# DNA Tandem Repeats as Iterable Objects to Count Cell Divisions: A Computational Model 

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#### Abstract

Cell lineages of nematodes are completely known: the adult male of Caenorhabditis elegans contains 1031 somatic cells, the hermaphrodite 959, not one more, not one less; cell divisions are strictly deterministic (as in the great majority of invertebrates) but so far nothing is known about the mechanism used by cells to count precise numbers of divisions. In vertebrates, each species has its invariable deterministic numbers of somites, vertebrae, fingers, and teeth: counting the number of iterations is a widespread process in living beings; nonetheless, it remains an unanswered question and a great challenge in cell biology. This paper introduces a computational model to investigate the possible role of satellite DNA in counting cell divisions, showing how cells may operate under Boolean logic algebra. Satellite DNA, made up of repeated monomers and subject to high epigenetic methylation rates, is very similar to iterable sequences used in programming: just like in the "iteration protocol" of algorithms, the epigenetic machinery may run over linear tandem repeats (that hold cell-fate data), read and orderly mark one monomer per cell-cycle (cytosine methylation), keep track and transmit marks to descendant cells, sending information to cell-cycle regulators.


## Keywords

Satellite DNA, Tandem-Repeats, Epigenetics

## 1. Introduction

Scott F. Gilbert, in his Developmental Biology, wrote: "How do our cells know when to stop dividing? If each cell in our face were to undergo just one more cell division, we would be considered horribly malformed. If each cell in our arms underwent just one more round of cell division, we could tie our shoelaces without bending over. Our arms are generally the same size on both sides of the

## body. How is cell division so tightly regulated?" [1]

The lineage of the roundworm Caenorhabditis elegans is absolutely invariable, autonomous, and genetically established without exceptions [2]: its midgut progenitor (E blastomere), isolated and cultured in vitro in the absence of morphogens, undergoes five (incomplete) division rounds as in the whole embryo [3]; an embryonic intestine of 20 cells is generated, not $32\left(2^{5}\right)$ because only 4 cells (unerringly the same) carry out the fifth division, while the remaining 12 stop dividing after four rounds. Similarly, in the embryo of the fruit fly Drosophila melanogaster, the early 13 divisions are precisely counted [4] without active morphogens, that will work later [5] or other molecular clocks (early divisions occur in a syncytium without cell-cell communication involving transmembrane signaling; maternal morphogenetic mRNAs have different concentration gradients along the developmental syncytium). Abouchar et al. [6] measured the precision of left-right and inter-individual fly wing vein patterns: "wing vein patterns are specified with identical spatial precision and are reproducible to within a single-cell width. The early fly embryo operates at a similar degree of reproducibility, suggesting that the overall spatial precision of morphogenesis in Drosophila performs at the single-cell level. Could development be operating at the physical limit of what a biological system can achieve?" Interestingly, the genome of Caenorhabditis elegans [7] contains large amounts of well-conserved Tandem Repeats (TRs) and $21 \%$ of the Drosophila genome is made of TRs [8]. "How is cell division so tightly regulated?" No tricks and no magic: an intrinsic and autonomous molecular mechanism capable of counting cell divisions at the physical limit of single-cell level must exist: precise, independent from external signalings (morphogens), and unaffected by environmental noise.

### 1.1. Satellite DNA and DNA Methylation

Satellite DNA (satDNA), so named because DNA centrifugation over a density gradient forms a "satellite" band, consists of huge arrays of TRs, i.e. sequences of non-coding DNA, each one made up of the repetition of its characteristic monomer, sometimes not perfectly identical [9] [10] [11]; many TR sequences are evolutionarily well conserved [12] indicating they play a role in executing biological tasks. TRs have a relevant part in morphogenesis in invertebrates and vertebrates: in butterflies, wing patterns show a basic plan controlled by noncoding DNA to realize the diversity of wings of different species: TRs involved in wing patterning [13] have been conserved over millions of years allowing wing patterns to evolve fast [14]; in dogs, different TRs, found in genes that control morphological variations, cause a rapid breed evolution [15]: anatomical differences appear strictly linked to numbers of cell divisions and TRs mutations [16]. SatDNA is a strong accelerator of evolution because of its surprising self-remodeling rate: monomer similarity [9] [10] [11] causes frequent DNA-polymerase slippages [17]; homologous recombinations [18] of tandem repeated monomers constitute a major impediment for the accurate DNA repair of nucleotide loss in
double-strand breaks: TRs polymorphism and rate of errors during DNA replication is $10^{5}$ times higher than single-point-mutations [16]. During evolution, DNA-polymerase slippages, non-disjunctions, recombinations, unequal cross-ing-over, rolling-circle replications, inversions, and multiple transpositions have generated diverse satDNA families, highly polymorphic and variously assembled in clusters of different textures and patterns [19] [20].

SatDNA is actively methylated [21] and plays an important role in epigenetics [22] [23] [24]; epigenetic tools comprise specialized enzymes [25] able to recognize the particular 3D structure of their target sequences: "writers" [26] catalyze biochemical modifications (e.g. DNA methyltransferases: methylation on the 5th position of the pyrimidine ring of cytosine, $C \rightarrow 5 \mathrm{mC}$ ), "readers" [27] recognize and interpret the significance of such modifications and recruit a set of methyl CpG-binding proteins and "erasers" [28] remove methylations [29] [30] [31]. Hypomethylated TR arrays have been found in different human diseases, cancer, and psychiatric disorders [32].

### 1.2. Boolean Algebra

Boolean algebra differs from elementary algebra in that its only values are "true" and "false" not numbers as in elementary arithmetic and algebra. Boolean algebra does not executearithmetic operations such as the addition of numbers but uses logical operators (e.g. conjunction): Boolean algebra takes advantage of logical operations, as elementary algebra uses numerical operations; in cell biology, Boolean operations are realized by the 3-dimensional (3D) structural shapechanging of a protein (e.g. a receptor) after binding with its ligand (binding may assume the only values of "true" or "false" and so the correlative 3D structural change). SatDNA is made up of sequences of similar monomers as pearl necklaces: it resembles arithmetic tools like abaci, where it is possible to count without using numbers. Following these fundamentals, iterations on linear substrates from a start point to a stop (i.e. counting a fixed number of repetitions) are common: DNA replication (strictly time-bound to DNA methylation: see further), DNA transcription, mRNA translation, or dyne in movements on microtubules provide some clear examples.

## 2. Materials and Methods

The program "Lineages" takes as INPUT the following in silico 3 ' $\rightarrow$ ' DNA sequence (named "sat_DNA"):

## ATTCCAACGGCTTAATTCCCACGGCTTAATTCCGACGGCTTAATTC CCACGGCTTAATTCCGACGGCTTAATTCCCACGGCTTAATTCCCACGG CTTAATTCCTACGGCTTA

This sequence matches with many eukaryote genomes (fish, arthropods) sharing up to $78 \%$ similarities (Fasta: 2.9E-37): in the worm Alitta virens it matches with five different loci of chromosome 4 , six loci of chromosome 7 , and eleven loci of chromosome 9. The "sat_DNA" sequence is made up of 8 quasi-palindromic
monomers, each one of 14 nucleotides, almost (but not completely) identical, with palindromic terminal consensus motives ( $3^{\prime}$ ATTC...CTTA 5'), necessary as recognition sites for epigenetic "readers" [33].

The program returns as OUTPUT cell lineages, shown step by step, division after division, obtainable through simulated possible evolutionary events that occurred on "sat_DNA" (see below).
Symbols in the program "Lineages"
" $M$ " = 5-methyl-cytosine ( 5 mC );
" $H$ " = 5-hydroxymethyl-cytosine (5HmC)
Python is a high-level language, very popular among software developers: its "for...in" and "while" loops perform repetitive tasks, traverse and scan iterable objects like lists and strings, one item at a time, simulating processive cellular functions executed step by step on linear templates (replication, transcription, translation); also "if/elif/else", "break", "continue", "and", "or", "not", "pop()" simulate biological functions: they reproduce (in silico) biological (in vivo) pathways that may result "True" or "False" (e.g. ligand/receptor matching, concentration level reaching); on the contrary, "classes, matrices, multidimensional arrays, recursive functions", indispensable in programming, are too different from biological mechanisms: for these reasons, only the aforementioned "biocompatible" statements have been usedin the program "Lineages": the code is not "elegant", but reproduces correctly biological operative modes.

## 3. The Program "Lineages": Purpose

The program "Lineages" shows that three common biochemical pathways: $i$ ) step-by-step progression, ii) cytosine methylation, iii) epigenetic marks) can work on TRs of satDNA to count the number of cell divisions. During the process of replication, the epigenetic machine copies (i.e. replicates on the still unmethylated new strand) cytosine methylations, and, at the same time, looks for marks (Cor 5 mC as the4th nucleotide of each monomer, C or T in the 5 th position of each monomer: read further).Changes introduced into the "satDNA" sequence by epigenetic methylations have beencomputed to simulate how epigenetic "writers" and "readers" could manage the count of cell divisions.

The standard monomer ( 3 ' $\rightarrow 5$ ') of "sat_DNA"
$\begin{array}{llllllllllllll}\mathrm{A} & \mathrm{T} & \mathrm{T} & \mathrm{C} & \mathrm{C} & \mathrm{A} & A & C & G & G & C & T & T & A\end{array}$
$\begin{array}{llllllllllll}1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 & 11 & 12 \\ 13 & 14\end{array}$
has a consensus motif $\underline{A C G G C T T A}$ (nucleotides 7 th $\rightarrow 14$ th) recognized by epigenetic enzymes and non-coding RNAs [33] [34] and three remarkable nucleotides (the 4th, 5th and 6th). In eukaryotes, DNA-methyltransferases are very sensitive, precise, and nucleotide specific: with the cooperation of non-coding RNAs [35] they can recognize 3, 2, or even only 1 nucleotide [36] [37]. In the above monomer, the 4th, 5th and 6th nucleotides have each their epigenetic peculiar task:

- the 4th nucleotide (cytosine) is a mark used in the program as a "flag-point":
when a monomer is read (checked to know cell fate information) it is "flagged" and silenced (turned OFF) [38] [39] by methylation: $\mathrm{C} \rightarrow 5 \mathrm{mC}, ~ " C$ " $\rightarrow$ " $M$ "in the code
- the 5th nucleotide of each monomer indicates the division fate of the cell: thymine " $T$ " states the entrance in quiescence i.e. no more divisions [40] while cytosine " $C$ " states the entry into a new division cycle.

In the first monomer, the $5^{\text {th }}$ nucleotide may be different: " $T$ " states the entrance in quiescence (G0-phase), " $C$ " indicates a new symmetric division into two "twin sister" cells, " $H$ ", hydroxymethylated cytosine ( 5 HmC ) more stable and less subject to spontaneous deamination (Hahn et al., 2020) indicates the asymmetric division of a stem cell [41] [42] that produces an identical self-renewed stem cell and one "unique daughter" differentiating cell.

- the 6th nucleotide of the first monomer of the "sat_DNA" sequence, adenine, is the START signal.
- the 6th nucleotide of the last monomer, thymine, is the STOP signal. In the other monomers, the 6th nucleotide may be guanine or cytosine.

As output the program shows the lineages achievable when the user of the program operates a possible evolutionary " $C$-to- $T$ " mutation by changing (overwriting) a " $C$ " with a " $T$ " in the 5 th position of one or more monomers before running the program (Figure 1).
"C-to-T" transition [44] is a consequence of spontaneous cytosine deaminations, whose rate is estimated to be 100 to 500 cytosines per cell per day [45]; cytosine deamination is a strong modulator of genomic potential: it produces uracil, that, if not immediately repaired and substituted, will be replaced by thymine after two rounds of replication, resulting in a " C -to-T" mutation.
(A)

(B)


Figure 1. (A) The lineage that results without any $C / T$ change. (B) In the 3rd monomer of "sat_DNA" the cytosine in the 5th position has been replaced by a thymine. the resulting lineage is shown. [Warning: 2D printed lineages do not respect the real 3D geometrical distribution of cells [43]: like genealogical trees and pedigrees, cell lineages trace progenitors and offspring of each cellin two dimensions (2D)].

The substitution of " $C$ " with " $T$ ", artificially operated by the user of the program in the 5thposition of any monomer, simulates casual evolutionary events that realize different cell lineages: the cell fate coded by the 5th nucleotide changes from "keep dividing" (" $C$ ") to "stop dividing (" $T$ "): natural selection conserves the better suitable cell lineages.

### 3.1. Running the Program "Lineages"

Epigenomic molecular mechanisms involved in DNA methylation work on linear nucleotide sequences in a fashion like the "iteration protocol" of algorithms. The computational model introduced here aims to show how these epigenetic mechanisms can count cell division rounds and keep track of their progressive movements on satDNA sequences: they can proceed linearly on TRs, read one monomer at a time, and leave a flag (cytosine-methylation) as a mark-record for descendant cells, sharing the same Boolean logic algebra that governs algorithms [46] [47] [48] [49].

The program "Lineages" simulates a possible cellular upstream preliminary check-point for entrance in mitosis: DNA replication (S-phase) generates two copies of the genome that must be converted back to symmetrically methylated DNA (reproducing the methylation pattern of mother cell DNA before the next S-phase, to avoid the loss of previous marks). Methylation maintenance, which occurs during chromatin re-assembly, is a fundamental epigenetic process for ensuring that methylations are copied and transmitted to daughter cells: methyltransferase enzymes are recruited to DNA methylation sites where they recognize specific epigenetic modifications on DNA strands and histones [49] [50]; the "while loop" at code line 208 (after the comment "\# RUNNING THE PROGRAM") simulates the work of epigenetic enzymes [51] that first recognize the start point on satDNA sequences [32] [36] [37] [52] [53] and then traverse the repetitive sequence:
i) readers and writers [25] run (iterate) the sequence, step by step, linearly, one monomer at a time, to check the state of each monomer (4th nucleotide)
ii) 4 th nucleotide $=$ " $M$ " means: "already read" (worked and silenced by methylation, 5 mC : please, go on and check the next monomer)
iii) 4 th nucleotide $=$ " $C$ " means: "not yet read" (not methylated because not yet checked: please, stop here and see for the fate of the present cell stated by the 5th nucleotide).
iv) The "sat_ $D N A$ " sequence is read sequentially by the epigenetic "readers" until the first "not yet read" monomer (not switched-OFF, i.e. 4th nucleotide = " $C$ ") is found: at this point, the processive machine stops, the monomer is transcribed into a non-coding RNA and its information about cell-fate (5th nucleotide $=$ " $C$ " or " $T$ ") is carried to cell-cycle regulators; a new "flag", methylation of cytosine in4th position (" $C$ " $\rightarrow$ " $M$ " in the code) is left by the "writer" enzyme (de novo methylation [24]); so, the current monomer changes its state to "already read" and updates the "bookmark-flags" (epigenetic cellular memory for
the offspring). After reading and sending fate information to cell-cycle regulators, the current round of epigenetic work stops, and the epigenetic complex disengages from DNA.

To simulate the asymmetric process of methylation maintenance (that distinguishes the fate of sister cells, see further) in the code it is supposed that one cell, named "first cell", inherits the DNA containing the maternal filament with the old-methylated "sat_DNA" sequence, whereas the other daughter cell "second cell" receives the identical filament but newly synthesized and newly methylated; in the program, it is supposed that the process of methylation maintenance (see further) in the "second cell" methylates one more monomer to generate asymmetry between the two sister cells: so, the position of flags ( 5 mC ) differ for one monomer between the two daughter-sister cells (in the "second cell" the flag, 5 mC , is shifted one monomer forward): this way, sister cells start reading "sat_DNA" from different positions, and find their private fate on different monomers.

### 3.2. How to Use the Program

The program "Lineages" may be easily executed by "not-pythonist" users: Google Colaboratory, commonly known as Google Colab, is a free cloud-based Python environment that provides a platform for writing and executing Python codes through one"s own browser without installing (the only requirement is a free Google account). For a short guide to Google Colab visit Dataquest: https://www.dataquest.io/blog/getting-started-with-google-colab-for-deep-learni ng/:
A) access to Colab B) once you're on the Google Colab interface, click on File >> New notebook to create a new notebook C) on the left-hand side, click on the folder icon to open the file browser D) copy the Word version of "Lineages" attached in Supplementary Material E) paste it ( $\mathrm{Ctrl}+\mathrm{V}$ ) on the line marked by a little black circle with a white triangular arrow inside F) press "Runtime" to run the program.

### 3.3. General Instructions

1) Copy the program "Lineages" attached in Supplementary Material and paste it into a Python environment.
2) Run the program: in a few seconds the complete lineage will appear on the shell: going back on the screen it is possible to follow each division round, cell by cell, and read the epigenetic modification (methylations) on the "sat_DNA" sequence.
3) Cancel (remove) " $C$ " in the 5 th position in one or more monomers and insert (write) " $T$ " in the same position: the relative lineages will appear. (The last two monomers, useful for the next generation, do not affect the lineage because the program stops after four division rounds, i.e. after generating four sets of offspring: 30 cells is the maximum population).
4) Cancel " $C$ " in the 5thposition in the first monomer and write " $H$ " to show the division of a stem (mother) cell in one (daughter) differentiating cell and one identical stem cell.

Instructions are also reported at the beginning of the program.

## 4. Results

"Lineages" is a computational model that wants to investigate the basic principle and steps of a possible biological pathway to count cell divisions and understand what are the actual molecular mechanisms.

In algorithms, iterators traverse data containers and access their elements one at a time, returning data and keeping track of the visited items; "Lineages" follows this same paradigm: it tests an in silico DNA repeated sequence ("sat$D N A$ ") as a molecular "iterable object" capable of counting cell divisions and returns the resulting lineages. "Lineages" tries to ascertain if TRs can follow the rules of Boolean algebra, necessary to count deterministic numbers of cell divisions without using numbers. "Lineages" operates with the same linear processivity as DNA replication, simulating epigenetic reactions (DNA methylation) whose patterns and paradigms are known to serve also as a template for histone reassembly during S phase [54].

SatDNA structure and composition suggest that it could effectively be used to count: made up of a sequence of similar monomers as a "pearl necklace", its resemblance with arithmetic tools is very attractive and appealing: abaci and ancient religious objects used to count prayers, such as rosaries and misbahahs, all comprised of sequences of similar "monomers", allow counting without numbers: if the user is equipped with many different tools of different lengths, each corresponding to some particular task, once the proper sequence has been chosen, the user enumerates its objects, one by one, and executes a precise amount of iterations without numbers. Indeed, satDNA comprises many sequences of different lengths, each distinguishable through its characteristic monomer. SatDNA is the only molecular tool that possesses a structure compatible with counting deterministic numbers of iterations without numbers. Indeed, the hypothesis that tandem repeats might work as genomic counters came up years ago [55]: telomere shortening, a much debated and controversial subject (telomeres, the ends of linear chromosomes, are made up of TRs) had been suspected to work as a "replicometer" [56]; telomere shortening occurs during cell division and has been associated with the replicative capacity of cells, in the sense that its shortening could limit the remaining number of divisions, causing cell senescence [57].

The computational model "Lineages"tests a different method to count cell divisions: no losses of material (telomere shortening is a consequence of rough nucleotide depletions), rather precise, intrinsic, and autonomous molecular mechanisms capable of running over TRs, using, as said, the same Boolean logic-algebra that governs algorithms [46] [47] [48] [49]. To count cell divisions, ordered natural numbers are not necessary for cells: different lengths of satDNA se-
quences [58] [59], selected during evolution to satisfy the iterative needs of the species, are enough to count different amounts of iterations. Big numbers of cell divisions may be counted on long CpG and non-CpG islands or on TR sequences that are re-counted several times: in the model organism Danio rerio (zebrafish), methylations of the "TGCT" monomer are inherited from maternal and paternal gametes, erased in mid-blastula transition, and de novo re-established in gastrulation in all embryonic layers [60]. The process of DNA methylation is coupled with replication and drives the movements of the replication fork (see: Ryba et al.: "The Temporal Order of DNA Replication Shaped by Mammalian DNA Methyltransferases" [61]). Like a metronome, cell divisions scan growth and cell fate [62]; rather than a simple metronome that controls cell divisions, a "differential metronome" must be considered: during mitosis, mother cell DNA is copied and both sister cells inherit an identical copy of DNA, nonetheless one of the two sister cells can stop dividing while the other goes on; no labels or addresses are necessary to identify cells (in silico as in vivo): each cell is distinguished by the inherited DNA, then, in sister cells, DNA must show some "notgenetic" difference due to the epigenetic (asymmetric) pathway of DNA methylation and methylation maintenance, very active on TRs [63].

## The Question of Asymmetry

Epigenetic machinery, if on one hand can count the numbers of cell divisions, on the other cannot provide differential information for the (different) fate of somatic sister cells: somatic cell lineages are not symmetrical in that, as just said, one of the two cells can stop dividing, whereas her sister continues to divide: how can daughter cells, that inherit identical copies of DNA, find different instructions for their future development? Epigenetic mechanisms are the actors of symmetry breaking because they can change DNA expression without affecting DNA sequence [42]. Because of sister cell DNA identity, differences must necessarily originate from an asymmetry in the process of DNA methylation maintenance that drives sister cells to "read instructions" in different DNA positions. DNA methylation maintenance is a complex mechanism (see: Ming et al.: "Mitotic inheritance of DNA methylation: more than just copy and paste" [64]): during DNA replication, cytosine methylations are maintained by a plethora of enzymes (see: Jones and Liang "Rethinking how DNA methylation patterns are maintained" [95]); methylation maintenance is coupled with replication and occurs concomitantly [25] [64]-[72]; at the replication fork UHRF1 (an epigenetic reader with high affinity for 5 mC ) recruits DNA-methyltransferases to equalize methylations on both strands [73] [74]; this process is asymmetric on TRs: on the leading filament DNA Pol $\delta$, with the cooperation of helicase, primase and a primer, meets monomers in opposite direction than DNA Pole on the lagging filament: a sequence of ten monomers (ideally numbered $3^{\prime} \_1,2$, $3 . . .9,10 \_5$ ' on the old-template-filament) meets DNA Pol $\delta$ from 1 to 10 (i.e. each monomer is met and copied from head to tail), but DNA Pols, with the
cooperation of ligase, primase and several RNA primers, synthesizes the new monomers in Okazaki fragments as 321,654 , 987 (i.e. each monomer is met and copied from tail to head). The role played by H3K9 histone confirms a differential action on the different (leading and lagging) filaments [71]; asymmetries in DNA methylation maintenance have been described in stem cells, where DNA methylation may be asymmetrically inherited by the two daughters [42]; timing differences during methylation maintenance have been found between the leading and lagging strands [64] [71] [72].

In the program "Lineages" a possible symmetry breaking mechanism has been simulated as follows: at the end of replication and methylation maintenance, two DNA molecules arise, one containing the old inherited, already methylated, filament and one with the newly synthesized and newly methylated strand: to carry different information to the offspring, these two molecules should contain a difference in the methylation pattern; the cell that inherits the "old" maternal filament receives all the previously methylated cytosines, while the other cell receives the newly synthetized strand (joined Okasaki fragments) that contains newly methylated cytosines; only this filament has been undergone to the process of methylation maintenance: in the program it is supposed that in its "sat_DNA" one more monomer is methylated and silenced to generate a difference between sister cells (see: Khristich and Mirkin "On the wrong DNA track: Molecular mechanisms of repeat-mediated genome instability, 2020 [18]); each cell finds its future fate (" $C$ " or " $T$ " in 5 th position) on a different monomer. This way, a biological mechanism is outlined, autonomous, automatic, and unsupervised, able to provide, from identical DNA sequences, different fates for sister cells and asymmetric lineages.

## 5. Discussion

Satellite DNA can no longer be considered "junk" (useless) DNA: its role is not restricted to maintaining chromosome structure or heterochromatin establishment, in many species satDNA is present within euchromatin [75] and many TRs are actively transcribed into non-coding RNAs; in C. elegans different regions of satDNA are differently transcribed in different tissues [7], unveiling tissue-specific roles of TRs. CpG dinucleotides frequently occur in DNA regions called "CpG islands": there are about 29,000 CpG islands in human euchromatin, and 50,000 if CpG islands in TRs are also counted [76]. Methylated cytosines, in CpGs and non-CpGs, are everywhere in genomes: repetitive sequences, enhancers, promoters, and gene bodies [77]. Cytosine methylation, in CpG dinucleotides but also non- CpG sites ( $\mathrm{CpA}, \mathrm{CpT}$, and CpC ), plays a crucial role as an epigenetic mark in animals, silencing or activating genes depending on tissues [77].

As already said, the role of satDNA in morphogenesis has been studied in invertebrates and vertebrates: the small gene diversity between chimpanzees and humans cannot explain the three times brain expansion and the twofold increase in neuron number in human cerebral cortex [78]: the real genomic disparity is in

TRs [79]. A small difference in the genome of Homo sapiens relative to Neanderthals, a single-point-mutation in the transketolase-like 1 gene (TKTL1) [80] leads to an impressive increase of basal radial glial cell divisions, that, in turn, boost the output of upper layer projection neurons: TKTL1 is involved in DNA methylation [81] [82] particularly in micro-satellite methylation [83]. SatDNA diversity correlates with relevant anatomical differences (see: Fondon and Garner "Molecular origins of rapid and continuous morphological evolution", 2004 [16]; Myers "Tandem Repeats and Morphological Variation", 2007 [84]); the abundant variety of TRs of different lengths can sustain interindividual morphological diversity: in the retina of the model organism Danio rerio, neurogenic progenitors produce two daughter cells with different fates, one deterministic and one probabilistic; interference with the deterministic branch of the lineage affects lineage progression, in contrast, the probabilistic branch has a large range of fates [85]; lineage flexibility fits well with the large number of TRs of different lengths.

Many finding supports the hypothesis that TRs, together with DNA methylation, play a role in cell divisions and in early embryo cleavage (where counting cell division rounds is crucial): TRs and epigenetic enzymes control Polycomb [86] [87]: Polycomb group, a family of chromatin remodeling protein involved in epigenetic pathways, is known to downregulate CyclinA in Drosophila, plants and vertebrates [38] [88] [89] [90] [91] [92]; a non-coding Alu (ashort interspersed repetitive element) controls the expression of cell cycle genes in human fibroblasts [93]: its overexpression promotes the transition from G1 to S phase; long-non-coding-RNAs, associated with satellite repeats [94], modulate methylation in eukaryotes [96]: permanent exit from the cell cycle is associated with epigenetic methylations [97]. Immediately after fertilization, in mammalian embryos a strong wave of de novo methylation sustains and supports zygote cleavage [24]; mammalian embryonic stem cells have large amounts of not methylated CpG dinucleotides in their satDNA: many CpG cytosines adenines in roundworms [98]) are actively methylated and demethylated [32].

## 6. Conclusion

"Lineages" is an efficient program to simulate processive cellular functions that, executed step by step on linear templates, enumerate precise and invariable numbers of cell divisions: TRs are run over by accurate epigenetic complexes involved in DNA methylation showing that TRs together with the epigenetic machine may be the molecular basis cells use to count deterministic numbers of division rounds. "Lineages"helps to mimic evolutionary processes that have established convenient pedigrees of cell divisions, approved by natural selection as the most suitable and adapted for each species.

## Conflicts of Interest

The author declares no conflicts of interest regarding the publication of this paper.

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## Appendix

```
# LINEAGES
#
# GENERAL INSTRUCTIONS:
#
# 1) Change "C" with "T" in 5 5h position of one or more sublists/monomers of 'sat_DNA' to silence one or more
monomers:
# run the program and, division after division, look, step by step, at the printed cell-lineage.
#
# 2) "H" in 5}\mp@subsup{5}{}{\mathrm{ th }}\mathrm{ position of the first sublist of 'sat_DNA' starts "asymm_mitosis".
```

```
from copy import deepcopy
import sys
sat_DNA = [['A','T','T,''','C','A','A','C','G','G','C','T','T','A'],['A','T','T','C','C','C','A','C','G','',''C','T','T','A'],\
    ['A','T','T,''C','C','G','A','C','G','G','C','T','T','A'],['A','T','T','C','C','C','A','C','G','G','C','T','T','A'],\
    ['A','T",'T,''C','C','G','A','C','G','G','C','T','T','A'],['A','T','T','C','C','C','A',''','G','G','C','T','T','A'],\
    ['A','T','T','C','C','C','A','C','G','G','C','T','T','A'],['A','T','T','C','C','T','A','C','G',''','C','T','T','A']]
```

```
cells = {0:sat_DNA } # dictionary of cells (founder cell has "key" = '0') and their methylated satDNA ("value")
arisen_cells = [0] # list of generated cells
existing_cells = [0] # after division, the mother cell is replaced by its two daughters; [n]
                        # (a number between squared brackets)is the number (label) of a cell in G0
dividing_cells = [0] # list of cells allowed to divide
not_div_cells = [] # list of cells banned to divide
Stop = 0 # setting Stop at '1' avoids infinite loops
pos =0 # 'pos' is the label (key) of each cell in 'cells' dictionary
num_mit = 1
```


## \# FUNCTION "SYMMETRICAL MITOSIS"

\#Division of a somatic (mother) cell in two (daughter) sister cells
def symm_mitosis(cell_num):
global num_mit
global Stop
global existing_cells
global dividing_cells
global arisen_cells
global pos
pos $=2^{*}$ cell_num \# from cell $n^{\circ}$ ' $x$ ', cells $n^{\circ}$ ' $2 x+1$ ' and ' $2 x+2$ ' arise
if Stop $==1$ :
return
elif sat_DNA $[0][3]==$ 'C' and sat_DNA[0][4] == 'H' and sat_DNA[0][5] == 'A':
asymm_mitosis(cell_num) \# "H" in 5th position of the first sublist/monomer of 'sat_DNA' indicates
"asymm_mitosis"
else:
cells $[$ pos +1$]=[] \quad$ \# list for deepcopying DNA
cells[pos+2] = []
arisen_cells.append(pos+1)
arisen_cells.append(pos+2)
print()
print('Mitosis $\mathrm{N}^{\circ}$, num_mit)
print()
print('from division of cell $\mathrm{N}^{\mathrm{o}}$, cell_num, 'cells $\mathrm{N}^{\circ}$, pos $+1, \backslash$
'and',pos+2,'arise')
\# First cell
cells[pos+1] = deepcopy(cells[cell_num]) \# satDNA is copied and associated as 'value' of this cell in
'cells' dictionary
$\mathrm{i}=0 \quad$ \# 'first cell' starts reading from the first sublist
(monomer)of 'sat_DNA'
while $\mathrm{i}<$ len(cells[pos+1]): \# step by step satDNA is traversed
if cells[pos+1][i][3] != ' $\mathrm{M}^{\prime}$ and cells[pos+1][i][3] != 'C': \# check for not recognized se-
quences
Stop $=1$
break
if cells[pos+1][i][4] != 'C' and cells[pos+1][i][4] != 'T': \# check for not recognized sequences Stop $=1$
break
if cells[pos+1][i][3]== 'M': \# this monomer has been silenced, check the next
i+=1 \# processive mechanism activation
continue
elif cells[pos+1][i][3]=='C'and cells[pos+1][i][4]=='C' and cells[pos+1][i][5]== 'T': \# this is the last monomer
dividing_cells.append(pos+1)
cells $[$ pos +1$][\mathrm{i}][3]=$ ' $\mathrm{M}^{\prime} \quad$ \# methylate this monomer and enter G0
Stop $=1$
break
elif cells[pos+1][i][3]!= 'M' and cells[pos+1][i][4] == 'C':
cells $[$ pos +1$][\mathrm{i}][3]=$ ' M ' \# methylate this monomer and divide
dividing_cells.append(pos+1)
break
elif cells[pos+1][i][3]!= 'M' and cells[pos+1][i][4]== 'T': \# methylate this monomer and enter
G0
cells $[$ pos+1][i][3] = 'M'
not_div_cells.append(pos+1)
break
$\mathrm{i}+=1$
\# Second cell
cells $[\operatorname{pos}+2]=$ deepcopy $($ cells $[p o s+1])$ \# at the replication fork the second cell DNA is bookmarked:
\# so, in the program, second cell DNA = first cell DNA
$\mathrm{i}=1 \quad$ \# 'second cell' starts reading from the second sublist of 'sat_DNA'
while $i<\operatorname{len}($ cells $[p o s+2])$ :
if cells[pos+2][i][3] != 'M' and cells[pos+2][i][3] != 'C':

$$
\text { Stop }=1
$$ break

if cells[pos+2][i][4]!= 'C' and cells[pos+2][i][4] != 'T':
Stop $=1$
break
if cells[pos+2][i][3] == ' M ':
$\mathrm{i}+=1 \quad$ \# processive mechanism activation continue
if cells $[$ pos +2$][\mathrm{i}][3]==$ ' C 'and cells $[$ pos +2$][\mathrm{i}][4]==$ ' C ' and cells $[\operatorname{pos}+2][\mathrm{i}][5]==$ ' T ': \# this is the last monomer
not_div_cells.append(pos+2)
cells[pos+2][i][3] = 'M' \# methylate this monomer and enter G
Stop $=1$
break
elif cells[pos+2][i][3]!= 'M' and cells[pos+2][i][4] == 'C': \# methylate this monomer and divide

$$
\operatorname{cells}[\text { pos+2][i][3] = 'M' }
$$

dividing_cells.append(pos+2)
break
elif cells[pos+2][i][3]!= 'M' and cells[pos+2][i][4] == 'T': \# methylate this monomer and

```
enter G0
        not_div_cells.append(pos+2)
        break
        i}+=
num_mit +=1
print('cell N}\mp@subsup{}{}{\circ}\mathrm{ ',cell_num,'satDNA:\n',cells[cell_num]) # 'sat_DNA' and its metilations are printed
print()
print('cell N'',pos+1,'satDNA:\n',cells[pos+1]) # " "
print()
print('cell N}\mp@subsup{}{}{\circ},\mathrm{ ,pos+2,'satDNA:\n',cells[pos+2]) # " " 
print()
print()
existing_cells = dividing_cells + not_div_cells
existing_cells.sort()
for q in range(len(existing_cells)):
    if existing_cells[q] in not_div_cells:
        existing_cells[q] = [existing_cells[q]]
print('existing_cells', existing_cells)
print('dividing_cells',dividing_cells )
print('not_div_cells', not_div_cells)
print()
if dividing_cells == []:
    Stop = 1
```

\# FUNCTION ASYMMETRICAL MITOSIS (Stem cells)
\# Division of a stem (mother) cell in: one (daughter) differentiating cell and one identical stem cell
def asymm_mitosis(cell_num):
global num_mit
global Stop
global arisen_cells
global existing_cells
global not_div_cells
global dividing_cells
global pos
if Stop $==1$ :
return
pos $=$ pos +1
cells[pos] = []
print()

```
print()
print()
print('Mitosis N',}\mp@subsup{}{}{\prime},\mathrm{ num_mit)
print()
print('Division of cell N}\mp@subsup{}{}{\circ}\mp@subsup{}{}{\prime}\mathrm{ , cell_num)
print()
print('from division of cell N}\mp@subsup{N}{}{\circ}\mathrm{ , cell_num, 'cell No 0 remains and the new cell N'0', pos,' arises')
print()
```

\# 'First cell' is an invariant "immortal" stem cell that maintains its original sat_DNA
if cell_num not in dividing_cells:
dividing_cells.append(cell_num)
num_mit +=1
\#Second cell
cells[pos] = deepcopy(cells[cell_num]) \# 'second cell' copies the sat_DNA from stem cell
cells[pos][0][4] = 'C' \# 'second cell' erases hydroxyimethylation in the first sublist
dividing_cells.append(pos) \# 'second cell' will divide
arisen_cells.append(pos)
print('cell $\mathrm{N}^{\circ} 0$ satDNA: $\mathrm{n}^{\prime}$ ', cells[0])
print()
print('cell ${ }^{0}{ }^{\circ}$,pos,'satDNA: $\mathrm{nn}^{\prime}$, cells[pos])
print()
print()
if pos $==30$ :
existing_cells $=$ dividing_cells + not_div_cells
existing_cells.sort()
print('existing_cells', existing_cells[:31])
print('dividing_cells',dividing_cells )
print('not_div_cells', not_div_cells)
print()
Stop $=1$
return
asymm_mitosis(0)

## \# RUNNING THE PROGRAM

while pos < 30:
if Stop $==0$ and dividing_cells $[0]<15$ :
$\mathrm{a}=$ dividing_cells.pop(0)
symm_mitosis(a)
else:
break
if sat_DNA[0][4] == 'H':
continue
\# LINEAGE GRAPHICS
print('cell lineage:')
print()
for i in range $(0,1)$ :
$\operatorname{print}\left(14^{* 1}\right.$ ','_'*22,'0','_'*23)
for i in range $(1,2)$ :
if i in arisen_cells:
print(' '×13,i,' '×47,end=")
else:
print(' ' $* 60$,end =")
for $i$ in range $(2,3)$ :
if i in arisen_cells:
print(i)
else:
print( ${ }^{\prime} \times 20$ )
for i in range $(3,4)$ :
if i in arisen_cells:
print(' '*5,i,'_'*5,'|','_*5,end=')
else:
print( ' '*18,end=")
for i in range $(4,5)$ :
if $i$ in arisen_cells:
print(",i,' ' 25, end=")
else:
print(' '*31,end=")
for i in range $(5,6)$ :
if $i$ in arisen_cells:
print(' '*5,i,'_'*5,'|','_*5,end=")
else:
print( ' '*19,end=")
for $i$ in range $(6,7)$ :
if $i$ in arisen_cells:
print(",i,' '*4)
else:

```
    print(")
    for i in range(7,8):
    if i in arisen_cells:
        print(' ',i,'_','|','_'*1,end=')
    else:
        print(' '*5,end=")
for i in range(8,9):
    if i in arisen_cells:
        print(",i,' '*4,end=")
    else:
        print(' '*11,end=")
for i in range(9,10):
    if i in arisen_cells:
        print(' ',i,'_','|','_',end=')
    else:
        print(",end=")
    for i in range(10,11):
    if i in arisen_cells:
        print(",i,' '*19,end=")
    else:
        print(' '*32,end=')
for i in range(11,12):
    if i in arisen_cells:
        print(' ',i,'_','|','_'*1,end=")
    else:
        print(' '*6,end=")
for i in range(12,13):
    if i in arisen_cells:
        print(i,' '*3,end=")
    else:
        print(' '*10,end=")
for i in range(13,14):
    if i in arisen_cells:
        print(' ',i,'_','|','_',end=")
    else:
        print(",end=")
for i in range(14,15):
    if i in arisen_cells:
        print(i)
    else:
        print(' ')
```

if 15 in arisen_cells and 17 in arisen_cells and 19 in arisen_cells and $\backslash$ 21 in arisen_cells and 23 in arisen_cells and 25 in arisen_cells and $\backslash$

27 in arisen_cells and 29 in arisen_cells:

$$
\begin{aligned}
& \text { ' ' } \times 5 \text {,'|', ' ' } 5 \text {,'|') } \\
& \text { print('15-16',' ','17-18',', ','19-20',' ','21-22',' ' '×18,'23-24','। } \\
& \text { ' ','25-26',' ','27-28',' ','29-30') }
\end{aligned}
$$

if 15 in arisen_cells and 17 in arisen_cells and 19 in arisen_cells and $\backslash$
21 in arisen_cells and 23 not in arisen_cells and 25 not in arisen_cells and $\backslash$
27 not in arisen_cells and 29 not in arisen_cells:
print(' ','|', '*3,' ','|',' '*5,'|',' '*6,'|')
print('15-16',' ','17-18', ' ','19-20',' ','21-22')
if 15 in arisen_cells and 17 in arisen_cells and 19 in arisen_cells and $\backslash$
21 in arisen_cells and 23 in arisen_cells and 25 not in arisen_cells and $\backslash$
27 not in arisen_cells and 29 not in arisen_cells:
print(' ','|',' '*3,' ','|',' '*5,'|',' '*6,'|',' '*21,'|')
print('15-16',' ','17-18',' ','19-20',' ','21-22',' '×18,'23-24')
if 15 in arisen_cells and 17 in arisen_cells and 19 in arisen_cells and $\backslash$
21 in arisen_cells and 23 in arisen_cells and 25 in arisen_cells and $\backslash$
27 not in arisen_cells and 29 not in arisen_cells:
print(' ','|', ' '*3,' ','|', '*5,'|', ' '*6,'|', ' '*21,'|',' '*5,'|')
print('15-16',' ','17-18',' ','19-20',' ','21-22',' '×18,'23-24',
' ','25-26')
if 15 in arisen_cells and 17 not in arisen_cells and 19 not in arisen_cells and $\backslash$
21 not in arisen_cells and 23 not in arisen_cells and 25 not in arisen_cells and $\backslash$
27 not in arisen_cells and 29 not in arisen_cells:
print(' ','|')
print('15-16')
if 15 in arisen_cells and 17 in arisen_cells and 19 not in arisen_cells and $\backslash$
21 not in arisen_cells and not 23 in arisen_cells and 25 not in arisen_cells and $\backslash$
27 not in arisen_cells and 29 not in arisen_cells:
print(' ','|', ' ' 3 ,' ','|')
print('15-16',' ','17-18')
if 15 in arisen_cells and 17 in arisen_cells and 19 in arisen_cells and $\backslash$
21 not in arisen_cells and 23 not in arisen_cells and 25 not in arisen_cells and $\backslash$
27 not in arisen_cells and 29 not in arisen_cells:
print(' ','|',' '*3,' ','|', ' ' 5 ,'|')
print('15-16',' ','17-18',' ','19-20')
if 15 in arisen_cells and 17 in arisen_cells and 19 in arisen_cells and $\backslash$
21 not in arisen_cells and 23 in arisen_cells and 25 not in arisen_cells and $\backslash$
27 not in arisen_cells and 29 not in arisen_cells:
print(' ','|',' '*3,' ','|',' '*5,'|',' '*30,'|')
print('15-16',' ','17-18',' ','19-20',' '*26,'23-24')
if 15 not in arisen_cells and 17 not in arisen_cells and 19 not in arisen_cells and 21 not in arisen_cells and 23 in arisen_cells \}
and 25 in arisen_cells and 27 in arisen_cells and 29 in arisen_cells:
print(' '*48,'|', ' ${ }^{2}$,'|', ' ${ }^{\prime} 5$,'|', ' ' ${ }^{2}$,'|')
print(' '*46,'23-24',' ','25-26', ' ',27-28',' ','29-30')
if 15 not in arisen_cells and 17 not in arisen_cells and 19 not in $\backslash$
arisen_cells and 21 not in arisen_cells and 23 in arisen_cells \}
and 25 in arisen_cells and 27 in arisen_cells and 29 not in arisen_cells:
print(' '*50,'|', ' $* 5$,'|',' ' ${ }^{\prime}$,'|')
print(' '*48,'23-24',' ','25-26',' ','27-28')
if 15 not in arisen_cells and 17 not in arisen_cells and 19 not in\}
arisen_cells and 21 not in arisen_cells and 23 in arisen_cells \}
and 25 in arisen_cells and 27 not in arisen_cells and 29 not in arisen_cells:
print(' ' $\times 50$,'|', ' ' $\times 5$,'|')
print(' '*48,'23-24', ' ',25-26')
if 15 not in arisen_cells and 17 not in arisen_cells and 19 not in \}
arisen_cells and 21 not in arisen_cells and 23 not in arisen_cells $\backslash$
and 25 not in arisen_cells and 27 in arisen_cells and 29 \}
not in arisen_cells:
print(' '*66,'|')
print(' ' * 64, '27-28')
if 15 not in arisen_cells and 17 not in arisen_cells and 19 not in \}
arisen_cells and 21 not in arisen_cells and 23 not in arisen_cells $\backslash$
and 25 not in arisen_cells and 27 not in arisen_cells and 29\}
in arisen_cells:
print(' ' $* 74$, '|')
print( ${ }^{\prime} \times 72$, '29-30')
if 15 not in arisen_cells and 17 not in arisen_cells and 19 not in \}
arisen_cells and 21 not in arisen_cells and 23 not in arisen_cells $\backslash$
and 25 not in arisen_cells and 27 in arisen_cells and 29 in arisen_cells:
print(' '*66,'|', ' '*5,'|')
print(' '*64,'27-28', ' ',29-30')
if 15 not in arisen_cells and 17 in arisen_cells and 19 not in \}
arisen_cells and 21 not in arisen_cells and 23 not in arisen_cells $\backslash$
and 25 not in arisen_cells and 27 in arisen_cells and 29\}
in arisen_cells:
print(' '*9,'|',' '*54,'|', ' '*5,'|')
print(' ' $\neq 7, ' 17-18$ ',' ' $* 50, ' 27-28$ ',' ','29-30')
if 15 not in arisen_cells and 17 in arisen_cells and 19 not in $\backslash$
arisen_cells and 21 not in arisen_cells and 23 not in arisen_cells $\backslash$
and 25 not in arisen_cells and 27 in arisen_cells and $29 \backslash$
not in arisen_cells:
print(' '*9,'|',' '*54,'|')
print(' '×7,'17-18',' '×50,'27-28')
if 15 not in arisen_cells and 17 in arisen_cells and 19 not $\backslash$
in arisen_cells and 21 not in arisen_cells and 23 not in $\backslash$
arisen_cells and 25 not in arisen_cells and 27 not in arisen_cells \}
and 29 not in arisen_cells:
print( ${ }^{\prime *}$ 9, '|')
print( ' '*7,'17-18')
if 15 not in arisen_cells and 17 not in arisen_cells and 19 not \}
in arisen_cells and 21 not in arisen_cells and 23 in \}
arisen_cells and 25 not in arisen_cells and 27 not in arisen_cells \}
and 29 not in arisen_cells:
print(' '*48,'|')
print( ' '*46,'23-24')
if 15 not in arisen_cells and 17 not in arisen_cells and 19 not \}
in arisen_cells and 21 not in arisen_cells and 23 not in $\backslash$
arisen_cells and 25 in arisen_cells and 27 not in arisen_cells \}
and 29 not in arisen_cells:
print(' ' $\times 57$,'|')
print(' '*55,'25-26')
if 15 not in arisen_cells and 17 not in arisen_cells and 19 not \}
in arisen_cells and 21 not in arisen_cells and 23 in \}
arisen_cells and 25 not in arisen_cells and 27 not in arisen_cells \}
and 29 in arisen_cells:
print(' '*48,'|',' '*20,'|')
print(' '丸46,'23-24',' '丸16,'29-30')
if 15 not in arisen_cells and 17 not in arisen_cells and 19\}
not in arisen_cells and 21 in arisen_cells and 23 not in $\backslash$
arisen_cells and 25 not in arisen_cells and 27 not in arisen_cells \}
and 29 not in arisen_cells:
print(' ' $\times 25$,'|')
print(' ' '×23,'21-22')
if 15 not in arisen_cells and 17 not in arisen_cells and 19 in \}
arisen_cells and 21 in arisen_cells and 23 not in arisen_cells \}
and 25 not in arisen_cells and 27 not in arisen_cells and 29 not in arisen_cells:
print(' '*17,'|', ' ' 6 ,'|')
print(' '×15,'19-20',' ' 2, '21-22')
if 15 not in arisen_cells and 17 not in arisen_cells and 19 in \}
arisen_cells and 21 not in arisen_cells and 23 not in arisen_cells \}
and 25 not in arisen_cells and 27 not in arisen_cells and 29 not in arisen_cells:
print(' '×17,'|')
print( ' ' ${ }^{15} 15,19-20$ ')
if 15 not in arisen_cells and 17 in arisen_cells and 19 not in arisen_cells and $\backslash$
21 not in arisen_cells and 23 not in arisen_cells and 25 not in arisen_cells and $\backslash$
27 not in arisen_cells and 29 in arisen_cells:

```
print(' ',' ',' '*3,', ','',', '*62,'|')
print(' '*5,' ','17-18 ',' '*57,'29-30')
```

if 15 in arisen_cells and 17 in arisen_cells and 19 in arisen_cells and $\backslash$
21 in arisen_cells and 23 in arisen_cells and 25 in arisen_cells and $\backslash$
27 in arisen_cells and 29 not in arisen_cells:


```
    ' '*5,'|')
print('15-16',' ','17-18',' ','19-20',' ','21-22',' '×18,'23-24',
    ' ','25-26',' ','27-28')
```

if 15 in arisen_cells and 17 in arisen_cells and 19 in arisen_cells and $\backslash$ 21 not in arisen_cells and 23 in arisen_cells and 25 not in arisen_cells and $\backslash$
27 not in arisen_cells and 29 in arisen_cells:

```
print(' ','|',' '*3,' ','|',' '*5,'|',' '*31,'|',' '*19,\
    '|')
print('15-16',' ','17-18',' ','19-20',' ',' '*25,'23-24',\
    ' '*15,'29-30')
```

if 15 in arisen_cells and 17 not in arisen_cells and 19 not in arisen_cells and $\backslash$ 21 in arisen_cells and 23 not in arisen_cells and 25 in arisen_cells and $\backslash$
27 not in arisen_cells and 29 not in arisen_cells:

```
print(' ','|',' '*21,'',', '*21,'|')
print('15-16 ',' ',' '*14,'21-22 ',' '*16,\
```

    '25-26')
    if 15 in arisen_cells and 17 not in arisen_cells and 19 not in arisen_cells and $\backslash$ 21 in arisen_cells and 23 not in arisen_cells and 25 not in arisen_cells and $\backslash$

27 not in arisen_cells and 29 not in arisen_cells:

```
print(' ','',', '*22,'|')
print('15-16 ',' '*17,'21-22')
```

