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#### Abstract

The systematic characterization of gene functions in species recalcitrant to Agrobacterium-based transformation, like Pisum sativum, remains a challenge. To develop a high throughput forward and reverse genetics tool in pea, we have constructed a reference ethylmethane sulfonate mutant population and developed a database, UTILLdb, that contains phenotypic as well as sequence information on mutant genes. UTILLdb can be searched online for TILLING alleles, through the BLAST tool, or for phenotypic information about mutants by keywords.


## Background

Mutational approaches have been widely exploited in breeding and basic research. In the genomic era, the completion of the sequencing of several plant genomes has enabled the development of reverse genetics strategies, where one first identifies a target gene based on the functional annotation of its sequence, and then proceeds with the phenotypic characterization of mutant alleles. Several mutagenesis techniques are dedicated to this approach, notably RNA interference suppression [1,2] and insertional mutagenesis by transposon tagging [3,4] or Agrobacterium T-DNA insertion [5]. These methods, however, are still mainly based on Agrobacterium T-DNA vectors and, thus, rely on the ability of a given plant species to be transformed. On the other hand, chemical mutagenesis based on an alkylating agents like ethylmethane sulfonate (EMS) [6] provides an easy and cost-effective way to
saturate a genome with mutations. TILLING (targeting induced local lesions in genomes) uses EMS mutagenesis coupled with gene-specific detection of single-nucleotide mutations [7-9]. This reverse genetic strategy encompasses all types of organisms [10-14] and can be automated in a high throughput mode, which is an absolute necessity to match the speed of candidate gene discovery.

The success of the TILLING approach relies on the construction of high quality mutant libraries. Ideally, the mutant population is phenotyped so that in silico analysis of the mutant lines can be carried out. To date, phenotypic databases can be found for tomato [15], rice [16], Lotus japonicus [13] and Arabidopsis [17], and a searchable collection of phenotypic mutants is available for Zea mays [18], Pisum sativum [19] and Arabidopsis thaliana [20].

Pea ( $P$. sativum) belongs to the Leguminoseae family, which provides excellent dietary components with health-promoting benefits and offers the important ecological advantage of contributing to the development of low input farming systems by fixing atmospheric nitrogen and further minimizing the need for external inputs when used as a break crop. Since Gregor Mendel's groundbreaking work on the theories of heredity, pea has been extensively used for basic research, in particular in the fields of seed biology and plant architecture. In many studied examples, legume genes were shown to have novel functions compared to those described for related Arabidopsis genes. Detailed characterization of these legume genes will help our understanding of cross-species gene function [21]. However, functional gene validation by transformation is impractical due to the difficulty of transforming pea using Agrobacterium. This situation renders pea an ideal candidate for TILLING. Although several pea EMS mutant populations already exist, they are unsuitable for a genomic approach as they have not been prepared or maintained under rigorously controlled conditions and suffer from crosscontamination. Hence, there is a need for a high-quality $P$. sativum genetic mutant reference collection, which could be used for both forward and reverse genetics studies. Within the frame of the European Grain Legumes Integrated Project [22], we have developed such a population by mutagenizing $P$. sativum cultivar Caméor with EMS, and establishing an associated TILLING platform and phenotype database, UTILLdb.

## Results

## Production of Caméor mutant population

Caméor is an early-flowering garden pea cultivar that completes its reproductive cycle within four months, permitting three successive generations a year under greenhouse conditions. Although pea is predominantly self-fertilizing, some residual cross-pollination can occur. In order to avoid contamination, 100 Caméor plants, derived from single seeds, were analyzed for genetic uniformity using a set of 16 short sequence repeat markers distributed over every arm of the seven predicted pea chromosomes [23] and left to set seeds in insect-proof greenhouses. In total, 10,000 Caméor seeds were produced and used to create the mutant population.

In order to balance maximum mutation density with acceptable plant survival rate, we first conducted a 'kill-curve' analysis on batches of 100 seeds, using a range of doses from 8 to 57 mM EMS. Most treated first generation mutant (M1) plants exhibited retarded growth at an early seedling stage, but all of them recovered. Thirty plants from each treatment were then grown until maturity and assessed for fertility and seed production. A high loss of fertility was observed at the highest doses, with less than $30 \%$ of plants fertile at doses higher than 32 mM EMS. The highest EMS doses allowing $50 \%$ of plants to set seeds, 16 mM and 24 mM , were retained and tested on large batches of seeds (Table 1). Little difference
was observed between these two doses with a tendency toward higher seed production with 16 mM EMS, so a final dose of 20 mM EMS was used for population production. The mean number of seeds per pod was also slightly higher for the plants treated with 16 mM than for those treated with 24 mM EMS. The high rate of arrested embryos in pods of M1 plants treated with EMS doses of $16-24 \mathrm{mM}$ attested to its good mutagenesis efficacy. Out of $8,600 \mathrm{M} 1$ plants, more than 4,817 lines that had produced more than 5 M 2 seeds each were individually harvested. To produce M3 seeds, four M2 seeds per M1 plant were sown in two-liter pots and M3 seeds were harvested from two sister plants, referred to as A and B. Leaf material was harvested from the healthiest looking plant, referred to as A (Figure 1). Seed stocks were sent to the Grain Legumes stock center in Dijon for multiplication, distribution and long-term storage of the lines.

## Phenotyping of the Caméor mutant population

As we intended to create a reference mutant collection that could be used for forward and reverse genetics, we carried out a systematic phenotyping of the mutant population. Our phenotype scoring was based on visual characterization of four plants per M2 family at key developmental stages, from germination until fruit maturation. To facilitate the phenotype scoring we defined a phenotype ontology adapted to pea. This phenotyping tool does not cover all phenotypic alterations (for example, no root evaluation was carried out) and was constructed for high-throughput scoring of many mutant lines in a relatively short growing season. The vocabulary used to describe the mutant plants was organized in a hierarchical tree and is composed of 107 subcategories of phenotypes clustered at different levels. The complete list of the vocabulary used is shown in Additional data file 1 and the number of lines found in each major phenotype category is shown in Table 2.

Out of the $4,817 \mathrm{M} 2$ families, 1,840 showed a visible phenotype, which represents $38 \%$ of the lines. Among the lines that showed a visible phenotype, $45 \%$ were scored for a single phenotype and $55 \%$ displayed multiple phenotypes, that is, they fall into more than one major phenotype category (Figure 2a). This rate of pleiotropy is an underestimation as the phenotypic characterization is based on high-throughput visual observation of only four mutant lines per M2 family. Detailed morphological and biochemical characterization of higher numbers of plants per M2 family would result in more phenotypic effects per mutant and, thus, a higher rate of pleiotropy. The most commonly observed phenotypes are related to stem size, leaf and plant architecture, followed by those related to cotyledons, stipules and seeds, with the least abundant phenotypes being related to flowers, plantlet architecture and petiole morphology (Figure 2b). Examples of phenotypes corresponding to the primary categories described are shown in Figure 3.


Figure I
Establishment of pea EMS mutant library. Caméor seeds were EMS mutagenized. Out of $8,600 \mathrm{MI}$ plants self-fertilized in an insect-proof glasshouse, 4,817 produced more than 5 M2 seeds each. Four M2 seeds, referred to as A-D, per MI parent were grown to maturity and scored for phenotypes. DNA was extracted from the plants referred to as A, which were left to set M3 seeds. As a backup, M3 seeds were harvested from the sister B plants. The collected M3 seeds were sent to the Grain Legumes Biological Resource Center for distribution, maintenance of the lines and long-term storage of the mutant library.

## Caméor TILLING platform

To set up the pea TILLING platform, DNA samples were prepared from 4,704 M2 plants, each representing an independent family and organized in pools of 8 M 2 families. One key factor in TILLING is the availability of the annotated genomic sequence of the gene to be tilled. Even though the pea genome
has not yet been sequenced, acquisition of the genomic sequences of target genes is facilitated by the high degree of synteny between pea and the model plant Medicago truncatula, which is being sequenced [24]. The CODDLE program (Codons Optimized to Discover Deleterious Lesions [25,26]) combined with the PRIMER3 tool [27] are used to define the best amplicon for TILLING. PCR products used for TILLING have a maximum size of about 1,500 bp and, therefore, longer genes are divided into several amplicons. To reduce variation in the quality and the quantity of the PCR amplification product due to the pea genome complexity and low amount of genomic DNA used in PCR, nested PCR is performed. Mutations are detected in the amplified targets using the mis-match-specific endonuclease ENDO1, as described previously [28]. Individual mutant lines are identified following a pool deconvolution step, and then the mutated base is identified by sequencing.

A primary objective in a mutagenesis project is to generate a saturated resource where every locus is mutated and represented by multiple alleles. To evaluate the existence of multiple alleles per locus, we screened for mutations in the pea Methyl transferase 1 gene (PsMet1) [29]. Three amplicons of $1,383,1,310$ and $1,149 \mathrm{bp}$ were tilled (Figure 4) and 96 mutants were identified (Figure 5). Sequence analysis of the mutations showed that 6 were intronic, 37 silent, 50 missense and 3 nonsense mutations (Figure 4b). Although characterization of PsMet1 mutants is beyond the scope of this article, we found that retrieval of the mutant alleles from the A plant M3 seed stocks was successful, without the need to use backup M3 seed stocks collected from the sister B plants (Figure 1). The exonic mutants were mostly present as heterozygotes ( 79 out of 90 mutations), but 11 lines were homozygous for the mutations. As expected with EMS mutagenesis, these mutations were distributed relatively evenly within the screened amplicons (Figure 4b).

To further evaluate the quality of the mutant population, we extended the TILLING screen to another 19 genes and identified 371 point mutations in those genes (Table 3). As expected for EMS, all the mutations were G:C to A:T transitions [6,30]. Induced mutations discovered in exons consisted of $66.75 \%$ missense, $28.51 \%$ silent and $4.74 \%$ stop mutations (Table 4). Although the number of observed missense mutations was bigger than the amount predicted by CODDLE (63.80\%), we recovered stop mutations in a slightly lower proportion than predicted (6.90\%). As many tilled amplicons harbor intronic segments, some recovered mutations were intronic. Although some of these could potentially affect the efficiency of mRNA splicing, such an impact is unpredictable. Thus, intronic mutants were not characterized further. In contrast, the large number of non-synonymous mutations recovered is of interest as they may lead to gain- or loss-of-function phenotypes. Such mutations will also permit dissection of the function of the protein with respect to its sub-domain structure.

Table I

## Effect of EMS

| Dose of EMS | 0 mM | 16 mM | 20 mM | 24 mM |
| :--- | :---: | :---: | :---: | :---: |
| Total MI seeds sown | 100 | 1000 | 4000 | 3600 |
| Percentage of MI plants setting seeds | $100 \%$ | $61 \%$ | $63 \%$ | $58 \%$ |
| Percentage of MI plants yielding more than 5 seeds | $100 \%$ | $56 \%$ | $52 \%$ | $39 \%$ |
| Percentage of arrested embryos in pods of MI plants | $3 \%$ | $45 \%$ | $49 \%$ | $52 \%$ |
| Mean number of seeds per pod ( $\pm$ SD) | $4.83 \pm 0.91$ | $2.00 \pm 0.86$ | $0.91 \pm 1.30$ | $0.79 \pm 1.93$ |

Effect of the concentration of EMS on M2 seed setting, on the frequency of arrested embryos (data are expressed as the percentage of total seeds analyzed) and on the mean number of seeds per pod in the MI generation (200 pods analyzed per dose). SD, standard deviation.

We calculated the mutation frequency in the 20 targeted genes (Table 3) according to Greene et al. [6]: mutation frequency equals the size of the amplicon multiplied by the total number of samples screened divided by the total number of identified mutants. We estimated the average mutation rate to be one mutation every 200 kb . This mutation density is 1.5 times higher than the rate of one mutation per 300 kb reported for Arabidopsis, the best characterized TILLING mutant population to date [6]. Therefore, the $16-24 \mathrm{mM}$ dose of EMS used to create the pea mutant population appears to be an adequate dose for TILLING. On average, we identified 34 alleles per tilled gene (after normalization to TILLING of the entire population). Considering that about half of missense mutations should have a deleterious effect on a typical

Table 2

Number of M2 families affected in the major categories and subcategories of phenotypes

| Major category | Subcategory | No. of families |  |
| :--- | :--- | :--- | :---: |
| I | Cotyledon | Color | 172 |
|  |  | Shape | 32 |
| 2 | Plantlet architecture | Architecture | 7 |
| 3 | Plant architecture | Architecture | 316 |
|  |  | Branching type | 205 |
| 4 | Leaf | Color | 610 |
|  |  | Shape and arrangements | 387 |
|  |  | Appearance | 253 |
| 5 | Stipule | Size/color/shape | 81 |
| 6 | Petiole | Petiole | 77 |
| 7 | Stem | Stem size | 6 |
|  |  | Shape | 1,447 |
| 8 | Flower | Flower morphology | 36 |
|  |  | Flowering time | 24 |
|  |  | Reproductive organs | 4 |
| 9 | Seed | Seed color | 12 |
|  |  | Shape | 2 |

protein [31], 25 alleles per tilled kilobase would be sufficient for phenotypic analyses.

## Setup of the UTILLdb database

We scored 4,817 lines in the mutant population for phenotypic alterations using 107 subcategories of phenotypes. In TILLING screens we searched for mutations in 20 genes and identified 467 alleles. In order to manage and integrate the expanding data from both the phenotype recordings and TILLING target genes, we implemented the database UTILLdb. UTILLdb was developed according to a relational database system, interconnecting four main modules: lines, phenotype categories, sequences and mutations. Two main types of data are accessible, the morphological phenotypes of mutants and the sequences of tilled genes and corresponding alleles, when available. UTILLdb may be searched using a sequence, through a BLAST tool [32] or for a phenotypic feature using a keyword search. The outcome of the search is shown as a table of results that displays the phenotype of each line, with associated pictures and mutated sequence if it exists. Thus, the user could ask whether lines that share mutations in a specific gene share the same phenotypes and vice versa. As we expect the phenotypic characterization of the TILLING mutants to become more detailed as they are analyzed by UTILLdb users, UTILLdb was designed so that the passport data of the mutant lines can be extended or modified as needed. UTILLdb is publicly accessible through a web interface [33]. A link is implemented to facilitate seed ordering. UTILLdb serves also as an entry point for users wishing to have their favorite gene tilled on the Caméor TILLING platform. Results from those screens as well as the phenotype of the mutants identified will be implemented in UTILLdb.

## Discussion

## Mutant population for forward and reverse genetics

EMS-mutagenized populations have been created for different crops with, in many cases, multiple populations per crop. Information on the quality of the mutagenesis and the production and maintenance of the seed stocks are, however, often unavailable. We have constructed a reference EMS mutant population from $P$. sativum cultivar Caméor under


Figure 2
Distribution of phenotypic characteristics of the mutant population and rate of pleiotropy. (a) Number of M2 families in each phenotypic group. The $x$-axis indicates the nine major phenotypic categories, listed in Table 2 , and the $y$-axis indicates the total number of M2 families. Each bar represents the number of mutants in the corresponding category. The blue bar represents the quantity of pleiotropic mutants (having more than one phenotype), given by the first number in the category label. The red bar represents the non-pleiotropic mutants and is given by the second number in the category label. (b) Total number of M2 families ( $y$-axis) sharing l-5 major phenotypic categories (x-axis). The bar for one phenotypic category indicates how many mutants are categorized in only one phenotypic group (non-pleiotropic mutants), and the bars for the 2-5 phenotypic categories represent the number of mutants that share two to five phenotypes, respectively. In each case, the total number of mutants is indicated on the top of the bar.
controlled conditions and developed a database, UTILLdb, which presents phenotypic data based on visual characterization of M2 plants from young seedling to fruit maturation stages. A hierarchical categorization of mutant phenotypes was used to describe the mutant plants. To facilitate the phenotype description, digital images were also recorded. We did not implement the previously published
plant phenotype ontology [34,35], a hierarchical description intended to develop a vocabulary that describes anatomy, morphology, and growth and developmental stages of a flowering plant, for the main reason that the plant phenotype ontology vocabulary is not yet adapted to describe mutant morphological traits in a crop like pea. Instead, the vocabulary used to describe the pea mutant plants was inspired from previous investigations of mutant collections (tomato [15], lotus [13], barley [36]) and adapted to pea.

Tto exploit the mutant population using reverse genetics, genomic DNA was prepared from the mutant lines via highthroughput automated protocols, and organized in pools for bulked screening. Individuals with mutations in the gene of interest were isolated by systematic pool deconvolution. Genes and mutations were integrated in UTILLdb through a web interface, which allows for global analysis of the TILLING mutants in the collection. This database also serves as a portal for users to request materials or TILLING experiments.

## Saturation of the mutation screen

EMS mutagenesis causes primarily G:C to A:T transitions [30]. In the TILLING screen for mutations in PsMet1, we identified 90 independent exonic mutations in a sequence that contains 1,434 cytosines and guanines and this in a mutant population of $4,704 \mathrm{M} 2$ families. Based on this we estimated the average frequency of mutations to be $1.33 \times 10^{5}$ ( $90 / 4,704 \times 1,434$ ). Given a genome size of $5,000 \mathrm{Mb}$ and a $43.23 \% \mathrm{G}: \mathrm{C}$ content in the coding sequence of the pea genome [37], there are $2.2 \times 10^{9} \mathrm{bp}$ susceptible to EMS mutagenesis. Assuming that all $\mathrm{G}: \mathrm{C}$ base pairs are equally sensitive to EMS, we would expect approximately $2.93 \times 10^{4}$ mutations in each EMS-mutagenized $\mathrm{M}_{1}$ plant $\left(\left(1.33 \times 10^{-5}\right) \times\left(2.2 \times 10^{9}\right)\right)$. We used the binomial distribution, $P=1-(1-F)^{N}$, to calculate the probability of finding a mutation in a given $\mathrm{G}: \mathrm{C}$ base pair in our mutant population. In this formula, $P$ is the probability of finding the mutation, $F$ is the mutation frequency per base pair ( $1.33 \times 10^{-5}$ ), and $N$ is the number of M1 mutant lines (4,704). Using this formula we estimated the probability of finding one mutation in any given G:C base pair in the genome as $0.06 \%$. Increasing the size of the mutant population to $50,000 \mathrm{M} 2$ plants raises the probability of finding one mutation in any given G:C base pair in the genome to $52 \%$. This number is relatively small and could be managed by our platform. In fact, 50,000 independent lines represent 65 DNA pool plates ( 96 -wells) or only 16 plates ( 384 -wells). This purely theoretical example shows that EMS mutagenesis coupled with TILLING is a very powerful tool for creating genetic diversity, especially if one considers that routine transformation of $P$. sativum has not yet been achieved and, hence, insertional mutagenesis is not an option.

## Analysis of mutants identified through TILLING

The calculated overall mutation rate of one mutation every 200 kb found in our population is intermediate between the


Figure 3 (see legend on next page)

Figure 3 (see previous page)
Examples of mutant phenotypes representing the nine major phenotypic groups. (a) Plant 566: cotyledon color, albino. (b) Plant 939: plantlet architecture, bushy; plant architecture, hyper compact; leaf color, pale green; stem size, extreme dwarf. (c) Plant 54: plant architecture, determinate growth. (d) Plant I,236: plant architecture, basal branching; leaf color, pale green, yellow; leaf size, medium; stem size, dwarf. (e, f) Plant 903: leaf, cone shaped at leaf base; flowers, sterile flowers. (g) Plant I,567: leaf, distorted; stipule, silver-argentous. (h) Plant 630: flowers, cauliflower type inflorescence; flowers, abnormal all; stem, dwarf; leaf, upcurling.
rate of one mutation per 300 kb reported for Arabidopsis [6] or Cenorhabditis elegans (1/293 kb) [38] and rice [39], and 2.5 -fold higher than the rate of two mutations per megabase for TILLING in maize [40]. A much more saturated mutation density has been observed in tetraploid wheat ( $1 / 40 \mathrm{~kb}$ ), hexaploid wheat ( $1 / 24 \mathrm{~kb}$ ) [41] or Brassica napus ( $1 / 10 \mathrm{~kb}$; unpublished data); however, such species are able to withstand much higher doses of EMS without obvious impact on survival or fertility rates, due to multiple gene redundancies in their polyploid genomes.

In the TILLING screen, we recovered from 8 (Sym29) to 96 mutants (PsMetI) per tilled gene. Some genes (End1, TL) are obviously much more mutated than others (DOF2, $e \operatorname{IF}($ iso $) 4 e$ ), despite the similarity of their GC content ( $36.6 \%$ for $D O F 2,34 \%$ for $T L$ ). Of course, the propensity of a gene to withstand mutations without the resulting protein causing deleterious effects on the plant plays a major role and gametophytic lethal mutations will never be found in the population. However, we could see that some primer pairs used for screening gave a higher background noise than others, which affects the discrimination between true mutants and false positives on the polyacrylamide gel image, and reduces the number of mutants recovered. Nevertheless, our average score of 34 mutant alleles identified per tilled gene is higher than the 10 mutations per gene of Arabidopsis [6] or rice [39].

Screening for mutations in PsMet1 resulted in 96 alleles, of which 50 were missense and 3 non-sense mutations; in this case, the large number of mutations recovered is, at first sight, impressive, but the large gene size and targeted region ( $3,842 \mathrm{bp}$ ), together with the fact that we tilled the entire population ( 4,704 lines), accounts for this result. On the other hand, this example illustrates the strength of TILLING when it comes to finding a specific point mutation.

Because of the high number of alleles we routinely identify, the possible impact of missense mutations on the function of a protein is assessed before systematic phenotyping of the mutant plants, using two different programs: SIFT (Sorting Intolerant From Tolerant) [42], which uses PSI-BLAST alignments, and PARSESNP (Project Aligned Related Sequences and Evaluate SNPs) [43], which provides a posi-tion-specific scoring matrix based on alignment blocks (Figure 4 d ). In the case of PsMet1, 13 out of the 50 missense mutations (23\%) were predicted to have a major impact on
the function of the protein. Thus, the corresponding 13 mutant lines are characterized first.

In Arabidopsis, the MetI gene controls maintenance of CpG methylation [29]. It was previously shown that point mutations in AtMetI can lead to genome hypomethylation [29,44] with a variable impact on plant development, ranging from a late-flowering phenotype to reduced embryo viability. P. sativum has a genome mainly composed of non-coding repeated sequences [45], which are typically subjected to chromatinmediated epigenetic suppression of transcription [46], in which an elevated rate of DNA methylation plays a major role. We intend to investigate the stability of those regions in a hypomethylated context, that is, in PsmetI lines for which CpG methylation is altered. As we are currently amplifying our mutant lines in order to get homozygous mutants and characterize their phenotypes and DNA methylation levels, it is still too early to speculate on the observed versus predicted effect of the mutations according to SIFT.

## Conclusion

In the 21st century, the need for crop improvement in order to face the growing demand of modern agriculture is increasing, while the social acceptance of so-called genetically modified or transgenic crops remains low. Besides, many plant species of agronomic importance are still unsuitable for Agrobacterium-based insertional mutation techniques, including pea. The development of TILLING technology, based on EMS mutagenesis, can contribute to overcoming this deficiency. Furthermore, as EMS generates an allelic series of the targeted genes it becomes possible to investigate the role of essential genes that are otherwise not likely to be recovered in genetic screens based on insertional mutagenesis. We have developed a complete tool that can be used for both forward (EMS saturated mutant collection and the associated phenotypic database) and reverse (high-throughput TILLING platform) genetics in pea, for both basic science or crop improvement. Hence, by opening it to the community, we hope to fulfill the expectations of both crop breeders and scientists who are using pea as their model of study.

## Materials and methods EMS treatment

EMS was diluted to the chosen dose in deionized water. Bottles (Schott type) each containing 900 seeds immersed in 450


Figure 4
Comparison between predicted and obtained mutations. (a) Output of the CODDLE program using as an example the PsMetl genomic sequence. Exons are represented by white boxes and introns by red lines. The CODDLE program was used to identify those regions of the gene in which G:C to A:T transitions are most likely to result in deleterious effects on the encoded protein (represented by the probability curve traced in tourquoise). The CODDLE algorithm is based on an evaluation of protein sequence conservation from comparison of database accessions of homologous proteins. For PsMetl, three fragments were chosen based on these CODDLE results (blue lines). External and internal primers were designed to amplify each region by nested PCR. (b) Graphic representation of mutations identified in the three regions of the gene PsMetl. This drawing was made using the PARSESNP program [43], which maps the mutation on a gene model to illustrate the distribution of mutations. Purple triangles represent silent mutations and black and red triangles represent missense and truncation mutations, respectively.


Figure 5

Figure 5
TILLING screen. Example of a PsMetl TILLING screen on eightfold pooled pea DNA. The image of the cleavage reaction is collected from both channels (dyes IRD700 and IRD800). The sizes of the cleavage products (circled) from the two dye-labeled DNA strands (red or green) add up to the size of the full-length PCR product (top of the gel). PCR artifacts are distinguishable from true mutants by yellow points (red and green added) as they appear at the same size in both channels. The size of the cleavage product (the sizing ladder can be seen at the left and middle of the image) indicates approximately where the single nucleotide polymorphism is located in the fragment.
ml of deionized water-EMS solution were placed on a rotary shaker ( 50 rpm ) overnight ( 15 h soaking). The EMS solution was then removed and seeds were rinsed extensively 12 times for 30 minutes with gentle shaking.

## Plant growing conditions

Pea (cultivar Caméor) seeds were sown in pots filled with sterile pouzzolane (inert medium, light volcanic grit) at a sowing depth of about 2 cm followed by abundant watering in greenhouse conditions. Plants were then automatically watered with a solution of $3 \cdot 5: 3 \cdot 1: 8.6 \mathrm{~N}: \mathrm{P}: \mathrm{K}$. The temperature was maintained between $14^{\circ} \mathrm{C}$ at night and $30^{\circ} \mathrm{C}$ during daytime, with supplementary lighting to provide a 16 h day.

## Genomic DNA extraction and pooling

Four pea leaf discs (diameter 10 mm ) were collected in 96well plates containing 2 steel beads ( 4 mm ) per well, and tissues were ground using a bead mill. Genomic DNA was isolated using the DNeasy 96 Plant Kit (Qiagen, Hilden, Germany). All genomic DNA was quantified on a o.8\% agarose gel using $\lambda$ DNA (Invitrogen, Carlsbad, CA, USA) as a concentration reference. DNA samples were diluted tenfold and pooled eightfold in a 96-well format. A population of 4,704 arrayed DNAs from mutagenized individuals is presently available for screening.

## PCR amplification and mutation detection

PCR amplification was based on nested-PCR and universal primers [14]. The first PCR amplification was a standard PCR reaction using target-specific primers and 4 ng of pea genomic DNA. One microliter of the first PCR served as a template for the second nested PCR amplification, using a mix of gene-specific inner primers carrying a universal M13 tail (CACGACGTTGTAAAACGAC for forward primers; GGATAACAATTTCACACAGG for reverse primers), in combination with M13 universal primers, M13F700 (CACGACGTTGTAAAACGAC) and M13R800 (GGATAACAATTTCACACAGG), labeled at the 5 'end with infra-red dyes IRD700 and IRD8oo (LI-COR ${ }^{\circledR}$, Lincoln, NE, USA), respectively. This PCR was carried out using $0.1 \mu \mathrm{M}$ of each primer, using the following two step cycling program: $94^{\circ} \mathrm{C}$ for 2 minutes, 10 cycles at $94^{\circ} \mathrm{C}$ for 15 s , primer-specific annealing temperature for 30 s and $72^{\circ} \mathrm{C}$ for 1 minute, followed by 25 cycles at $94^{\circ} \mathrm{C}$ for $15 \mathrm{~s}, 50^{\circ} \mathrm{C}$ for 30 s and $72^{\circ} \mathrm{C}$ for 1 minute, then a final extension of 5 minutes at $72^{\circ} \mathrm{C}$. Muta-

Table 3

Tilled genes and mutation density in Caméor mutant population

| Tilled genes | Amplicon size (bp) | \% of GC in exons | Identified mutants | Screened M2 families | Mutation frequency |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Ps CONSTANS-like a (PsCOLa) | 1,012 | 46.30\% | 11 | 1,536 | I/I4I Kb |
| LectineA | 971 | 40.80\% | 13 | 1,536 | 1/II5 Kb |
| Sucrose transporter (SUTI) | 1,014 | 52.40\% | 12 | 1,536 | 1/I30 Kb |
| Cell wall invertase (cwlNV) | 1,612 | 41.50\% | 12 | 1,536 | 1/206 Kb |
| Serine-threonine proteine kinase (Sym29) | 2,457 | 44.00\% | 8 | 768 | 1/236 Kb |
| Phosphoenolpyruvate carboxylase (PepC) | 1,009 | 44.40\% | 25 | 3,072 | 1/I24 Kb |
| Lecl-like (LIL) | 870 | 39.10\% | 21 | 4,608 | 1/191 Kb |
| DOF transcription factor 2 (PsDOF2) | 1,200 | 36.60\% | 9 | 3,072 | 1/410 Kb |
| Trypsine inhibitor (TII) | 712 | 34.20\% | 13 | 3,840 | 1/210 Kb |
| Pea albumine (PA2) | 746 | 38.50\% | 9 | 3,072 | 1/255 Kb |
| Anther specific protein (Endl) | 851 | 40.50\% | 31 | 3,072 | 1/84 Kb |
| MADS box gene (PMIO) | 1,302 | 34.60\% | 20 | 4,608 | 1/300 Kb |
| MADS box gene (PM2) | 1,390 | 31.30\% | 28 | 4,608 | 1/229 Kb |
| Tendril-less transcription factor (TL) | 1,104 | 34.00\% | 28 | 3,072 | 1/I21 Kb |
| Eukaryotic translation initiation factor (eiF4e) | 1,383 | 36.90\% | 36 | 4,608 | $1 / 177$ Kb |
| Eukaryotic translation initiation factor (elF(iso)4e) | 772 | 36.70\% | 10 | 4,608 | 1/356 Kb |
| Methyl transferase I (Metl) | 3,842 | 40.20\% | 96 | 4,704 | 1/188 Kb |
| Retinoblastoma related (RBR) | 2,959 | 40.80\% | 72 | 4,608 | 1/1/2 Kb |
| Late embryogenesis abundant protein (PsLEAM) | 952 | 44.00\% | 17 | 4,608 | 1/258 Kb |
| Heat shock protein 22 (HSP22) | 622 | 45.66\% | 18 | 4,608 | 1/159 Kb |
| Total/mean | 26,780 | 40.12\% | 467 | - | 1/200 Kb |

Part or all of the Caméor mutant population was screened for mutations in the genes listed. The size of the screened amplicon, the number of mutants identified and the mutation frequency for each amplicon are indicated. The average mutation frequency was estimated to one mutation per 200 kb and is calculated as in Greene et al. [6], except that we have summed the sizes of all the amplicons and we divided by the total number of identified mutants.
tion detection was carried out as described previously [28]. The nature of the mutations was identified by sequencing.

## Abbreviations

CODDLE, Codons Optimized to Discover Deleterious Lesions; EMS, ethylmethane sulfonate; PARSESNP, Project Aligned Related Sequences and Evaluate SNPs; SIFT, Sorting Intolerant From Tolerant; TILLING, targeting induced local lesions in genomes.

## Authors' contributions

CLS, FM, JB, GA and RT performed the EMS mutagenesis and took care of the plants; MD extracted the DNA; TILLING screens and analysis were done by MD and JS; VS, VB, YDO and CG set up UTILLdb; AB coordinated the study. The manuscript was written by JS, MD, CLS, VB and AB.

## Additional data files

The following additional data are available with the online version of this paper. Additional data file 1 is a table providing

Table 4
Mutation types

|  | All | Silent | Missense | Truncation |
| :---: | :---: | :---: | :---: | :---: |
| Percent expected (CODDLE) | 100 | 29.30 | 63.80 | 6.90 |
| Percent observed | 100 | 28.51 | 66.75 | 4.74 |
| Percent heterozygouscacro | 86.60 | 27.7 | 56.4 | 2.5 |
| Percent homozygous | 13.34 | 4.78 | 8.06 | 0.5 |

Comparison of expected and observed types of mutations in tilled exonic regions; distribution between heterozygous and homozygous states in the mutant lines. The percentage of expected mutations was calculated by adding the results of CODDLE analysis, on the amplified regions only, for each gene.
the pea mutant phenotype list used for describing and recording M2 mutant plant phenotypes in UTILLdb.

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