GeneWeaver: a web-based system for integrative functional genomics

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ABSTRACT

High-throughput genome technologies have produced a wealth of data on the association of genes and gene products to biological functions. Investigators have discovered value in combining their experimental results with published genome-wide association studies, quantitative trait locus, microarray, RNA-sequencing and mutant phenotyping studies to identify gene-function associations across diverse experiments, species, conditions, behaviors or biological processes. These experimental results are typically derived from disparate data repositories, publication supplements or reconstructions from primary data stores. This leaves bench biologists with the complex and unscalable task of integrating data by identifying and gathering relevant studies, reanalyzing primary data, unifying gene identifiers and applying ad hoc computational analysis to the integrated set. The freely available GeneWeaver (http://www.GeneWeaver.org) powered by the Ontological Discovery Environment is a curated repository of genomic experimental results with an accompanying tool set for dynamic integration of these data sets, enabling users to interactively address questions about sets of biological functions and their relations to sets of genes. Thus, large numbers of independently published genomic results can be organized into new conceptual frameworks driven by the underlying, inferred biological relationships rather than a pre-existing semantic framework. An empirical ‘ontology’ is discovered from the aggregate of experimental knowledge around user-defined areas of biological inquiry.

INTRODUCTION

An increasing number of investigators have integrated genome-wide experimental data across studies (1–5), but have lacked the data resources, algorithms and tools for widespread deployment of this approach. The data comes largely from gene expression microarray and now RNA-sequencing experiments, quantitative trait locus (QTL) mapping and genome-wide association studies (GWAS). Additional functional genomic data come from mutation and perturbation screen analyses (6–8) and from the vast array of individually curated biological relationships that have been associated with an array of biological ontology terms (e.g. (9–11). There are many powerful applications of this strategy, including meta-analysis or convergent analysis across microarrays (4), the refinement of QTL positional candidates through gene-centered evidence (2,12), cross-species translation of gene expression (13), mapping model organism findings onto human disease (14), using similarity of functional associations to identify the function of poorly characterized genes and using the similarity of biological associations to identify relations among biological processes.

GENEWEAVER: A DATA REPOSITORY AND ANALYSIS SYSTEM FOR FUNCTIONAL GENOMIC DATA INTEGRATION

GeneWeaver, powered by the Ontological Discovery Environment (15), is a freely available web-based software system that enables biologists to perform integrative functional genomics on their own collections of experimental data in combination with the system’s incorporated database. The system enables users to integrate genome-wide experimental studies across multiple protocols and species (1–5), using convergent evidence to find consensus among the noise in reported associations of genes and biological functions and to classify functions...
based on their underlying biological entities. The system addresses a major hurdle to the widespread implementation of integrative functional genomics by integrating a curated data repository with accompanying analysis tools (16).

The GeneWeaver data repository currently contains over 48 000 gene sets consisting of over 80 000 genes from seven species, derived largely from gene expression microarrays, RNA-sequencing experiments, QTL mapping and GWAS, mutation and perturbation screen (6–8), and curated biological relationships (e.g. (9–11)). In addition, GeneWeaver integrates other large community data sources, such as the drug related gene database of the Neuroscience Information Framework (NIF) (17), GeneNetwork (18) and the Comparative Toxicogenomics Database (CTD) (19). Inclusion of these diverse data resources, often with incompatible formats, demonstrates how GeneWeaver can promote the identification of common biological processes of poorly annotated genes through data integration without the overhead of extensive reanalysis.

Unlike ontology over-representation analysis systems [e.g. DAVID (20)] that match a single gene set result to many curated associations of genes to ontology terms or pathways, GeneWeaver infers relations among genes and processes by set–set matching across all gene sets in a query. This enables users to examine similar and distinct results among a large number of related experiments (including those compiled into ontology annotations), and to compare the role of genes to each of the functions assessed in the various experiments.

At its core, GeneWeaver achieves this data convergence through use of discrete gene-function associations, where gene lists are converted into an edge list of gene-to-biological function relationships where gene set descriptors are loosely defined as a ‘biological function.’ The resulting pair-wise relationships between genes and functions are queried using multivariate statistical methods and novel algorithms based on graph data mining including biclique, clique and paraclique analysis, clustering of the similarity matrix, Jaccard and hypergeometric tests, graph clustering algorithms such as dominating set and a novel algorithm for inference of an empirically derived ontology of gene functional relations, the Phenome Map (15).

APPLICATIONS AND USE OF GENEWEAVER

A visitor to GeneWeaver lands on a public page with links to data management, advanced search and analysis tools. All users have access to numerous tutorials, including videos, How To guides and our standards documentation (see Supplemental Data, http://geneweaver.org/index.php?action=help). While anonymous users have access to the GeneWeaver’s curated data repository, an account is required to store uploaded data, view data shared directly among registered users and their groups and to create and share individual projects. The system consists of a database containing curated gene set data and community metadata, both of which interact with a variety of discovery and data management tools (Figure 1). A user of the site typically begins their analysis with a search for a term or gene of interest. When a user enters a free text term, the default is a full-text search of meta-data fields including descriptions, publication information and NCBO Annotator (21) derived MeSH and Disease Ontology (22) terms. When a user searches for a gene symbol or other gene identifier, any gene set containing that gene symbol, its homologs or any identifier mapped onto that identifier by major model organism databases is retrieved. Users have tremendous flexibility in specifying searches through field-restricted queries and Boolean search.

For example, a search for ‘Drd2’ retrieves gene sets including expression in the lateral septal complex from the Allen Mouse Brain Atlas (GS127928), positional candidates for mechanical sensitivity and for methamphetamine response in a mouse QTL analyses (GS84202), QTL analysis for alcohol consumption (GS84212, GS84216), mutant phenotypic alleles for abnormal alcohol consumption from Mouse Genome Informatics (GS87659), Gene Ontology annotations for post synaptic density (GS97454), paraquat interacting genes from CTD (GS121514) and 712 other gene sets. A query for ethanol retrieves 540 Gene Sets including genetic studies of alcoholism in humans (e.g. GS46979, GS46980, GS46982, GS46984, GS46985), mice (GS37188) and rats (GS37196, GS37197). Clicking on a Gene Set name opens a detail page that enables users to view the metadata or contents of a gene set, translate the identifiers, execute overrepresentation analyses via GAGGLE (23) and find similar gene sets. Users can select appropriate gene sets using check boxes and add them to a ‘Project’ using ‘Add Selected to Project.’ The user should then select ‘Analyze Gene Sets’ to perform integrative analyses using tools described below (Figure 2). The forms under ‘Manage Gene Sets’ allow users to upload their own gene sets and add them to new or existing projects for seamless integrative analysis with the data in the repository.

GENEWEAVER ANALYSIS TOOLS AND FUNCTIONS

User selected gene sets are combined through identifier cross-mapping to enable analysis of discrete relationships between genes and functions using a wide variety of tools. Each of the embedded tools functions independently to explore and prioritize genes and gene sets of interest. The inputs and outputs are described as genes or gene sets (Table 1). The modular strategy allows users to arrange tools in numerous distinct workflows. Most of these tools operate on the representation of a group of gene lists as a bipartite graph consisting of genes as one set of vertices and gene sets as a second set of vertices. The edges between these vertices are weighted by scores or unweighted discrete values (15). To execute an analysis from the ‘Analyze Gene Sets’ page, first select a project using the check box next to its name, or expand the project to select individual gene sets within it.
Then select a tool using the icons to the left. Clicking a tool icon immediately executes the default analysis, but tool options may be set by first expanding the options by clicking on the plus sign next to the icon. Most tools have options for the inclusion of homologous genes and for setting thresholds and analysis parameters. Several tools also allow users to highlight nodes containing ‘emphasis genes,’ a user specified list of genes of particular interest.

The gene set graph

The gene set graph tool is a module that enables users to identify genes and gene sets that are highly connected among an input group of gene sets. Gene Sets are represented as a central column of vertices, with high degree (highly connected) gene vertices plotted to the right and lower degree gene vertices plotted to the left (Figure 3, Supplementary Figure S1). Users control the minimum degree threshold and can incorporate the ‘emphasis genes’ feature.

The phenome graph

The phenome graph is a hierarchical network of multiway gene set intersections (Figure 4, Supplementary Figure S2). The result is a graph of gene set intersections of very high order, enabling users to find genes connected to all populated subsets of an input set of gene lists. In algorithmic terms, these intersections are created from the overlap of maximal bicliques in discrete bipartite structures (15). This tool creates an ‘empirical ontology’ of function based on the underlying genes within curated and user-defined gene sets in that similar gene sets, regardless of semantic annotation, are joined through similarity of their contents. Our fast implementation of bipartite algorithms (24) allows GeneWeaver to create very large hierarchies from over 100 typical inputs and our integration of homology and identifier translation tables enables integration across species and experimental systems. To ensure robustness of the result and reduce the size of the output graph, users may prune resulting hierarchies using bootstrapping procedures and stopping rules based on leaf contents (such as the minimum number of genes or gene sets in each node) or graph depth.

Anchored bicliques of biomolecular associates

The ABBA tool is an analysis module used to find genes that have similar functional associations to the members of a query set of genes. When a list of genes is input into ABBA, the tool generates a list of all gene sets from within the GeneWeaver database that contain a user determined number of the input genes. ABBA then returns a ranked list of similar genes that are enriched among the same Gene Sets as the input genes (see Supplementary Figures S3 and S4 for sample use cases). The user may designate a connectivity threshold for the number of gene sets that contain the predicted genes. This tool, for example, has been used to successfully identify genes that may be
functionally similar to genes associated with autism within the MGI database (25).

**Gene set similarity**

Gene Set similarity is estimated for a single set against the entire database, executed from an individual gene set page, or it can be analyzed using pair-wise similarity tools, Jaccard Similarity (Supplementary Figure S5) and Hypergeometric test. These two tools operate on a user-selected set of gene sets and produce both a matrix of similarity statistics and a matrix of pair-wise Venn diagrams for all Gene Sets in the analysis. Each operates on a reference gene set consisting only of those genes which can be mapped across identifiers from one gene set to the other.

**Boolean gene set functions**

Although our bipartite graph algorithms are both complete and scalable, the ability to handle large query results, construct inputs, interpret and visualize large graphical results is still a challenge for human users. A set of Boolean Gene Set Logic features enable users to reduce large query results. For example, a user may identify a set of overlapping quantitative trait loci, each for a related biological phenotype. A common positional candidate is most likely, so using the gene set intersection feature, users can reduce the group of QTL candidate gene lists to a single gene set containing only those genes in a high order intersection of the QTLs. Likewise, users may want to distill large sets of genes for comparison of related functions. By taking the union of all genes that are represented among groups of user selected gene sets, comparisons of smaller numbers of gene sets can be performed. For example, to identify the similarities and differences among genes associated with broad classes of psychological disorders, a user could take the union of all genes associated with major classes of conditions: mood disorders, alcoholism, anxiety disorders and chronic stress and then use the Phenome Graph function to examine high-order intersections among them (Supplementary Figure S6).
Table 1. Analysis tools and basic functions available in GeneWeaver

| Analysis tools                        | Gene sets               | Genes               | Input | Output | Description                                                                 | Settings            |
|---------------------------------------|-------------------------|---------------------|-------|--------|------------------------------------------------------------------------------|---------------------|
| Anchored bicliques of biomolecular    | X                       | X                   | X     | Genes  | Highly similar genes, highly connected phenotypes                           | Degree threshold    |
| associates                            |                         |                     |       |        | Find genes that are connected to similar gene sets as your input genes.       |                     |
| Similar gene sets                     | X                       |                     | 1     | Gene set| Ranked list of similar gene sets (top 500)                                   |                     |
| Phenome map                           | X                       | X                   | X     | Gene sets| Graph of hierarchical intersections of gene sets                           |                     |
| Gene set graph                        | X                       | X                   | X     | Gene sets| High degree connectivity of genes to your gene sets                         |                     |
| Jaccard similarity                    | X                       | X                   |       | Gene sets| Matrix of pair-wise Venn diagrams and Jaccard similarity scores             |                     |
| Hypergeometric tests                  | X                       |                     | X     | Gene sets| Pair-wise similarity of gene sets.                                           |                     |
| Jaccard clustering                    | X                       |                     |       | Gene sets| Dendogram and cluster heat map revealing gene set similarity.               |                     |
| Combine                                | X                       |                     |       | Gene sets| Adjacency matrix.                                                            |                     |
| Emphasis genes                        |                         |                     |       | Genes  | Highlight genes of interest on GeneSet graph and Phenome map output         |                     |
| Boolean gene set logic                | X                       |                     |       | Gene sets| Condense a large number of gene sets into a single gene set based on connectivity, e.g. to find the intersection of multiple QTL positional candidates, or the union of all genes annotated to a set of related biological functions. | Connectivity threshold |

All functions are available from the ‘Analyze GeneSets’ page, except for ABBA and ‘Find similar gene sets,’ which are accessed from the ‘Search for Genes’ menu and ‘GeneSet Pages,’ respectively. Emphasis genes may be accessed from either ‘Analyze GeneSets’ or any ‘Gene Set Page’.

Gene set similarity clustering

GeneWeaver employs multivariate clustering of the Jaccard gene set similarity matrix to rapidly collapse large gene sets and arrange them into hierarchical clusters. A number of common clustering algorithms are included (Supplementary Figure S7). This tool enables users to rapidly identify groups of similar and dissimilar inputs that can be used to perform important quality control for redundant inputs among a set of user-submitted data.

DATA CURATION

It is critical for any bioinformatics resource to contain traceable data, ample documentation and a mechanism for reproducible results. The GeneWeaver database consists of curated data from a variety of sources (Supplementary Figures S9–S14, Supplementary Tables S4 and S5), and all data shared publicly on the site must pass curatorial review. As a final check on the accuracy of data in the database, users are given sufficient information to be able to retrieve or regenerate the underlying Gene Set and may submit their corrected version to our curator.

Each gene set in GeneWeaver is stored with descriptive metadata at several levels, including publication information, reference gene identifiers, free text descriptions and structured term annotations for disease, biological process and experimental context. Controlled description of the gene sets are user submitted by tree-based term navigation and selection from the Open Biomedical Ontologies [Mouse Anatomy, Gene Ontology, Mammalian Phenotype Ontology (26), MeSH]; additional terms are suggested using the NCBO Annotator tool. GeneWeaver’s curator reviews these terms for accuracy. These levels provide human and machine interpretable information about each functional genomics experiment to describe the analysis process by which a result was originally generated. Free-text ‘Gene Set names’ succinctly describe the contents of a set. A fuller ‘description’ field provides a detailed description of the data generation and rules for inclusion on a gene set list. An ‘abbreviation’ field
provides a compact recognizable label for graphics. Finally, publication information is obtained to enable users to readily retrieve complete information about the underlying experiment.

GeneWeaver has a systematic curation process to supply validated secondary data for analysis tools. Internal curation is driven by directed research, largely related to alcoholism, addiction and behavioral disorders, but the workflow may be applied to multiple interest groups. From literature, data curation starts with term identification, which is subsequently crossed with genome experimentation key words such as ‘GWAS,’ ‘QTL,’ ‘microarray’ and ‘RNAseq.’ Combined terms are used in MEDLINE searches to identify publications which may contain secondary functional genomics results. Each publication is checked against the existing database and manually scanned for secondary data, including supplemental files. Appropriate data sets are uploaded along with descriptive metadata. Descriptions are written to contain sufficient information to enable users to readily retrieve the original source data and to access documentation for the individual studies.

Figure 3. The gene set graph. The gene set graph reveals the highly connected genes among the nine gene sets selected in Figure 2. This analysis reveals DDX5 as the most highly connected gene, connected to both human and mouse alcohol-related measures. Inset: clicking on a gene node executes a search for gene sets containing the featured gene or its homologs. Clicking on a gene set node reveals the contents and metadata for that gene set.
The NCBO Annotator is then applied to suggest MeSH terms, Disease Ontology, Gene Ontology, Mammalian Phenotype Ontology and other structured vocabulary terms that may be appropriate to the data set. These terms are reviewed by the curator to comprise the final entry.

Gene sets may also be submitted by users for public access via the GeneWeaver database. Each of these is marked provisional until the curator reviews them by first searching for similar sets from the same publication, and searching for similar sets by contents to identify duplicate entries. The curator then determines whether

Figure 4. The phenome graph. The phenome graph drawn from nine inputs selected in Figure 2. The phenome graph is a directed acyclic graph of the intersections of gene sets. Each node represents gene sets and the genes they share. Higher order intersections are represented in the root nodes at the top, and individual gene sets in the leaves at the bottom. Inset: clicking a node opens a page showing the intersections among gene sets in list form. Results from this page can be sent to other tools for annotation, including GAGGLE.
the description is compliant with our standards document (see Supplementary Data), and finally assesses whether the gene set conforms to the literature and/or description provided. Finally, the NCBO Annotator is applied to identify structured annotations to the gene set. Ultimately, the user may wish to reanalyze data from the original source and evaluate the appropriateness of particular source data for their analyses. Sufficient details and links to external sites are provided for this purpose.

A large amount of GeneWeaver data comes from major bioinformatics resources including NCBI, ENSEMBL and various model organism databases, including MGD (9), Rat Genome Database [RGD (27)], HUGO Gene Nomenclature Committee [HGNC (28)], Saccharomyces Genome Database [SGD (29)], FlyBase (30), WormBase (31) and the Zebrafish Model Organism Database [ZFIN (32)]. Some of these data are converted to gene sets, including GO and MP annotations, Comparative Toxicogenomics Database (33) associations and QTL positional candidates from RGD and MGI. These data sources are updated every 6 months. A large accessory data warehouse contains the gene identifier mappings obtained from Homologene and the model organism databases. A table of update dates is found at http://geneweaver.org/index.php?action=help&cmd=updates, and acquisition dates of individual gene sets are indicated at the level of the gene set. When a newer version of a gene set in a user project exists, the older version is flagged as ‘deprecated,’ and the user has the opportunity to update the stored set to the latest version.

Each time the data warehouse is updated, gene identifiers are mapped onto one another using the mapping provided by each species’ model organism database, and assigned a unique ID. Users specify the source species of their upload, and the uploaded list is compared to the gene identifiers in the database for that species to identify the appropriate mapping to gene identifiers. Gene symbols are ambiguous and users are cautioned not to use this method if at all possible. When an ambiguous symbol, i.e. one for which there are multiple gene IDs from the Generic Model Organism Database (GMOD), is uploaded, the user is notified and all of the corresponding gene IDs are assigned to the set. This inherent noise in the user input is filtered out through convergent data analysis, provided that multiple instances of the precise identifier occur, whereas spurious associations are unlikely to recur by chance.

A multitiered system denotes the level of curation applied to each data set, allowing users to limit the scope of their analysis as desired. All data obtained from established community resources such as MP and GO annotations to genes and Kyoto Encyclopedia of Genes and Genomes pathway members are classified as Tier I, Public Research Resource data and is updated on a 6-month cycle. Pathway data (e.g. in KEGG), is converted to gene lists representing the gene products in each pathway. Gene Ontology annotations are converted to gene lists through transitive closure such that the genes annotated to any child node are also annotated to parent nodes successively up each branch of the ontology. A similar process is used to convert phenotypic alleles annotated to the Mammalian Phenotype Ontology. Each allele is assigned to the gene symbol for which it is allelic, and that symbol is associated to gene lists corresponding to each MPO term. Transitive closure is again used to assign gene symbols to parents of any child term in the MPO. Data version and accession date are integrated into the metadata for these records.

Tier II consists of human-curated machine-generated data, such as phenotype to gene correlation data in reference populations and QTL positional candidates or GWAS positional candidates. GeneWeaver’s curation team examines the fidelity of these large data sets to our standards documentation and assigns appropriate metadata to them. Tier III data consists of curated results of individual functional experimental studies, typically obtained from publications. Tier IV, or provisional data, is submitted data, advanced by users for public sharing but still pending review. It is exposed to search and analytic modules, but can be filtered or excluded readily. Curators examine provisional data for its adherence to our standards documentation for metadata quality, redundancy with existing data or unusual threshold artifacts, before moving it to Tier III, human curated. Tier V consists of data explicitly for private, personal or group use and is not shared with the larger community.

Users and groups

In order to tightly couple analysis to the collation and curation of user-submitted data, ODE allows guests to create individual accounts, user-defined groups and associated projects. Automated graded-access is built into analysis tools to control data sharing across users and groups. Any gene set may be maintained as private and excluded from our curation Tiers, although it is strongly suggested that users make every effort to meet requirements outlined in the standards documentation (see Supplementary Data) to facilitate interpretation of analysis results by users and group members. Projects provide an effective means to sort and store related data sets and query results over time. Version control of data in user accounts is also provided. Data curation is a dynamic process. It is possible that user selected genes or gene sets may change status over time with some data being deprecated. Legacy data are not removed from the repository, but, data sets containing deprecated genes or gene-function associations are clearly noted and may or may not be updated at the user’s discretion to maintain continuity and reproducibility of results through time.

Community contribution to the GeneWeaver repository

Users are encouraged to augment the repository by uploading data specific to their research question. Data upload is in the form of a simple two-column tab delimited text file consisting of one header row. The left column is a list of the gene symbols or identifiers, and the second column is a score that can later be used to threshold genes of interest. These scores may be the results of statistical analyses such as $P$-values, $q$-values, correlation metrics and effect sizes or they may be binary values
RESOURCE DISSEMINATION

The GeneWeaver analysis system is made available on the GitHub community code repository under the GPL license. The system includes HTML, AJAX, SVG pages, a PostgreSQL database, Python and C modules for analysis. Outputs include PDF graphics export, along with XML, GAGGLE micro-formats and tab delimited text for data interoperability. A standalone GeneWeaver package is currently unavailable because the data, tools and schema framework are highly system-dependent and intended as an Internet-based analysis framework. However, we are developing a set of APIs to allow users to operate on and contribute to our secondary data environment. Tools are documented in a variety of formats including a movie demo, tutorial exercise, quick start PDF, tool tips and interactive help. A web-based form and email links provide individualized user support.

FUTURE DIRECTIONS

The underlying framework of GeneWeaver is extensible on several fronts. Tool development is centered on providing users better approaches to managing the scope of output that the site provides. This will be accomplished through further development of simple text output that can be analyzed using other tools, and in the development of additional software for local processing including FGPU assisted processing of larger visual output. GeneWeaver may be augmented by our user community through the development of additional modules that operate on the gene-phenotype graph which can be piloted on machine readable output generated from the GeneWeaver site. New species are included through the incorporation of additional annotated genomes and homology tables, and new gene identifiers, particularly for microarray platforms, are incorporated through the use of the GEO IDs (34). GeneWeaver has a relational back-end designed to scale vastly beyond the current data and to accommodate biomolecular entities other than genes. We are working to expand the underlying data structure to accept discrete relationships between biological objects that are outside of the concept of gene. For example, ongoing development will allow users to submit RNA, SNP or methylation data in conjunction with assigned biological functions without having to generalize these data to gene symbols. Our intention is to generalize the secondary database to a solution that will incorporate most relevant current and future technologies for genome-wide function characterization.

Data sets from publications representing genome-wide studies in functional biology are continuously being added and curated. This includes the active and passive solicitation of user-submitted data. The goal is to continue to identify quality public research resource data for their potential incorporation into ODE. Rich documentation and active training of users will facilitate the best use of the analysis tools by biological researchers.

CONCLUSIONS

GeneWeaver provides a real-time, interactive and extensible environment for user-initiated integrative analysis of functional genomics data. It contains a significant repository of curated multispecies gene sets assigned to biological function and associated with community metadata. Access to GeneWeaver allows users to efficiently ask questions about the underlying functional genomics of complex biological systems by spanning secondary data covering multiple species and experimental protocols. By leveraging our large curated repository of discrete gene set relationships users can easily integrate their own functional genomics studies and apply robust analysis tools to discover new knowledge without the burden of having to collect, reanalyze and collate appropriate comparative studies.

When systematic data curation is applied in conjunction with GeneWeaver’s analysis tools, a consistent framework can be imposed on the perceived disorder of diverse secondary data in any area of functional genomics analysis, ensuring scalability, data quality and a common vocabulary. By rendering these data computable, we retrieve valuable biological information from functional genomic studies that would otherwise be relegated to the challenges of diverse, poorly standardized, yet publicly available data.

SUPPLEMENTARY DATA

Supplementary Data are available at NAR Online: Supplementary Tables 1–5, Supplementary Figures 1–14.

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REFERENCES

1. Guo, A.Y., Webb, B.T., Miles, M.F., Zimmerman, M.P., Kendler, K.S. and Zhao, Z. (2009) ERGR: an ethanol-related gene resource. Nucleic Acids Res., 37, D840–D845.

2. Le-Niculescu, H., Patel, S.D. and Niculescu, A.B. (2010) Convergent integration of animal model and human studies of bipolar disorder (manic-depressive illness). Curr. Opin. Pharmacol., 10, 594–600.

3. Li, C.Y., Mao, X. and Wei, L. (2008) Genes and (common) pathways underlying drug addiction. PLoS Comput. Biol., 4, e2.

4. Mulligan, M.K., Ponomarev, I., Hitzemann, R.J., Belknap, J.K., Tabakoff, B., Harris, R.A., Crabbe, J.C., Blednov, Y.A., Grahame, N.J., Phillips, T.J. et al. (2006) Toward understanding the genetics of alcohol drinking through transcriptome meta-analysis. Nat. Protoc. Sci. USA, 103, 6368–6373.

5. Nissenbaum, J., Devor, M., Seltzer, Z., Gebauer, M., Michaelis, M., Tal, M., Dorfman, R., Abibul-Yarkoni, M., Lu, Y., Elahipanah, T. et al. (2010) Susceptibility to chronic pain following nerve injury is genetically affected by CACNG2. Genome Res., 20, 1180–1190.

6. Austin, C.P., Battey, J.F., Bradley, A., Bucan, M., Capecchi, M., Collins, F.S., Dove, W.F., Duyk, G., Dymecki, S., Eppig, J.T. et al. (2004) The knockout mouse project. Proc. Natl Acad. Sci. USA, 101, 176.

7. Li, Z., Vizeacoumar, F.J., Bahr, S., Li, J., Warringer, J., Vizeacoumar, F.S., Min, R., Vandersluis, B., Bellay, J., Devit, M. et al. (2011) Systematic exploration of essential yeast gene function with temperature-sensitive mutants. Nat. Biotechnol., 29, 361–367.

8. Maimnezh, S., Davierwala, A.P., Haynes, J., Moffat, J., Peng, W.T., Zhang, W., Yang, X., Postooleajo, Chua, G., Lopez, A. et al. (2004) Exploration of essential gene functions via titratable promoter alleles. Cell, 118, 31–44.

9. Blake, J.A., Bult, C.J., Kadin, J.A., Richardson, J.E. and Eppig, J.T. (2011) The Mouse Genome Database (MGD): premier model organism resource for mammalian genomics and genetics. Nucleic Acids Res., 39, D842–D848.

10. Smith, C.L. and Eppig, J.T. (2009) The mammalian phenotype ontology: enabling robust annotation and comparative analysis. Interdiscip. Rev. Syst. Biol. Med., 1, 390–399.

11. The Gene Ontology Consortium. (2010) The Gene Ontology in 2010: extensions and refinements. Nucleic Acids Res., 38, D331–D335.

12. Guan, Y., Ackert-Bicknell, C.L., Kell, B., Troyanskaya, O.G. and Hibbs, M.A. (2010) Functional genomics complements quantitative genetics in identifying disease-gene associations. PLoS Comput. Biol., 6, e1000091.

13. Neely, G.G., Hess, A., Costigan, M., Keene, A.C., Goulas, S., Langeslag, M., Griffin, R.S., Belfer, I., Dai, F., Smith, S.B. et al. (2010) A genome-wide Drosophila screen for heat nociception identifies alpha2delta3 as an evolutionarily conserved pain gene. Cell, 143, 628–638.

14. McGary, K.L., Park, T.J., Woods, J.O., Cha, H.J., Wallingford, J.B. and Marcotte, E.M. (2010) Systematic discovery of nonobvious human disease models through orthologous phenotypes. Proc. Natl Acad. Sci. USA, 107, 6544–6549.

15. Baker, E.J., Jay, J.J., Philip, V.M., Zhang, Y., Li, Z., Kirova, R., Langston, M.A. and Chester, E.J. (2009) Ontological Discovery Environment: a system for integrating gene-phenotype associations. Genomics, 94, 377–387.

16. Chesler, E.J. and Baker, E.J. (2010) The importance of open-source integrative genomics to drug discovery. Curr. Opin. Drug Discov. Dev., 13, 310–316.

17. Gardner, D., Akil, H., Ascoli, G.A., Bowden, D.M., Bug, W., Donohue, D.E., Goldberg, D.H., Grethe, J.S., Gupta, A. et al. (2008) The neuroscience information framework: a data and knowledge environment for neuroscience. Neuroinformatics, 6, 149–160.

18. Wang, J., Williams, R.W. and Manly, K.F. (2003) WebQTL: web-based complex trait analysis. Neuroinformatics, 1, 299–308.

19. Davis, A.P., King, B.L., Mockus, S., Murphy, C.G., Saraceni-Richards, C., Rosenberg, M., Wiegert, T. and Mattingly, C.J. (2011) The Comparative Toxicogenomics Database: update 2011. Nucleic Acids Res., 39, D1067–D1072.

20. Dennis, G. Jr, Sherman, B.T., Hosack, D.A., Yang, J., Gao, W., Lane, H.C. and Lempicki, R.A. (2003) DAVID: Database for Annotation, Visualization, and Integrated Discovery. Genome Biol., 4, F3.

21. Jonquét, C., Shah, N.H., and Musen, M.A. (2009) The Open biomedical annotator. Summit on Transl. Bioinformat. AMIA, San Francisco, pp. 56–60.

22. Osborne, J.D., Flatow J., Holko, M., Lin, S.M., Kibbe, W.A., Zhiu, L.J., Danila, M.I., Fung, O. and Chisholm, R.L. (2009) Annotation of the mammalian genome with disease ontology. BMC Genomics, 10(Suppl. 1), S6.

23. Shannon, P.T., Reiss, D.J., Bonneau, R. and Baliga, N.S. (2006) The Gaggle: an open-source software framework for integrating bioinformatics software and data sources. BMC Bioinformatics, 7, 176.

24. Zhang, Y., Chesler, E.J. and Langston, M.A. (2007) On finding Biclques in Bipartite graphs: a novel algorithm with application to the integration of diverse biological data types. Hawaii International Conference on System Sciences, Waikoloa, H1.

25. Meehan, T.F., Carr, C.J., Jay, J.J., Bult, C.J., Chesler, E.J. and Blake, J.A. (2011) Autism candidate genes via mouse phenomics. J. Biomed. Inform., March 11 (doi:10.1016/j.jbi.2011.03.003; epub ahead of print).

26. Smith, C.L., Goldsmith, C.A. and Eppig, J.T. (2005) The Mammalian Phenotype Ontology as a tool for annotating, analyzing and comparing phenotypic information. Genome Biol., 6, R7.

27. Twigger, S.N., Shimoyama, M., Bromberg, S., Kwitek, A.E. and Jacob, H.J. (2007) The Rat Genome Database, update 2007—Easing the path from disease to data and back again. Nucleic Acids Res., 35, D658–D662.

28. Brufor, E.A., Lush, M.J., Wright, M.W., Sneddon, T.P., Povey, S. and Birney, E. (2008) The HGNC Database in 2008: A resource for the human genome. Nucleic Acids Res., 36, D445–D448.

29. Cherry, J.M., Adler, C., Ball, C., Chervitz, S.A., Dwight, S.S., Hester, E.T., Jia, Y., Juvik, G., Roe, T., Schroeder, M. et al. (1998) SGD: Saccharomyces Genome Database. Nucleic Acids Res., 26, 73–79.

30. Tweedie, S., Ashburner, M., Fallis, K., Leyland, P., McQuilton, P., Marygold, S., Millburn, G., Osmi-Sutherland, D., Schroeder, A., Seal, R. et al. (2009) FlyBase: enhancing Drosophila Gene Ontology annotations. Nucleic Acids Res., 37, D555–D559.

31. Harris, T.W., Antoshechkin, I., Bieri, T., Blasiar, D., Chan, J., Chen, W.J., De La Cruz, N., Davis, P., Duesbury, M., Fang, R. et al. (2010) WormBase: a comprehensive resource for nematode research. Nucleic Acids Res., 38, D463–D467.

32. Sprague, J., Bayraktaroglu, L., Bradford, Y., Conlin, T., Dunn, N., Fashena, D., Frazer, K., Haendel, M., Howe, D.G., Knight, J. et al. (2008) The Zebrafish Information Network: the zebrafish model organism database provides expanded support for genotypes and phenotypes. Nucleic Acids Res., 36, D768–D772.

33. Mattingly, C.J., Colby, G.T., Forrest, J.N. and Boyer, J.L. (2003) The Comparative Toxicogenomics Database (CTD). Environ. Health Perspect., 111, 793–795.

34. Burretti, T., Troup, D.B., Wilhite, S.E., Ledoux, P., Rudnev, D., Evangelista, C., Kim, I.F., Soboleva, A., Tomashovsky, M. and Edgar, R. (2007) NCBI GEO: mining tens of millions of expression profiles—database and tools update. Nucleic Acids Res., 35, D760–D765.