Identification of genes specifically or preferentially expressed in maize silk reveals similarity and diversity in transcript abundance of different dry stigmas

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Abstract

Background: In plants, pollination is a critical step in reproduction. During pollination, constant communication between male pollen and the female stigma is required for pollen adhesion, germination, and tube growth. The detailed mechanisms of stigma-mediated reproductive processes, however, remain largely unknown. Maize (Zea mays L.), one of the world’s most important crops, has been extensively used as a model species to study molecular mechanisms of pollen and stigma interaction. A comprehensive analysis of maize silk transcriptome may provide valuable information for investigating stigma functionality. A comparative analysis of expression profiles between maize silk and dry stigmas of other species might reveal conserved and diverse mechanisms that underlie stigma-mediated reproductive processes in various plant species.

Results: Transcript abundance profiles of mature silk, mature pollen, mature ovary, and seedling were investigated using RNA-seq. By comparing the transcriptomes of these tissues, we identified 1,427 genes specifically or preferentially expressed in maize silk. Bioinformatic analyses of these genes revealed many genes with known functions in plant reproduction as well as novel candidate genes that encode amino acid transporters, peptide and oligopeptide transporters, and cysteine-rich receptor-like kinases. In addition, comparison of gene sets specifically or preferentially expressed in stigmas of maize, rice (Oryza sativa L.), and Arabidopsis (Arabidopsis thaliana [L.] Heynh.) identified a number of homologous genes involved either in pollen adhesion, hydration, and germination or in initial growth and penetration of pollen tubes into the stigma surface. The comparison also indicated that maize shares a more similar profile and larger number of conserved genes with rice than with Arabidopsis, and that amino acid and lipid transport-related genes are distinctively overrepresented in maize.

Conclusions: Many of the novel genes uncovered in this study are potentially involved in stigma-mediated reproductive processes, including genes encoding amino acid transporters, peptide and oligopeptide transporters, and cysteine-rich receptor-like kinases. The data also suggest that dry stigmas share similar mechanisms at early stages of pollen-stigma interaction. Compared with Arabidopsis, maize and rice appear to have more conserved functional mechanisms. Genes involved in amino acid and lipid transport may be responsible for mechanisms in the reproductive process that are unique to maize silk.
Background

In angiosperms, stigmas discriminate between compatible and incompatible pollen grains by preventing the adhesion, hydration, penetration, and growth of incompatible pollen [1]. After landing on the stigma, compatible pollen grains hydrate and germinate to produce pollen tubes that penetrate the stigma surface. These pollen tubes then utilize signals and nutrients from the stigma or style for successful pollination and fertilization. To carry out these reproductive processes, two types of stigmas have developed in flowers. One type is the wet stigma, which deposits fluid secretions or exudates on its surface. The other is the dry stigma, which deposits fluid secretions or exudates on its surface. With the exception of self-incompatibility (SI), relatively little is known about the molecular mechanisms of plant reproduction in dry stigmas.

Because many genetic, cellular, and biotechnological tools, as well as a maize reference genome (inbred line B73), are currently available, maize has been proposed as a model grass to study the function of the dry stigma in reproduction [5]. Maize silk is a specialized elongated tissue that is functionally equivalent to the stigma and style of a typical pistil [6,7]. Hairs that are located on the silk surface serve as receptive structures to support pollen adhesion, hydration, and germination. To possess this unique function, maize silk must express genes and produce proteins required for successful pollination, and many of these genes might be specifically or preferentially expressed in silk. A genome-wide identification of silk-specific/preferential genes might therefore serve as an initial step in the analysis of the molecular pathways involved in reproduction. To obtain silk-specific/preferential genes at the whole genome level, we performed a deep sequencing analysis in maize silk using next-generation sequencing [8] and microarray analysis.

Several mechanisms related to pollination and fertilization have been implicated in pollen germination, pollen tube guidance, and pollen-ovule interactions [9,10]. In contrast, little is understood about the molecular mechanisms of stigmatic reproductive functions. A major challenge to the study of reproductive functions of the stigma is the dearth of informative mutants, especially for grasses [4]. Exploring stigma function through comparative genomics is likely to yield valuable information and build a link between morphological and genetic analyses. Candidate genes thus identified can be selected for further reverse genetic analysis. To date, genome-wide transcriptome analysis of dry stigmas has been conducted in Arabidopsis, rice, and maize [11-15]. The Arabidopsis stigma dataset was established by cDNA subtraction and microarray analysis of stigma tissues [11,12]. In rice, the stigma dataset was generated using a 57 K Affymetrix rice whole genome array and 10 K cDNA microarray [13]. Microarray and RNA-seq analyses have recently been performed on different maize tissues of inbred line B73 at vegetative and reproductive stages and on silk at the anthesis stage [14,15].

In this study, RNA-seq was performed to analyze transcript abundance profiles of mature silk (MS), mature pollen (MP), mature ovary (MO), and 6-day-old seedling (SL). To build a profile of candidate silk genes that are involved in early pollination events, genes expressed in maize silk were compared with those expressed in the other three tissues. We identified 1,427 genes that were specifically or preferentially expressed in maize silk. Bioinformatic analysis of the maize silk dataset not only identified pathways and genes that have been shown to be essential for pollination, but also a number of genes with novel or previously-unidentified functions, such as those encoding amino acid transporters, peptide and oligopeptide transporters, and cysteine-rich receptor-like kinases (CRKs). In addition, we characterized conserved and distinct features of dry stigmas by comparing stigma-specific/preferential datasets of Arabidopsis, rice, and maize.

Results and discussion

Transcriptome analysis of reproductive organs and seedlings in maize

To identify genes involved in reproductive processes, an RNA-seq analysis was performed on MS that emerged from the husk over 3 days (Figure 1A). To eliminate transcripts involved in vegetative growth and basic metabolism in silk tissue, the transcriptomes of MP (Figure 1B), MO (Figure 1C), and SL (Figure 1D) were also sequenced. This enabled us to distinguish transcripts specific to silk from transcripts that contribute to common plant functions. mRNAs from the four tissues were used to construct libraries, which were then sequenced on an Illumina HiSeq 2000 system. After sequencing quality control (Additional file 1) and removal of the “dirty” raw reads (see Methods), the number of purity-filtered reads ranged from 6,145,170 to 6,764,608 per library (Additional file 2).

To identify genes corresponding to reads in each library, the purity-filtered reads were mapped to version 2 of the maize B73 reference genome (AGPv2) [16] with the Short Oligonucleotide Alignment Program 2 (SOAP2) aligner (Additional files 2 and 3) [17]. To make the libraries meaningful, reads that appeared less than four times per million unique mapped reads were eliminated from further statistical analysis. As a result, a total of 16,710 genes were expressed in MS, 5,119 in MP, 17,903 in MO,
and 18,466 in SL (Figure 1E, Additional file 4). As shown in Figure 1E, the extent of gene overlap was determined by comparing the transcript libraries of the four experimental tissues. Many genes (62.07%) expressed in MS were also detected in MO and SL but not in MP, implying that the transcriptome of MS is closer to MO and SL than to MP.

To investigate the difference between the transcriptome profile of MS and those of MP, MO, and SL, a Pearson correlation coefficient (PCC) analysis was performed on the sequencing libraries derived from the four tissues (Figure 1F). Overall, PCC values ranged from 0.0093 to 0.49 (Figure 1F, Additional file 5). As expected, SL tissue was poorly correlated with MP (PCC = 0.012) and MS (PCC = 0.23). Interestingly, however, MS showed a higher correlation with MO (PCC = 0.38) than with MP and SL. The results indicate that the expression patterns of genes in MS were similar to those in MO, in which many genes may be involved in female reproduction. To further clarify the functional difference between the four tissue transcriptomes, a comparative gene ontology (GO) analysis (Parametric Analysis of Gene Set Enrichment, PAGE) [18] was performed on the genes expressed in each tissue. The results show that MS-expressed genes were distinct from those expressed in vegetative tissues (SL) and other reproductive organs (MP and MO).

**Comparison between RNA-seq data and microarray and real-time PCR data**

The transcript abundance profiles of MS, MP, and MO were analyzed using the Affymetrix 18 K maize genome array, and the results were compared with the RNA-seq data. Spearman rank correlation coefficients (SCCs) of log2-transformed expression values were calculated between the three corresponding tissue pairs in the two datasets. The SCC value between the dissimilar tissue pair MS and MP was calculated as a negative control (Figure 2). As a result, SCC values between the same tissues (0.73–0.84) were greater than those between dissimilar tissue pair (0.33). In addition, we used a scatter diagram to plot the log2-transformed expression values from the microarray and RNA-seq analyses of the three same tissue pairs and the one dissimilar tissue pair (Figure 2). These plots illustrate that similar expression patterns were detected for the same tissue using the two different techniques. Transcript abundance values generated by the RNA-seq analysis were distributed much more broadly than those from the microarray analysis. This difference may be due to the wider range of expression detected by RNA-seq compared with the microarray. Overall, these results indicate that data from RNA-seq and microarray analyses were comparable.

Real-time PCR analysis was employed to further validate the expression of genes examined by RNA-seq in MS, MP, MO, and SL (Figure 3). We randomly selected 27 genes, including eight with putative functions belonging...
to different gene families, nine encoding hypothetical proteins, five encoding expressed proteins, and five without annotation (Additional file 7). Results showed that the expression patterns of these genes were well correlated with the RNA-seq results ($R = 0.846$), supporting the reliability of our RNA-seq data.

Identification and functional classification of genes specifically or preferentially expressed in maize silk

Analysis of genes specifically or preferentially expressed in MS might lead to a better understanding of the involvement of maize silk in plant reproduction. To this end, 725 genes specific to the MS library, compared with those from the other three tissues, were identified (Figure 1E). A significance of digital gene expression analysis [19] was then conducted to compare MS vs. MP, MS vs. MO, and MS vs. SL. Using fold change $\geq 2$, false discovery rate (FDR) $< 1e-05$, and $P$-value $< 1e-05$ as criteria for each comparison, 702 genes were identified from all three comparisons, and their transcript abundances were significantly greater in MS compared with MP, MO, and SL. These genes were considered to be preferentially expressed in MS. A total of 1,427 genes specifically or preferentially expressed in the silk were thus identified (Additional file 8A). These genes were subsequently divided into 21 categories and one unassigned gene category using MapMan [20]. The genes involved in RNA binding, processing, and transcription, protein targeting and degradation, signal transduction, miscellaneous enzyme families, and transport were well represented (Figure 4). In the category of RNA binding, processing, and transcription, 128 genes encoding transcription factors (TFs)
were identified. Most (58.6%) belonged to six families: MYB, WRKY, NAC, C2C2-CO-like, bHLH, and AP2-EREBP (Figure 5A, Additional file 8A). Expression and genetic analyses have substantiated the involvement of all six TF families in flower development and reproduction [21-24]. It is unclear, however, whether these TFs are directly involved in pollen-stigma interaction.

Genes encoding ubiquitin-proteasome-system (UPS) proteins constituted a large proportion (63.5%) of the protein targeting and degradation category (Figure 5B, Additional file 8A). Recent studies have shown that UPS proteins function in both sporophytic and gametophytic SI [25]. An E3 ubiquitin ligase, ARMADILLO REPEAT CONTAINING 1 (ARC1), is required for the rejection of incompatible pollen in sporophytic SI in *Brassica* [26], and F-box proteins (SCF) assist degradation of S-RNase in non-self pollen tubes in gametophytic SI [27]. Although a large number of UPS-related genes were identified in maize silk, SI has not been reported in maize; gametophytic SI has been found, however, in wild relatives of maize [5]. Moreover, ubiquitin has been shown to be involved in human sperm discrimination during fertilization [28,29]. Abnormal sperm is ubiquitinated on its surface, and then phagocytosed by epididymis cells [29]. Our findings provide new insights into the function of ubiquitin during pollen-stigma interaction in plants. The roles of genes encoding UPS-related proteins in maize silk deserve further investigation.

Seventy-one genes encoding receptor-like kinases (RLKs) were identified from the signal transduction category. These genes were classified into 13 subfamilies (Figure 5C, Additional file 8A). The largest subfamily includes 26 receptor-like kinases (DUF26 RLK) or CRKs, which play important roles in regulation of pathogen defense and programmed cell death [30-32]. Despite the essential roles of cysteine-rich proteins (CRPs) in SI and in pollen tube growth and guidance, the function of CRKs in pollen-pistil interaction has not been reported. However, based on a recent study involving PDLP1 with two DUF26 domains, we tentatively propose that the enrichment of CRKs in maize silk suggests an enhanced cell-to-cell trafficking role in this species [33]. The second largest category of genes encodes cytoplasmic receptor-like kinases (RLCKs). These kinases are directly activated by calcium/calmodulin that modulates pollen tube growth and pollen-pistil interaction [34,35]. The function of stigmatic RLCKs in compatible pollination is still unknown. The third category consists of genes

**Figure 3 Validation of RNA-seq results by real-time PCR.** Transcript abundance represented by the two heat maps are the average of transcript abundance values from three independent real-time PCR experiments (left) and RPKM values of the RNA-seq analysis (middle). For each gene, the tissue with the maximum transcript abundance was set to 100, and relative transcript abundances in the other three tissues were calculated based on this maximum. Relative expression is represented by color scales as indicated (right). R is the correlation coefficient value between the two platforms.
encoding leucine-rich repeat RLKs that are involved in regulation of pollen tube growth in petunia and tomato [36-38]. Examples in this category include LePRK1 and LePRK2, which are two well-defined pollen kinases that act in pollen-pistil interaction primarily by perceiving signals from pistil to regulate pollen tube growth. Additionally, six S-locus receptor kinase (SRK) proteins were identified from the MS-specific/preferential dataset. In

![Figure 4: Classification of stigma specifically/preferentially expressed genes in maize, rice, and Arabidopsis.](BMCGenomics_2012_13_294_Fig4)

![Figure 5: Distribution of TF, RLK, and UPS-related genes in the MS-specific/preferential dataset.](BMCGenomics_2012_13_294_Fig5)
Brassica and Arabidopsis, SRKs are usually localized in stigma papilla cells, and interact with S-locus cysteine rich proteins to control SI response [39]. We therefore suggest that genes encoding RLKs are probably important participants in stigma-mediated reproductive processes, but their functions in compatible pollination require further genetic elucidation.

Bioinformatic analysis revealed a number of maize silk genes that may act in reproductive processes

To identify GO terms significantly enriched in genes specifically or preferentially expressed in MS, a singular enrichment analysis (SEA) [18] was performed. Genes were classified based on their cellular component, molecular function, or biological process, and 22 GO terms were overrepresented in MS based on P-value < 0.001 and FDR ≤ 0.05 cutoffs (Table 1). Most of the overrepresented GO terms were involved in transport, cell wall regulation, signaling, and lipid metabolism, which are essential in stigmatic reproductive processes [1,40].

This suggests that the identified MS-specific/preferential dataset reflects the reproductive characteristics of maize silk.

Based on GO cellular components, the two GO terms designated as integral to membrane and peroxisome were overrepresented in the MS-specific/preferential datasets (Table 1). Several studies have shown that proteins integral to membranes play important roles in pollination [1]. On the other hand, the functions of all five peroxisome-associated proteins are unknown (Table 1). We therefore identified their homologous genes in Arabidopsis by performing a BlastP search against the Arabidopsis protein database [41]. Two identified genes encode peroxins, two gene products may be related to peroxisome biogenesis, and one gene product is involved in fatty acid metabolism. Among these genes, peroxin is essential for female and male gametophytic recognition, according to an analysis of the loss-function mutant abstinence by mutual consent [42].

Based on the GO molecular function and biological process analyses, terms with lipid-related annotation

| GO Term                        | Term type | Query item | Query total | Bg item | Bg total | P-value | FDR     |
|-------------------------------|-----------|------------|-------------|---------|----------|---------|---------|
| amino acid transport          | P         | 10         | 900         | 52      | 39203    | 2.80E-07| 0.00021 |
| protein amino acid phosphorylation | P         | 93         | 900         | 2387    | 39203    | 2.70E-06| 0.00082 |
| oligopeptide transport        | P         | 12         | 900         | 105     | 39203    | 6.10E-06| 0.0011  |
| fatty acid metabolic process  | P         | 14         | 900         | 187     | 39203    | 0.00013 | 0.014   |
| lipid transport               | P         | 10         | 900         | 104     | 39203    | 0.00015 | 0.014   |
| cell wall macromolecule catabolic process | P         | 8          | 900         | 73      | 39203    | 0.00028 | 0.022   |
| chitin catabolic process      | P         | 6          | 900         | 43      | 39203    | 0.00044 | 0.03    |
| drug transport                | P         | 11         | 900         | 154     | 39203    | 0.00097 | 0.05    |
| integral to membrane          | C         | 75         | 900         | 2156    | 39203    | 0.00055 | 0.029   |
| peroxisome                    | C         | 5          | 900         | 33      | 39203    | 0.0009  | 0.029   |
| protein serine/threonine kinase activity | F         | 92         | 900         | 2248    | 39203    | 4.40E-07| 0.00013 |
| monoxygenase activity         | F         | 33         | 900         | 553     | 39203    | 1.30E-06| 0.00029 |
| heme binding                  | F         | 43         | 900         | 876     | 39203    | 6.20E-06| 0.00007 |
| acyltransferase activity      | F         | 16         | 900         | 187     | 39203    | 8.80E-06| 0.00008 |
| amino acid transmembrane transporter activity | F         | 6          | 900         | 28      | 39203    | 3.60E-05| 0.0024  |
| electron carrier activity     | F         | 42         | 900         | 914     | 39203    | 3.50E-05| 0.0024  |
| transferase activity, transferring hexosyl groups | F         | 28         | 900         | 542     | 39203    | 9.40E-05| 0.0053  |
| sugar binding                 | F         | 12         | 900         | 150     | 39203    | 0.00021 | 0.0095  |
| secondary active transmembrane transporter activity | F         | 22         | 900         | 421     | 39203    | 0.00042 | 0.018   |
| chitinase activity            | F         | 6          | 900         | 43      | 39203    | 0.00044 | 0.018   |
| lipid binding                 | F         | 13         | 900         | 202     | 39203    | 0.00093 | 0.035   |
| drug transmembrane transporter activity | F         | 11         | 900         | 154     | 39203    | 0.00097 | 0.035   |

GO term classifications: P, Biological Process; C, Cellular Component; F, Molecular Function.

Query item number in MS preferential expressed genes.

Total annotated query item number in agrigo.

Query item number in maize genome version 5a.

Total annotated item number in maize genome version 5a.

Determined by Fisher exact test.

Determined by Benjamini–Hochberg–Yekutieli procedure.

GO terms with P-value < 0.001 and FDR ≤ 0.05 were regarded as overrepresented terms.
were overrepresented, including lipid binding, lipid transport, and fatty acid metabolic processes. This is consistent with previous findings that lipids play important roles in pollination [43-45]. We therefore further analyzed lipid metabolism-related genes in the MS-specific/preferential gene dataset using MapMan [20] (Additional file 8A). Nearly 30% of the genes were involved in fatty acid synthesis and elongation. These overrepresented genes encode members of 3-ketoacyl-CoA synthase (KCS), long-chain acyl-CoA synthase, acyl-activating enzyme, and acyl-CoA binding protein families. KCS proteins are responsible for the synthesis of very-long-chain fatty acids (VLCFAs), which are major components of the cuticle (wax and cutin) covering the stigma [46]. In Arabidopsis, mutations in FIDDLEHEAD and CER genes have been shown to disrupt VLCFA biosynthesis, and thus affect pollen-stigma interaction, indicating important roles for VLCFAs in reproductive processes [45,47-49]. We also identified a number of genes involved in phospholipid synthesis in the MS-specific/preferential gene dataset. Phospholipids regulate pollen germination and tube growth [36,50], but their functions in the stigma are unknown. Moreover, genes involved in lipid transport and degradation were also overrepresented. Most of the genes encoding lipid binding proteins and lipid transfer proteins (LTPs) that have been shown to be essential to successful pollen adhesion and tube guidance were studied in wet stigma species [51-53]. In addition, we found many genes involved in lipid degradation and desaturation, such as genes encoding triacylglycerol lipase, acyl-CoA thioesterase, phospholipase D, GDSL lipases, and fatty acid desaturase. According to several recent studies [1], some of these proteins may support pollen hydration and germination. For example, EXL4, a member of the GDSL family, is involved in rapid initiation of pollen hydration on the stigma [54]. Because application of unsaturated triacylglycerols to stigma-less tobacco plants promotes pollen hydration and germination [43], fatty acid desaturase might be required for reproductive success in the stigma.

Genes involved in various transport processes, including transport of amino acids, oligopeptides, lipids, and drugs, were overrepresented based on GO biological process analysis (Table 1). The stigma is a specialized tissue that secretes materials supporting the growth of compatible pollen tubes and inhibiting the growth of incompatible pollens [55]. To investigate transport-related genes in the MS preferential gene dataset in detail, a MapMan functional analysis [20] was performed (Figure 6, Additional file 8A). Unexpectedly, genes encoding amino acid, peptide, and oligopeptide transporters constituted a large proportion of the transport-related genes. Results from a recent study showing that amino acids transported from the pistil regulate glutamate receptor-like proteins, which in turn control pollen tube Ca\(^{2+}\) concentration and growth, provide an insight into amino acid transporter function in maize silk [56]. Furthermore, AtPTR5, a gene encoding a peptide transporter in Arabidopsis, was found to facilitate peptide uptake into germinating pollen [57]. Another peptide transporter gene, AtpTRI, is also expressed in the stigma and style, and may function in a similar manner to supply peptides for initial pollen tube growth [58]. In addition, as shown in Figure 6, genes for potassium, unspecific cation, divalent, gated, and calcium channel proteins were also well represented. Among these, potassium and calcium channel proteins were overrepresented. In the stigma, Ca\(^{2+}\) is required for pollen germination and pollen tube guidance [59,60]. Although potassium channel proteins and K\(^+\) have been implicated in regulation of pollen tube growth [61], their roles in the pistil are yet to be investigated.

Approximately 14% of genes identified in the maize genome encode proteins with putative N-terminal signal peptides, and nearly 19% of the genes specifically or preferentially expressed in MS encode proteins involved in the secretory pathway (Additional file 9). The genes enriched in maize silk are likely participants in extracellular signaling. Various secreted polypeptides have been reported to be involved in signaling during pollen-pistil interaction. These include pollen tube attractants, factors for pollen germination and tube growth, and determinants of SI. Interestingly, most of these are CRPs [62], but so far, no CRPs have been identified from stigmas in the Poaceae. In this study, we used sequence motif models generated from each CRP subfamily [63] to identify CRPs in the MS preferential datasets (see methods and Additional file 10). Identified CRP sequences have N-terminal signal peptides, and CRP subcellular locations include the cell wall, plasma membrane, and endomembrane system (Additional file 10). Classes of CRPs identified in the maize silk preferential dataset include LTPs, lipid binding proteins, and serine-type endopeptidase inhibitors. Style/stigma cysteine-rich protein, a plant LTP, is involved in adhesion of the pollen tube to the transmitting tract [51-53]. The large number of LTPs identified from maize silk may provide an explanation for how pollen tubes can adhere to the epidermal surface of the transmitting tract during their long journey through the maize silk. Lipid binding proteins are also involved in pollen-pistil interactions. For example, glycine-rich proteins with lipid binding domains reduce pollen hydration rates [64]. Although serine-type endopeptidase inhibitors have not been implicated in pollen-pistil interactions, these inhibitors are known to participate in cell-to-cell communication during plant-pathogen interactions [65].
Dry stigmas may share similar mechanisms in the early stages of pollen-stigma interaction

Dry stigma preferential datasets have currently been generated for two species: Arabidopsis and rice [12,13]. To reveal the conservation or divergence of stigma preferential genes among different dry stigmas, we compared gene sequences generated from these earlier studies with our MS-specific/preferential gene sets.

For this purpose, protein sequences of 1,427 maize silk-specific/preferential genes were extracted and blasted against the MSU rice protein database [66] and Arabidopsis protein database [41], with homologous proteins filtered using an E-value ≤ 1e-10 cutoff. We then selected homologous protein-encoding genes that were also included in rice and Arabidopsis stigma-specific/preferential datasets. For each maize gene, only the best hit (lowest homologous protein E-value) of these genes in the rice or Arabidopsis stigma-specific/preferential datasets was selected as the homologous gene. We found that 471 genes specifically or preferentially expressed in maize silk (32.96% of the MS-specific/preferential dataset) were homologous to 213 rice stigma-specific/preferential genes (38.87% of the rice stigma-specific/preferential dataset) (Additional file 11A). We also blasted the protein sequences of the 548 rice and 115 Arabidopsis stigma-specific/preferential genes against the maize AGPv2 5b filtered gene set peptide database [16]. Homologous maize genes were identified using the same criterion as above. As a result, 221 rice stigma-specific/preferential genes (40.32% of the rice stigma-specific/preferential dataset) matched 179 maize silk-specific/preferential genes (12.54% of the MS-specific/preferential dataset) (Additional file 11C), and 53 Arabidopsis stigma-specific/preferential genes (46.09% of the Arabidopsis stigma-specific/preferential dataset) matched 48 MS-specific/preferential genes (3.77% of the MS-specific/preferential dataset) (Additional file 11D). From these BLAST results, we found that the three dry stigma-specific/preferential datasets match each other well, suggesting conserved mechanisms across monocots and dicots.

We found that 107 maize silk-specific/preferential genes had homologous matches in both rice (45 genes) and Arabidopsis (23 genes) stigma-specific/preferential datasets (Additional file 12A). The proteins encoded by some of these genes (17 out of 107) have been shown to be involved in the early events of pollen-stigma interaction. For example, S-locus glycoproteins participate in pollen adhesion [67], aquaporins and GDSL lipases regulate the process of pollen hydration [54,68], fatty acid-elongation-related proteins participate in pollen...
hydration and pollen tube germination [43,44], and expansins and pectinesterases function in initial growth and penetration of the pollen tube into the stigma surface [69-71]. Despite differences in their morphology and transcript abundance patterns, dry stigmas of maize, rice, and Arabidopsis may share similar mechanisms in the early stages of stigma-mediated reproductive processes.

Maize stigma shares more conserved functional mechanisms with rice than with Arabidopsis

To compare the characteristics of stigma-specific/preferential genes among maize, rice, and Arabidopsis, we classified the genes in all three dry stigma datasets into 21 categories and one unassigned gene category using MapMan [20] analysis, and compared their functional profiles (Figure 4). Functional distributions of the genes in maize and rice were very similar to each other, but differed from those of Arabidopsis. Moreover, genes related to transcription and hormone functions constituted relatively larger proportions in maize and rice than in Arabidopsis. These results suggest that transcriptional regulation networks and hormone-mediated regulation play more prominent roles in stigmas of maize and rice than in those of Arabidopsis.

The expression profile of maize silk-specific/preferential genes shared more similarity with that of rice than with that of Arabidopsis (Additional file 12B). A large number of genes were conserved between maize (364) and rice (185) stigma-specific/preferential datasets, but the same genes were not present in the Arabidopsis dataset. To further analyze common characteristics shared between maize and rice, we performed GO analysis on these conserved genes [18]. Gene terms related to protein serine/threonine kinase activity, protein amino acid phosphorylation, and oligopeptide transport were overrepresented (Table 2). Because maize and rice are both members of Poaceae while Arabidopsis belongs to Brassicaceae, we suggest that mechanisms related to stigma function are more similar in species with closer genetic backgrounds.

|GO Term* | Term type* | Query itemb | Query totalb | Bg itemd | Bg totale | P-valuef | FDRg |
|----------|------------|-------------|--------------|----------|----------|----------|------|
|protein amino acid phosphorylation | P | 67 | 294 | 2387 | 39203 | 1.50E-17 | 8.60E-15 |
|oligopeptide transport | P | 12 | 294 | 105 | 39203 | 4.10E-11 | 2.90E-09 |
fatty acid biosynthesis process | P | 10 | 294 | 159 | 39203 | 5.00E-07 | 2.60E-05 |
apoptosis | P | 8 | 294 | 168 | 39203 | 5.00E-05 | 0.002 |
|regulation of transcription | P | 46 | 294 | 3352 | 39203 | 0.00024 | 0.0069 |
multidrug transport | P | 5 | 294 | 79 | 39203 | 0.00035 | 0.0076 |
|protein serine/threonine kinase activity | F | 67 | 294 | 2248 | 39203 | 8.30E-19 | 2.80E-16 |
|ATP binding | F | 85 | 294 | 5133 | 39203 | 1.50E-09 | 6.30E-08 |
|acyltransferase activity | F | 11 | 294 | 187 | 39203 | 2.70E-07 | 5.70E-06 |
sugar binding | F | 10 | 294 | 150 | 39203 | 2.90E-07 | 5.80E-06 |
|transcription factor activity | F | 28 | 294 | 1390 | 39203 | 7.50E-06 | 0.00012 |
|sequence-specific DNA binding | F | 22 | 294 | 985 | 39203 | 1.40E-05 | 0.00022 |
|transporter activity | F | 37 | 294 | 2111 | 39203 | 0.00011 | 0.0015 |

*GO term classifications: P, Biological Process; C, Cellular Component; F, Molecular Function.

bQuery item number in MS preferential expressed genes.

cTotal annotated query item number in agriGO.

bQuery item number in maize genome version 5a.

eTotal annotated item number in maize genome version 5a.

fDetermined by Fisher exact test.

gDetermined by Benjamini–Hochberg–Yekutieli procedure.

GO terms with P-value < 0.001 and FDR ≤ 0.05 were regarded as overrepresented terms.

Amino acid and lipid transport-related genes are involved in distinctive mechanisms of maize silk-mediated reproductive processes

Comparison of functional profiles of the three different dry stigmas (Figure 4) revealed that genes involved in transport and in protein targeting and degradation represented a larger percentage of the MS-specific/preferential dataset than those of rice and Arabidopsis. In addition, we identified 923 genes that were distinctively represented in MS-specific/preferential dataset (Additional file 12C). GO analysis of these genes indicated that genes involved in amino acid transmembrane transporter activity and in amino acid and lipid transport were overrepresented (Table 3). A previous study has shown that nutrients in pollen can support only about 2 cm tube growth in maize silk [7]. For successful fertilization, maize silk must provide enough nutrients and appropriate signals to support pollen tube growth and guidance for a much longer distance. These supporting activities are apparently carried out by the distinctively-enriched transporters in maize silk. In accordance with
this idea, we identified 13 amino acid transporter genes in the maize silk-specific/preferential dataset, but only one gene in rice and none in Arabidopsis. As shown in previous studies [56], amino acid-mediated communication is required for pollen-pistil interaction. The transporters may also transfer amino acids into the pollen tube to regulate its growth. In addition, the lipid transport-related genes uniquely represented in the maize silk-specific/preferential dataset included a number of genes encoding LTPs that may function in the adhesion of pollen tubes to the surface of transmitting tract during their growth in the silk. We therefore suggest that the distinctive expression of genes encoding amino acid and lipid transporters is consistent with the extended length of maize silk.

A number of genes showed presence/absence variations and different expression patterns between different maize inbred lines

Two recent transcriptomic studies on maize inbred line B73 have revealed important information about temporal and spatial gene expression patterns in different tissues, and highlighted the utility of microarray and RNA-seq methods in the generation of databases for gene discovery and functional characterization [14,15].

In one of the studies, RNA-seq analysis generated transcriptome profiles of different reproductive tissues, including cobs and pollen, as well as developing seeds and leaves [15]. To make a meaningful comparison between inbred lines B73 and Zheng58, genes exclusively or preferentially expressed in silk compared with those in post-emergence cob, pollen, and leaf were extracted from the previously-reported data [15], and were searched against our MS-specific/preferential dataset. A total of 72 specific and two preferential genes in our dataset were found to be specific to previous data, and 154 specific and 153 preferential genes were preferentially expressed in the tissues of the previous studies [15] (Additional file 13A). On the other hand, 24 out of 138 silk-exclusive genes of inbred line B73 were absent from our RNA-seq dataset, and 25 of our 1,427 MS-specific/preferential genes were absent from the data derived from inbred line B73. Presence/absence variations (PAVs) between inbred lines B73 and Zheng58 have been found for a number of genes [72]. Our results indicate that besides the previously-reported PAVs, many genes show different expression patterns between these two inbred lines.

We also compared our MS-specific/preferential dataset to datasets derived from tissues absent in our study but present in the two previous studies. A comparison between our MS-specific/preferential dataset and those of leaves, cobs, ovule, tassels, anthers, pollen, seeds, embryo, and endosperm from the previously-reported microarray data identified nine specific and 27 preferential genes

| GO Term                        | Term type | Query item | Query total | Bg item | Bg total | P-value | FDR  |
|-------------------------------|-----------|------------|-------------|---------|----------|---------|------|
| lipid transport               | P         | 9          | 476         | 104     | 39203    | 6.00E-06 | 0.0034 |
| amino acid transport          | P         | 6          | 476         | 52      | 39203    | 4.20E-05 | 0.0079 |
| chitin catabolic process       | P         | 5          | 476         | 43      | 39203    | 0.00018  | 0.025  |
| tetracycline transport         | P         | 6          | 476         | 75      | 39203    | 0.00033  | 0.029  |
| response to antibiotic         | P         | 6          | 476         | 84      | 39203    | 0.00006  | 0.038  |
| amino acid transmembrane transporter activity | F | 6          | 476         | 28      | 39203    | 1.00E-06  | 0.00019 |
| chitinase activity             | F         | 5          | 476         | 43      | 39203    | 0.00018  | 0.022  |
| small conjugating protein ligase activity | F | 13         | 476         | 324     | 39203    | 0.00023  | 0.022  |
| tetracycline transporter activity | F     | 6          | 476         | 75      | 39203    | 0.00033  | 0.022  |
| lipid binding                  | F         | 9          | 476         | 202     | 39203    | 0.00006  | 0.049  |
| coenzyme binding              | F         | 20         | 476         | 734     | 39203    | 0.001    | 0.049  |
| integral to membrane          | C         | 51         | 476         | 2156    | 39203    | 1.90E-05  | 0.0016 |
| peroxisome                    | C         | 5          | 476         | 33      | 39203    | 4.80E-05  | 0.0016 |

*GO term classifications: P, Biological Process; C, Cellular Component; F, Molecular Function.
*Query item number in MS preferential expressed genes.
*Query item number in AGPv2.
*Total annotated query item number in AGPv2.
*Total annotated item number in AGPv2.
*Determined by Fisher exact test.
*Determined by Benjamini–Hochberg–Yekutieli procedure.

GO terms with P-value < 0.001 and FDR ≤ 0.05 were regarded as overrepresented terms.
In this study, 1,427 genes expressed predominantly or specifically were identified in maize silk using RNA-seq. A number of novel gene sets might be involved in stigma-mediated reproductive processes. A comparative analysis of the stigma-specific/preferential gene datasets of Arabidopsis, rice, and maize suggest that 1) different dry stigmas share similar mechanisms in the early stages of pollen-stigma interaction; 2) maize shares more conserved functional mechanisms with rice than with Arabidopsis; and 3) genes involved in amino acid and lipid transport might be responsible for the distinct functions of reproductive processes in maize silk. Further functional analysis of the identified genes will aid our understanding of stigmatic mechanisms taking place during reproduction.

Methods
Plant materials, growing conditions, and RNA extraction
Maize inbred line Zheng58 was used for RNA-seq and microarray analysis. For SL materials, seeds of Zheng58 were surface-sterilized with 3% sodium hypochlorite for 10 min and rinsed well with distilled water. Sterilized seeds were pre-germinated, and their pollen tubes were long and healthy. After the silks had grown from the husks for approximately 3 days, the upper parts (about 4 cm) were harvested as MS tissues. After removing surrounding glumes and upper silk, whole MO were harvested from the middle portion of the ear. Reproductive tissues (including MP, MO, and MS) used for microarray analysis were similarly collected from other plots at the Shandong Agricultural University experimental station. Samples for each biological replicate were collected from three maize plants and combined. Visible MS, SL, and MO growth conditions were photographed with a Canon digital camera. MP images were obtained using an Olympus DX51 microscope. All samples were immediately frozen in liquid nitrogen and stored at −80°C for later use.

Total RNA from MS, MP, MO, and SL was extracted according to the CTAB protocol [79] and purified using an RNaseasy MinElute Cleanup Kit (Qiagen, Valencia, CA). Purified RNA samples were then quantified by Nanodrop spectrophotometry (Nanodrop Technologies, Wilmington, DE) and agarose gel electrophoresis. RNA integrity was assessed qualitatively using an Agilent 2100 Bioanalyzer (Agilent Technologies, Böblingen, Germany).

RNA sequencing library construction and Illumina sequencing
Approximately 5 to 8 μg total RNA was used to construct each RNA-seq library of MS, MP, MO, and SL. Poly(A) RNAs were isolated from total RNA using oligo (dT) magnetic beads (Illumina, San Diego, CA). RNA fragmentation, cDNA synthesis, and PCR amplification were performed according to the Illumina RNA-seq protocol (Cat # RS-100-0801). Sequencing was carried out on an Illumina HiSeq 2000 sequence analyzer at Beijing Genomics Institute (Shenzhen, China). Single-end 49-bp sequence reads were generated. Image data acquired from the sequencing run were base-called and quality-analyzed with Illumina Genome Analysis Pipeline version 1.6. Raw reads with adaptors, containing more than 10% of unknown bases, or with more than 50% of low quality bases (quality value ≤ 5) were excluded from further analysis. To evaluate the sequencing data, we performed a sequence saturation analysis for the data from each of the four tissues. All sequence data are available in the ArrayExpress database [80] under accession number E-MTAB-964.

For reads-to-reference genome/genes alignment, SOAP2 [17] was used, allowing no more than two mismatched bases. Sequence reads for each tissue sample were generated from statistical analysis, and aligned to the AGPv2 maize B73 reference genome [16]. Distributions of reads...
mapped to genomes and genes from the four tissues are shown in Additional file 2 and Additional file 3. For all mapped genes with unique matched reads, transcript abundance was normalized by using the RPKM (Reads Per kb per Million reads) method [81]. If there was more than one transcript for a gene, the longest one was used to calculate its transcript abundance and coverage.

PCC analyses of log2-transformed RPKM values between the four RNA-seq libraries were performed using the R package [82], and log2-transformed RPKM values of some genes less than zero were still set to zero. Hierarchical clustering of genes expressed in at least one sample was performed using average linkage cluster analysis (Cluster 3.0), and the results were visualized using Java Treeview 1.0 software (Stanford University School of Medicine, Stanford, CA). A heat map of correlation values was drawn using Scalable Vector Graphics.

Target preparation, Affymetrix GeneChip hybridization, and statistical analysis of microarray data
In accordance with the Affymetrix GeneChip Expression Analysis Technical Manual [83], 8 μg of total RNA was biotin-labeled for each biological replicate. Target hybridization, washing, staining, and microarray scanning were conducted according to the standard GeneChip® Expression Analysis Technical Manual [83]. Original signal data from microarrays were processed using the GeneChip Operating software (GCOS 1.4). The Affymetrix library RMA (Robust Microarray Analysis) [84] tool was used to normalize the raw data. Microarray data are available in the ArrayExpress database [80] under accession number E-MEXP-3292.

Real-time PCR analysis
RNA extraction and purification were performed as described above. Genomic DNA contamination was eliminated by RNase-free DNase I treatment (Promega, Madison, WI). Total RNA (4 μg) was reverse-transcribed with oligo (dT) primer using M-MLV reverse transcriptase according to the manufacturer’s instructions (Promega, Madison, WI, USA). All primer sequences are listed in Additional file 7. Quantitative PCR assays in triplicate were performed using SYBR Green Real-time PCR Master Mix (Toyobo, Osaka, Japan) with a Bio-Rad CFX96 Real-Time Detection System. Quantitative variation in the different replicates was calculated using the delta-delta threshold cycle relative quantification method. Amplification of 18S rRNA was used as an internal control for data normalization.

Comparative analysis of RNA-seq, microarray, and real-time PCR data
To assess the reliability of the RNA-seq results, a comparison between RNA-seq and 18 K Maize Genome array data was carried out. Because the microarray was based on NCBI GenBank (September 29, 2004) and Z. mays UniGene Build 42 (July 23, 2004) databases, each probe set was assigned to one EST accession number in GenBank. For the microarray, 17,734 probes represented 13,339 genes in the maize genome [85]. To improve the reliability of the comparison, the corresponding nucleotide sequences were extracted from the Affymetrix GeneChip Maize Genome Array sequence file [85] and blasted against the maize AGPv2 5b filtered gene set database [16]. Using identity > 0.8 and E-value < 1e-20 for the cutoff, the best hit obtained for a given maize transcript was assigned to the gene represented by each probe set. Prior to performing the comparison between RNA-seq and microarray data, genes with more than one predicted isoform were removed from both datasets. RPKM values and RMA-normalized probe set expression values were log2-transformed, and SCC values were calculated to account for the discrepancy in scale between the two platforms. In addition, we calculated the correlation coefficient between real-time PCR and RNA-seq results.

GO analysis
To facilitate comparison of overrepresented GO terms in the four experimental tissues, PAGE [18], a comparative GO analysis tool, was adopted. For genes specifically/preferentially expressed in MS, genes common to maize and rice stigma-specific/preferential datasets, and genes distinct in the MS-specific/preferential dataset, another GO analysis tool, SEA [18], was used to identify overrepresented GO terms.

Gene annotation, classification, and sub-cellular location prediction
Genes were first annotated using AGPv2 5b.60 [16], and then classified according to MapMan annotation [20] on the longest transcripts of each gene. The sub-cellular location of each gene was predicted using TargetP [86]. To see whether a sub-cellular location was overrepresented, the proteins encoded by the longest transcripts of maize AGPv2 5b filtered datasets [16] were set as a background for the analysis.

Comparison of stigma-specific/preferential datasets among maize, rice, and Arabidopsis
The specifically or preferentially expressed datasets of maize, rice, and Arabidopsis were compared to identify the similarity and diversity of transcript abundance profiles in their stigmas. The specifically/preferentially expressed genes of rice and Arabidopsis stigmas were acquired directly from published data [12,13], and their annotations were updated by using Rice Genome Annotation Project release 7 [66] and Arabidopsis Information Resource version 10 [41]. Comparisons between maize
and rice/Arabidopsis stigma-enriched genes were carried out using two statistical methods. First, protein sequences of maize silk-enriched genes were extracted and blasted against rice and Arabidopsis [41,66] protein databases, with E-values ≤ 1e-10 used to identify their homologous genes in rice and Arabidopsis. Using the same criterion, protein sequences of rice and Arabidopsis stigma specifically or preferentially expressed genes were blasted against the maize AGPv2 5b filtered gene set peptide database [16] to find their homologous genes in maize silk. If a homologous gene in one stigma dataset was found in one of the other two stigma datasets, it was considered to be conserved between the two stigma datasets. Secondly, we used function-classification methods to identify conserved and novel gene families among the three species based on MapMan annotation [20].

Identification of CRPs in the maize silk-specific/preferential dataset
The sequence motif models of each group of CRPs constructed by Silverstein et al. [63] were used as queries to blast against the AGPv2 5b filtered gene set peptide database [16]. Using E-value < 1e-10 as the criterion, we obtained a large number of genes with high similarities to the input sequence motif models. We then identified candidate genes belonging to the MS-specific/preferential dataset from these homologs to examine whether they truly encoded CRPs. Genes that met the standards set by Silverstein et al. [63] were considered to be novel CRPs in maize silk.

Comparative analysis of maize transcriptomic data between different inbred lines
We first extracted genes exclusively or preferentially expressed in silk compared with those in post-emergence cob, pollen, and leaf from previously-reported RNA-seq data [15]. Genes expressed in silk but not in the other three tissues were considered to be MS-specific, and genes expressed in maize silk with transcript abundances twice greater than all the other three tissues were designated as MS-preferential. Our MS-specific/preferential dataset was then searched for these specific/preferential genes. Only identical sequences with the same ID numbers were selected.

To compare the MS-specific/preferential dataset with those of tissues absent in our study but analyzed in published RNA-seq data [15], the published data set was searched for our MS-specific/preferential genes, and identical sequences with the same ID numbers were selected. Transcript abundance of these genes in silk was compared with that in leaves, cobs, ovule, tassels, anthers, pollen, seeds, embryo, and endosperm in the published data set. Genes expressed in silk and absent in other tissues were considered to be specifically expressed in silk. Silk-preferential genes were selected using a criterion of transcript abundances in silk twice greater than in all the other tissues.

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To compare the MS-specific/preferential dataset with those of tissues absent in our study but present in previously-reported microarray data [14], we first searched for our MS-specific/preferential genes in the microarray data and selected identical genes with the same ID numbers. Using the previously-reported data, the transcript abundance of these genes in silk was then compared with that in roots, whole seedling, internodes, cobs, tassels, anthers, leaves, endosperms, embryos, and pericarp using Significance Analysis of Microarrays software [87]. Genes that were not expressed in other tissues were considered to be specifically expressed in silk. Genes up-regulated in silk with fold change ≥ 2 and q-value ≤ 0.05 were regarded as preferentially expressed.

For comparisons between the MS-specific/preferential dataset and those of tissues absent in our study but analyzed in published RNA-seq data [15], the published data set was searched for our MS-specific/preferential genes, and identical sequences with the same ID numbers were selected. Transcript abundance of these genes in silk was compared with that in leaves, cobs, ovule, tassels, anthers, pollen, seeds, embryo, and endosperm in the published data set. Genes expressed in silk and absent in other tissues were considered to be specifically expressed in silk. Silk-preferential genes were selected using a criterion of transcript abundances in silk twice greater than in all the other tissues.
Additional file 8: Stigma specifically/preferentially expressed genes in maize, rice and Arabidopsis. Rice and Arabidopsis stigma specifically/preferentially expressed genes were derived directly from the previous studies [12,13]. Gene category of maize, rice and Arabidopsis stigma-specific/preferential genes were based on MapMan annotation.

Additional file 9: Subcellular localization prediction of MS-specific/preferential genes and the longest transcripts of all filtered genes in maize AGPv2 5b.

Additional file 10: Distribution of cysteine-rich proteins in maize, rice and Arabidopsis stigma-specific/preferential datasets.

Additional file 11: Sequence comparison among maize, rice and Arabidopsis stigma-specific/preferential datasets. A, 471 maize silk specifically/preferentially expressed genes hit 213 rice stigma specific/preferential genes; B, 140 maize silk specifically/preferentially expressed genes hit 37 Arabidopsis stigma-specific/preferential genes; C, 221 rice stigma specifically/preferentially expressed genes hit 179 maize silk-specific/preferential genes; D, 53 Arabidopsis stigma specifically/preferentially expressed genes hit 48 maize silk-specific/preferential expressed genes.

Additional file 12: Distribution of the homologs of maize silk specifically/preferentially expressed genes. A, Common genes in the three species; B, Maize silk-specific/preferential genes that can only hit rice stigma-specific/preferential dataset; C, Distinct stigma-specific/preferential genes that only expressed in maize silk.

Additional file 13: Comparative analysis of maize transcriptomic data between inbred lines Zheng58 and B73.

Additional file 14: Mature pollen grains germinate well in liquid medium.

Competing interests
The authors declare that they have no competing interests.

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Authors’ contributions
XSZ was responsible for design and coordination in this study. XHX, HC, and JPM performed molecular experiments (material collection, RNA extraction, real-time PCR). XHX and HC carried out the data analysis. XSZ, HC, XHX, and YLS draft this manuscript. XSZ, XQG and FW revised the manuscript. All the authors read and approved the final manuscript.

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