**An analysis of intrinsic base substitution patterns in diverse species reveals connections to cancer and metabolism**

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**Abstract**

**Mutations in genomic DNA are the starting material for evolution by natural selection. Analyses of large-scale sequencing data have revealed that mutagenic processes often create distinctive patterns of base substitutions, called mutational signatures. In this work, we analyzed the base substitution patterns within mutational data from a large number of strains or individuals among seven model species (totalling >784 million mutations), as well as single nucleotide polymorphisms (SNPs) in 41 species from the NCBI SNP database (>603 million SNPs). We found that the intrinsic base substitution pattern for most of these species closely matches mutational signature 5 from the Catalog of Somatic Mutations in Cancer (abbreviated as COSMIC). Signature 5 is ubiquitous in cancers and normal human cells, suggesting that the similar patterns of mutation across many species are likely due to conserved biochemical processes. Finally, we show that a similar pattern of base substitutions can be obtained using a yeast model system that allows controlled generation of genomic single-stranded DNA. Taken together, we propose that intrinsic biochemical processes in cells are coupled to the continuous generation of mutations, which in turn, are acted upon by natural selection to drive the evolution of species.**

1  Mutagenesis is often thought of as a stochastic process, wherein the temporal and positional occurrence of each individual mutational event is not predictable (*1*).  And yet, sets of mutations induced by a given DNA damaging agent or process do create reproducible, characteristic patterns, called mutational signatures (*2*).  Mutational signatures derived from cancer genomics data have been published, which have yielded considerable insight into specific sources of DNA damage in humans (*3*).  These signatures provide a composite picture of the recurrent patterns of base substitutions found in cancers, and in some cases, normal tissues as well.  The observations in humans led us to investigate whether similar patterns of base substitutions are found when comparing genetic variation within and between other species.

We analyzed mutational data (784,102,480 base substitutions) from seven model species, comprising:  1,011 varieties of budding yeast (*Saccharomyces cerevisiae*) (*4*); 161 varieties of fission yeast (*Schizosaccharomyces pombe*) (*5*); 152 isolates of roundworm (*Caenorhabditis elegans*) (*6*); 85 isolates of fruit fly (*Drosophila melanogaster*) (*7*); 36 strains of mouse (*Mus musculus*) (*8*); 42 strains of rat (*Rattus norvegicus*) (*9*); and 49 individual dogs (*Canis familiaris*) (*10*).  When the data were plotted using the convention for reporting mutational signatures (six substitution types, each subdivided into 16 contexts with one flanking base on either side, 5′ and 3′) (*11*), it became clear that the majority of mutations in all seven species were either C→T or T→C transitions (see Figure 1A).  To assess the similarity of these base substitution patterns to known

mutational signatures, we computed the cosine similarity to the set of published

COSMIC signatures (see Figure 1B) (*2*).  For context, a cosine similarity value of

1 is obtained when comparing identical signatures.  The substitution patterns for

all seven model species were each most similar to COSMIC signature 5 (referred

to hereafter as COSMIC 5), which is one of two 'clock-like' mutational signatures

found in human, as the numbers of such mutations essentially always increase

monotonically with age (*12*).  COSMIC 5 also accounts for ~75% of de novo mu-

tations in healthy human children (*13*).  These results show that, just as COSMIC

5 is found ubiquitously in human, intrinsic base substitution patterns similar to

COSMIC 5 are found in multiple model species.

　　　　Given that COSMIC 5-like base substitution patterns are intrinsic to the

model species, we then investigated if similar patterns are observed among

SNPs in a larger, more diverse set of species.  We downloaded sets of available

VCF (variant call format) files from NCBI SNP for 44 species (*14*).  Two data sets

were obviously invalid, as they completely lacked two substitution types (C→G

and T→A, likely due to a data processing error), while another data set was

dropped from analysis because of probable sequencing artefacts (see Supple-

mentary Figure 1).  Among the remaining 41 species (yielding 603,722,210

SNPs, see Supplementary Figure 2), the large majority showed the highest co-

sine similarity to COSMIC 5 (see Figure 2).  A particularly interesting exception is

the mosquito (*Anopheles gambiae*), with the highest similarity to COSMIC signa-

ture 3, which is attributed to deficiency in homologous recombination DNA repair

(*3*).  In contrast to COSMIC 5, the mosquito base substitution pattern is relatively

'flat' across substitutions and trinucleotide contexts (see Supplementary Figure

2).  The bee (*Apis mellifera*) and the grape (*Vitis vinifera*) are also notable outliers

whose intrinsic base substitution patterns do not particularly resemble any COS-

MIC signature.  Such exceptions notwithstanding, intrinsic base substitution pat-

terns that resemble COSMIC 5 do tend to be shared among collections of SNPs

in many other organisms.  Taken altogether, the observations from prior studies

in human and our results here suggest the likelihood that conserved biochemical

processes in many species are responsible for the set of COSMIC 5-like intrinsic

base substitution patterns.

To quantify and visualize the extent of similarity or divergence among base

substitution patterns, we generated profiles showing means and 95% confidence

intervals of relative frequency for each mutation type (see Figure 3A).  For the

seven model species, the 95% confidence intervals for most substitution types

were small, i.e., indicating low variance (see Supplementary Figure 3A).  The

most highly variable trinucleotides with respect to differences in frequencies of

mutations were TTA and ATT for T→A transversions (see also Figure 1A).  Fre-

quencies of C→T at ACG and CCG motifs are also quite varied, in parallel with

T→C at ATG and CTG.  These motifs were particularly mutable in rat and dog,

but less so in the other model species.  Despite these instances of (relatively)

high variability, the mean values do provide an estimated consensus COSMIC 5-

like pattern among the seven model species. The cosine similarity between this

consensus pattern and COSMIC 5 itself is 0.928 (see Supplementary Figure 3B).

Likewise, we applied this analysis to the SNP data for the 41 species from

the NCBI database (see Figure 3B). The 95% confidence intervals for the means

were more compact, reflecting the larger number of species examined and yield-

ing a lower variance. Again, the substitution types with higher mean relative fre-

quencies showed more variability (see Supplementary Figure 3C). The consen-

sus pattern from SNP data very closely matched the corresponding pattern from

mutations among the seven model organisms, with a cosine similarity of 0.981

(see Figure 3C). All four NCG motifs exhibit more common C→T transitions in

the SNP data, suggestive of possible contributions from deamination of 5-methyl-

cytosine at those motifs in a subset of species (e.g., goat, horse, macaque,

sheep, etc.), similar to COSMIC signature 1 (*3*). Also similar to the model organ-

ism data, the consensus SNP pattern has a cosine similarity of 0.931 compared

to COSMIC 5 (see Supplementary Figure 3D). Taken together, we propose that

for many species, their respective sets of intrinsic base substitutions form a range

of related patterns, and COSMIC 5 is simply the corresponding human counter-

part in this continuum. Human is a special case in the sense that our species' in-

trinsic base substitution pattern also includes a ~25% contribution of COSMIC

signature 1 (*13*, *15*). If there is a COSMIC 1-like component to the intrinsic base

substitution patterns of other species as well, then it is comparatively minor and

outweighed by the COSMIC 5-like contribution.

In principle, many species can be used as models to study the molecular

mechanisms underlying these COSMIC 5-like mutational patterns, as these bio-

chemical mechanisms are likely to be well conserved.  We previously used a

budding yeast system which features controlled generation of genomic single-

stranded DNA (ssDNA) to elucidate the relative contributions of the APOBEC3A

and APOBEC3B cytidine deaminases to cancer mutagenesis (*16*).  This system

relies on the *cdc13-1* temperature sensitive mutation (*17*), so that shifting to 37°C

causes telomere uncapping, extensive end resection, and cell cycle arrest in $G_2$

via the DNA damage checkpoint (*18*).  Since the exposed ssDNA is much more

prone to damage than double-stranded DNA, this system can acquire mutations

for analysis much more readily.  When we sequenced the genomes that were al-

lowed to accumulate mutations, we obtained a pattern of base substitutions that

resembled the consensus base substitution pattern in the seven model species,

the consensus pattern from SNPs of 41 species, and COSMIC 5 (cosine similari-

ties = 0.941, 0.931, and 0.933, respectively, see Figure 4A and Supplementary

Figure 4).  This further underscores the generality of COSMIC 5-like intrinsic

base substitution patterns.

We then carried out genetic experiments to investigate molecular mecha-

nisms responsible for the mutational pattern.  Mutagenesis is dependent on the

translesion DNA synthesis (TLS) polymerases zeta and Rev1 (*19*), i.e., on the er-

ror-prone bypass of damaged bases (*20*) (see Figure 4B).  In contrast, deletion of

*RAD30* (encoding polymerase eta) had no effect, since Rad30 specializes in

error-free bypass of ultraviolet light-induced cyclobutane dimer photoproducts

(*21*) and apparently plays no role in mutagenesis opposite DNA damage that

gives rise to the COSMIC 5-like pattern.  Deletion of genes important for error-

free bypass also had no significant effect (see Supplementary Figure 5).  Further,

we found that energy metabolism is key for this form of mutagenesis.  Cells main-

tained in glucose (a fermentative, glycolytic energy source (*22*)) acquired muta-

tions most frequently, while cells in acetate (a carbon source that must be utilized

by aerobic respiration (*23*)) had the fewest acquired mutations (see Figure 4C).

Moreover, increasing the concentration of glucose from 2% (the amount normally

used in media) to 8% led to a five-fold increase in the frequency of mutations

(see Figure 4D).  Finally, blocking the last step of glycolysis using auxin-induced

degron alleles of *CDC19* (encoding pyruvate kinase) caused a large decrease in

mutagenesis (see Figure 4E), further corroborating the central role of energy me-

tabolism to generate the DNA damage which produces a COSMIC 5-like base

substitution pattern.

Taken altogether, our results suggest that the intrinsic base substitution

patterns of many organisms are linked to energy metabolism.  This would provide

a plausible explanation for why such similar patterns are found across many spe-

cies, including human, since the main pathways of energy metabolism are well

conserved (*24*).  This model would also imply that processes of energy metabo-

lism will inevitably produce DNA damage, which leads to mutations that fit a simi-

lar pattern across many species.  In turn, these mutations are the starting

material for evolution by natural selection.  Since energy generation is a prerequi-

site for living, the continuous basal level of mutagenesis which results would en-

sure that natural selection will have an ever-renewing repertoire of genetic vari-

ants to act upon.  We propose that the DNA damage itself is likely very similar

between species, but the resulting patterns will be more or less divergent de-

pending on the similarity of the respective sets of error-prone translesion synthe-

sis polymerases when comparing between species.  Species with polymerases

that are more closely related would have more similar patterns than those from

less related species.  Finally, a coupling between metabolism and COSMIC 5-like

mutational patterns could help explain why the extent of that type of mutagenesis

can be highly variable in cancers, when comparing between different patients

(*12*) or temporally within the same patient (*25*):  namely, such variability could be

linked to differences in energy metabolism between individuals or over the course

of disease progression in particular individuals.

Materials and Methods

Data Acquisition

VCF files were downloaded from the following sources: for *Saccharomy-*

*ces cerevisiae*, http://1002genomes.u-strasbg.fr/files/; for *Schizosaccharomyces*

*pombe*, ftp://ftp.pombase.org/pombe; for *Caenorhabditis elegans*, https://www.el-

egansvariation.org/data/release/20160408; for *Drosophila melanogaster*,

https://trace.ncbi.nlm.nih.gov/Traces/sra/sra.cgi?study=SRP050151; for *Mus*

*musculus*, ftp://ftp-mouse.sanger.ac.uk/; for *Rattus norvegicus*,

ftp://ftp.rgd.mcw.edu/pub/strain_specific_variants/Hermsen_et_al_40Ge-

nomes_Variants/; for *Canis familiaris*, http://www.dog10kgenomes.org/; and for

NCBI SNP data, ftp://ftp.ncbi.nih.gov/snp/organisms/archive/.

Mutational Analyses

Where possible, we used the MutationalPatterns R package (version

1.10.0) to extract mutational and trinucleotide contexts from VCF files, to plot mu-

tational signatures, to calculate cosine similarities, and to plot heatmaps (*11*).

We constructed BSgenome (*26*) reference genome packages in order to use Mu-

tationalPatterns to process VCF files if none of the stock reference genomes

were appropriate. In cases where an organism's reference genome proved to be

incompatible with BSgenome (or more precisely, its dependency GenomeInfoDB

(*27*)), we used a script that extracted mutational and trinucleotide context data.

This script used a combination of Unix shell functions, R (*28, 29*), and Samtools

(*30*). Summed matrices of mutational counts were loaded into

MutationalPatterns for graphics rendering in order to produce a consistent style for figures. Most analyses were run on a PC operating Ubuntu (Linux) version 18.04, with the remaining analyses run on a MacBook Pro operating macOS version 10.14.5 or 10.14.6. R version 3.6.0 was used on both computing platforms.

Reagents

Bacto peptone and yeast extract were purchased from Becton, Dickinson and Company (Franklin Lakes, New Jersey). Auxin and canavanine were purchased from MilliporeSigma (St. Louis, Missouri). Agar, carbon sources (i.e., glucose, galactose, raffinose, and potassium acetate), hygromycin, and G418 were purchased from BioBasic (Markham, Ontario).

Yeast Genetics

All yeast strains were constructed in the ySR127 background, a MAT alpha haploid with the *cdc13-1* temperature sensitive mutation. In addition, there is a cassette of three reporter genes (*CAN1*, *URA3*, and *ADE2*) near the de novo left telomere of chromosome V. Further details on ySR127 have been described previously (*31*).

Gene knockouts were constructed by one step gene replacement using antibiotic resistance marker cassettes (*32*). To construct the CDC19 degron strains, we first introduced the *Oryza sativa* TIR1 gene cassette (OsTIR1) into the HO locus of ySR127, using the protocol described in (*33*). We then inserted the 3mAID degron tag (containing three tandem repeats of the core domain of the IAA17 protein from *Arabidopsis thaliana*), using the protocol described in (*34*).

Mutagenesis experiments were initiated by inoculating single colonies

each into 5 mL of rich media (2% Bacto peptone, 1% Bacto yeast extract, 2%

carbon source, supplemented with 0.001% adenine hemisulfate) in round bottom

glass tubes.  Cells were grown at permissive temperature (23°C) for three days.

Then, cells were diluted ten-fold into fresh media and shifted to restrictive tem-

perature (37°C).  Auxin (iodoacetic acid) was added to 1 mM final concentration

for degron experiments.  After the temperature shift, cells were collected by light

centrifugation, washed in water, and plated onto synthetic complete media to as-

sess survival and onto canavanine-containing media to select for mutants.  Fur-

ther details of this procedure are described in (*35*).  Genomic DNA from mutants

of interest were extracted and sequenced essentially as described previously

(*16*).  Illumina sequencing was outsourced to Genome Quebec (McGill Univer-

sity).  We used Bowtie 2 (*36*), SAMtools (*30*), and bcftools (*37*) to map reads and

call variants.

Figure Legends

Figure 1.  (A) 96-bar graphs showing the intrinsic base substitution pat-

terns of seven model species, derived by analysis of data from (*4–10*).  The pat-

terns tended to show mostly C→T or T→C substitutions.  Interestingly, both in-

vertebrate species and fission yeast had prominent T→A peaks at ATT and TTA

motifs.  (B) Heatmap showing cosine similarity values to 30 signatures from

COSMIC.  For all seven species, the highest cosine similarity values suggested

closest resemblance to COSMIC signature 5.

Figure 2.  Heatmap depicting cosine similarity values when comparing sets

of SNPs from 41 species vs. the 30 COSMIC signatures.  Most species had the

highest cosine similarity values compared to COSMIC signature 5.  See Supple-

mentary Figure 2 for the 96-bar substitution patterns of these species.

Figure 3.  (A) 96-bar graph showing mean and 95% confidence interval for

relative frequencies of each substitution type among the seven model species.

(B) 96-bar graph showing mean and 95% confidence interval for relative frequen-

cies of each SNP type among the 41 species from NCBI.  (C) The mean values

of the two profiles are shown, along with the difference in relative frequencies

when one is subtracted from the other.  The cosine similarity is quite high (0.981).

Figure 4.  (A) 96-bar graph showing the base substitution pattern resulting

from damage to exposed single-stranded DNA in yeast.  (B) The inactivation fre-

quency of the *CAN1* reporter was dependent on *REV3* (encoding the catalytic

subunit of polymerase zeta) and *REV1*, i.e., on error-prone translesion DNA

synthesis.  Deletion of *RAD30* (encoding polymerase eta) had no effect.  Cells

were temperature shifted to 37°C for 6 hours.  (C) The frequency of mutagenesis

was dependent on carbon source.  Acetate (which must be metabolized by aero-

bic respiration) produced the fewest mutations.  Glucose (which is metabolized

fermentatively) produced the most.  Galactose and raffinose (which have a mixed

'respirofermentative' metabolism were intermediate.  (D) Increasing glucose con-

centration to 8% led to a significant increase in mutagenesis within single-

stranded DNA.  Controls at 23°C did not show any significant differences

throughout the tested range of glucose concentrations.  Cells were shifted to

37°C for 24 hours.  (E) Mutagenesis was dependent on completion of glycolysis.

Inducible degradation of Cdc19 (pyruvate kinase) blocked the last step of the gly-

colytic pathway and caused over a ten-fold decrease in mutagenesis.  Cells were

shifted to 37°C for 24 hours in the presence or absence of 1 mM auxin.

Supplementary Figure 1.  (A) Analysis of the SNP dataset for apple (*Malus*

*domestica*) revealed an apparent data processing issue that deleted all C→G

and T→A substitutions.  (B) The SNP dataset for flycatcher (*Ficedula albicollis*)

was also affected by this issue.  (C) The SNP pattern for *Nomascus leucogenys*

(the northern white-cheeked gibbon, but referred to simply as "primates" on

NCBI) was also suspiciously unusual with very high peaks at the reciprocal motifs

ACA and ATA, as well as ACG and ATG, for both C→T and T→C substitutions.

As this was very much an outlier, this dataset was dropped for further considera-

tion due to suspected sequencing artefact.

Supplementary Figure 2.  96-bar profiles showing the intrinsic base substi-

tution patterns inferred from SNPs of 41 species downloaded from NCBI.  Most

patterns showed, to varying extents, a preponderance of C→T and T→C substi-

tutions where these substitution types were similar in overall proportion, forming

a range of COSMIC 5-like patterns.  Exceptions to this overarching meta-pattern

were mosquito, bee, grape, and corn.

Supplementary Figure 3.  (A) Scatter plot of 95% confidence interval sizes

vs. their corresponding mean relative frequency for the 96 substitution types

shown in Figure 3A (model species).  Most of the variance among the patterns

for the seven model species was found among substitution types that have over

0.01 relative frequency.  (B) Comparison of the mean relative frequencies among

the seven model species to COSMIC 5.  Root sum of squares (RSS) and cosine

similarity values are shown.  (C) Scatter plot of 95% confidence interval sizes vs.

their corresponding mean relative frequency for the 96 substitution types shown

in Figure 3B (NCBI SNP data).  Notice the 95% confidence intervals are smaller

than in (A) because of the larger number of species among the SNP data analy-

sis.  (D) Comparison of the mean relative frequencies among the 41 species with

NCBI SNP data to COSMIC 5.

Supplementary Figure 4.  Comparison of the base substitution pattern ob-

served in the yeast single-stranded DNA vs.: (A) the consensus pattern from

seven model species; (B) the consensus pattern from the 41 species from NCBI

SNP; and (C) COSMIC 5.  Root sum of squares (RSS) and cosine similarity val-

ues are shown as well.

Supplementary Figure 5.  Deletion of genes involved in homologous re-

combination (*RAD51*, *RAD52*) or template switching to bypass base lesions

(*RAD5*, *RAD18*, *MMS2*) had no significant effect on mutagenesis after correction

for false discovery from multiple testing.

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# Figure 1A

# Figure 1B

# Figure 2

# Figure 3C

## Figure 4A

## Figure 4B

## Figure 4C

# Figure 4D

Glucose Concentration

# Figure 4E

# Supplementary Figure 1

# Supplementary Figure 2

# Supplementary Figure 2, continued

# Supplementary Figure 2, continued

# Supplementary Figure 2, continued

# Supplementary Figure 2, continued

# Supplementary Figure 2, continued

# Supplementary Figure 3A

# Supplementary Figure 3B

## Supplementary Figure 3C

# Supplementary Figure 3D

# Supplementary Figure 4A

# Supplementary Figure 4B

# Supplementary Figure 4C

# Supplementary Figure 5