# **Article Title**

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#### 2 ABSTRACT

- 3 Abstract length and content varies depending on article type. Refer to http://
- 4 www.frontiersin.org/about/AuthorGuidelines for abstract requirement and length
- 5 according to article type.

### 1 INTRODUCTION

7 Hello, how are we doing?

# 2 METHODS

### 8 **2.1 Clean up**

- 9 All functional annotation sets were cleaned up the following way (using definitions from the Gene
- 10 Ontology version 2019-07-01):
- 1. Any annotations where the GO accession was marked as obsolete were removed.
- 12 2. Some terms in the GO have 'alternative ids'. When naively removing duplicates, two entries will not
- be recognized as duplicates if they have different accessions pointing to the same GO term. Therefore,
- all GO accessions were changed to their respecitve 'main id' and the dataset was again scanned for
- duplicates.
- Table 1 provides information on the number of annotations that were removed this way from each dataset.
- 17 All further analyses were performed on the cleaned datasets since we assume the user will only be interested
- 18 in still valid and non-redundant functional annotations.

## 19 2.2 Choosing the right evaluation metric

- 20 A plethora of different metric to evaluate the quality of functional annotation predictions is available using
- 21 different approaches and there seems to be no clear standard yet. TODO: THIS IS WEAK Additionally,

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**Table 1.** Number of removed annotations during cleanup.

Genome	Dataset	Obsolete Annotations	Duplicates
Arachis hypogaea	GOMAP	0	0
Glycine max	GOMAP	203	0
Hordeum vulgarum	GOMAP	101	0
Medicago truncatula A17	GOMAP	0	0
Medicago truncatula R108	GOMAP	0	0
Oryza sativa	GOMAP GoldStandard	111 38	2 556
Отуда ѕанға	Gramene61-IEA	10	14
Phaseolus vulgaris	GOMAP	0	0
	GOMAP	285	0
Triticum aestivum	GoldStandard Gramene61-IEA	0 47	10 48
Vigna unguiculata	GOMAP	0	0
	GOMAP	1107	70
7. a aa D722	GoldStandard	1	0
Zea mays B73.v3	Gramene49	94	2
	Phytozome	54	0
	GOMAP	752	83
Zea mays B73.v4	GoldStandard	55	174
	Gramene61-IEA	99	157
Zea mays Mo17	GOMAP	726	77
Zea mays PH207	GOMAP	798	76
Zea mays W22	GOMAP	754	82

each of the metrics has a different focus and lalala so choosing a metric for quality evaluation is not trivial. When we first published GOMAP (Wimalanathan et al., 2018), we used a modified version of the hierarchical evaluation metrics originally introduced in (Verspoor et al., 2006) because they were simple, clear, and part of an earlier attempt at unifying and standardizing GO annotation comparisons (Defoin-Platel et al., 2011). In the meantime, Plyusnin et al. (2018) have published an approach for evaluating different metrics showing substantial differences within the robustness of different approaches. Todo Describe Their Approach We have applied their method on the Gold Standards available to us to determine which evaluation metric is the most appropriate in our case. The results of this analysis can be seen in Todo.

We then evaluated our predictions and the other annotation sets using the best performing metrics as well as the one we previously used. TODO

### 3 RESULTS

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33 ... a quantitative comparison of the datasets in Table.

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**Table 2.** Quantitative metrics of the cleaned functional annotation sets. CC, BF, MP, and A refer to the aspects of the GO: Cellular Component, Biological Function, Molecular Process, and Any/All.

			Annotations <sup>a</sup> Annotated Genes [%] <sup>b</sup>		Median Ann. per G.c									
Genome	Genes	Dataset	CC	BF	MP	A	CC	BF	MP	A	CC	BF	MP	A
Arachis hypogaea		GOMAP	153433	132944	493799	780176	57667	56855	67123	67124	2	2	6	10
Glycine max		GOMAP	129215	113827	417555	660597	46020	47034	52871	52872	2	2	6	11
Hordeum vulgarum		GOMAP	88130	80282	272823	441235	35237	36470	39733	39734	2	2	5	10
Medicago truncatula A17		GOMAP	107362	99719	364065	571146	42325	43736	50443	50444	2	2	6	10
Medicago truncatula R108		GOMAP	112343	108031	382322	602696	40332	50220	55706	55706	1	2	5	9
		GOMAP	72780	64685	248700	386165	28619	29853	35824	35825	2	2	6	9
Oryza sativa Oryza sativa	100	GoldStandard	7730	11060	19378	38176	5725	7383	9031	11387	1	1	1	3
Oryza sanva		GoldStandard	7730	11060	19378	38176	5725	7383	9031	11387	1	1	1	3
		Gramene61-IEA	14633	32787	39105	86529	10771	15537	16705	21446	1	1	1	3
Phaseolus vulgaris		GOMAP	72005	64583	229630	366218	25934	25539	27432	27433	2	2	6	11
Ü		GOMAP	267741	218623	785960	1272324	95604	98187	107890	107891	2	2	6	10
Triticum aestivum	100	GoldStandard	1590	923	4807	7323	965	620	1662	1866	1	0	2	3
	100	Gramene61-IEA	38975	109319	109518	257832	28849	59596	52564	75785	0	1	1	2
Vigna unguiculata		GOMAP	75867	68313	243278	387458	27173	27124	29772	29773	2	2	6	11
0 0		GOMAP	135211	87420	291251	513882	34866	38073	39468	39469	3	2	6	11
7		GoldStandard	1565	65	299	1929	1548	60	151	1634	1	0	0	1
Zea mays B73.v3		Gramene49	20072	31056	30089	81217	11834	17991	15800	21926	1	1	1	3
		Phytozome	4787	19044	13100	36931	4524	13728	11365	16132	0	1	1	2
		GŎMAP	88827	82251	278719	449797	36717	37337	39323	39324	2	2	6	10
Zea mays B73.v4		GoldStandard	11510	15019	25737	52428	8349	10067	12120	14971	1	1	1	3
-		Gramene61-IEA	20265	47657	58110	126525	14774	22064	23965	29152	1	1	2	3
Zea mays Mo17		GOMAP	87567	79214	277787	444568	33618	35105	38619	38620	2	2	6	10
Zea mays PH207		GOMAP	90617	85500	288677	464794	35170	36762	40556	40557	2	2	6	10
Zea mays W22		GOMAP	95390	85039	289780	470209	36987	37685	40689	40690	2	2	6	10

<sup>&</sup>lt;sup>a</sup> How many annotations in the CC, BF, and MP aspect does this dataset contain? A = How many in total? A = CC + BF + MP

**Table 3.** Quality evaluation of the used GO annotation sets.

Genome	Dataset	SimGIC2 score
Oryza sativa	GOMAP Gramene61-IEA	0.253680 0.330437
Triticum aestivum	GOMAP Gramene61-IEA	0.218996 0.175564
Zea mays B73.v3	GOMAP Gramene49 Phytozome	0.052182 0.091475 0.028721
Zea mays B73.v4	GOMAP Gramene61-IEA	0.257543 0.328777

#### 3.1 **Quality Evaluation**

TODO If it turns out that our predictions are good with hF but bad with more approriate metrics, explanation would be that score thresholds for the prediction tools used in the GOMAP pipeline have been chosen to maximize this hF value. It now seems reasonable to re-adjust these thresholds to maximize a different metric which will likely result in a drop in hF score but increase in other metrics. Again emphasizes the importance of choosing the right evaluation metric.

#### REFERENCES

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- Defoin-Platel, M., Hindle, M. M., Lysenko, A., Powers, S. J., Habash, D. Z., Rawlings, C. J., et al.
- (2011). AIGO: Towards a unified framework for the Analysis and the Inter-comparison of GO functional 41
- 42 annotations. BMC Bioinformatics doi:10.1186/1471-2105-12-431

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b How many genes in the genome have at least one GO term from the CC, BF, MP aspect annotated to them? A = How many at least one from any aspect?

 $<sup>(</sup>A = CC \cup BF \cup MP)$ <sup>c</sup> Take a typical gene that is present in the annotation set. How many annotations does it have in each aspect? A = How many in total? Ask your favorite statistician why  $A \neq CC + BF + MP$ 

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Plyusnin, I., Holm, L., and Töoröonen, P. (2018). Novel Comparison of Evaluation Metrics for Gene

- Ontology Classifiers Reveals Drastic Performance Differences. *bioRxiv* , 427096doi:10.1101/427096
- Verspoor, K., Cohn, J., Mniszewski, S., and Joslyn, C. (2006). A categorization approach to automated
  ontological function annotation. *Protein Science* doi:10.1110/ps.062184006
- 47 Wimalanathan, K., Friedberg, I., Andorf, C. M., and Lawrence-Dill, C. J. (2018). Maize GO Annotation-
- 48 Methods, Evaluation, and Review (maize-GAMER). Plant Direct 2, e00052. doi:10.1002/pld3.
- 49 52