha^2H_n^2}{n^2(n-1)^2(n-2)(n-3)}(\alpha^2(\alpha+1)^2+4\alpha^2H_\alpha^2-4\alpha^2(\alpha+1)H_\alpha)\\ &+\frac{4\alpha^2H_n^2}{n^2(n-1)^2(n-2)(n-3)}(\alpha^2(\alpha+1)^2+4\alpha^2H_\alpha^2-4\alpha^2(\alpha+1)H_\alpha)\\ &+\frac{4\alpha^2H_n^2}{n^2(n-1)^2(n-2)(n-3)}(\alpha^2(\alpha+1)^2+4\alpha^2H_\alpha^2)\\ &+\frac{4\alpha^2H_$$

$$\begin{split} &(\alpha^2(\alpha+1)^2+4\alpha^2H_{\alpha}^2-4\alpha^2(\alpha+1)H_{\alpha})\\ &+\frac{H_n}{n^2(n-1)^2(n-2)(n-3)}(8\alpha^2n^3-8\alpha n^2(2\alpha^2-4\alpha+3-\alpha H_{\alpha})\\ &+8\alpha n(4\alpha^2+4\alpha+3-5\alpha H_{\alpha})-24\alpha^2(\alpha+1-2H_{\alpha}))-\frac{4\alpha^2H_n^2}{n^2(n-1)^2})\\ &=\gamma_{1,1}\Theta(\frac{\alpha}{n}+\frac{\alpha^3\log\frac{n}{\alpha}}{n^3})+\gamma_{2,2}\Theta(\frac{\alpha}{n^2}+\frac{\alpha^2\log n}{n^3}+\frac{\alpha^3\log\frac{n}{\alpha}}{n^4})\\ &+(\gamma_{1,2}+\gamma_{2,1})\Theta(\frac{\alpha}{n}+\frac{\alpha^3\log\frac{n}{\alpha}}{n^3}) \end{split}$$

The final variance of \hat{J} is given by:

$$\begin{split} Var[\hat{J}] &= \frac{n^2(n-1)^2}{(2\alpha n - 2\alpha H_\alpha - \alpha(\alpha+1) + 2\alpha H_\alpha)^2} Var[\Psi] \\ &= \gamma_{1,1}\Theta(\frac{n}{\alpha} + \frac{\alpha\log\frac{n}{\alpha}}{n}) + \gamma_{2,2}\Theta(\frac{1}{\alpha} + \frac{\log n}{n} + \frac{\alpha\log\frac{n}{\alpha}}{n^2}) \\ &+ (\gamma_{1,2} + \gamma_{2,1})\Theta(\frac{n}{\alpha} + \frac{\alpha\log\frac{n}{\alpha}}{n}) \end{split}$$