

Appendix for:

Fragmentation patterns and personalized sequencing of cell-free DNA in urine and plasma of gliomas.

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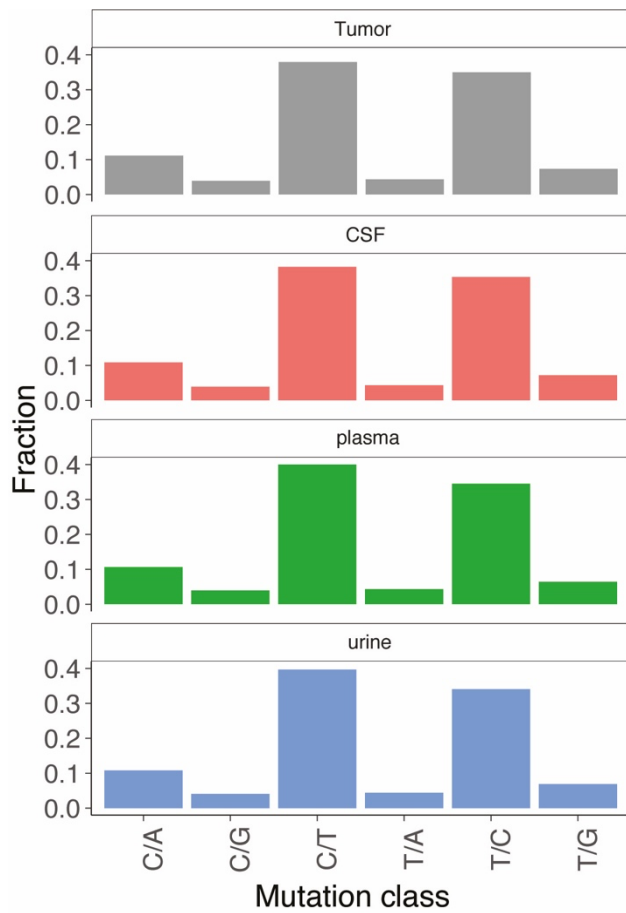
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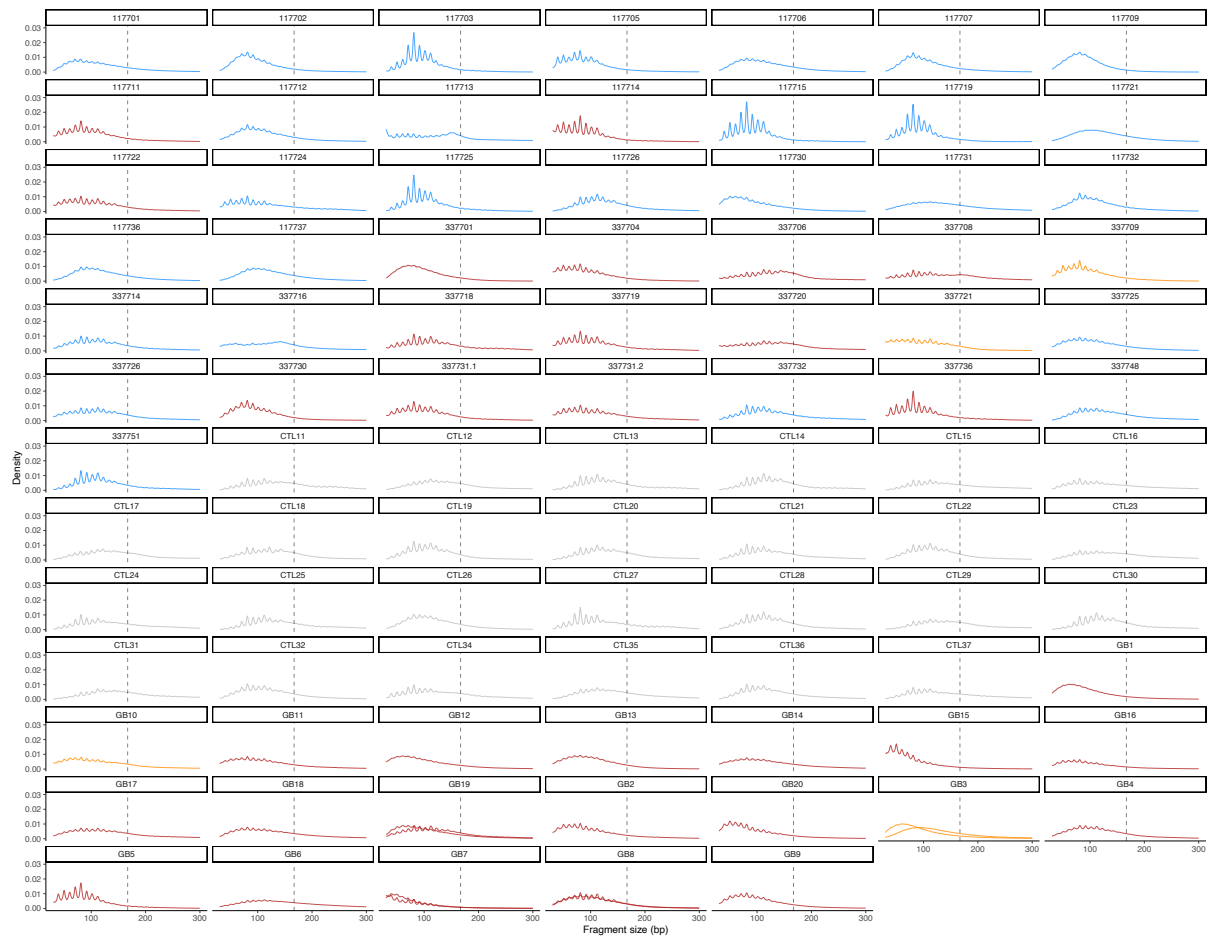
Appendix Figure S2:



Appendix Figure S2: mutation class depending on the tissue or body fluid sequenced.

In grey are displayed tumor samples, in red are displayed CSF samples, in green plasma samples and in blue urine samples.

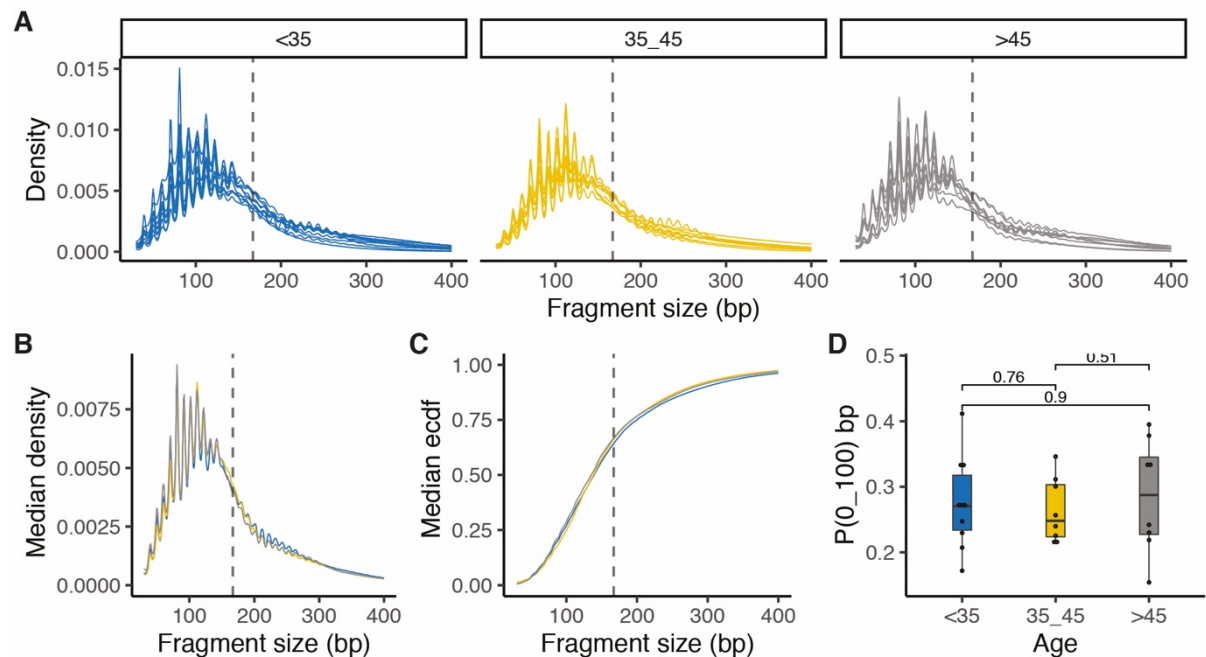
Appendix Figure S4:



Appendix Figure S4: cfDNA fragmentation of urine samples determined using SWGS per individuals.

The colors indicate HGG (in red), LGG (in orange), healthy cases (in grey) and other CNS diseases (in blue).

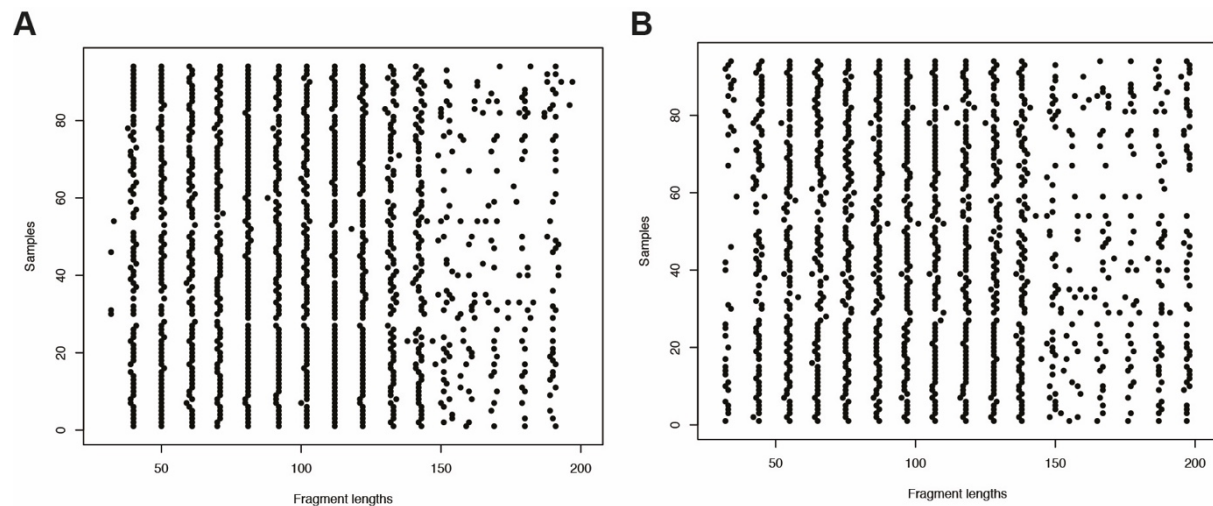
Appendix Figure S5:



Appendix Figure S5: Influence of the age on the urine cfDNA fragmentation from healthy individuals.

Colors represent individuals <35 years old (n=8, blue), between 35 and 45 years old (n=9, yellow), and >45 years old (n=8, grey). Age is unknown for 1 individual (not shown). The median age of the healthy individuals is 41 years old (range: 23-61). **A**: cfDNA size distribution per age group. **B**: median cfDNA size distribution. **C**: median empirical cumulative distribution (ecdf). KS-test showed no significant difference. **D**: proportion of cfDNA fragments <100bp depending on the age group. No significant difference between the group can be detected (Wilcoxon-test).

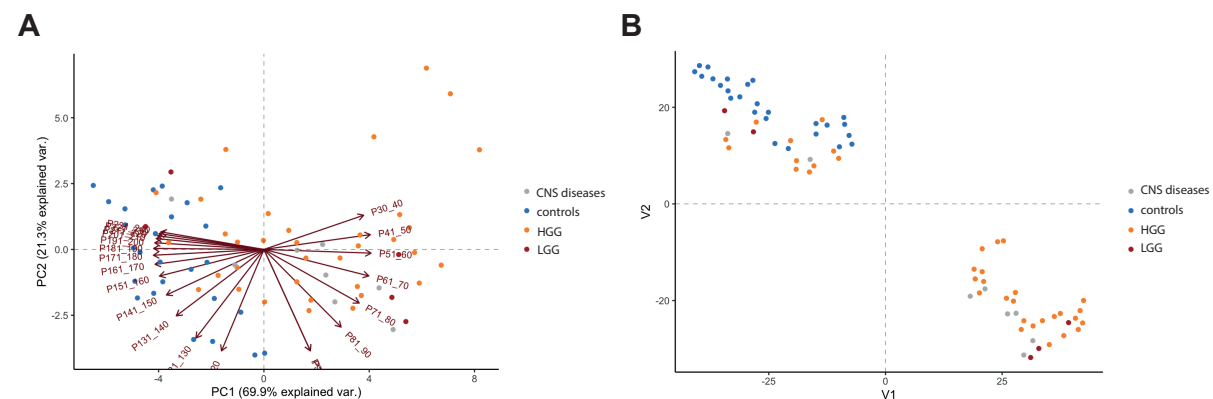
Appendix Figure S6:



Appendix Figure S6: Selecting “peaks” and “valleys” for the determination of the 10 bp oscillations amplitude.

Each dot represents the local maxima (“peak”) and local minima (“valley”) depending on cfDNA fragment size determined from the sWGS of the urine samples (see Methods). **A**: local “peak” depending on the fragment length. **B**: local “valley” depending on the fragment length. The value of “peaks” and “valleys” are conserved across samples in the size range 50-140 (corresponding to a sub-nucleosome structure), but not for smaller or larger values.

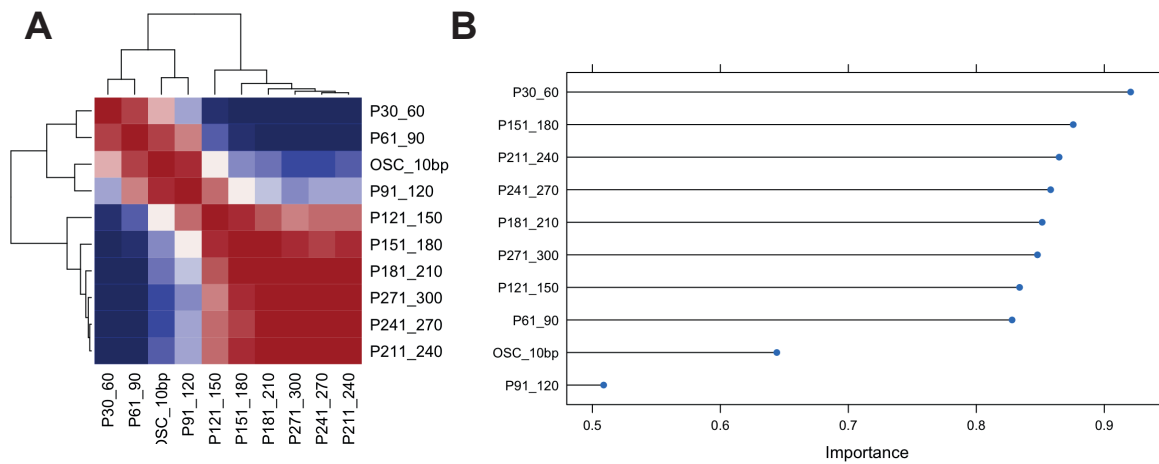
Appendix Figure S7:



Appendix Figure S7: Clustering of cfDNA fragmentation features recovered from sWGS using 10bp binning.

A: Principal component analysis comparing cancer (HGG and LGG) and control samples (healthy and other CNS diseases) using data from the urine fragmentation features. Red arrows indicate features tested during the predictive analysis. Fragmentation features were calculated from the cfDNA fragments size (The proportion of cfDNA fragments was calculated every 10 bp bins between 30 and 300 bp); and the amplitude of the 10bp oscillations: OSC_10bp). **B**: tSNE analysis comparing cancer and control samples using data from the same urine fragmentation features.

Appendix Figure S8:



Appendix Figure S8: Evaluation of the fragmentation features determined from sWGS of urine samples using the 30 bp binning. A: Correlation matrix of the 10 fragmentation features determined by sWGS from the 74 urine samples included in the training and validation dataset of the classifier models. The correlation score was estimated for each cross-comparison, and the value displayed on as a color intensity (red = 1, blue = -1). **B:** Ranking of the individual features importance calculated with a Learning Vector Quantization (LVQ) model.