The language–number interface in the brain: a complex parametric study of quantifiers and quantities

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INTRODUCTION

No species but ours can pride itself in the possession of both mathematical and linguistic abilities. Some primates possess numerical abilities, but only we humans can talk about what we count. Are these two abilities governed by the same principles and supported by the same neural structures? Philosophers from Plato to Frege have pondered the relation between mathematical and natural languages, as these seem to share properties, having alphabets and combinatorial rules, allowing for recursion, as well as ambiguous expressions. Chomsky (1988), for instance, has speculated that "the number faculty developed as a by-product of the language faculty" (p. 169; see Changeux and Connes, 1995; for a recent similar position). In contrast to these speculations, experimental evidence, mostly from double-dissociations observed in brain-damaged patients, has suggested that neurologically, the two capacities are distinct (Dehaene and Cohen, 1997; Cohen and Dehaene, 2000; Brannon, 2005; Gelman and Butterworth, 2005). We report an fMRI experiment that addressed the language/math question from a new angle, and helped uncover the anatomical loci of linguistic and numerical operations. Our behavioral and imaging data seem to provide a fresh perspective on this perennial debate. Guided by current models of linguistic and mathematical capacity, we separated the neural underpinnings of complex language processes from those of numerosity-related ones. Within the latter, moreover, we were successful in identifying two distinct neural systems (one for quantity estimation, the other, for comparison). This three-way distinction between neurocognitive components emerged as subjects were evaluating linguistic statements about quantity against visual scenarios. Critically, it emanated from analyses of data from a single cross-modal parametric fMRI experiment that concomitantly probed numerical cognition and language.

The current view on numerical cognition and its brain basis is rather refined (Dehaene and Cohen, 1997; Butterworth, 1999; Cantlon and Brannon, 2007), drawing on complex computational models, and on empirical results from humans and monkeys. These models parse basic arithmetical operations into distinct steps – notably estimation (the encoding of numerical size), and comparison (the calculation of numerical distance), driven by Weber’s Law, that states that the ability to perceive a minimal difference between the quantities of two stimuli is invariant across the range of possible stimulus sizes. The consequence of this law in the present context is, roughly, that the organism’s ability to compare two quantities is a function of the proportion, rather than the distance, between them (Meck and Church, 1983; Dehaene and Changeux, 1993; Verguts and Fias, 2004). Experiments on
numerosity typically present a sequence of images to participants. The first image depicts a number of objects (the reference numerosity \( r \)), and is followed by other images, of which the crucial one contains the *comparandum* numerosity \( c \). In most instances, \( r \) is kept constant across trials, while \( c \) is a systematically varied parameter. Typically, this sequence is not accompanied by linguistic material.

The implementation of tasks that require the estimation of \( r \) and \( c \) and then \( r - c \) comparison, has led to the discovery of neural correlates of these two operations in the macaque brain. There, neurons in lateral intraparietal cortex (LIP) in both hemispheres that respond to a relatively broad range of numerosities were distinguished from anteriorly adjacent (VIP) and frontal neurons that code for numerical distance, namely carry out comparisons (Nieder and Miller, 2003; Piazza et al., 2004; Roitman et al., 2007). Monkeys’ numerical judgment behavior is, moreover, said to be asymmetrical, best described by Gaussian curves on a logarithmically compressed scale, whose variability is fixed across numerosities (Dehaene et al., 2003).

In humans, bilateral parietal and frontal regions modulated by numerical distance have also been identified, and shown to have similar properties: in a seminal study by Piazza et al. (2004), participants viewed a sequence of images that contained a fixed number of objects \( r \), thereby adapting to \( r \); the critical part then presented an image with the same objects, but whose numerosity \( c \) was parameterized. Using an fMRI adaptation paradigm (i.e., using the habituation of the fMRI signal as an indicator of repeated involvement of a region in the same kind of cognitive process), they found regions in right angular gyrus, right intraparietal sulcus (IPS), and right superior parietal lobe (SPL) that follow the relation between \( r \) and \( c \), indicating involvement in numerical comparison: These regions present an adaptation response pattern to a numerical distance parameter (manipulated in a comparison task) that resembles monkeys’ behavioral and neural patterns.

For language to be directly related to numerical cognition, aspects of language processing that pertain to quantity would be expected to have a role in numerical cognition. Past studies have either investigated linguistic and arithmetical abilities separately (Dehaene et al., 2003), or focused on the relation between number words and their denotation (Dehaene and Cohen, 1997; Cohen and Dehaene, 2001; Varley et al., 2005). Investigations into the relation between the processing of sentences about numerical relations and corresponding number-containing scenarios are few and far between (McMillan et al., 2005; Troiani et al., 2009), despite the fact that we as speakers do just this naturally and frequently, as we evaluate statements we hear (or speak) against their real-life contexts. As a result, how exactly the neural substrate for numerical calculation relates to language is not entirely clear. Our project therefore involved a rather natural task, one that required participants to relate linguistic expressions to scenarios whose analysis necessitated numerical Estimation and Comparison: the scenarios contained quantities of objects, whereas the linguistic materials were sentences with quantity denoting expressions, that is, natural language quantifiers like *many, few, most, or less-than-half*. Successful performance required proper parsing of the visual scenarios, as well as an analysis of sentence meaning. Our contrasts, then, involved quantifiers that differed in meaning from one another.

We probed circuits for linguistic and mathematical processing in healthy participants in an fMRI experiment whose design featured two parameters, one that served as a proxy for numerical estimation, and the other, for comparison. It also featured a linguistic contrast that was orthogonal to numerosity (Polarity, the difference between positive quantifiers like *more-than-half* and *less-than-half*), which we used to detect certain aspects of the analysis of linguistic meaning which are orthogonal to numerosity. This design enabled us to measure the BOLD response related to each of these operations separately on the same data set.

A parametric proportion paradigm (PPP henceforth) was introduced: participants were asked to verify auditory sentences about proportion against visual scenarios that contained two quantities of objects, and do so under time pressure. The relevant meaning representations were created through the inclusion of a proportional (or degree) quantifier in each sentence, which was either positive, e.g., *many of the circles are blue*, or negative, e.g., *few of the circles are blue*. Each sentence was immediately followed by a visual image with an array of 50 quasi-randomly positioned circles. While the number of circles was fixed at 50 (a numerosity chosen to avoid subitizing), their radii were varied, and they were divided into two contiguous clusters of blue and yellow circles. The truth or falsity of each sentence depended on the meaning of the sentence and its match to the blue/yellow proportion (Figure 1A). To verify the sentence above, for example, participants had to first estimate \( c \), the number of circles in the target color mentioned in the sentence (here blue), and \( r \), the number of circles in the other color (here yellow). Next, they had to compare \( c \) to \( r \). They then had to judge the truth-value of the sentence against the representation of the scenario (semantic truth-value judgment, SVJT). A sentence was deemed true just in case if \( c > r \). Alternatively, as \( r + c = 50 \) in every case, a comparison of \( c \) to \( 25 = 50/2 \) would also suffice, as they sentence is true when \( c > 25 \). Either way, the successful completion of the task in a limited time window required both estimation and comparison. BOLD response as well as reaction time (RT) were measured.

The PPP design systematically varied the proportion between the colors across scenarios. Participants thus judged each sentence against eight different scenarios (created in Mathematica™), each with a different blue/yellow proportion (Figure 1B). Properties of the Images and How They Were Created in Supplementary Material. The manipulation of \( c \) (= Target Color, TarCol) modulated RT: a change in \( c \) it coupled to a change in \( r \) (because \( c + r = 50 \)), and affects the numerical distance between \( c \) and \( r \); this, in turn, affects the difficulty of the comparison component of the task—the closer \( c \) is to \( r \) (or to 25), the more difficult the task becomes, and RT is elevated relative to trials in which \( c \) is distant from \( r \) (or from 25). The PPP used sentences conveyed weak statements that are true in multiple scenarios, and so truth-value was kept constant across several values of \( c \). Moreover, they were contrasted along the linguistic dimension of Polarity, as some contained negative quantifiers (*few*) whereas others were positive (*many*).

This design allows correlating the BOLD response separately with two parameters, and one contrast:
FIGURE 1 | Experimental design. (A) Overall design: Auditory sentence stimuli were of the form "Quantifier of the circles are color," where Quantifier was one of six proportional quantifiers, and color was blue or yellow. Each of the 12 resulting sentences was followed by a visual array of a fixed numerosity $T = 50$ blue and yellow circles, in which blue/yellow ratio was varied parametrically: the comparandum numerosity $c$ of circles in the $TarCol$ was a parameter, taking the values 5, 10, ... , 45. As $T$ was fixed, the numerosity of circles in the other color $r$ varied accordingly (45, 40, ... , 5 out of 50). Three different tokens were generated for each proportion, and combined with the sentences. (B) Time course of a trial: The auditory part was 2.8 s long (sentence duration was $\leq$ 2.8 s). Subsequently, an image was presented (1.2 s). Participants were asked to make a truth-value judgment by a button press. RTs were time-locked to the visual phase of the trial, where overall trial duration was 6 s. Speeded responses were mostly $< 2$ s, which enabled a short rest period before the next trial began.

- The Estimation parameter $c$ (i.e., Number of elements of Target Color): an increase in $c$ forces the estimation of a larger numerosity. Changing $c$ should incrementally tax the process of coding for numerical size, and activate brain regions that support it. Thus $c$ can be used as a proxy to Estimation, and to identify regions involved in numerical estimation.

- The Comparison parameter RT: The closer $c$ gets to $r$, the more difficult the comparison becomes, affecting RT (Figure 2). RT can therefore be used as an index of numerical comparison. Individual trial RTs were therefore used to identify regions involved in the comparison between $c$ and $r$ (or $c$ and 25).

- The Semantic Polarity contrast: linguistic stimuli were divided along a clear non-numerical axis that related to their meaning – Polarity, characterized by whether or not the meaning representation of the quantifier contained a negation (cf. The Meaning of the Proportional Quantifiers Used in Supplementary Material for details). Half of the quantifiers were positive (many, most, more-than-half), and the other half negative (few, fewest, less-than-half). Contrasting these two types revealed brain regions involved in the linguistic analysis of quantifier meaning during the PPP task. Crucially, this contrast is orthogonal to both Estimation and Comparison, ensuring that numerosity circuits are excluded.

The experimental goal, then, was to obtain a three-way dissociation within the same fMRI data set. This goal was accomplished...
by identifying voxels responding to the one or the other of the uncorrelated parameters (c the monotonic estimation parameter, and RT, the non-monotonic comparison parameter), or to the semantic contrast.

The present design differs from previous numerosity experiments, and experiments that used linguistic stimuli to study numerical cognition (Cohen and Dehaene, 2001; McMillan et al., 2005; Troiani et al., 2009) by the following major aspects: first, we used two (c, RT) instead of a single parameter; second, a variable comparandum c is typically compared to a fixed reference numerosity r, and so r + c is not constant. As we used scenarios that were suited to express proportion, r and c were related, such that r + c = 50. Our paradigm and the standard one, however, are inter-translatable, and the relationship between them is formally well-defined (see The Mapping Between the Current PPP and Standard Numerosity Experiments in Supplementary Material for a mathematical translation between the two paradigms).

Finally, the language materials included six different proportion or degree quantifiers – words like many, most, and few – that (1) denote relations between cardinalities of sets (as opposed to specific numbers), (2) that presuppose both estimation and comparison, and (3) that are weak, in that their truth-value is preserved across multiple scenarios. This last property makes them suitable to parametric designs, because it allows a sentence that contains such a quantifier to keep its truth-value constant across different numerosities (Table 1). We used several quantifiers, distinguished along the Polarity dimension (positive vs. negative), which enabled us to correlate the BOLD signal with a contrast that is purely linguistic, and does not bear on numerosity, even if both sides of this contrast come into contact with numerosity-related scenarios (see Materials and Methods, and The Meaning of the Proportional Quantifiers Used in Supplementary Material for further discussion).

Table 1 | German quantifiers (with English translations) used in the study.

| Quantifier                  | Polarity  |
|-----------------------------|-----------|
| Viele                       | Positive  |
| Wenige                      | Negative  |
| Die meisten                 | Positive  |
| Die wenigsten               | Negative  |
| Mehr als die Hälfte        | Positive  |
| Weniger als die Hälfte      | Negative  |

The trial schema (Figure 1) involved the presentation of the sound file containing one of the stimulus sentences. Sound files had a duration of below 2.8 s. At a trial time of 2.8 s after the onset of the sound file, the visual scenario containing blue and yellow circles was presented. The subjects’ responses were time-locked to the onset of the visual scenario. Each trial had an overall duration of 6 s, after which time the next trial started.

fMRI DATA ACQUISITION

The fMRI experiment was carried out on a 3-T Trio scanner (Siemens, Erlangen, Germany). A standard birdcage head coil was used with foam paddings to reduce head motion. The functional data were recorded from 40 axial slices using a gradient-echo EPI sequence with echo time (TE) = 30 ms, flip angle = 90°, and repetition time (TR) = 3 s. The field of view (FOV) was 256 mm, with an in-plane resolution of 3 mm × 3 mm. The slice thickness was 3 mm with an inter-slice gap of 1 mm. A time series containing a total of 684 images was recorded, amounting to a total functional scanning time of 34 min.

fMRI DATA ANALYSIS

Pre-processing

Data analysis was performed using SPM5 (Wellcome Department of Cognitive Neurology, UK) running on MATLAB 7 (The Mathworks Inc., Natick, MA, USA). Pre-processing involved the standard procedures of realignment to the mean image of the EPI time series, normalization of functional data to the MNI template using the unified segmentation procedure provided in SPM5, spatial smoothing with a Gaussian kernel of 8 mm FWHM, and highpass filtering at 1/128 Hz in order to correct for slow drifts in the BOLD signal.

Statistical analyses

Each of the different processing steps involved in quantification, i.e., Composition, Estimation, and Comparison, was addressed by its own event-related analysis (note that the differential analysis of the auditory phase vs. the visual phase in a trial is enabled by
the temporal spacing of the onset times of more than 1 s: Boyn-
ton et al., 1996; Dale and Buckner, 1997). Moreover, subsequent
analysis of Semantic Structure was performed. Data from individ-
ual first level analyses were then entered into random-effects group
analyses at the second level using the flexible factorial option for
repeated-measures designs in SPM5.

1. In a first analysis for Composition, the set of brain regions was
assessed which was activated while listening to the auditory sen-
tence that contained the quantifying expression and building up
a semantic representation of the according scenario (data type: beta
estimates of BOLD amplitude during the auditory presentation
phase).

2. The next, parametric, analysis was run for Estimation, i.e., cod-
ing the circles in the TarCol for numerical size (data type: beta
estimates of the (monotonically increasing) parametric mod-
ulator of the amount of circles of the TarCol during the visual
presentation phase).

3. The third, again parametric, analysis was done for Comparison,
i.e., coding for numerical distance between |c| and its comple-
ment color (data type: beta estimates of the (non-monotonic)
RT regressor during the visual presentation phase) and com-
paring it to the mental representation created during Composi-
tion. To this end, we identified voxels which (1) parametrically
responded to the RT regressor and (2) were contained in the
Composition phase during which the mental representation
was first generated. The resulting Comparison effect was thus
calculated by a conjunction analysis of second level $T$-statistic
maps obtained from Composition and the RT parameter. Since
the effect for Composition was obtained in the auditory domain
and that for the RT regressor in the visual domain, the conjunc-
tion analysis reveals such areas that are involved in a-modal
processing of semantic representations not solely related to the
auditory sentence or the visual scenario. This type of analysis
is valid despite the different nature of the raw data, since it is
calculated on the $T$-statistic maps which are all at the same ($T$
scale.

Composition. Composition (i.e., creation of a semantic repre-
sentation) refers to the first, auditory phase of the trial when subjects
listened to a sentence containing a quantifier expression. There
were six conditions, i.e., one for each quantifier.

1. At the first level, an event-related general linear model (GLM)
analysis was performed. The duration for each condition was
set to 2.8 s, beginning with sentence-onset and covering the
entire auditory phase. Each condition was convolved with the
canonical hemodynamic response function (HRF) and its first
temporal derivative. For subsequent ANOVA at the second
level, the beta weights for the six Composition conditions were
contrasted against the implicit resting baseline by calculating
contrasts of the type ‘$T$’.

2. At the second (group) level, the random-effects repeated-
measures $1 \times 6$ ANOVA was calculated in order to obtain an $T$-
statistic map of the main effect for Composition as contrast of
$T$-statistic map of the main effect for Composition as contrast of
type ‘$T$’, reported at an uncorrected threshold
of $P < 0.001$. Since activation during listening was not con-
trasted against some high-level baseline, but only against rest,
this analysis is the most permissive, including all regions poten-
tially (but not necessarily) involved in building up a semantic
expectation of the upcoming visual array.

Estimation. Estimation refers to the visual phase in a trial when
the display containing yellow and blue circles was presented after
the subject listened to the sentence in the auditory Composition
phase. In particular, the Estimation effect is operationalized as lin-
ear increase in BOLD signal with linear increase of the number of
circles of the TarCol.

1. At the first level, the parametric increase of the BOLD signal
with increasing number of circles of the respective color-of-
mention was assessed separately for each quantifier on a trial-
by-trial basis. The event-related GLM analysis for individual
data sets involved 12 ($2 \times 6$) orthogonal conditions, one for
each color-of-mention (2) and quantifier (6). For each con-
dition, a stick function (i.e., duration $= 0$, onset time $= \text{trial
onset}$) was convolved with a canonical HRF and its first tem-
poral derivative. Stick functions with duration $= 0$ were chosen
in order to analyze the initial matching of the visual sce-
nario with the mental representation generated in the auditory
Composition phase before, and to address this process in the
GLM independently of the actual duration of this matching
process (this latter aspect relates to processing difficulty and is
dressed with the analysis described next).

2. In order to model the parametric BOLD increase as a function
of the number of circles (i.e., the data relevant for subsequent
analysis), the percentage of circles in the TarCol was entered
as monotonic, continuous regressor of interest into the model
on a trial-by-trial basis separately for each condition. For sub-
sequent ANOVA at the second level, we were interested in the
parametric BOLD increase for each quantifier independent of
the actual color (blue, yellow). Therefore, the beta weights for
the two regressors (blue, yellow) for the same quantifier were
collapsed pair-wise over colors, resulting in one average beta
estimate per quantifier and subject. This was achieved using
contrast vectors of the type ‘$1 \ 1 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0$
’ at the first level.

3. At the second level, these individual contrast images were
submitted to a random-effects analysis, realized as a repeated-
measures $1 \times 6$ ANOVA. In order to assess which brain regions
uniformly responded with increasing activation to increasing
 numerosity, the $T$-statistic map for main contrast for Esti-
mation was calculated as contrast of type ‘$1 \ 1 \ 1 \ 1 \ 1 \ 1$
’, reported at a threshold of $P < 0.001$ (uncorrected).

Comparison. Comparison is a process that calculates the numer-
ical distance between the number of TarCol circles obtained dur-
ing Estimation and its complement (the non-$\text{TarCol}$ color). The
matching between the outcome of the Comparison process and
the meaning of the sentence will dictate the decision (e.g., $1/2$ in
the case of ‘more-than-half of the circles are blue’) created during
Composition on the other hand. As outlined in the Results section
above, the processing of this numerical distance becomes the more
difficult, the smaller this distance is. The difficulty, in turn, is represented in the RTs, with high RTs indicating high difficulty of Comparison.

Thus, the Comparison effects mainly reflect regions parametrically responding to increasing RT. Technically, identification of the Comparison was achieved as follows.

1. At the first level, a stick function (duration = 0) for each of the six conditions (one for each quantifier), was convolved with a canonical HRF and its first temporal derivative.
2. Additionally, for each trial, the RT for the STVJ task was entered as regressor of interest for the corresponding condition on a trial-by-trial basis. These regressors represent the parametric, non-monotonic increase of the BOLD response reflecting increase of processing difficulty. For subsequent ANOVA at the second level, the beta weights for each of the six parametric regressors of processing difficulty were contrasted against the implicit resting baseline in order to obtain T-statistic maps by calculating contrasts of the type ‘1 0 0 0 0 0’.
3. At the second level, these individual contrast images were submitted to a random-effects analysis, realized again as a repeated-measures 1 × 6 ANOVA. The T-statistic map for the main effect of the parametric RT regressor was computed as a contrast of type ‘1 1 1 1 1 1’, reported at an uncorrected threshold of P < 0.001.
4. In order to identify those voxels that were commonly involved in semantic Composition and responded to semantic processing difficulty (RT regressor), the T-statistic maps (at P < 0.001 uncorrected) obtained for the main effects for (auditory) Composition and (visual) Comparison part were submitted to a conjunction analysis in order to identify brain regions involved in a-modal semantic processing of quantities. We will henceforth refer to this conjunction analysis as the effect of Comparison.

**Semantic structure.** We used a set of quantifiers for the present study that we classified along the semantic dimension of Polarity, the absence or presence of linguistic negation. Negation is absent in positive quantifiers (many, most, more-than-half), but present in negative ones (few, fewest, less-than-half) quantifiers.

Effects of Semantic Structure – Polarity – can be examined in all three conditions of the parametric RT regressor of interest for the corresponding condition on a trial-by-trial basis. These regressors represent the parametric, non-monotonic increase of the BOLD response reflecting increase of processing difficulty. For subsequent ANOVA at the second level, the beta weights for each of the six parametric regressors of processing difficulty were contrasted against the implicit resting baseline in order to obtain T-statistic maps by calculating contrasts of the type ‘1 0 0 0 0 0’.

In order to identify those voxels that were commonly involved in semantic Composition and responded to semantic processing difficulty (RT regressor), the T-statistic maps (at P < 0.001 uncorrected) obtained for the main effects for (auditory) Composition and (visual) Comparison part were submitted to a conjunction analysis in order to identify brain regions involved in a-modal semantic processing of quantities. We will henceforth refer to this conjunction analysis as the effect of Comparison.

1 In order to fully appreciate the effects of the monotonic regressor for Estimation and the non-monotonic RT regressor for Comparison, these two effects were assessed in two separate GLM analyses for each subject at the first level. Effects due to shared variance of the two regressors, if present, are identified as overlaps of the resulting T-statistic maps at the second level analyses (i.e., technically, as logical conjunctions).

**Localization of effects with cytoarchitectonic probability maps**

For the anatomical localization of the activations we used cytoarchitectonic probability maps, which are based on an observer-independent analysis of the cytoarchitecture in a sample of 10 post-mortem brains (Zilles et al., 2002; Schleicher et al., 2005). They provide information about the location and variability of cortical regions in standard MNI reference space. For the assignment of MNI coordinates to the cytoarchitectonically defined regions we used the SPM Anatomy Toolbox (Eickhoff et al., 2005) available with all published cytoarchitectonic probability maps and references from http://www.fz-juelich.de/inm/inm-l/spm_anatomy_toolbox.

**RESULTS**

**BEHAVIORAL DATA**

The following behavioral results were found: First, participants’ responses (truth-value judgments) presented a step-function along the “yes–no” axis (where “yes” indicates a sentence-scenario match, Figure 2), indicating high overall performance. For all six quantifier conditions, the Estimation parameter c (the number of circles in the Target Color) was highly correlated with mean response type (yes/no; |r| > 0.8, P < 0.001 in every instance, with quantifiers’ polarity determining sign; Table 4). Second, mean RTs, when plotted against c in the images, formed non-monotonic functions, with maxima around the middle (n = 25, Figure 2). The correlation between the c and mean RT was low (not significant for four quantifier conditions, and significant but low (i.e., r < 0.30) for more-than-half and many; Table 4). Third, mean response type (yes/no) and RT were uncorrelated for three quantifier conditions; correlation for the other three conditions was significant, but low in every instance (Table 4).

**BRAIN REGIONS ACTIVATED IN ESTIMATION AS COMPARED TO COMPARISON**

Brain regions responsive to Estimation were identified via the monotonic parameter c, which varied incrementally (c = 5, 10, 40, 45) and was correlated with the BOLD fMRI signal. Effects of this parametric analysis (at Puncorrect < 0.001) were observed in both hemispheres, mainly in parietal and frontal regions including bilateral inferior parietal lobule (IPL), bilateral IPS, left SPL, left inferior frontal and right middle frontal gyrus (IFG; MFG), and the supplementary motor area (SMA; Figure 3, Table 2 for a full list). The cytoarchitectonic probabilistic brain atlas (Amunts et al., 2004) implemented in the SPM Anatomy toolbox (Eickhoff et al., 2005) revealed that parietal effects overlapped with areas hIP1-3 of the IPS (Choi et al., 2006; Schepersjans et al., 2008), and areas PGa, PFm, PF of the IPL (Caspers et al., 2006). Frontal effects extended to Broca’s region (areas 44 and 45; Amunts et al., 1999) and area 47 (Talairach daemon: http://www.nitrc.org/projects/tal-daemon; see also Amunts et al., 2010) in the IFG as well as premotor area 6 in the precentral gyrus (Geyer, 2003).

Regions responsive to Comparison were identified via the RT parameter. It varied non-monotonically and was made a regressor for the BOLD fMRI response. Responses to Comparison
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FIGURE 3 | Estimation and Comparison regions. Parametric effects for the Estimation of the numerosity of circles in target color (blue areas), and for the Comparison as manifest through RT (red areas). All effects are main effects over the six quantifiers. Estimation clusters: brain regions in which the BOLD signal intensity co-varies with the linearly increasing number of circles in target color (F-test). Comparison clusters: brain regions in which the BOLD signal intensity co-varies with RT – a reflection of processing difficulty. Regions commonly tapped by the Estimation and Comparison parameters are shown in purple (red plus blue). IPS, intraparietal sulcus; IFG, inferior frontal gyrus; IFG/orb, pars orbitalis of the IFG.

also involved a bilateral fronto-parietal set of regions. Parietal effects included bilateral IPL, bilateral IPS, and left SPL. Frontal regions comprised bilateral IFG and SMA. In addition, the putamen was activated (Figure 3; Table 3 for a full list). Cytoarchitectonic probability maps revealed the involvement of frontal areas 44, 6, and 4a, and parietal areas hIP1-3 and PF, PFm, and PFcm.

Although both the Estimation effect and the Comparison effect recruited fronto-parietal regions, they overlapped only in part having centroids that were separate in each region (Figure 3). In the inferior parietal lobule, the Comparison effect was located anterior–inferior to the Estimation effect. Both effects involved areas in the IPS (hIP1-3) and inferior parietal lobule (PFm, PF). However, whereas Comparison uniquely involved area PFcm, Estimation recruited inferior parietal area PGa as well as superior parietal areas 7A and 7PC, allowing the cytoarchitectonic distinction between Estimation and Comparison regions within the parietal cortex. In the frontal cortex, a similar pattern of effects was observed with Estimation recruiting more anterior (and slightly inferior) aspects of the IFG, precentral gyrus, and SMA than Comparison. These were mostly distinct from the more posterior Comparison effects at the pre-defined significance threshold. Again, Estimation and Comparison conjointly recruited only some areas (i.e., 44 and 6). Interestingly, however, the Comparison effect also extended more posteriorly, covering parts of motor area 4a, whereas the Estimation effect reached more anteriorly into area 45.

SEMANTIC POLARITY CONTRAST

We compared the fMRI effect of Polarity (negative > positive quantifiers) in all three sets of parameters, i.e., for the non-monotonic Comparison parameter, for the monotonic Estimation parameter, and also for the BOLD signal in the auditory phase of the trial when the quantifier sentence was presented. Consistency of a Polarity effect in all three parameter sets was tested with a conjunction analysis revealing only voxels responding (at P < 0.05; extent k = 100 voxels) likewise in all three parameter sets (i.e., an overall significance level of P < 0.000125; see Supplementary Material for details). The most prominent effect was in the left inferior frontal gyrus (cytoarchitectonic area 45; Figure 4).

DISCUSSION

The separation between semantic Polarity, numerical Estimation, and numerical Comparison was achieved via decisions that participants made on visual scenarios in which numerosity (and subsequently proportion) was parameterized. These scenarios were presented in contrasting linguistic contexts. Though embedded in a tightly controlled design, the PPP was implemented in a rather naturalistic verification task: we verify sentences daily, in communicative acts that require us to answer a yes/no question. These can range from the most mundane topics (Are you wearing a clean shirt?) to complex ones that contain quantifiers (Is every man in the room wearing a tie?). We also engage in verification when we evaluate statements, just like in the PPP task (Most of the paintings in this gallery are by Lucian Freud). The choice of proportional quantifiers, whose calculation requires both Estimation and Comparison, helped in keeping the probing method constant while systematically varying both the numerosities and task difficulty so as to allow for the bi-parametric analysis through the PPP.

There are two novel aspects here: First, these data were obtained for the first time from the same set of subjects in a single study,
Table 2 | Main clusters in which signal increased as a function of parametric increase of stimlation (number of circles of the mentioned color), with their cytoarchitectonic reference.

| Cluster size (voxels) | Local maximum in macroanatomical structure | x  | y  | z  | T_max | Percent of cluster volume in cytoarchitectonic area |
|-----------------------|-------------------------------------------|----|----|----|-------|--------------------------------------------------|
| Cluster 1 (1930)      | Left inferior parietal lobule              | -45| -53| 57 | 5.30  | 32.2 hIP1                                       |
| Cluster 2 (1685)      | Left medial frontal gyrus                 | 0  | 24 | 44 | 4.99  | 16.6 hIP3                                       |
| Cluster 3 (1404)      | Left precentral gyrus (cluster extends into IFG) | -39| 5  | 33 | 6.77  | 43.2 Area 44                                    |
| Cluster 4 (373)       | Right middle frontal gyrus               | 39 | 14 | 38 | 4.46  | 16.8 Area 45                                    |
| Cluster 5 (386)       | Left inferior frontal gyrus              | -48| 20 | 8  | 5.18  | 56.9 Area 44                                    |
| Cluster 6 (304)       | Right supramarginal gyrus (cluster extends into right angular gyrus and IPS) | 48 | -42| 45 | 4.76  | 48.4 PFm                                        |
| Cluster 7 (291)       | Left inferior frontal gyrus              | -53| 41 | -8 | 6.02  | 19.1 PF                                          |
| Cluster 8 (232)       | Right angular gyrus                      | 36 | -60| 45 | 3.94  | 276 PGa                                         |
| Cluster 9 (223)       | Left middle temporal gyrus               | -51| -38| -8 | 4.21  | 13.8 hIP3                                       |
| Cluster 10 (159)      | Right inferior occipital gyrus           | 32 | -95| -11| 4.52  | 69.2 hOC3v                                      |
| Cluster 11 (139)      | Left inferior occipital gyrus            | -26| -98| -11| 4.18  | 48.9 hOC3v                                      |
| Cluster 12 (82)       | Right medial frontal gyrus              | 8  | 48 | 44 | 4.15  | 41.7 Area 18                                    |
| Cluster 13 (74)       | Left middle occipital gyrus              | -32| -69| 32 | 4.17  |                                                 |
| Cluster 14 (58)       |                                           | 44 | 17 | -15| 3.70  |                                                 |
| Cluster 15 (36)       | Right middle orbital gyrus              | 27 | 41 | -14| 3.94  |                                                 |
| Cluster 16 (202)      | Right cerebellum                        | 21 | -86| -35| 3.65  |                                                 |
| Cluster 17 (26)       | Left SMA                                | -15| 3  | 65 | 3.51  | 96.2 Area 6                                      |
| Cluster 18 (21)       | Left superior frontal sulcus            | -21| 42 | 18 |       |                                                 |

References to cytoarchitectonic maps: area 2: Grefkes et al. (2001); areas 4a/4p: Geyer et al. (1998); areas hIP3/7A/7M/7P/7PC: Scheperjans et al. (2008); area 6: Geyer (2003); areas 17/18: Amunts et al. (2000); areas 44/45: Amunts et al. (1999); areas PGa/PGp/PFm/PF/PFcm: Caspers et al. (2008); areas hIP1/hIP2: Choi et al. (2006); areas hOC3v/hOC4v: Rottschy et al. (2007); areas HIPP(SUB)/HIPP(CA): Amunts et al. (2005); area OP2: Eickhoff et al. (2006). Further abbreviations: T_max, T value at local maximum. Cluster overlap with cytoarchitectonic areas is listed if it exceeds 10%.

instead of merging together disparate studies of different scopes and quality. Second, and most importantly, it is the spatial relation of the semantic Polarity regions to the arithmetical parts, exposed through the parametric analyses of the PPP’s Proportion and RT parameters that served as proxies for Estimation and Comparison, respectively. Not only were these two neurocognitive components successfully separated in humans for the first time on the same data set (and in keeping with earlier findings – Piazza et al., 2004; Cohen Kadosh et al., 2005 – and sketches – Dehaene et al., 2003), but also, their spatial relation to activations during semantic analysis was demonstrated: Bilateral parietal regions were shown to be entrusted with numerical Estimation (via the c parameter), and adjacent regions on the opposite side of the IPS that support numerical Comparison (via the RT parameter, uncorrelated with the previous one). A new picture of a processing sequence along the “dorsal stream” seems to be emerging. The original notion of the dorsal stream was that of “where” pathway in visual perception, relevant for localizing objects in space. The concept of a functional processing pathway connecting occipito-temporal brain regions to the frontal cortex via the parietal lobe was supported by an anatomical connectivity study (Caspers et al., 2011). In this study, analogies between white matter tracts in humans and different kinds of monkeys were established. Interestingly, though, the existence of such anatomical connectivity does not imply information about the role of the dorsal stream in cognitive processing. A recent hypothesis (Friederici et al., 2012) relates the maturation of the dorsal stream to linguistic, and in particular syntactic, development from childhood to adulthood. Parallel to that notion, research in the field of numerosity processing (Dehaene et al., 2003) suggested a functional parcelation within the parietal aspect of the dorsal stream: (1) The angular gyrus is regarded as a region for manipulation of verbal number representations. (2) The horizontal aspect of the IPS is thought to house the mental number line. (3) The superior parietal cortex, finally, might relate to attentional processing when orientation along the mental number line is concerned.

Our study relates to the hypothesis by Dehaene et al. (2003) in that it provides an empirical basis for a bi- (or even tri-) partition of the parietal aspect of the dorsal stream, which was obtained.
Table 3 | Conjunction analysis representing brain areas involved in comparison (abstract representation of numerosity).

| Cluster size (voxels) | Local maximum in macroanatomical structure | x  | y  | z  | T_max | Percent of cluster volume in cytoarchitectonic area |
|-----------------------|-------------------------------------------|----|----|----|-------|---------------------------------------------------|
| Cluster 1 (6059)      | Left SMA                                   | −5 | 6  | 53 | 7.72  | 28.0 Area 6                                      |
| Cluster 2 (4124)      | Right putamen                              | 20 | 14 | −5 | 7.36  |                                                   |
| Cluster 3 (3964)      | Left putamen (cluster extends into IFG)    | −18| 17 | −8 | 6.44  | 10.4 Area 44                                     |
| Cluster 4 (2163)      | Right inferior parietal lobule             | 63 | −42| 17 | 7.16  | 24.6 PFm                                          |
| Cluster 5 (831)       | Left inferior parietal lobule              | −48| −47| 50 | 4.29  | 23.0 hIP2                                         |
|                      | (cluster extends into IPS)                 |    |    |    |       |                                                   |
| Cluster 6 (571)       | Right precentral gyrus                     | 41 | 0  | 50 | 4.53  | 29.1 Area 6                                      |
| Cluster 7 (317)       | Right middle frontal gyrus                | 38 | 42 | 27 | 4.14  |                                                   |
| Cluster 8 (184)       | Left supramarginal gyrus (cluster extends into left angular gyrus) | −57| −41| 27 | 4.38  | 77.2 PF                                           |
| Cluster 9 (177)       | Left middle temporal gyrus                | −56| −50| 9  | 4.27  |                                                   |
| Cluster 10 (163)      | Left inferior frontal gyrus               | −41| 42 | 15 | 4.27  |                                                   |
| Cluster 11 (119)      | Left calcareous sulcus                    | −23| −71| 8  | 4.21  | 41.2 Area 17                                      |
| Cluster 12 (110)      | Right inferior frontal gyrus              | 44 | 8  | 24 | 3.88  | 54.5 Area 44                                      |
| Cluster 13 (92)       | Right inferior frontal gyrus              | 56 | 11 | 20 | 4.01  | 98.9 Area 44                                      |
| Cluster 14 (90)       | Left thalamus                              | −8 | −8 | 3  | 3.68  |                                                   |

Parametric increase in brain activation as a function of processing difficulty (reaction time) is intersected with brain activation while listening to sentences containing a quantifier word and building up a semantic representation of the upcoming visual display. For further details see legend of Table 2.

results are consistent with similar ones from the macaque (Nieder and Miller, 2003; Roitman et al., 2007; Piazza and Izard, 2009; Caspers et al., 2011), our data, obtained from humans instead of animals, for the first time establish a clear connection to linguistic analysis of quantity, which builds upon and extends previous findings: Whereas a fronto-parietal network emerges for Estimation and Comparison, linguistic–semantic analysis is most prominent in the left inferior frontal cortex. Taken together, our data characterize processing along the dorsal stream from inferior parietal to inferior frontal regions when numerical processing is linked to linguistic–semantic processing. With respect to the hypothesis by Friederici et al. (2012), the present study may suggest that the dorsal stream is indeed involved in linguistic processing. However, over and above its supposed role for the emergence of syntactic abilities, our data suggest its additional importance for the transformation from visual percepts of numerosities into cognitive and linguistic formats. Interestingly, this account parallels the syntax hypothesis in so far as it stresses the difference between monkey and man with respect to linguistic abilities.

We thus found a potential human/monkey homology for successive steps of numerosity assessment, but moreover, established a direct connection to language. At the same time, we provided evidence for the neural modularity of language and arithmetic, revealed through a task that matches linguistic representations with numerical ones. Such matching has been used before to distinguish numbers from numerosities (Cohen and Dehaene, 2004).

with the same set of participants in the same experimental session. Regions for numerical Estimation and Comparison were found, which anatomically correspond to the angular and SPL regions referred to by Dehaene et al. (2003) and which are indeed separated by the IPS. Moreover, while the present numerosity
Table 4 | Correlations between the Estimation parameter $c$, average response rate in the STVJ task, and corresponding RT for each quantifier.

| Quantifier    | $r$  | $P$     |
|---------------|------|---------|
| Correlation of Estimation parameter $c$ with average response |       |         |
| Few           | −0.825 | <0.001 |
| Fewest        | −0.849 | <0.001 |
| Less-than-half | −0.822 | <0.001 |
| Many          | 0.869  | <0.001 |
| More-than-half | 0.884  | <0.001 |
| Most          | 0.869  | <0.001 |
| Correlation of Estimation parameter $c$ with RT |       |         |
| Few           | 0.403  | 0.531   |
| Fewest        | 0.108  | 0.114   |
| Less-than-half | −0.096 | 0.158   |
| Many          | 0.263  | <0.001  |
| More-than-half | 0.184  | 0.007   |
| Most          | 0.097  | 0.157   |
| Correlation of Average response with RT |       |         |
| Few           | −0.002 | 0.972   |
| Fewest        | −0.043 | 0.531   |
| Less-than-half | 0.135  | 0.047   |
| Many          | 0.178  | 0.009   |
| More-than-half | 0.198  | 0.003   |
| Most          | 0.126  | 0.064   |

2001), and in the context of previous work on quantification (McMillan et al., 2005; Troiani et al., 2009). Our contribution, supporting the modularity of language and arithmetic, stems from the fact that the task we used recruits not only lexical or sentential, but also complex compositional, resources, thereby better approximating the much discussed Chomsky’s “mental organ for language.” Whereas parietal regions were involved only in arithmetical but not semantic analysis, frontal regions, by contrast, appear to be involved in both. This pattern suggests that the left inferior frontal cortex contributes to the evaluation of the numerical representations provided in the parietal regions. This is consistent with the fact that in the macaque brain, number-selective neurons (Estimation) in inferior frontal cortex respond later than number-sensitive (Comparison) parietal neurons (Piazza and Izard, 2009). This temporal delay suggests an order, whereby frontal modules evaluate the output from parietal modules. In humans, this frontal contribution relates to semantic structure and may thus additionally be linked to linguistic processing.

Interestingly, functional division between Estimation and Comparison similar to that in parietal cortex was also found in both frontal lobes. Likewise, the semantic Polarity analysis was prominent in the left frontal lobe in area 45. Evidence for frontal involvement in numerical cognition has been available (Piazza et al., 2007), but the present results are surprising, as they seem to suggest commonalities in the functional organization between the inferior frontal cortex and the peri-IPS regions. McMillan et al. (2005), who studied the relation between natural language quantifiers and numerosity, have argued for a functional distinction, such that frontal regions house working memory, and parietal ones are entrusted with knowledge of numbers. The present study demonstrated the involvement of both frontal and parietal cortex areas likewise in response to Estimation and Comparison, even though these processes may require working memory resources to different extents. Thus, the present data provide no evidence in support of the hypothesis by McMillan et al. (2005). Moreover, their distinction that was made with respect to the underlying anatomy, however, is possibly not fine-grained enough. Given the significant architectonic difference between areas 44 and 45 (involved in the Comparison network) on the one hand, and Brodmann’s area 47 (involved in the Estimation network) on the other hand, it seems difficult to maintain that all frontal clusters uniformly support the same function, be it working memory or other, during numeric cognition. Certainly, there is a wealth of evidence for structural and functional connectivity between anterior IPS areas and ventral–posterior inferior frontal cortex (Uddin et al., 2010; Caspers et al., 2011; Mars et al., 2011). But whether these really involve working memory loops (Vigneau et al., 2011), or rather a house a sequential evaluation algorithm for the evaluation of previously estimated numerosities, as proposed here on the basis of a well-controlled experiment, will be the objectives of future research of the underlying structural and functional connectivity. The significance of this work, then, is in exposing a rich array of regions that provide a rare window into the intriguingly complex neural system that handles the flow of information between neural substrates for linguistic and numerical cognition.

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SUPPLEMENTARY MATERIAL
The Supplementary Material for this article can be found online at http://www.frontiersin.org/Evolutionary_Neuroscience/10.3389/fnevo.2012.00004/abstract

Supplement S1 | Properties of the images and how they were created.

Supplement S2 | The mapping between the current PPP and standard numerosity experiments.

Supplement S3 | The meaning of the proportional quantifiers used.

Supplement S4 | Correlations between the proportion parameter $c$, average response rate in the STVJ task, and corresponding RT for each quantifier.
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