

Somatic hypermutations signatures

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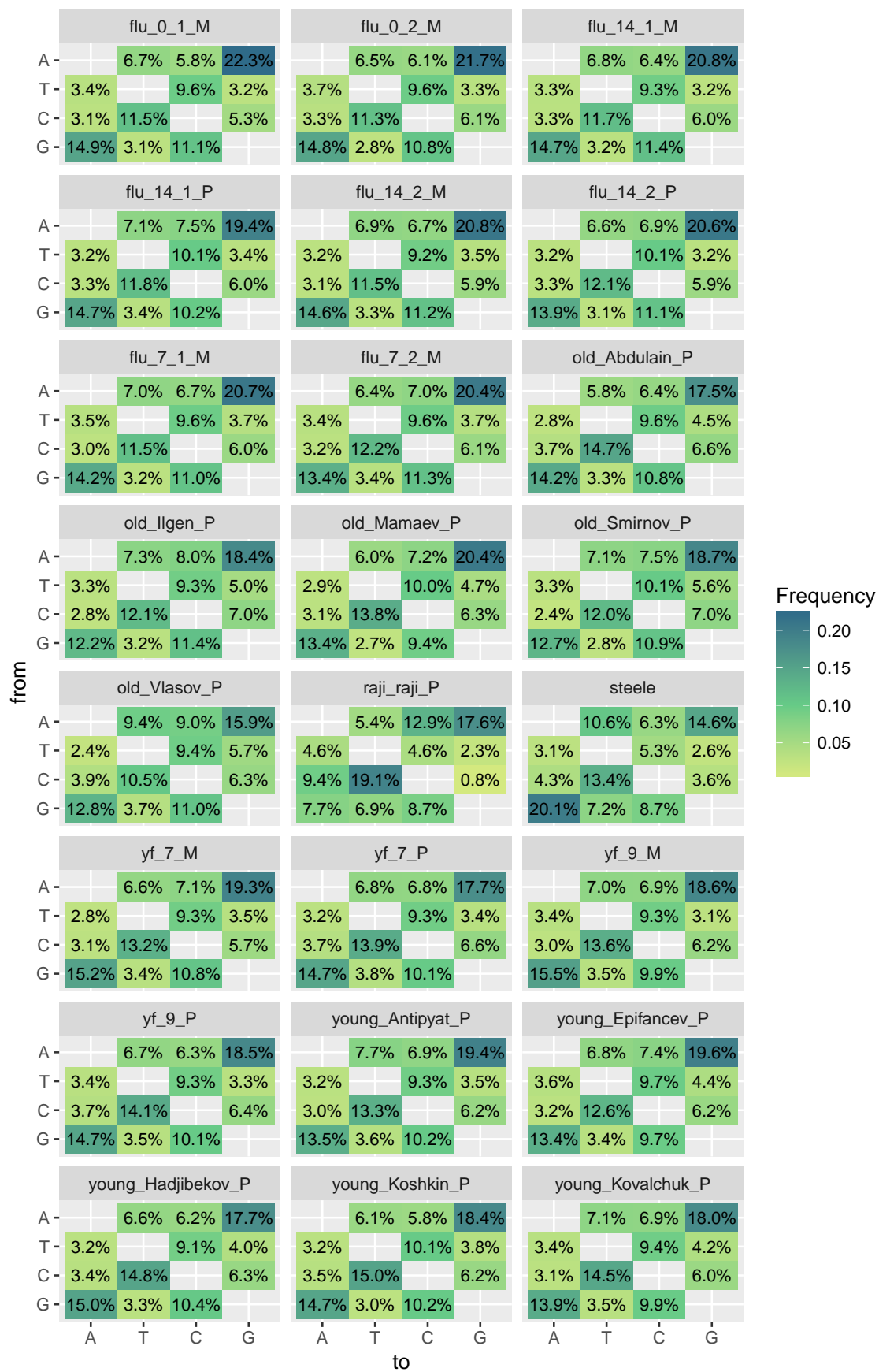
Exploratory data analysis

Load all our data, plot relative substitution frequencies. Here is some summary of what we currently have:

- We have flu (`flu`) and yellow fever (`YF`) vaccination time-courses which track plasma (`P`) and memory (`M`) B-cells.
- We also have `old` and `young` donors vaccinated against yellow fever, `P` cells only and no controls unfortunately.
- Raji cell line (`raji`) and data from Steele 2009 (`steele`) are included for reference.

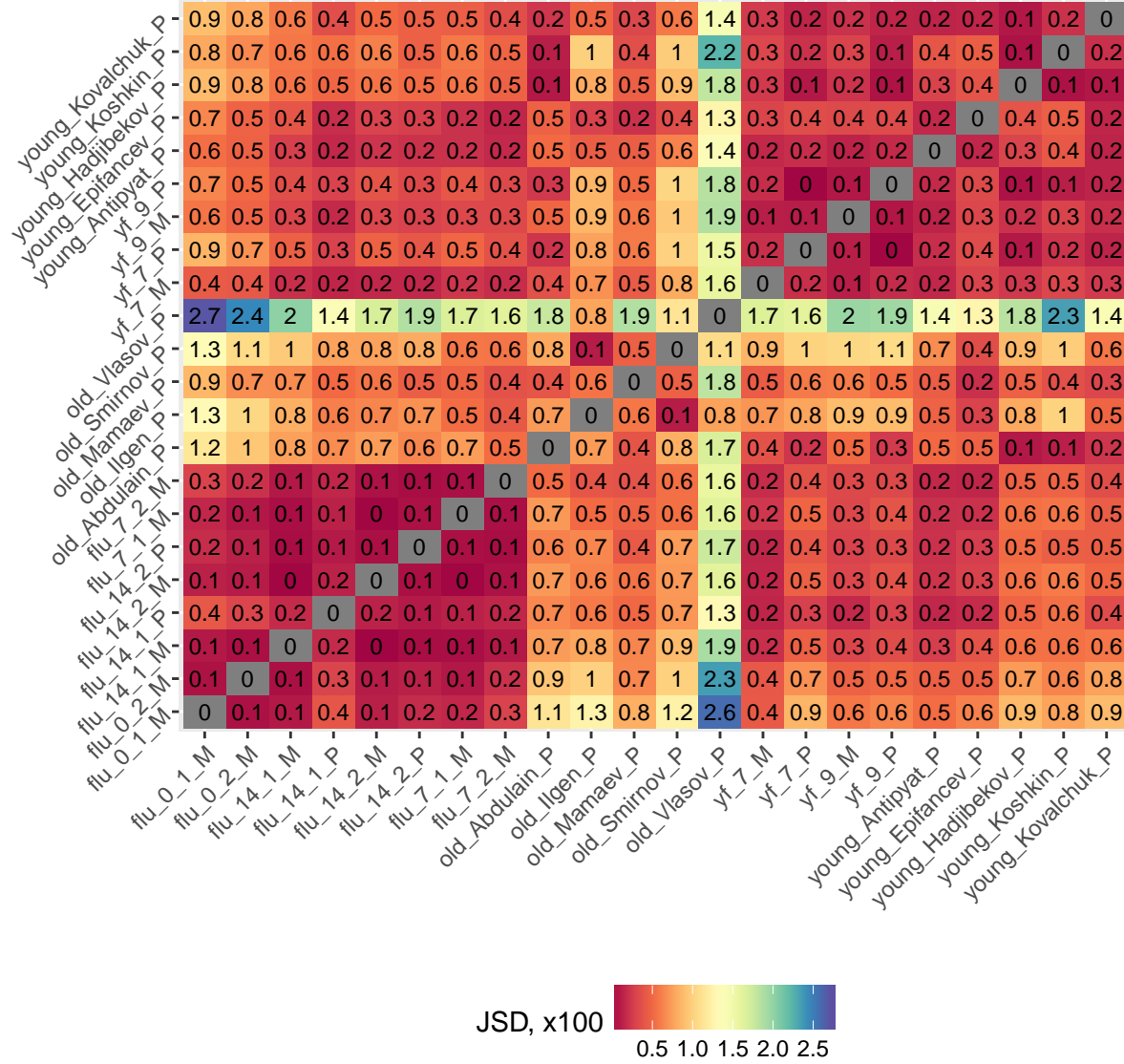
Note that we work with relative fractions of substitutions which is computed as follows. Let the number of substitutions from base B_i to base B_j be $\#(B_i \rightarrow B_j)$, the absolute substitution frequency is then $F_{ij} = \#(B_i \rightarrow B_j) / \#B_i$ where $\#B_i$ is the total number of occurrences of base B_i in a sample of sequences. The relative frequency is given by normalizing all F_{ij} to $\sum_{ij} F_{ij} = 1$ (to 100%), i.e. $f_{ij} = F_{ij} / \sum_{lk} F_{lk}$.

Also note that here we ignore abundancies of individual B-cell clonotypes and count each of them only once when summing over substitutions. This is reasonable as it removes substitution frequency biases coming from preferential expansion of B-cell clonotypes with certain hypermutations.

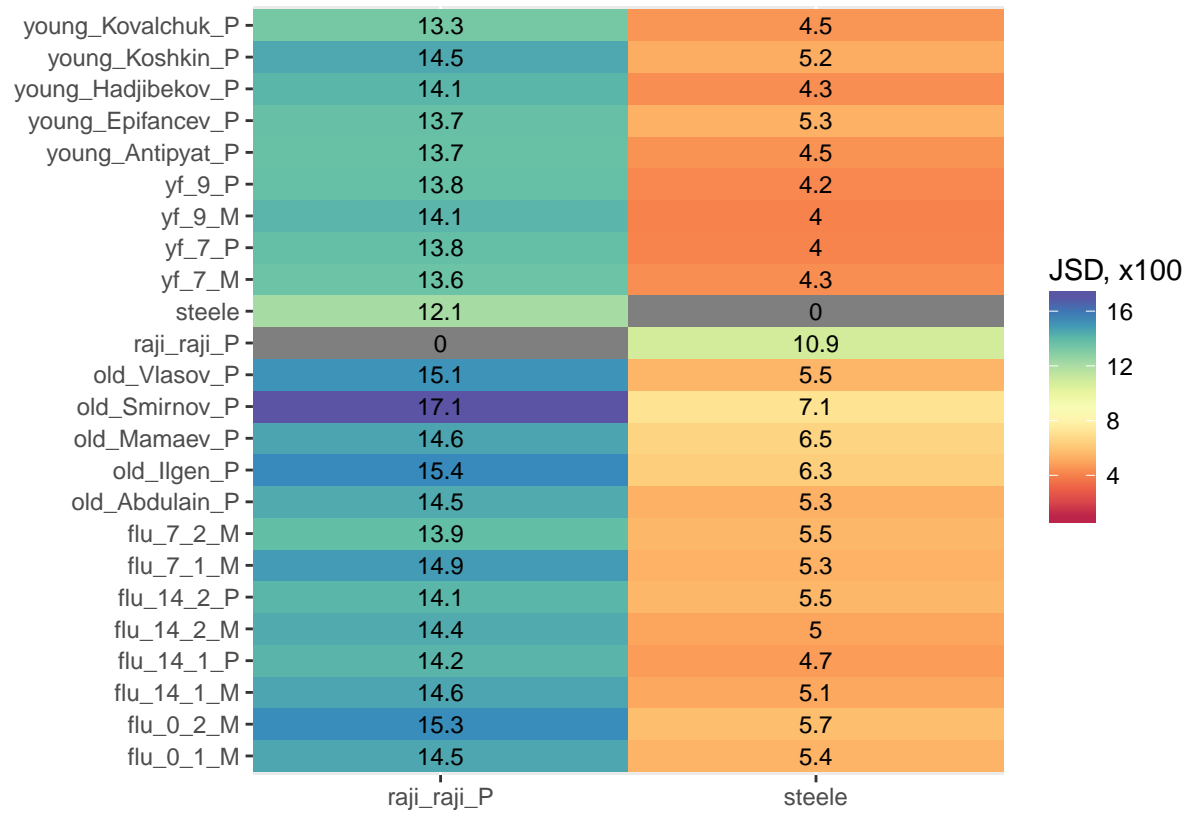


We compute Jensen-Shannon divergences (JSD), a metric that is commonly used to compare frequency distributions. The smaller the divergence, the closer are substitution frequency distributions. Of note:

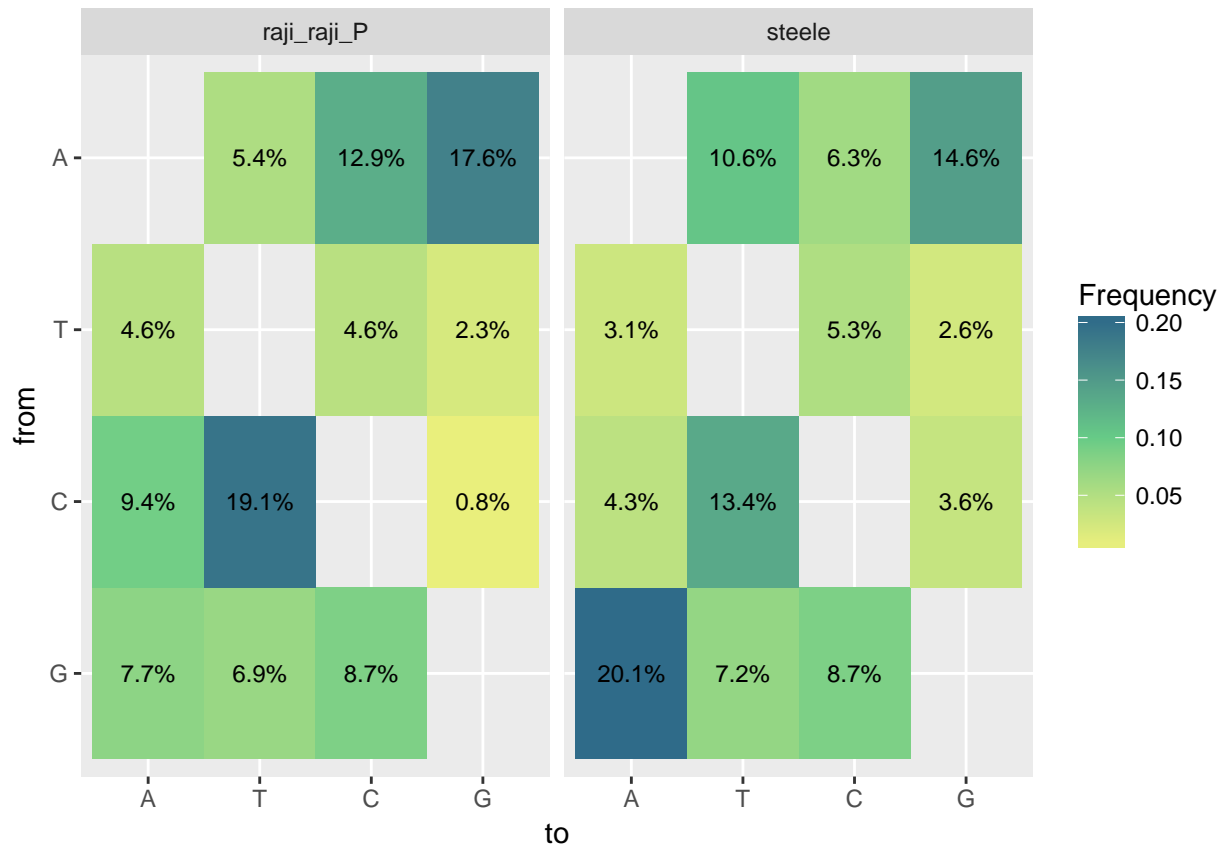
- Old donors appear to be outliers at this plot, but we cannot rule out batch effect in the absence of controls
- Samples for `flu` are highly correlated. Unfortunately this also includes control. All these samples come from the same donor.



Compare substitution frequency distributions of our samples with `steele` reference and `raji` cell line. Note that `raji` is an extreme outlier, this is quite obvious from the substitution frequency matrices given above. The data from `steele` is far more similar to our results, but still more than 2 times farther in terms of JSD distance from each sample than the sample if from its most distant counterpart in our vaccinated donor set.

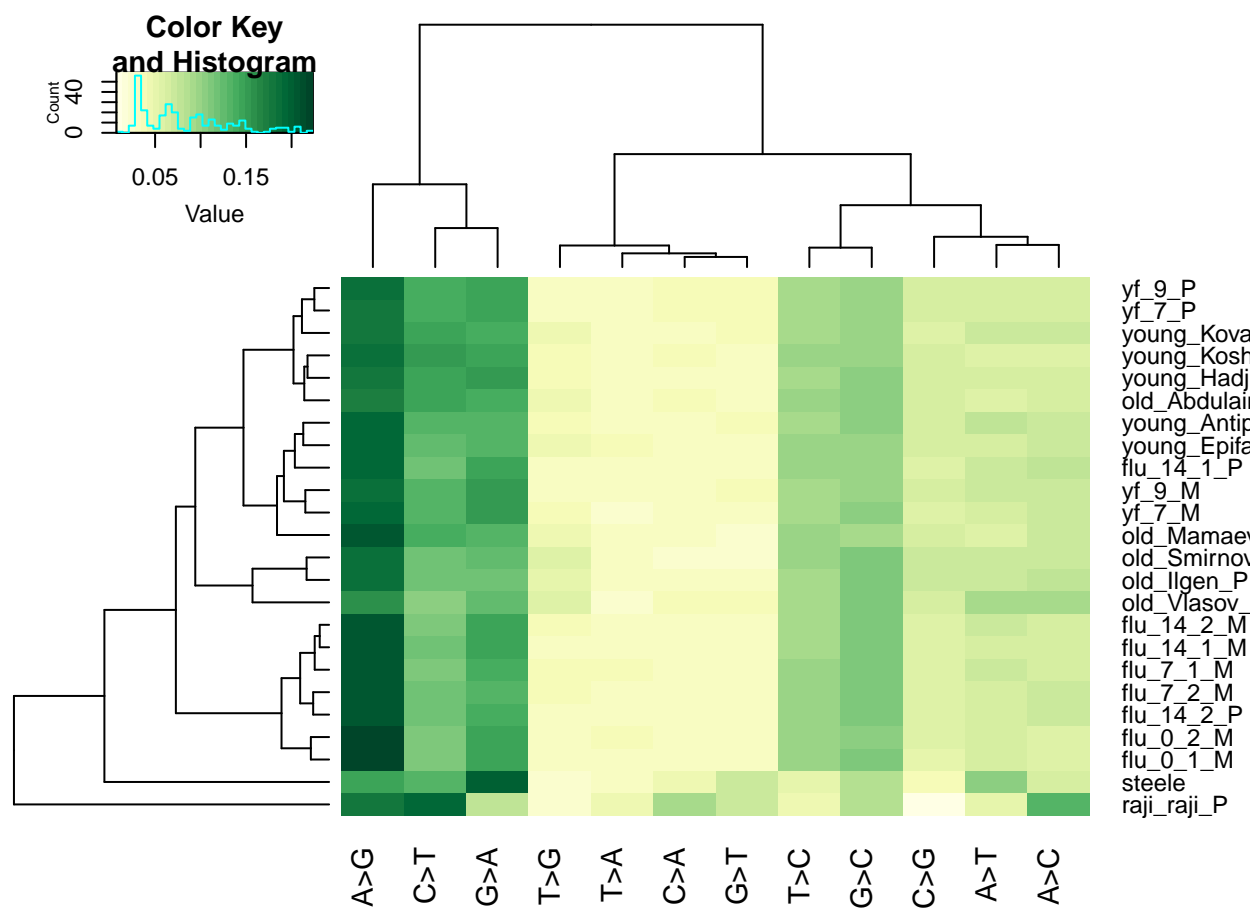


Once more **raji** and **steele** substitution frequencies side-by-side. The **C>>G** rule does not hold for **raji** sample.

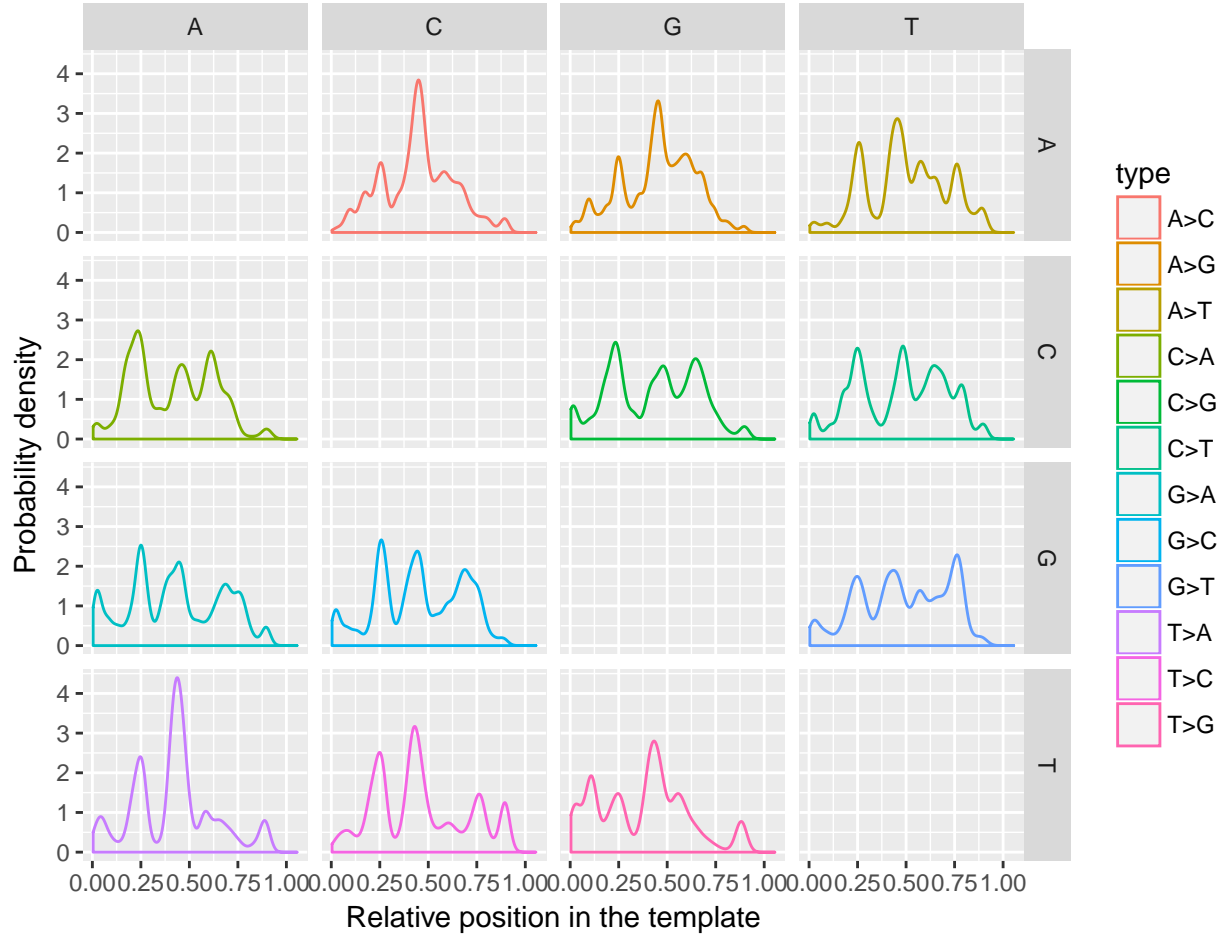


Clustering samples based on mutation profile

Using freq as value column: use value.var to override.



Distribution of substitutions by position in template



Mining for ADAR and AID signatures

Let us first define a set of four variables corresponding to ADAR/AID signatures:

The fraction of mutations originating from a given base type i is $f_i = \sum_{j \in A, T, G, C} f_{ij}$

- AID prevalence $AID_p = f_C + f_G$.
- AID strand bias $AID_s = f_G / (f_C + f_G)$.
- ADAR prevalence $ADAR_p = f_A + f_T$.
- ADAR strand bias $ADAR_s = f_A / (f_A + f_T)$.

The plot below shows the aforementioned values for each sample. Reference values for **raji** and **steele** are shown in **red** and **blue** respectively.

```
## Using proj, sample, cells, name as id variables
```

