The *Physalis peruviana* leaf transcriptome: assembly, annotation and gene model prediction

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**Abstract**

**Background:** *Physalis peruviana* commonly known as Cape gooseberry is a member of the Solanaceae family that has an increasing popularity due to its nutritional and medicinal values. A broad range of genomic tools is available for other Solanaceae, including tomato and potato. However, limited genomic resources are currently available for Cape gooseberry.

**Results:** We report the generation of a total of 652,614 *P. peruviana* Expressed Sequence Tags (ESTs), using 454 GS FLX Titanium technology. ESTs, with an average length of 371 bp, were obtained from a normalized leaf cDNA library prepared using a Colombian commercial variety. De novo assembling was performed to generate a collection of 24,014 isotigs and 110,921 singletons, with an average length of 1,638 bp and 354 bp, respectively. Functional annotation was performed using NCBI’s BLAST tools and Blast2GO, which identified putative functions for 21,191 assembled sequences, including gene families involved in all the major biological processes and molecular functions as well as defense response and amino acid metabolism pathways. Gene model predictions in *P. peruviana* were obtained by using the genomes of *Solanum lycopersicum* (tomato) and *Solanum tuberosum* (potato). We predict 9,436 *P. peruviana* sequences with multiple-exon models and conserved intron positions with respect to the potato and tomato genomes. Additionally, to study species diversity we developed 5,971 SSR markers from assembled ESTs.

**Conclusions:** We present the first comprehensive analysis of the *Physalis peruviana* leaf transcriptome, which will provide valuable resources for development of genetic tools in the species. Assembled transcripts with gene models could serve as potential candidates for marker discovery with a variety of applications including: functional diversity, conservation and improvement to increase productivity and fruit quality. *P. peruviana* was estimated to be phylogenetically branched out before the divergence of five other Solanaceae family members, *S. lycopersicum*, *S. tuberosum*, *Capsicum spp*, *S. melongena* and *Petunia spp*.

**Keywords:** *P. peruviana*, Solanaceae, ESTs, Functional annotation, Gene model, Phylogenetics
sequences in public databases, mainly as a result of phylogenetic studies in the Solanaceae family [5,6]. Therefore, there is a pressing need for efforts to obtain global genetic and genomic information from the Cape gooseberry, *P. peruviana*.

Advances in next generation sequencing (NGS) technology over the past few years have made it possible to rapidly perform *de novo* transcriptome and even genome assembly for non-model organisms with no or little prior genomic data available [7]. However, polyploidy and the large size of many plant genomes, which is predominantly due to amplification of repetitive elements or sometimes partial genome duplication [8], pose challenges to *de novo* whole genome assembly of plants. As such, EST sequencing, which avoids non-coding and repetitive DNA components, is a cost-effective and commonly used strategy to analyze the transcribed portion of a genome. Availability of ESTs represent a valuable resource for research as they provide comprehensive information regarding the transcriptome facilitating gene discovery and genome annotation and aiding in the determination of phylogenetic relationships [9]. An increasing number of successful studies have been published describing EST sequencing and *de novo* transcriptome assembly for large-scale gene discovery [9-18].

Here we describe the sequencing and assembly of the first *P. peruviana* leaf transcriptome from its cDNA-derived ESTs using the 454 GS-FLX Titanium technology, as well as in silico functional annotation and gene model prediction of the assembled transcriptome. The overall workflow of the project is represented in Figure 1. This first transcriptome draft will provide valuable resources for the development of molecular genetic tools that can be used in agronomic trait related marker discoveries, in addition to studies that aim to solve phytosanitary, fruit quality and production problems.

**Results and discussion**

**EST sequencing and assembly**

We performed three fourths 454 GS FLX Titanium run on one normalized cDNA library constructed from *P. peruviana* leaf tissue, generating approximately 336 Mbp of sequence data from 652,614 reads with an average length of 375 bp (Figure 2). After a trimming process by SeqClean [19], which removes adaptors, primer sequences, poly-A tails, as well as short, longer and low quality sequences, a total of 641,512 high quality reads were obtained with an average length of 371 bp. *De novo* transcriptome assembly was performed using Newbler 2.5.3 [20], which has been shown to perform better than a number of other commonly used assemblers [21]. Table 1 shows the transcriptome sequencing and assembly statistics, 79.66% of the reads were assembled into 29,911 contigs, and then further into 24,014 isotigs, with an average assembled length of 1,638 bp. The isotig N50 length was 2,504 bp. All isotigs that share common contigs belong to the same isogroup, presumably equivalent to one gene locus containing multiple alternatively spliced transcripts. The 24,014 assembled isotigs are part of 14,049 isogroups (equivalent to an average 1.7 transcripts per gene), among which 9,655 isogroups have only one isotig each. Isotigs whose length exceeded 200 bp (23,964 in total) were kept for further analysis. The remaining 20.34% reads are singletons, which cannot be connected with any other reads. The 110,921 singletons were kept for further analysis. The average coverage of assembled isotigs is estimated to be 9.1X. The number goes down to 3.9X if we include all the singletons as the effective transcribed portion. Isotigs and singletons together will be referred as cDNAs in the rest the manuscript. The raw data files are available at the National Center for Biotechnology Information (NCBI) Sequence Read Archive (SRA) accession number SRP005904. The assembled reads were deposited in the Transcriptome Shotgun Assembly (TSA) Database [GenBank:JO124085-JO157957].

**Functional annotation**

As the first step for assigning putative functions to the *P. peruviana* transcriptome, BLASTX searches [22,23] were used to align the cDNAs to the UniProtKB/Swiss-Prot and NCBI RefSeq databases. A total of 19,162 isotigs and 35,428 singletons had a BLAST hit (with an
expectation value < 1e-5) to known proteins and matched 8,721 and 15,192 unique protein accessions, respectively. More than 99% of the BLASTX hits from both isotigs and singletons were from plant proteins. Compared to isotigs, a much greater percentage of singletons do not have any significant hits (68%), which could be mainly due to their short lengths. Using Blast2GO [24], we retrieved gene ontology (GO) terms and enzyme commission numbers (EC) for the *P. peruviana* cDNAs (Table 2) from the BLASTX output described above. A total of 33,105 GO terms were assigned to 12,672 cDNAs (including isotigs and singletons). Among all the GO terms extracted, 13,935 (42%) belong to the Molecular Function class, 10,375 (31%) to Biological Process class and 8,795 (27%) to Cellular Component class. There are 7,519 cDNAs assigned to multiple GO terms.

The biological process (BP) GO category comprise different types of metabolic processes which are the most represented categories: there are 4,620 sequences associated with metabolic processes (GO level 2), which is expected, since the metabolic network in plants is by far more extensive compared to other organisms [25]. We found GO terms associated with primary metabolites, which include the universal building blocks of sugars, amino acids, nucleotides, lipids, and energy sources that are essential for plant survival. Additionally, we found GO terms associated with secondary metabolites that play key roles in maintaining plant fitness including ones that function in the protection of plants against microbial, viral infections and UV radiation. Shown in Figure 3 are a number of GO terms (BP category, level 4) that are abundant and relevant to plant physiology, like the metabolic processes of nitrogen compound, nucleotide, carbohydrate, amine and phosphorus. Another category worthy to mention is “response to stimulus” (BP category, level 2). We found 1,120 sequences associated with this category, which include candidate genes for resistance to pathogen attacks. Shown in Figure 3 are a number of level 4 GO categories including: response to organic substance, defense response and response to hormone stimulus. In the molecular function (MF) category, 30% of the *P. peruviana* cDNAs have high similarity to proteins with transferase or hydrolase activity (GO level 3) that includes genes associated with secondary metabolic synthesis pathways [9,10]. Other abundant level 3 MF categories include: nucleotide binding, ion binding and oxidoreductase binding (Figure 3).

We were able to assign 129 unique enzyme commission (EC) numbers to 1,671 *P. peruviana* cDNAs, where 25 unique EC numbers were in turn assigned to 52 metabolic pathways linked to 1,255 cDNAs (Table 3). We found 187 cDNAs involved in thiamine metabolism in addition to 84 sequences associated with secondary metabolite biosynthesis and 53 assigned to the phenylpropanoid biosynthesis pathway. These pathways are of particular interest in *Physalis* as thiamine has been known to induce defense response in plants through the salicylic acid and Ca\(^{2+}\)-related signaling pathways [26,27] and may play roles in biotic or abiotic stress [28]. Furthermore, secondary metabolites such as phenylpropanoids play important roles in resistance mechanisms to pathogens and recently have also been used in medicinal applications including antioxidants, anticancer and anti-inflammatory [2,28].

### Table 1 *P. peruviana* transcriptome assembly overview

| Filtered EST reads | Contigs | Isotigs | Isogroups | Singletons |
|--------------------|---------|---------|-----------|------------|
| 641,512            | 29,911  | 24,014  | 14,049    | 110,921    |

| Average length (bp) | 371 | 743 | 1,638 | 354 |
|---------------------|-----|-----|-------|-----|
| N50 size (bp)       | 1,438 | 2,504 |      |     |

**Figure 2** Length distributions of *P. peruviana* EST reads (left), assembled contigs (center) and isotigs (right). Data obtained after sequencing with three fourths run of 454 GS FLX Titanium of the normalized leaf cDNA library.

**Table 1** *P. peruviana* transcriptome assembly overview

A total of 12,974 *P. peruviana* cDNAs were found to have significant similarities to 3,117 protein domains present in the NCBI CDD (Conserved Domain Database) [30]. The most abundant domain present in proteins encoded by the *P. peruviana* transcriptome is the

**Protein domains encoded by the *P. Peruviana* leaf transcriptome**
pentatricopeptide repeat domain (PPR), found in 350 cDNAs. The PPR containing proteins are commonly found in the plant kingdom and although its function is still unclear, the PPR domain has been found in proteins involved in RNA editing in a number of recent studies [31-34]. Following the PPR domain, the next three most commonly found domains in the \textit{P. peruviana} transcriptome are: protein kinase domain (294 cDNAs), NB-ARC domain (190 cDNAs) and WD40 domain (123 cDNAs). Protein kinases are one of the largest protein families in plants, involved in a wide variety of physiological processes [35], like calcium-dependent protein kinases and MAP kinases which are involved in the recognition of elicitors or pathogens and the subsequent activation of defense response in plants [36]. The NB-ARC domain is a nucleotide-binding motif shared by plant resistant gene products involved in regulated cell death [37,38]. The WD40 domain, whose common function is coordinating multi-protein complex assemblies, is found in a large number of eukaryotic proteins that cover a wide variety of functions including adaptor and regulatory modules in signal transduction, pre-mRNA processing and cytoskeleton assembly [39,40]. Additionally, the WD40 domain is critically involved in the ubiquitin proteasome pathway which regulates photomorphogenesis, flowering and abiotic stress response in plants [41].

Other frequently found domains include: RNA recognition motif (115 cDNAs), RING-finger domain (96 cDNAs), Leucine rich repeat N-terminal domain (89 cDNAs), tyrosine kinase catalytic domain (84 cDNAs), all of which are important for various cellular processes.
which are commonly found in eukaryotic cells and involved in a broad range of biological processes. The data is summarized in Table 4.

Out of the 110,921 singletons, there are 9,909 of them (length >200 bp) where GO term(s) were assigned to the sequence through Blast2GO (see Materials) or where a significant similarity to a well-characterized protein domain from NCBI CDD was found. We deposited the 9,909 singletons described above, in addition to the 24,024 assembled isoforms in the NCBI’s TSA (Transcriptome Shotgun Assembly) sequence database, which is available at GenBank (accessions JO124085-JO157957). Those sequences with their functional annotations, including GO terms and domain similarity related description, are also provided as Additional file 1: ‘Cape gooseberry cDNAs’.

In silico SSR marker identification

The presence of Simple Sequence Repeats (SSRs) in the P. peruviana transcriptome was identified in silico using Phobos [42]. A total of 5,971 SSR loci were found in the Cape gooseberry cDNAs, where imperfect motifs were the most abundant (5,568), in contrast to 403 loci representing perfect motifs (Table 5). Microsatellites were searched in cDNAs avoiding redundant results in isoforms, considering that searching in the alternative transcripts could lead us to predict the same SSRs in different isoforms corresponding to the same isogroup. Trinucleotide (1,068) and hexanucleotide (3,036) motifs were the most commonly found repetitions in the P. peruviana leaf transcriptome, accounting for 68% of the SSRs identified, in contrast to other plant studies where tri- and dinucleotides were the most commonly found repeat units [18,43,44].

We recently reported the first set of microsatellite markers developed for P. peruviana and related species [45] where the large majority of SSR loci was found in untranslated regions (UTRs) of transcripts with similarity to known proteins in public databases, leading to the identification of two novel polymorphic SSRs related to proteins involved in pathogen defense response. SSRs prioritization for plant breeding programs can be done via functional annotation of cDNAs associated with predicted SSRs and Gene Ontology annotations like ones involved in plant defense. Here we used and updated functional annotation of the transcriptome and the entire collection of assembled transcripts to report ten novel predictions for cDNA-derived SSRs in Cape gooseberry. These SSRs are associated with proteins with gene ontology annotations involved in plant defense to biotic stress such as defense response to fungus, programmed cell death, callose deposition in cell wall during defense response, plant hypersensitive response, and jasmonic acid, ethylene and salicylic acid hormones (Additional file 2: ‘Functional annotation of ten Physalis peruviana SSR markers related to plant defense’). The SRRs obtained in this study are the raw materials for future studies in genetic variation among Physalis populations, which can be used for: construction of genetic maps, quantitative trait loci (QTL) identification in this species and plant breeding programs focused on phytosanitary Cape gooseberry problems.

### Table 4 Protein domains identified in P. peruviana transcriptome

| CDD* Identifier | Domain name | Number of cDNAs |
|-----------------|-------------|-----------------|
| 193328, 188079  | Pentatricopeptide repeat domain (PPR motif) | 350              |
| 173623, 189373  | Protein kinase domain | 294              |
| 144508          | NB-ARC domain | 190              |
| 29257           | WD40 domain  | 123              |
| 128654          | RNA recognition motif | 115              |
| 29102           | RING-finger domain | 96               |
| 191981          | Leucine rich repeat N-terminal domain | 89               |
| 128515          | Tyrosine kinase catalytic domain | 84               |
| Others          |             | 11633           |

*CDD: Conserved Domain Database.

### Table 5 SSRs identified in P. peruviana cDNAs

| Motif          | PERFECT | IMPERFECT |
|----------------|---------|-----------|
| Dinucleotide   | 69      | 275       |
| Trinucleotide  | 150     | 918       |
| Tetranucleotide| 28      | 465       |
| Pentanucleotide| 22      | 1,008     |
| Hexanucleotide | 134     | 2,902     |
| **TOTAL**      | 403     | 5,568     |

### Gene model prediction in P. Peruviana

The genome of P. peruviana has not been sequenced yet, nevertheless it is possible to generate gene model predictions using the P. peruviana transcriptome and the genomes of Solanum lycopersicum (tomato) or Solanum tuberosum (potato), which are the two closest related species that have genome sequence available [46,47]. The cDNA to genomic DNA alignments were generated using the Splign software package [48] as described in Methods. All the assembled transcripts through the previous steps including 23,964 isoforms and 9,909 singletons, were mapped to the S. lycopersicum genome, resulting in 12,436 (36.7%) aligned cDNAs, representing 8,801 gene loci and 9,454 transcript models. On the other hand, 14,515 (42.9%) P. peruviana cDNAs were mapped to the S. tuberosum genome, representing 10,166 gene loci and 10,992 transcript models, as summarized in Table 6. Splign requires the consensus intron sequences (GT/AG
or GC/AG) at the splice sites; therefore strand orientation for the multiple-exon alignments from the Splign output can be decided by the 4-nucleotide sequences at the two intron(s) ends. At the moment, no strand orientation is assigned to single exon transcripts, as the query sequence gets aligned to a continuous region in the genome, unless there is strong polyadenylation signal.

The majority of aligned exons have an identity with the genomic sequence ranging from 70% to 95%, with an average identity of 87.6%. Figure 4 shows a number of features of the gene models from alignments of P. peruviana to S. tuberosum genome. Most of gene models contain less than 20 exons. The longest one has 51 exons. The average length of the aligned exons is 228 base pairs and that of the intron is 1,287 base pairs. The intron-exon boundaries as predicted by cDNA to genome alignments are highly conserved when both the S. tuberosum and S. lycopersicum genomes are used to align the P. peruviana transcriptome. We have generated General Feature Format (GFF) files for the gene models (Additional file 3: ‘Cape gooseberry gene model predictions using the tomato genome’ and Additional file 4: ‘Cape gooseberry gene model predictions using the potato genome’).

Further examination of the gene models revealed that there are 11,949 P. peruviana cDNAs mapped to both S. lycopersicum and S. tuberosum genome as shown in Figure 5 (panel A-a), among which 9,795 cDNAs have multiple-exon gene models on both genomes (panel A-b). 9,436 (96%) multiple-exon cDNAs have at least one intron occurring at exactly the same position on the cDNA when aligned to the S. lycopersicum and S. tuberosum genomes (panel A-c). Furthermore, there are 6,358 cDNAs having exactly the same set of intron positions on the cDNA when mapped to the two genomes (Figure 5 panel A-d). However, for those intron positions that have the same coordinates in the cDNA when mapped to the two genomes, the length and thus the sequence of corresponding introns in the gene models from the two genomes have large variances, as revealed in Figure 5 B.

Intron length variation is exemplified in Figure 5C, where a P. peruviana cDNA (ID Php00a06743.16696) was mapped to both the S. lycopersicum and S. tuberosum genomes, resulting in two identical sets of exons, but different sets of intron lengths (a, b). There is also a number of S. lycopersicum and S. tuberosum cDNAs that have the same predicted gene model in their own genome, respectively (all the cDNAs are aligned by Splign). Figure 5C (c) shows the nucleotide sequences around the first intron site of the 3 cDNAs from P. peruviana, S. tuberosum and S. lycopersicum. Primers targeting conserved flanking exonic regions as indicated can be used to amplify intronic fragments from all three species, P. peruviana, S. lycopersicum and S. tuberosum.

### Table 6 Cape gooseberry gene model prediction overview from alignments to the tomato and potato genomes

| Genomes          | Aligned cDNAs | Gene loci | Transcript models |
|------------------|---------------|-----------|-------------------|
| S. lycopersicum  | 12,436        | 8,801     | 9,454             |
| S. tuberosum     | 14,515        | 10,166    | 10,992            |

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**Figure 4** Predicted gene models features for alignments of P. peruviana to the S. tuberosum genome. A) Distribution of exon identities. B) Distribution of number of exons per transcript. C) Distribution of exon lengths. D) Distribution of intron lengths.
The conserved orthologous set (COS) markers are sets of genes conserved throughout evolution in both sequence and copy number [49,50] that have been used extensively in comparative genomic and phylogenetic studies in Solanaceae. The COS marker strategy involves design of universal exonic primers among closely related species based on ortholog identification and multiple sequence alignment to amplify intronic/exonic regions. In the present study, we present another convenient approach to find universal exon regions - gene model prediction by Splign using two or more related genomes to define common models. Given the fact that the *P. peruviana* genome is not available yet, and genomes of both *S. lycopersicum* and *S. tuberosum* are only in their initial versions, gene model predictions would be particularly valuable in obtaining specific intronic regions for marker and SNP discovery in non-model species, as well as for comparative genomic and phylogenetic studies.

We also aligned the *S. lycopersicum* transcriptome to its own genome (data from http://solgenomics.net/organism/solanum_lycopersicum/ genome) and also to the *S. tuberosum* genome [48] using Splign. We mapped

![Figure 5 Gene model prediction in *P. peruviana*.](image)}
34,704 from a total of 34,727 *S. lycopersicum* cDNAs to its own genome with 100% identity (data not shown). However, only 28,366 (81.7%) *S. lycopersicum* cDNAs can be mapped to the *S. tuberosum* genome with an average identity of 90%. In the Cape gooseberry case only 42.9% *P. peruviana* cDNAs get mapped to the *S. tuberosum* genome, suggesting that *P. peruviana* is evolutionarily more distant from *S. lycopersicum* and *S. tuberosum* than the two species from each other. We then conducted further analysis to estimate the phylogenetic location of *P. peruviana* in the Solanaceae family.

**Experimental validation of intron positions**

The experimental validation of predicted exon/intron boundaries in the assembled cDNAs was carried out in a small sample of cDNAs, which are putative homologous of plant disease resistance genes and can be mapped to both the potato and tomato genomes. For each of these cDNAs, a pair of COSII primers was designed to span one putative intron (based on the computational predicted gene model) for PCR amplification of the genomic DNA. The information of the primers used is summarized in Table 7. All the amplified PCR products had the expected length and then were sequenced using conventional Sanger sequencing. Comparison of the amplified genomic fragments to their corresponding cDNAs revealed that all the eight samples we sequenced indeed showed the exon/intron boundaries consistent with the gene models predicted by Splign. Three of them had the experimentally identified intron positions exactly the same as the predicted. In the other five samples, the predicted intron positions are a few base pairs (1–6 bp) away from the experimentally identified sites. The results are shown in Table 8.

**Phylogenetic relationship of *P. Peruviana* with other solanaceae species**

We found five putative orthologs among *P. peruviana*, *S. lycopersicum*, *S. tuberosum*, *Capsicum spp* (pepper), *S. melongena* (eggplant) and *Petunia spp*. The proteins are: xyloglucanase inhibitor containing pepskin_retropepsin superfamily domain, mitochondrial catalytic protein containing PP2Cc superfamily domain, mitochondrial small ribosomal subunit protein containing RPS2 superfamily domain, phosphate transporter and a functionally unknown protein.

To obtain the best accuracy of the phylogenetic tree to be built, we compared the five putative orthologous proteins to the NCBI’s plant RefSeq protein database. There are seven other species that have BLAST hits with an expect value < 1e-5 to all the five orthologs from the previous steps. These seven species are: *Arabidopsis thaliana*, *Populus trichocarpa* (black cottonwood), *Ricinus communis* (castor bean), *Vitis vinifera* (grape), *Oryza sativa* (rice), *Zea mays* (corn) and *Sorghum bicolor*, none of which belongs to the Solanaceae family. The phylogenetic tree was constructed between thirteen plants using the software Phyml [51] and MEGA [52]. The tree generated has good bootstrapping support at all of the branch points except for the position of *V. vinifera*. We removed the *V. vinifera* sequence and constructed the tree presented in Figure 6.

The phylogenetic relationship among *S. lycopersicum*, *S. tuberosum*, *Capsicum spp*, *S. melongena* and *Petunia spp* is consistent with a previous study by Wang Y et al. [53], in which the tree was constructed based on an unduplicated conserved syntenic segment in the genomes of the five plants. Our results showed that *P. peruviana* branched out before the divergence of the other five Solanaceae family

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**Table 7 Primers used for experimental validation of intron positions**

| COSII Marker | Primer sequences (5′ – 3′) | *P. peruviana* unique identifier | Primer position in the cDNA |
|--------------|---------------------------|---------------------------------|-----------------------------|
| C2_At3g07100 | ACGAACGATGCTGCTGGATATAC Php00a01046.06900 | F-1156/1178 R-997/1022 |
| C2_At3g07100 | AGACCTGCGGGGATCTAAGCTCTTCTG |
| C2_At2g35920 | TGCTGCAACACATAGCTGAG Php00a05845.15798 | F-424/446 R-172/195 |
| C2_At2g35920 | AAGCTCTTGTAGTGGTGTTCGAAG |
| C2_At3g06580 | TGCTCAACTCACATGAGTGTGAAAG Php00a06435.16388 | F-715/740 R-784/806 |
| C2_At3g06580 | AGCAAAACCCAGATTTGGCATAC |
| C2_At5g41480 | TATCTGTGCTGCTGAGATAGTC Php00a02812.11168 | F-836/859 R-646/669 |
| C2_At5g41480 | ATGATCTGCTGCATCGCCAATAC |
| C2_At5g27620 | ATCTACATGTGCTGGATGGAAC Php00a03129.12202 | F-776/799 R-720/740 |
| C2_At5g27620 | TCTCTCTGCGTGACAGCTGC |
| C2_At3g04870 | AGCGCGGTAGTATCAGGAGG Php00a02563.10671 | F-1226/1246 R-954/977 |
| C2_At3g04870 | TGACATGCGCCAGATTTCCCATAC |
| C2_At5g60160 | ACACAATGCTAATCAAGCTATGC Php00a01985.09526 | F-278/297 R-482/505 |
| C2_At5g60160 | TCATCCACCGGCCACATTTC |
members. Details of the phylogenetic analysis are summarized in Additional file 5: ‘Phylogenetic analysis workflow’.

Conclusions
This report constitutes the first genomic resource for the Physalis genus providing a large collection of assembled and functionally annotated cDNAs. The Physalis genus is part of the Solanaceae family, whose members are important sources of food, spice and medicine. However, genomic data for other members of the Physalis genus is limited. Therefore, this resource will enhance comparative studies within the family and the transcriptome will serve as a starting point for gene discovery in Physalis and for future annotations of the Physalis peruviana genome sequence. A number of the genes identified in this study provide candidates for resistance genes against viruses, fungal or bacterial pathogens. Additionally, this study is a potentially invaluable resource for mapping and marker-assisted breeding in Physalis peruviana and closely related species like Physalis philadelphica, commonly known as tomatillo, which are food staples in Central American countries.

Methods
cDNA synthesis and cDNA library normalization
Fresh leaf tissue from the Cape gooseberry Physalis peruviana Colombian ecotype plants from the Colombian in vitro germplasm bank (accession number 09U216-6) at the Corporacion Colombiana de Investigacion Agropecuaria (CORPOICA) were processed and flash frozen in liquid nitrogen. Tissues were immediately sent to Bio S&T Inc. (Montreal, QC, Canada) where RNA extraction, cDNA synthesis and normalization were performed. Briefly, RNA was extracted using a modified TRIzol method (Invitrogen, USA). cDNA synthesis was carried out using 16 μg total RNA by a modified SMART™ cDNA synthesis method and then were normalized by a modified normalization method [54,55] where full-length cDNA was synthesized with two set of primers for driver and tester cDNA. Single-stranded cDNA was used for hybridization instead of double-stranded cDNA. Excess amounts of sense-stranded cDNA hybridized with anti-sense-stranded cDNA. After hybridization, duplex DNA was removed by hydroxyapatite chromatography. Normalized tester cDNA was re-amplified and purified with tester specific primer L4N by failsafeTm PCR (Epicentre Biotechnologies, USA), while driver cDNA was unable to amplify using L4N primer. Size fractionation of re-amplified cDNA was done in a 1% agarose gel. Greater than 0.5 kb cDNA fragments were purified by electroelution and after determining the concentrations, purified cDNAs were precipitated and stored in 80% EtOH at −80°C.

cDNA sequencing and assembly
The normalized cDNA library was prepared for sequencing at Emory Genomics Center (Atlanta, GA, USA).

Table 8 Comparison of validated exon/intron boundaries between PCR results and the predicted gene models

| P. peruviana unique identifier | Position of boundary of the 2 adjacent exons at the mRNA | Match predicted gene model? | PCR results | Predicted model |
|-------------------------------|----------------------------------------------------------|----------------------------|-------------|----------------|
| Php0oa01046.06900             | 1077/1078                                                | Precisely                  | 1077/1078   | 1077/1078      |
| Php0oa05845.15798             | 297/298                                                  |                            | 294/295     | 294/295        |
| Php0oa02208.09960             | 653/654                                                  | 6 bp shift                 | 647/648     | 647/648        |
| Php0oa06435.16388             | 774/775                                                  | 774/775                    | Precisely   | Precisely     |
| Php0oa02812.11168             | 751/752                                                  | 2 bp shift                 | 749/750     | 749/750        |
| Php0oa03329.12202             | 756/757                                                  | Precisely                  | 756/757     | 756/757        |
| Php0oa02563.10671             | 1040/1041                                                | 3 bp shift                 | 1037/1038   | 1037/1038      |
| Php0oa01985.09526             | 416/417                                                  | 1 bp shift                 | 417/418     | 417/418        |

Figure 6 Phylogenetic relationship among Solanaceae species. The Cape gooseberry, P. peruviana is phylogenetically branched out before the divergence of the other five Solanaceae family members: S. lycopersicum, S. tuberosum, Capsicum spp, S. melongena and Petunia spp. The numbers are bootstrapping values. The tree was constructed using the amino acid substitution model with PhyML package.
Approximately 5 μg of purified cDNA was sheared into small fragments via Covaris E210 Acoustic Focusing Instrument and sequenced in three-fourths 454 plate run on a 454 GS-FLX Titanium platform (Roche). The SFF files containing raw sequences and quality scores were submitted to the NCBI Sequence Read Archive (accession number SRP005904).

SeqClean [19,56] was used before and after the assembly, for automated trimming and validation of the raw read files and the assembled file. SeqClean was launched with a minimum and maximum length cut-off of 50pb and 600pb. We used the Newbler software, GS de novo Assembler (Roche, version 2.5.3) with default parameters, to assemble reads into contigs, then further into isotigs. Isotigs within an isogroup represent putative alternatively spliced transcripts of a gene. Reads that cannot be connected with any others were defined as singletons.

**Functional annotation**

After assembly, a local BLASTX [22,23] was used to compare the assembled isotigs and singletons against the UniProtKB/Swiss-Prot database (released on April-2011) using an expect value threshold of 1e-5. The remaining cDNAs that did not get hits from UniProtKB/Swiss-Prot were compared against the NCBI RefSeq database (Release 47). The BLASTX output (XML format) was subjected to Blast2GO [24] for Gene Ontology (GO) analysis. Blast2GO retrieves the most significant GO terms associated with the obtained hits to the query sequence. When possible, Blast2GO also provides Enzyme commission (EC) numbers and the metabolic pathways they participate. We also compared all the *P. peruviana* cDNAs against the NCBI CDD database [30] using an expect value threshold of 1e-5 and selected all the hits where the aligned length is more than 2/3 of the targeted CDD length for domain identification.

**SSR identification**

Phobos (version 3.3.11) (http://www.rub.de/spezzoo/cm/cm_phobos.htm) was used to identify microsatellites (SSRs) in the publicly available collection of assembled transcripts and singletons [GenBank: JO124085-JO157957]. Perfect and imperfect searches were performed using default parameters.

**Gene model prediction**

Gene model prediction was carried out using the software package Splign [48], which has been proven to be able to accurately compute cDNA-to-Genome alignment with high efficiency. At the heart of the program is a compartmentalization algorithm which identifies possible gene duplications, and a refined alignment algorithm recognizing introns and splice signals. The complete genome of *P. peruviana* is not available yet. The two closest relatives of *P. peruviana* that have genomic sequences available are *S. lycopersicum* (tomato) (data from http://solgenomics.net/organism/solanum_lycopersicum/genome; ITAG Release 2.3) and *S. tuberosum* (potato; PGSC_DM_v3.4) [47]. Therefore we used the draft genomes of potato and tomato as the reference genome to map the assembled *P. peruviana* leaf transcriptome.

**Phylogenetic analysis**

We selected orthologous proteins using the all-to-all alignment and mutual best hits selection strategy [57]. Pairwise alignments were performed using BLASTP (expect value < 1e-5) using the RefSeq proteins from *S. lycopersicum*, *S. tuberosum*, *Capsicum spp*, *S. melongena* and *Petunia spp*. At the time of the analysis, the numbers of RefSeq proteins from the five species were: 4,788 for *S. tuberosum*; 6,008 for *S. lycopersicum*; 263 for *S. melongena*; 1,701 for *Capsicum spp* and 1,226 for *Petunia spp*. Fifteen putative orthologous proteins were found, which are present in all five species. Next, we aligned the assembled *P. peruviana* isotigs using BLASTX (expect value < 1e-5) against the database made of the fifteen orthologous groups obtained from the previous step (altogether 75 proteins). We identified eleven orthologous groups of proteins from all the five plants with hit(s) from the *P. peruviana* transcriptome. The best hit was chosen when multiple *P. peruviana* proteins hit a given group. We manually examined the alignments in eleven clusters and removed those with large length variation (the longest one is > 20% of the shortest one) and susceptible similarities (< 65%). Thereafter we ended up with five orthologous groups among the six species.

To obtain higher accuracy phylogenetic tree, we further compared the five orthologous groups against the entire plant RefSeq protein database using BLASTP. There are altogether seven more plants that have significant hit(s) (expect value < 1e-5) for all the five orthologous groups. To this step we have thirteen plants for the five orthologous groups. We concatenated the five proteins in each species (in the same order) and aligned them using the program MUSCLE [58]. The alignment results were manually refined and subjected to Phyml [51] and MEGA version 5 [52] for phylogenetic tree construction. Bootstrapping was repeated 1,000 times. Both programs produced the same results.

**Additional files**

Additional file 1: Cape gooseberry cDNAs. The annotated FASTA sequences of the assembled transcriptome, including singletons.

Additional file 2: Functional annotation of ten *Physalis peruviana* SSRs markers related to plant defense.

Additional file 3: Cape gooseberry gene model predictions using the tomato genome. Gene model predictions using Splign from cDNA to genome alignments to the tomato genome.

Additional file 4: Cape gooseberry gene model predictions using the potato genome. Gene model predictions using Splign from cDNA to genome alignments to the potato genome.

Additional file 5: Phylogenetic analysis workflow.
Competing interests

The authors declare that they have no competing interests.

Acknowledgments

Support for this research was provided by a grant from the Colombian Ministry of Agriculture Contract Nos. 054/08072-2008 L4787-3281 to Luz Stella Barrero and 054/08190-2008 L992-3322 to Victor Manuel Núñez Zaranas. Gina Garzón-Martínez was supported by a Colciencias* Joven Investigador* Fellowship during 2010. Leonardo Marino-Ramírez expresses his deepest gratitude to his friend and colleague Dr. Alba Marina Cotes Prado for all the advice and support she gave him to conduct this project. This research was supported in part by the Intramural Research Program of the National Institutes of Health, National Library of Medicine, and National Center for Biotechnology Information.

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Authors’ contributions

GG-M acquired the sequencing data, performed the genome assembly, the functional annotation, SSR marker identification and drafted the manuscript. ZIE participated in the genome assembly, performed functional annotation, gene model prediction, phylogenetic analysis and drafted the manuscript. DL participated in the design of the study and contributed to the phylogenetic analysis. LSR conceived of the study, and participated in its design and coordination and helped to draft the manuscript. LM-R conceived of the study, and participated in its design and coordination and helped to draft the manuscript. All authors read and approved the final manuscript.

Received: 23 November 2011 Accepted: 25 April 2012 Published: 25 April 2012

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