**Article**

**Inflammatory Biomarkers of Traumatic Brain Injury**

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**Abstract:** Traumatic brain injury (TBI) has a complex pathology in which the initial injury releases damage associated proteins that exacerbate the neuroinflammatory response during the chronic secondary injury period. One of the major pathological players in the inflammatory response after TBI is the inflammasome. Increased levels of inflammasome proteins during the acute phase after TBI are associated with worse functional outcomes. Previous studies reveal that the level of inflammasome proteins in biological fluids may be used as promising new biomarkers for the determination of TBI functional outcomes. In this study, we provide further evidence that inflammatory cytokines and inflammasome proteins in serum may be used to determine injury severity and predict pathological outcomes. In this study, we analyzed blood serum from TBI patients and respective controls utilizing Simple Plex inflammasome and V-PLEX inflammatory cytokine assays. We performed statistical analyses to determine which proteins were significantly elevated in TBI individuals. The receiver operating characteristics (ROC) were determined to obtain the area under the curve (AUC) to establish the potential fit as a biomarker. Potential biomarkers were then compared to documented patient Glasgow coma scale scores via a correlation matrix and a multivariate linear regression to determine how respective biomarkers are related to the injury severity and pathological outcome. Inflammasome proteins and inflammatory cytokines were elevated after TBI, and the apoptosis-associated speck like protein containing a caspase recruitment domain (ASC), interleukin (IL)-18, tumor necrosis factor (TNF)-α, IL-4 and IL-6 were the most reliable biomarkers. Additionally, levels of these proteins were correlated with known clinical indicators of pathological outcome, such as the Glasgow coma scale (GCS). Our results show that inflammatory cytokines and inflammasome proteins are promising biomarkers for determining pathological outcomes after TBI. Additionally, levels of biomarkers could potentially be utilized to determine a patient’s injury severity and subsequent pathological outcome. These findings show that inflammation-associated proteins in the blood are reliable biomarkers of injury severity that can also be used to assess the functional outcomes of TBI patients.

**Keywords:** inflammasome; inflammation; brain injury; biomarkers; cytokines; interleukin

1. **Introduction**

Traumatic brain injury (TBI) has a complex and chronic pathology that represents a significant public health concern in the United States and throughout the world [1–5]. It is estimated that in the United States there are 3.17 million people suffering with long term disability resulting from TBI, representing an annual economic impact in excess of...
$56 billion [6]. TBI presents as a biphasic pathology in which the effects of the initial traumatic insult results in persistent inflammation and the chronic activation of the innate immune system [7–10]. Primary injury involves the release of damage associated molecular patterns (DAMPS) from injured tissue resulting in the activation of the innate immune response and formation of the inflammasome [11–14]. Although the levels of DAMPs and PAMPs have been shown to gradually decrease over the first week after injury, chronic inflammatory activity is often seen months to years after injury, resulting in a secondary injury from chronically activated microglia and their subsequent release of inflammatory cytokines [15–17].

The inflammasome is a multi-protein complex that activates caspase-1 leading to the cleavage and release of the inflammatory cytokines interleukin (IL)-1β and IL-1, and the formation of a gasdermin-D pore as part of the programmed cell death mechanism of pyroptosis [18–23]. Inflammasome formation is triggered by numerous substances and has been shown to be activated after TBI in rodents and humans [24–26]. PAMPs and DAMPs are recognized by toll like receptors (TLR) initiating a cascade of events including TBI-induced cellular potassium efflux, increased intracellular calcium, subsequent mitochondrial dysfunction and excitotoxicity, which result in the formation and activation of the inflammasome [27]. The inflammasome activation involves cleavage of caspase-1 through the use of a scaffolding adaptor protein known as the apoptosis-associated speck-like protein containing a caspase recruiting domain (ASC), which oligomerizes to form an ASC speck. ASC specks bind to caspase-1, resulting in the formation of the inflammasome complex [28–31] and downstream activation of the pro-inflammatory cytokines interleukin (IL)-18 and IL-1β [32].

Biomarkers are specific proteins that are used as indicators of the status of different physiological processes in an individual [33–39]. Biomarkers are often used to determine the stage or severity of an underlying disease or injury [40–46]. TBI presents as a multifactorial series of events that affect a variety of cells within the central nervous system (CNS) including neurons, microglia, astrocytes, oligodendrocytes and endothelial cells [47,48]. However, a variety of biomarkers need to be identified in order to gain a better understanding of the different molecular events that affect different cell types after TBI in the clinical setting. Two biomarkers have been thus far approved by the FDA for the monitoring of TBI patients. These biomarkers are ubiquitin carboxy-terminal hydrolase (UCH-L1) and glial fibrillary acidic protein (GFAP) [49–55]. UCH-L1 is expressed in neurons, and it is highly upregulated after TBI, whereas GFAP is expressed in astrocytes [56]. Thus, these two approved biomarkers for TBI offer clinicians an assessment of the degree of neuronal degradation and astroglial activation after brain injury. However, to date, no approved fluid biomarker or series of biomarkers is available to determine the inflammatory response in the acute setting after TBI.

Previous studies have shown that inflammasome proteins are reliable biomarkers of the inflammatory response in several conditions such as stroke [66], Alzheimer’s disease [67], multiple sclerosis [68], age-related macular degeneration [69], psoriasis [70] and non-alcoholic steatohepatitis [71], indicating that the inflammasome plays a major role in the pathophysiology of a variety of diseases affecting the CNS and the
periphery. Moreover, those findings highlight the usefulness of inflammasome signaling proteins as biomarkers of injury and disease.

Despite ample evidence for the increased expression of a variety of inflammatory proteins in the CSF and blood of patients with TBI when compared to healthy uninjured controls, few studies have aimed to determine the biomarker characteristics of these inflammatory proteins, including the receiver operating characteristic (ROC) curve as well as the determination of cut-off points to identify the respective sensitivity and specificity of the different inflammatory biomarkers. In addition, previous studies have not compared the area under the curve (AUC) between different inflammatory biomarkers with the goal of identifying which inflammatory biomarkers are more suitable surrogates of the inflammatory response taking place acutely after TBI. Here, we measure the protein levels of inflammasome signaling proteins and inflammatory cytokines associated with TBI to then determine the biomarker characteristics of these proteins as well as the contribution of these inflammatory proteins to long term outcomes as determined by the Glasgow-Outcome Scale-Extended (GOS-E) and to injury severity as determined by the Glasgow-Comma Scale (GCS). Importantly, we follow a systematic approach to determine the suitability of each biomarker as a surrogate of inflammation following TBI and compare the AUC between each biomarker to identify which biomarkers have the potential to be more reliable biomarkers that can be used in the clinical setting.

2. Results
2.1. Inflammasome Proteins and Inflammatory Cytokines Are Elevated in TBI Patients

Increased inflammatory activity through inflammasome and cytokine signaling has been previously reported in animal and human TBI studies [27,72]. In order to determine which inflammatory proteins were elevated in this cohort of human TBI patients, we analyzed the levels of inflammasome proteins caspase-1 (Figure 1A), ASC (Figure 1B) and IL-18 (Figure 1C), and the inflammatory cytokines TNF-α (Figure 1D), IL-6 (Figure 1E), IL-4 (Figure 1F), IL-10 (Figure 1G), IL-8 (Figure 1H) and IL-2 (Figure 1I) in the blood serum of these patients and compared them to age-matched healthy controls. TBI patients had significantly elevated levels of inflammasome signaling proteins caspase-1 (Figure 1A), ASC (Figure 1B) and IL-18 (Figure 1C), as well as significantly elevated levels of cytokines TNF-α (Figure 1D), IL-6 (Figure 1E), IL-4 (Figure 1F), IL-10 (Figure 1G) and IL-8 (Figure 1H). In contrast, the levels of IL-2 were higher in the serum of healthy uninjured controls when compared to the serum of TBI patients (Figure 1I). Moreover, we found no significant difference in the levels of IL-12 when comparing the control with the TBI group (Figure S1). Therefore, these results indicate that TBI patients have sustained an acute increase in inflammatory activity after TBI.

2.2. Inflammatory Biomarkers of TBI

Previous studies have shown that inflammasome proteins are potentially promising biomarkers for determining TBI pathological outcomes [24,26,57]. In order to determine the biomarker reliability of the inflammasome proteins caspase-1 (Figure 2A), ASC (Figure 2B) and IL-18 (Figure 2C), and the inflammatory cytokines TNF-α (Figure 2D), IL-6 (Figure 2E), IL-4 (Figure 2F), IL-10 (Figure 2G), IL-8 (Figure 2H) and IL-2 (Figure 2I) in the context of TBI, we plotted the ROC curve for each protein (Figure 2). Of the inflammatory cytokines examined, IL-6 (Figure 2E) had the highest AUC 1.0 (Table 1) with a sensitivity of 100% and a specificity of 100% (Table 2). TNF-α had an AUC of 0.98 AUC (96% sensitivity, 95% specificity), IL-10 and IL-8 also presented high AUC values (0.97 and 0.95, respectively), whereas IL-4 had an AUC of 0.79 (74% sensitivity, 75% specificity) and IL-2 an AUC of 0.74 (95% sensitivity, 56% specificity).
Figure 1. Inflammatory cytokines and Inflammasome Proteins are Elevated after TBI. Simple Plex Assay and MSD-VPLEX Inflammatory Panel of blood serum from TBI patients and age-matched controls. Data were analyzed utilizing a two-tailed Mann–Whitney nonparametric test. Inflammatory cytokines and inflammasome proteins that showed a statistically significant increase after TBI were plotted. Box and whisker plots show mean and quartiles for each inflammatory protein of interest with respective $p$ values listed above. Dots correspond to data points outside the 5th and 95th percent confidence interval. Results showed that (A) Caspase-1: N: Control: 31, TBI: 78; (B) ASC: N: Control: 28, TBI: 91; (C) IL-18: N: Control: 31, TBI: 90; (D) TNF-α: N: Control: 21, TBI: 51; (E) IL-6: N: Control: 21, TBI: 46; and (F) IL-4: N: Control: 20, TBI: 50; (G) IL-10: N: Control: 19, TBI: 41; (H) IL-8: N: Control: 12, TBI: 52 were all significantly elevated in TBI patients when compared to controls. (I) IL-2: N: Control: 9, TBI: 19 was significantly decreased in TBI patients when compared to controls.

Table 1. ROC Analysis.

| Biomarker | Area | Std. Error | 95% C.I.       | $p$-Value |
|-----------|------|------------|----------------|-----------|
| Caspase-1 | 1.0  | 0          | 1.0 to 1.0     | <0.0001   |
| ASC       | 0.97 | 0.01384    | 0.9428 to 0.9971 | <0.0001  |
| IL-18     | 0.8143 | 0.04538 | 0.7254 to 0.9033 | <0.0001 |
| TNF-α     | 0.9776 | 0.01963 | 0.9391 to 1.000 | <0.0001 |
| IL-6      | 1.0  | 0          | 1.0 to 1.0     | <0.0001   |
| IL-4      | 0.7945 | 0.05424 | 0.6882 to 0.9008 | 0.0001   |
| IL-10     | 0.9538 | 0.02697 | 0.9009 to 1.0   | <0.0001   |
| IL-8      | 0.9696 | 0.2115 | 0.9281 to 1.0   | <0.0001   |
| IL-2      | 0.7398 | 0.1084 | 0.5274 to 0.9522 | 0.0437   |
| IL-12     | 0.5333 | 0.07812 | 0.3802 to 0.6864 | 0.6645   |
| IL-13     | 0.5126 | 0.1093 | 0.2985 to 0.7267 | 0.8858   |
Figure 2. ROC of Inflammatory Biomarkers. ROC and AUC were calculated for each inflammatory cytokine and inflammasome proteins that were significantly different when comparing healthy uninjured controls and TBI patients. (A) Caspase-1: N: Control: 31, TBI: 78; (B) ASC: N: Control: 28, TBI: 91; (C) IL-18: N: Control: 31, TBI: 90; (D) TNF-α: N: Control: 21, TBI: 51; (E) IL-6: N: Control: 21, TBI: 46; and (F) IL-4: N: Control: 20, TBI: 50; (G) IL-10: N: Control: 19, TBI: 41; (H) IL-8: N: Control: 12, TBI: 52; (I) IL-2: N: Control: 9, TBI: 19.

Table 2. Cut-off point in serum of TBI patients.

| Biomarker | Cut-Off Point (pg/mL) | Sensitivity (%) | Specificity (%) | LR    | PPV (%) | NPV (%) | Accuracy (%) |
|-----------|-----------------------|-----------------|-----------------|-------|---------|---------|---------------|
| Caspase-1 | >0.8150               | 100             | 100             | 100   | 100     | 100     | 100           |
| ASC       | >284                  | 92              | 93              | 12.92 | 98      | 79      | 92            |
| IL-18     | >156                  | 83              | 74              | 3.229 | 90      | 61      | 81            |
| TNF-α     | >2.202                | 96              | 95              | 20.18 | 98      | 91      | 96            |
| IL-6      | >6.443                | 100             | 100             | 100   | 100     | 100     | 100           |
| IL-4      | >0.03868              | 74              | 75              | 2.96  | 88      | 54      | 74            |
| IL-10     | >0.6527               | 88              | 100             | 100   | 100     | 79      | 92            |
| IL-8      | >29.18                | 92              | 100             | 100   | 100     | 75      | 94            |
| IL-2      | <0.5145               | 95              | 56              | 2.132 | 82      | 83      | 82            |
| IL-12     | <158.1                | 60              | 57              | 1.40  | 75      | 40      | 59            |
| IL-13     | >2.271                | 76              | 57              | 1.784 | 87      | 40      | 72            |
Of the inflammasome proteins examined, caspase-1 (Figure 2A) had the highest AUC at 1.0 (Table 1) with a sensitivity of 100% and a specificity of 100%, followed by ASC with an AUC of 0.97 with a sensitivity of 92% and a specificity of 93% (Table 2). IL-18 presented an AUC of 0.81 (sensitivity of 83%, specificity of 74%). These results indicate that caspase-1, ASC, IL-18, TNF-α, IL-6 IL-8 and IL-10 are reliable biomarkers of TBI with AUC values above 0.80.

2.3. Comparison between ROC Curves for Identified Inflammatory Biomarkers

To compare the ROC curves for caspase-1, ASC, IL-18, TNF-α, IL-6, IL-4, IL-10, IL-8 and IL-2, a Pearson correlation coefficient was first obtained from a correlation matrix (Figure 3A and Figure S2). The highest correlation was found between ASC and caspase-1 with a coefficient of correlation of 0.8, followed by a correlation of 0.58 between IL-6 and TNF-α, 0.57 between IL-6 and IL-8 and 0.53 between IL-6 and IL-4. After finding the coefficient of correlation, the ROC curves were compared according to the formula (Equation (1)):

$$z = \frac{(A_1 - A_2)}{\sqrt{SE_1^2 + SE_2^2 - 2rSE_1SE_2}}$$

(1)

ROC curve comparison analysis indicated that the ROC between caspase-1 and ASC ($p = 0.03$), caspase-1 and IL-18 ($p = 4.27 \times 10^{-5}$), caspase-1 and IL-4 ($p = 0.0001$) as well as caspase-1 and IL-2 ($p = 0.01$) were significantly different from each other (Figure 3B). Similarly, the ROC between ASC and other analytes differed from IL-18 ($p = 0.0009$), IL-6 ($p = 0.03$), IL-4 ($p = 0.001$) and IL-2 ($p = 0.04$). For IL-18, it also differed from TNF-α ($p = 0.002$), IL-6 ($p = 4.27 \times 10^{-5}$), IL-10 ($p = 0.02$) and IL-8 ($p = 0.004$). For TNF-α, the ROC curves also differed with IL-4 ($p = 0.0001$) and IL-2 ($p = 0.04$). For IL-6, the ROC curves differed with IL-4 ($p = 0.0002$) and IL-2 ($p = 0.02$). For IL-4, IL10 ($p = 0.002$) and IL-18 ($p = 0.0005$), and for IL-8 the ROC curve also differed to that of IL-2 ($p = 0.05$). Taken together, these analyses highlight caspase-1 and IL-6 as useful inflammatory biomarkers superior to all other biomarkers examined in this study; however, caspase-1 and IL-6 were not different from each other (Figure 3B). ASC and TNF-α were not different from each other but ASC was more reliable than IL-18 and IL-4. Similarly, IL-10 and IL-8 were not different from each other, and IL-8 was more reliable than IL-2.

2.4. Inflammatory Biomarkers of Injury Severity

Next, we screened inflammatory biomarkers to determine whether there was a difference in the levels of these proteins between patients that presented mild TBI and those who had moderate to severe TBI as determined by the GCS. Patients with mild TBI were those who presented a GCS between 13 and 15; whereas patients with a GCS between 3 and 12 were grouped in the moderate to severe cohort. Of all the analytes measured in this study, IL-13 was the only protein to be elevated in the moderate to severe group when compared to patients in the mild TBI group (Figure 4A). Moreover, we calculated the ROC curve for IL-13 and found that the AUC for IL-13 was 0.75 (Figure 4B) with a 95% confidence interval between 0.5815 to 0.9126 and an SEM of 0.085 ($p = 0.01$), indicating that IL-13 discriminates between mild and moderate to severe TBI. Furthermore, with a cut-off point of 3.12 pg/mL, the sensitivity and specificity of IL-13 were 71% and 79%, respectively. This resulted in a PPV of 85% and a NPV of 61% with an accuracy of 74%.
Figure 3. ROC Comparison among Inflammatory Biomarkers. (A) Correlation matrix using a Pearson correlation among inflammatory Biomarkers. (B) p-values of significance for the comparison among the inflammatory proteins analyzed: caspase-1, ASC, IL-18, TNF-α, IL-6, IL-4, IL-10 and IL-8.
We further fit a multivariate linear regression model using a stepwise approach to predict inflammatory biomarkers that contribute to the GCS. Accordingly, using as predictors the inflammatory proteins caspase-1, ASC, IL-18, TNF-α, IL-2, IL-4, IL-6, IL-8, IL-10, IL-12, IL-13 and IFN-γ.

The best model was chosen based on the AIC (14.37) by the stepwise method, and then the estimate (coefficients), standard error and \( p \)-values for each predictor and intercept (slope), as well as the BIC (35.52), residuals (Figure S3A), RMSE (1.65), mean of residuals (\( -6.25 \times 10^{-17} \)), confidence intervals, and the DW autocorrelation test for the best fit model, were calculated. An adjusted \( R^2 \) value (0.78) was obtained for the model to determine the approximate contribution of the fitted model to the GCS (Table 3). Thus, based on this model, we determined that the GCS score is contributed to in part by IL-12, IL-13 and the log (IL-12), considering an adjusted \( R^2 \) of 0.78 and a \( p \)-value of 0.03, consistent with our findings of IL-13 as a biomarker of injury severity after TBI based on the GCS.

Table 3. Linear Regression Model to predict GCS.

| GCS       | Estimate | Std. Error | \( p \)-Value | Confidence Interval                |
|-----------|----------|------------|--------------|-----------------------------------|
| Intercept | 21.245347| 2.221427   | 0.000668     | 15.07767794 to 27.413016974       |
| IL-13     | -1.519195| 0.332622   | 0.010282     | -2.44270084 to -0.595688402       |
| IL-12     | -0.016000| 0.004802   | 0.029063     | -0.02933364 to -0.002666043       |
| LOG(IL-2) | 2.095605 | 0.670269   | 0.035304     | 0.23463864 to 3.956571427         |

| Adjusted \( R^2 \) | 0.7798 |
|---------------------|--------|
| BIC                 | 35.51754 |
| RMSE                | 1.644861 |
| Mean of Residuals   | \( -6.25 \times 10^{-17} \) |
| DW Statistic        |        |
| rho ≠ 0             | \( p \)-value = 0.812 |
| rho < 0             | \( p \)-value = 0.443 |
| rho > 0             | \( p \)-value = 0.561 |

\[
\text{GCS} = 21.25 - 1.52 \text{ (IL-13)} - 0.02 \text{ (IL-12)} + 2.10 \times \text{LOG(IL-12)}.
\]
2.5. Inflammatory Biomarkers of Outcome

To determine the contribution of inflammatory proteins to outcomes according to the GOS-E, we first divided the outcomes as favorable and unfavorable and then determined if there was a statistically significant difference between the levels of the inflammatory proteins caspase-1, ASC, IL-18, TNF-α, IL-2, IL-4, IL-6, IL-8, IL-10, IL-12, IL-13 and IFN-γ with regard to favorable (GOSE scores of 5–8) vs. unfavorable (GOSE scores of 1–4) outcomes. Of the protein analyzed, caspase-1 (Figure 5A) and IL-10 (Figure 5B) were significantly elevated in patients that presented unfavorable outcomes.

Following the identification of caspase-1 and IL-10 as proteins that were elevated in patients with unfavorable outcomes, we aimed to identify whether these two analytes are good biomarkers of outcomes in TBI patients. The ROC curves of caspase-1 (Figure 5C) and IL-10 (Figure 5D) were 0.64 (p = 0.03) and 0.81 (p = 0.006), respectively. The SEM for the ROC of caspase-1 was 0.07 with a 95% confidence interval between 0.5127 and 0.7720. For IL-10, the SEM was 0.097 with a 95% confidence interval between 0.6182 and 0.9990. Moreover, with a cut-off point of 3.33 pg/mL, and a likelihood ratio of 1.3, the sensitivity and specificity of caspase-1 were 55% and 57%, respectively, with a PPV of 68% and an NPV of 44% with an accuracy of 56%. In contrast, for IL-10, with a cut-off point of 5.55 pg/mL, the sensitivity and specificity were 64% and 100%, respectively, with a PPV of 100% and an NPV of 61% with an accuracy of 77%.

We then fitted a multivariate linear regression model using caspase-1 and IL-10 as the predictors to explain the GOS-E. After analyzing residuals (Figure S3B) and adding a logarithmic transformation for caspase-1 levels, the adjusted R^2 value (0.65) was obtained.
for the model to determine the approximate contribution of IL-10 and caspase-1 to the GOS-E \((p = 0.03)\) (Table 4).

### Table 4. Linear Regression Model to predict GOS-E.

| GCS        | Estimate | Std. Error | \(p\)-Value | Confidence Interval            |
|------------|----------|------------|--------------|-------------------------------|
| Intercept  | 9.1238   | 0.7938     | 8.74 \times 10^{-05} | 7.0832670 to 11.1643303 |
| IL-10      | 0.2483   | 0.1416     | 0.1398       | -0.1156099 to 0.6122821    |
| \text{LOG(Caspase-1)} | -2.5997   | 0.6700     | 0.0116      | -4.3220676 to -0.8772845 |

Adjusted \(R^2\) 0.6524

BIC 24.35273

RMSE 0.8338086

Mean of Residuals 1.39 \times 10^{-17}

DW Statistic

| rho ! = 0 | \(p\)-value = 0.574 |
| rho < 0   | \(p\)-value = 0.716 |
| rho > 0   | \(p\)-value = 0.316 |

GOS-E = 9.12 + 0.25(\text{IL-10}) − 2.60 \times \text{LOG(caspase-1)}.

### 3. Discussion

Recent advancements in biomarker analytical technology have provided for the greater sensitivity and consistency across assays, thereby allowing the identification of substances in serum as potential biomarkers. In this study, we used two technologies: electrochemiluminescence (MESO QuickPlex SQ120, MSD, Rockville, MD, USA) and microfluidics (Ella, Protein Simple, San Jose, CA, USA) to measure inflammatory biomarkers in the serum of patients with TBI. Overall, we demonstrate that caspase-1, ASC, IL-6 and TNF were the most reliable pro-inflammatory biomarkers of the acute response after TBI, whereas IL-10 was the best anti-inflammatory biomarker. Therefore, the serum of TBI patients is an ideal source for the measurement of the signaling proteins that may be used for diagnostic and prognostic potential to estimate the degree of neuronal damage, astrocyte activation, and the inflammatory response involving microglial neutrophils and other inflammatory cells in TBI patients.

Previous studies have measured a variety of inflammatory proteins in healthy individuals and TBI patients [73]. Many of these studies did not evaluate the actual biomarker characteristics of these proteins. However, it is not sufficient to solely measure levels of proteins in control and TBI groups, but it is critical to calculate the ROC curve and to obtain the AUC by plotting the sensitivity in the \(y\)-axis and 1-specificity in the \(x\)-axis for each analyte. In addition, cut-off points should be identified with their respective sensitivity and specificity. AUC values between 0.9 and 1.0 correspond to an excellent biomarker; from 0.8 to 0.9, a good biomarker; from 0.7 to 0.8, a fair biomarker; from 0.6 to 0.7, poor and from 0.5 to 0.6, a failed analyte [74]. Our data show that caspase-1 and IL-6 with an AUC of 1.0 were the best inflammatory biomarkers of those examined in this study, followed by TNF-\(\alpha\) and ASC with an AUC of 0.98 and 0.97, respectively, and IL-8 and IL-10 with an AUC of 0.97 and 0.96, respectively. Furthermore, the most sensitive inflammatory biomarkers were caspase-1 and IL-6, followed by TNF-\(\alpha\), IL-2, ASC and IL-8, whereas the most specific biomarkers were caspase-1, IL-6, IL-8, IL-10, TNF-\(\alpha\) ASC. Moreover, caspase-1 and IL-6 presented an accuracy of 100\%, followed by TNF-\(\alpha\) with 96\%, IL-8, 94\% and ASC and IL-10 with an accuracy of 92\%. Taken together, these data indicate that caspase-1, ASC, TNF-\(\alpha\), IL-6, IL-10 and IL-8 are the most reliable diagnostic inflammatory biomarkers of TBI, among those studied. However, when there are several biomarkers that have similar AUC values, it is important to determine if the ROC curves differ among the different biomarkers. Thus, following the determination of the ROC curves for each biomarker, we
then compared the ROC for each of them, and identified significant differences between different biomarkers despite many of them having high AUC values. For instance, we found significant differences between caspase-1 and ASC. However, the ROC for caspase-1 and IL-6 were not significantly different, indicating that both of these analytes have the same biomarker potential based on their respective biomarker characteristics, yet this is not to say that each of these biomarkers does not provide different information pertaining to the acute inflammatory response after TBI. Similarly, the ROC curve for ASC and TNF were not found to be significantly different either, and the AUC of IL-6 was found to be superior to that of ASC, consistent with the significance found between caspase-1 and ASC. Overall, the findings of this study highlight the importance of caspase-1, ASC, IL-6 and TNF as pro-inflammatory biomarkers of the acute response after TBI, whereas IL-10 was the best anti-inflammatory biomarker.

Additionally, in this study, we dichotomized the GOS-E into favorable and unfavorable outcomes to identify whether inflammatory proteins were significantly different between patients with different outcomes after TBI. We found that caspase-1 and IL-10 were elevated in patients with unfavorable outcomes when compared to those with favorable outcomes. IL-10 is secreted by numerous cells of the CNS after injury and has been shown to play a protective role by reducing cytokine activity, proinflammatory activity, and apoptosis [75,76]. Additionally, IL-10 inhibits IL-2 activity, and has been shown in numerous studies to be increased after stroke or TBI [75,76]. Although IL-10 has a pro-survival purpose, increased IL-10 expression after injury has been associated with worsened pathological outcomes with higher expression associated with increased chance for mortality [75]. Our results support these observations and further suggest an interplay between pro-inflammatory (caspase-1) and anti-inflammatory (IL-10) proteins in the pathogenesis that play a role in long-term outcomes in patients with TBI. Biomarker analysis indicated that the AUC for caspase-1 was 0.64 and the AUC for IL-10 was 0.81, suggesting that IL-10 was a more reliable biomarker of long-term outcomes with a cut-off point of 5.55 pg/mL. Previous studies have shown that serum levels of TNF-α remain elevated for at least one-year post-injury [15,77]. Serum levels of IL-6 have been shown to be more elevated in more severe cases of TBI, and that elevated IL-6 levels were associated with worsened outcomes [77]. Similarly, IL-8 increased IL-8 expression in serum or CSF of TBI patients is associated with an increased chance for mortality and overall worsened pathological outcomes [78], and may be attributed to the chemoattractant properties of IL-8 that recruit and activate monocytes to the site of injury, increasing the overall inflammatory response after TBI or stroke [76,78].

Lastly, we found that IL-13 was elevated in patients with moderate to severe TBI as determined by the GCS. The exact role of IL-13 in CNS injury pathology is still up for debate [78,79], although some studies have suggested that IL-13 plays a neuroprotective role in that it reduces inflammatory activity, reduces axonal loss, and mediates microglia polarization, encouraging the adoption of the anti-inflammatory phenotype [76]. IL-13 treatment improves pathological outcomes in a murine model of TBI [79]. Furthermore, IL-2 has been shown to be decreased after TBI [80], and IL-13 has been shown to have shared functionality with IL-4, and synergizes with IL-2 to promote IFN production [76]. Our findings indicate that IL-13 had an AUC of 0.75 and a cut-off point of 3.12 pg/mL, with a sensitivity of 71% and a specificity of 79%; thus, IL-13 is a fair biomarker of TBI injury severity. Moreover, a multivariate linear regression model consisting of IL13, IL-2 and IL-12 indicated that, combined, these three biomarkers contribute to the GCS with an adjusted R² of 0.78, thus highlighting the importance of IL-13 and a key biomarker of injury severity.

4. Materials and Methods

4.1. Participants

Study specimens from TBI patients were acquired from Son Espases University Hospital (Palma de Mallorca, Spain). The study was approved by the Comité Ético de las Islas Baleares (IRB protocol number 3127/15). Written informed consent was obtained from
a family member or proxy according to the IRB (Table 5). Healthy age-matched controls were acquired from BiolVT (Hicksville, NY, USA). Informed consent was obtained from specimen donors. Control samples were obtained by donors participating in the study Prospective Collection of Samples for Research funded by SeraTrials, LLC. with IRB number 20170439. Blood samples from TBI patients used in this study were collected in the range of approximately 60 to 720 min after TBI with a median of 367.5 min (~6 h after TBI). Exclusion criteria consisted of patients with normal findings on the CT scan on admission, patients with a major extracranial trauma (defined as extracranial Injury Severity Score > 18 points), and patients with past medical history relevant to CNS pathology such as brain tumor, meningitis, cerebral vasculitis or stroke.

Table 5. Summary of demographic data and clinical characteristics in patients with TBI.

| TBI (N = 93)          |
|-----------------------|
| Gender (n, %)         |
| Male 74 (80%)         |
| Female 19 (20%)       |
| Age (years) median (Range) 47 (15–83) |
| Injury Mechanism (n; %) |
| Fall 52 (56%)         |
| Assault 5 (5%)        |
| MVA 36 (39%)          |
| Glasgow Coma Scale (n; %) |
| 3–8 37 (40%)          |
| 9–12 20 (21%)         |
| 13–15 34 (37%)        |
| Undetermined 2 (2%)   |
| Motor score (n; %)    |
| M6 38 (41%)           |
| M5 29 (32%)           |
| M4 4 (4%)             |
| M3 2 (2%)             |
| M2 1 (1%)             |
| M1 14 (15%)           |
| Undetermined 5 (5%)   |
| Pupillary Reactivity (n; %) |
| Both reactive 78 (84%) |
| 1 reactive 8 (9%)     |
| None reactive 7 (7%)  |
| Hospital length of stay (days) median (Range) 13 (1–149) |
| ICU length of stay (days) median (Range) 5 (1–90) |

TBI: Traumatic Brain injury; MVA: motor vehicle accident; ICU: Intensive Care Unit.

4.2. Data Collection

Patients’ clinical data were recorded and reviewed using the electronical medical records from the hospital (Power Chart; Millenium, 2011, Cerner Corporation, Kansas City, MO, USA). We collected all the variables included in the International Mission for Prognosis and Analysis of Clinical Trials in TBI (IMPACT) prognostic calculator for each patient. We also collected the GCS that first responders wrote in their prehospital report or the hospital admission GCS if the former was not available. The 6-month outcome was assessed using the extended version of the Glasgow Outcome Scale (GOSE) by a trained Neurosurgery Intensive Care Unit attending (JRP) by telephone consultation, and he was blinded to biomarker analysis.
4.3. Simple Plex Assay

The serum concentrations of inflammasome proteins (Caspase-1, ASC and IL-18) were measured in 93 TBI patients and in 31 age-matched controls via Ella System (Protein Simple) as described in [69]. Briefly, samples were loaded as 50 µL of diluted sample into sample wells of a CART with 1 mL of washing buffer loaded separately into respective buffer wells. An assay was run using the Runner Software (version 3.5.2.20, San Jose, CA, USA). Samples were then automatically analyzed utilizing the Simple Plex Explorer (version 3.7.2.0, San Jose, CA, USA) [70].

4.4. MSD V-PLEX Inflammatory Panel

Serum levels of the inflammatory cytokines TNF-α, IL-2, IL-4, IL-6, IL-8, IL-10, IL-12p-70, IL-13 and interferon (IFN)-γ were measured utilizing the V-PLEX Proinflammatory Panel 1 (MSD) as in [67]. All relevant controls, detection antibodies, standards, reagents, and dilutants were supplied by the manufacturer and prepared in accordance with manufacturer’s instructions. Briefly, samples were diluted 2-fold prior to loading into the plate. Plate wells were washed three times with wash buffer prior to sample loading. Then, 50 µL of sample was loaded into respective plate wells and allowed to incubate for 2 h at room temperature on a plate shaker. After incubation, plate wells were washed three times with wash buffer. Detection antibody was then added to plate wells and was allowed to incubate for 2 h at room temperature on a plate shaker. After antibody incubation, plate wells were washed three times with wash buffer. A 2X Read buffer was then added to each well and the plate was analyzed utilizing the MESO QuickPlex SQ120 (MSD, Rockville, MD, USA) and DISCOVERY WORKBENCH software (version 4.0.12, Rockville, MD, USA).

4.5. Statistical & Biomarker Analysis

Simple Plex and V-PLEX data from TBI and control samples were analyzed utilizing Prism 9 software (GraphPad). Outliers were removed prior to further statistical analyses using the Robust regression and Outlier removal (ROUT) method with a Q set to 1%. Descriptive statistics were run, and normality was determined by the Shapiro–Wilk test or the D’Agostino and Pearson Test. Non-parametric data were analyzed using a two-tailed Mann-Whitney test and parametric data were analyzed using a two-tailed t-test. p-value of significance was set to p < 0.05.

Receiver operating characteristics (ROC) were calculated to obtain the area under the curve (AUC) in order to obtain cut-off points and the respective specificity, sensitivity and likelihood ratio. The cut-off point for each analyte was chosen based on the highest likelihood ratio in the sensitivity vs. 1-specificity plot, favoring a higher sensitivity than specificity values, to obtain assays with a higher likelihood of reliability for each analyte [81]. Positive and negative predictive values were also calculated along with overall assay accuracy.

A comparison of ROC curves between inflammatory biomarkers was carried out as described in [82] using the Equation (1) to obtain a critical ratio z:

The p-value was determined using the following formula using Microsoft Excel (version 16.57, Redmond, WA, USA):

\[ Z = 2 \times (1 - NORMSDIST(z)). \]

A Pearson correlation was carried out to obtain r in order to calculate the z-score to allow for a comparison of ROC curves between analytes obtained from the same samples.

Linear regression analyses to explain the GCS and the GOS-E were fit using all the inflammatory proteins analyzed in this study through a stepwise approach based on the lowest Akaike information criterion (AIC) using RStudio/RMarkdown (Version 1.2.5033, Boston, MA, USA) and were then fitted to obtain the estimate, standard error and p-values for each predictor and the intercept. The Bayesian information criterion (BIC), residuals, root mean-square error (RMSE), mean of residuals, confidence intervals, and the
autocorrelation using the Durbin–Watson (DW) statistic were then calculated for the best fit model. The Durbin–Watson (DW) statistic was used to test for autocorrelation. After identifying a best fit model, data points underwent logarithmic transformation to normalize the distribution of the data. An adjusted r-squared value was obtained to determine the approximate contribution of these three proteins to either the GCS or the GOS-E. The final models were then further evaluated by residual analysis with and without logarithmic transformation.

5. Conclusions

In conclusion, we provide a systematic approach for inflammatory biomarker identification that includes: (1) measurement of the levels of inflammatory problems in the serum of affected and unaffected individuals to determine if there are statistical differences between groups; (2) determination of the diagnostic biomarker characteristics (AUC, sensitivity, specificity, likelihood ratio, accuracy, PPV and NPV) of each inflammatory protein or analyte that was statistically significant when comparing the levels between affected and unaffected individuals; (3) comparison of the ROC among the different biomarkers to identify potential biomarker differences between groups; (4) dichotomization of the GCS into mild and moderate to severe outcomes to determine if there are inflammatory biomarkers that meet the criteria as useful biomarkers of injury severity; and (5) dichotomization of the GOS-E into favorable and unfavorable outcomes to determine if there are inflammatory biomarkers that meet the criteria as useful biomarkers of long-term outcomes. Taken together, we identified the caspase-1, ASC, IL-18, TNF-α, IL-2, IL-4, IL-6, IL-8, IL-10 and IL-12 as surrogate biomarkers in serum of the inflammatory response acutely after TBI. Thus, the use of inflammatory biomarkers when combined with GFAP and UCH-L1 may offer clinicians a better understanding of the overall scope of injury and provide a probable prognosis and potential for disability considering a variety of mechanisms contributing to the TBI pathology, including neuronal damage (UCH-L1), reactive astrogliosis (GFAP) and inflammation (caspase-1, ASC, TNF-α and IL-6). Identification of clinically relevant biomarkers of the inflammatory response after TBI allow for future studies looking at how therapeutics affect these biomarkers, and how the effects of those therapeutics on biomarkers affect injury severity and functional outcomes in patients after TBI. In addition, the identification of these inflammatory biomarkers provides the opportunity of developing therapeutics that can be used to more specifically treat the inflammatory response associated with TBI. Taken together, with modern approaches for the measurement of biomarkers with higher accuracy and sensitivity than in the past, and with the identification of biomarkers of neuronal damage, reactive astrogliosis, and inflammation, personalized care for TBI patients is becoming a more tangible reality. Furthermore, in light of the results in this project, future studies in clinically relevant animal models of TBI should focus on understanding the individual and synergistic effects of therapeutically targeting the inflammatory proteins identified as relevant biomarkers of the inflammatory response after TBI for their ability to improve histopathological and functional outcomes.

Supplementary Materials: The following supporting information can be downloaded at: https://www.mdpi.com/article/10.3390/ph15060660/s1, Figure S1: IL-12 profile in serum. Figure S2: (A) Autocorrelation Plot of all analytes examined in this study; (B) p-values for the respective comparisons shown in the autocorrelation plot. Figure S3: Residual analysis of the multivariate linear regression models used to explain GCS (A) and GOS-E (B) were performed to determine the goodness of fit of each of the models.

Author Contributions: J.P.d.R.V. and J.P.-B. conceived and designed the study. J.P.-B., R.W.K., W.D.D. and J.P.d.R.V. provided materials. N.H.J., R.H., R.R.T., J.R.P., O.S., J.A.L.-P., J.P.-B. and J.P.d.R.V. were responsible for data collection. All authors contributed to data analysis and interpretation as well as the preparation of the manuscript. R.W.K., W.D.D., J.P.-B. and J.Pd.R.V. contributed to literature review. All authors have read and agreed to the published version of the manuscript.
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Institutional Review Board Statement: The study was conducted in accordance with the Declaration of Helsinki, and approved by the Ethics Committee of Comité Ético de las Islas Baleares (protocol code 3127/15 and date of approval: 29 November 2017).

Informed Consent Statement: Informed consent was obtained from all subjects involved in the study.

Data Availability Statement: Data is contained within the article and supplementary material.

Conflicts of Interest: J.P.d.R.V., R.W.K. and W.D.D. are co-founders and managing members of InflamaCORE, LLC and have licensed patents on inflammasome proteins as biomarkers of injury and disease as well as on targeting inflammasome proteins for therapeutic purposes. J.P.d.R.V., R.W.K. and W.D.D. are Scientific Advisory Board Members of ZyVersa Therapeutics.

References

1. Bachynski, K.E.; Goldberg, D.S. Youth sports & public health: Framing risks of mild traumatic brain injury in american football and ice hockey. J. Law Med. Ethics 2014, 42, 323–333. [CrossRef] [PubMed]

2. Dams-O’Connor, K.; Cantor, J.B.; Brown, M.; Dijkers, M.P.; Spiegelman, L.A.; Gordon, W.A. Screening for traumatic brain injury: Findings and public health implications. J. Head Traum. Rehabil. 2014, 29, 479–489. [CrossRef] [PubMed]

3. Othman, H.; Ludin, S.M.; Saidi, S.; Awang, M.S. The needs of traumatic brain injury survivors’ caregivers and the implication required during the COVID-19 pandemic: Public health issues. J. Public Health Res. 2021, 10, 2205. [CrossRef] [PubMed]

4. Peters, M.E. Traumatic brain injury in older adults: Shining light on a growing public health crisis. Int. Rev. Psychiatry 2020, 32, 1–2. [CrossRef]

5. Waltzman, D.; Haarbaure-Krupa, J.; Womack, L.S. Traumatic Brain Injury in Older Adults-A Public Health Perspective. JAMA Neurol. 2022, 79, 437–438. [CrossRef]

6. Titus, D.J.; Johnstone, T.; Johnson, N.H.; London, S.H.; Chapalamadugu, M.; Hogenkamp, D.; Gee, K.W.; Atkins, C.M. Positive allosteric modulation of the alpha7 nicotinic acetylcholine receptor as a treatment for cognitive deficits after traumatic brain injury. PLoS ONE 2019, 14, e0223180. [CrossRef]

7. Lassaren, P.; Lindblad, C.; Frostell, A.; Carpenter, K.L.H.; Guilfoyle, M.R.; Hutchinson, P.J.A.; Helmy, A.; Thelin, E.P. Systemic inflammation alters the neuroinflammatory response: A prospective clinical trial in traumatic brain injury. J. Neuroinflamm. 2021, 18, 221. [CrossRef]

8. Risbrough, V.B.; Vaughn, M.N.; Friend, S.F. Role of Inflammation in Traumatic Brain Injury-Associated Risk for Neuropsychiatric Disorders: State of the Evidence and Where Do We Go From Here. Biol. Psychiatry 2022, 91, 438–448. [CrossRef]

9. Visser, K.; Koggel, M.; Blauw, J.; van der Horn, H.J.; Jacobs, B.; van der Naalt, J. Blood-based biomarkers of inflammation in mild traumatic brain injury: A systematic review. Neurosci. Biobehav. Rev. 2022, 132, 154–168. [CrossRef]

10. Xu, W.; Yue, S.; Wang, P.; Wen, B.; Zhang, X. Systemic inflammation in traumatic brain injury predicts poor cognitive function. Immun. Inflamm. Dis. 2022, 10, e577. [CrossRef]

11. Irrera, N.; Russo, M.; Pallio, G.; Bitto, A.; Mannino, F.; Minutoli, L.; Altavilla, D.; Squadrito, F. The Role of NLRP3 Inflammasome in the Pathogenesis of Traumatic Brain Injury. Int. J. Mol. Sci. 2020, 21, 6204. [CrossRef]

12. Ismael, S.; Ahmed, H.A.; Adris, T.; Parveen, K.; Thakor, P.; Ishrat, T. The NLRP3 inflammasome: A potential therapeutic target for traumatic brain injury. Neuroal. Regen. Res. 2021, 16, 49–57. [CrossRef] [PubMed]

13. Mortezaei, K.; Khanlarkhani, N.; Beyer, C.; Zendedel, A. Inflammasome: Its role in traumatic brain and spinal cord injury. J. Cell Physiol. 2018, 233, 5160–5169. [CrossRef] [PubMed]

14. O’Brien, W.T.; Pham, L.; Symons, G.F.; Monif, M.; Shultz, S.R.; McDonald, S.J. The NLRP3 inflammasome in traumatic brain injury: Potential as a biomarker and therapeutic target. J. Neuroinflamm. 2020, 17, 104. [CrossRef] [PubMed]

15. Kumar, R.G.; Boles, J.A.; Wagner, A.K. Chronic Inflammation After Severe Traumatic Brain Injury: Characterization and Associations with Outcome at 6 and 12 Months Postinjury. J. Head Traum. Rehabil. 2015, 30, 369–381. [CrossRef] [PubMed]

16. Niu, X.; Bai, L.; Sun, Y.; Wang, Y.; Bai, G.; Yin, B.; Wang, S.; Gan, S.; Jia, X.; Liu, H. Mild traumatic brain injury is associated with effect of inflammation on structural changes of default mode network in those developing chronic pain. J. Neurosci. Biology 2021, 21, 135. [CrossRef] [PubMed]

17. Witcher, K.G.; Bray, C.E.; Chunchai, T.; Zhao, F.; O’Neil, S.M.; Gordillo, A.J.; Campbell, W.A.; McKim, D.B.; Liu, X.; Dziabas, J.E.; et al. Traumatic Brain Injury Causes Chronic Cortical Inflammation and Neuronal Dysfunction Mediated by Microglia. J. Neurosci. 2021, 41, 1597–1616. [CrossRef]

18. Kamajaya, L.J.; Boucher, D. Gasdermin D Cleavage Assay Following Inflammamosome Activation. Methods Mol. Biol. 2022, 2459, 39–49. [CrossRef] [PubMed]

19. Kayagaki, N.; Stowe, I.B.; Lee, B.L.; O’Rourke, K.; Anderson, K.; Warming, S.; Cuellar, T.; Haley, B.; Roose-Girma, M.; Phung, Q.T.; et al. Caspase-11 cleaves gasdermin D for non-canonical inflammasome signalling. Nature 2015, 526, 666–671. [CrossRef]
20. Liu, X.; Zhang, Z.; Ruan, J.; Pan, Y.; Magupalli, V.G.; Wu, H.; Lieberman, J. Inflammasome-activated gasdermin D causes pyroptosis by forming membrane pores. *Nature* 2016, 535, 153–158. [CrossRef]

21. Mascarenhas, D.P.; Cerqueira, D.M.; Pereira, M.S.F.; Castanheira, F.V.S.; Fernandes, T.D.; Manin, G.Z.; Cunha, L.D.; Zamboni, D.S. Inhibition of caspase-1 or gasdermin-D enables caspase-8 activation in the Naip5/Nlrc4/ASC inflammasome. *PLoS Pathog.* 2017, 13, e1006502. [CrossRef] [PubMed]

22. Wang, K.K.; Yang, Z.; Zhu, T.; Shi, Y.; Rubenstein, R.; Tyndall, J.A.; Manley, G.T. An update on diagnostic and prognostic biomarkers for traumatic brain injury. *Expert. Rev. Mol. Diagn.* 2018, 18, 165–180. [CrossRef] [PubMed]

23. Wu, J.; Sun, J.; Meng, X. Pyroptosis by caspase-11 inflammasome-Gasdermin D pathway in autoimmune diseases. *Pharmacol. Res.* 2021, 165, 105408. [CrossRef] [PubMed]

24. Adamczak, S.; Dale, G.; de Rivero Vaccari, J.P.; Bullock, M.R.; Dietrich, W.D.; Keane, R.W. Inflammasome proteins in cerebrospinal fluid of brain-injured patients as biomarkers of functional outcome: Clinical article. *J. Neurosurg.* 2012, 117, 1119–1125. [CrossRef] [PubMed]

25. de Rivero Vaccari, J.P.; Lotocki, G.; Alonso, O.F.; Bramlett, H.M.; Dietrich, W.D.; Keane, R.W. Therapeutic neutralization of the NLRP1 inflammasome reduces the innate immune response and improves histopathology after traumatic brain injury. *J. Cereb. Blood Flow Metab.* 2009, 29, 1251–1261. [CrossRef] [PubMed]

26. Kerr, N.; Lee, S.W.; Perez-Barcena, J.; Crespi, C.; Ibanez, J.; Bullock, M.R.; Dietrich, W.D.; Keane, R.W.; de Rivero Vaccari, J.P. Inflammasome proteins as biomarkers of traumatic brain injury. *PLoS ONE* 2013, 8, e101028. [CrossRef] [PubMed]

27. de Rivero Vaccari, J.P.; Dietrich, W.D.; Keane, R.W. Therapeutics targeting the inflammasome after central nervous system injury. *Transl. Res.* 2016, 167, 35–45. [CrossRef] [PubMed]

28. Diaz-Parga, P.; de Alba, E. Inflammasome regulation by adaptor isoforms, ASC and ASCb, via differential self-assembly. *J. Biol. Chem.* 2022, 298, 101566. [CrossRef]

29. Li, S.; Wang, L.; Xu, Z.; Huang, Y.; Xue, R.; Yue, T.; Xu, L.; Gong, F.; Bai, S.; Wu, Q.; et al. ASC deglutathionylation is a checkpoint for NLRP3 inflammasome activation. *J. Exp. Med.* 2021, 218, e20202637. [CrossRef] [PubMed]

30. Nagar, A.; Rahman, T.; Harton, J.A. The ASC Speck and NLRP3 Inflammasome Function Are Spatially and Temporally Distinct. *Front. Immunol.* 2021, 12, 752482. [CrossRef]

31. Wittmann, N.; Behrendt, A.K.; Mishra, N.; Bossaller, L.; Meyer-Bahlburg, A. Instructions for Flow Cytometric Detection of ASC Specks as a Readout of Inflammasome Activation in Human Blood. *Cells* 2021, 10, 2880. [CrossRef] [PubMed]

32. Tsuchiya, K.; Hara, H.; Fang, R.; Hernandez-Cuellar, E.; Sakai, S.; Daim, S.; Chen, X.; Dewan, C.M.; Wu, H.; et al. ASC deglutathionylation is a checkpoint for NLRP3 inflammasome activation. *J. Exp. Med.* 2021, 218, e20202637. [CrossRef] [PubMed]

33. Clark, A.L.; Sorg, S.F.; Schiehser, D.M.; Bigler, E.D.; Bondi, M.W.; Jacobson, M.W.; Jak, A.J.; Delano-Wood, L. White Matter Complicated Assessment. *J. Head Traum. Rehabil.* 2016, 6, 110. [CrossRef] [PubMed]

34. Di Battista, A.P.; Buonora, J.E.; Rhind, S.G.; Hutchison, M.G.; Baker, A.J.; Rizoli, S.B.; Diaz-Arrastia, R.; Mueller, G.P. Blood Biomarkers in Moderate-To-Severe Traumatic Brain Injury: Potential Utility of a Multi-Marker Approach in Characterizing Outcome. *Front. Neurol.* 2015, 6, 110. [CrossRef] [PubMed]

35. Huijbrigs, M.E.; Bazarian, J.J.; Shultz, S.R.; Kawata, K. The biological significance and clinical utility of emerging blood biomarkers for traumatic brain injury. *Neurosci. Biobehav. Rev.* 2021, 130, 433–447. [CrossRef] [PubMed]

36. Lewis, L.M.; Schloemann, D.T.; Papa, L.; Fucetola, R.P.; Bazarian, J.; Lindburg, M.; Welch, R.D. Utility of Serum Biomarkers in the Diagnosis and Stratification of Mild Traumatic Brain Injury. *Acad Emerg. Med.* 2017, 24, 710–720. [CrossRef] [PubMed]

37. Lifshitz, J.; Rowe, R.K.; Griffiths, D.R.; Evilsizer, M.N.; Thomas, T.C.; Adelson, P.D.; McIntosh, T.K. Clinical relevance of midline fluid percussion brain injury: Acute deficits, chronic morbidities and the utility of biomarkers. *Brain Inj.* 2016, 30, 1293–1301. [CrossRef] [PubMed]

38. Rodriguez-Rodriguez, A.; Egea-Guerrero, J.J. The utility of biomarkers in traumatic brain injury clinical management. *Crit. Care* 2016, 20, 376. [CrossRef] [PubMed]

39. Strathmann, F.G.; Schulte, S.; Goerl, K.; Petron, D.J. Blood-based biomarkers for traumatic brain injury: Evaluation of research approaches, available methods and potential utility from the clinician and clinical laboratory perspectives. *Clin. Biochem.* 2014, 47, 876–888. [CrossRef] [PubMed]

40. Al-Adli, N.; Akbik, O.S.; Rail, B.; Montgomery, E.; Caldwell, C.; Barrie, U.; Vira, S.; Al Tamimi, M.; Bagley, C.A.; Aoun, S.G. The Clinical Use of Serum Biomarkers in Traumatic Brain Injury: A Systematic Review Stratified by Injury Severity. *World Neurosurg.* 2021, 155, e418–e438. [CrossRef] [PubMed]

41. Anada, R.P.; Wong, K.T.; Jayapalan, J.J.; Hashim, O.H.; Ganesan, D. Panel of serum protein biomarkers to grade the severity of traumatic brain injury. *Electrophoresis* 2018, 39, 2308–2315. [CrossRef] [PubMed]

42. Carabias, C.S.; Gomez, P.A.; Panero, I.; Eiriz, C.; Castano-Leon, A.M.; Egea, J.; Lages, A. i+12 Neurotraumatology Group, C. Chitinase-3-Like Protein 1, Serum Amyloid A1, C-Reactive Protein, and Procalcitonin Are Promising Biomarkers for Intracranial Severity Assessment of Traumatic Brain Injury: Relationship with Glasgow Coma Scale and Computed Tomography Volumetry. *World Neurosurg.* 2020, 134, e120–e143. [CrossRef] [PubMed]

43. Chen, H.; Cao, H.L.; Chen, S.W.; Guo, Y.; Gao, W.W.; Tian, H.L.; Xue, L.X. Neuroglobin and Nogo-a as biomarkers for the severity and prognosis of traumatic brain injury. *Biomarkers* 2015, 20, 495–501. [CrossRef] [PubMed]
44. Czeiter, E.; Amrein, K.; Gravesteijn, B.Y.; Lecky, F.; Menon, D.K.; Mondello, S.; Newcombe, V.F.J.; Richter, S.; Steyerberg, E.W.; Vyvere, T.V.; et al. Blood biomarkers on admission in acute traumatic brain injury: Relations to severity, CT findings and care path in the CENTER-TBI study. *Elife* 2020, 9, e52785. [CrossRef]
45. Sharma, S.; Kumar, A.; Choudhary, A.; Sharma, S.; Khurana, L.; Sharma, N.; Kumar, V.; Bisht, A. Neuroprotective Role of Oral Vitamin D Supplementation on Consciousness and Inflammatory Biomarkers in Determining Severity Outcome in Acute Traumatic Brain Injury Patients: A Double-Blind Randomized Clinical Trial. *Clin. Drug Investig*. 2020, 40, 327–334. [CrossRef]
46. Zoltewicz, J.S.; Mondello, S.; Yang, B.; Newsom, K.J.; Kobeissy, F.; Yao, C.; Lu, X.C.; Dave, J.R.; Shear, D.A.; Schmid, K.; et al. Biomarkers track damage after graded injury severity in a rat model of penetrating brain injury. *J. Neurotrauma*. 2013, 30, 1161–1169. [CrossRef]
47. Harting, M.T.; Jimenez, F.; Adams, S.D.; Mercer, D.W.; Cox, C.S., Jr. Acute, regional inflammatory response after traumatic brain injury: Implications for cellular therapy. *Surgery* 2008, 144, 803–813. [CrossRef]
48. Mautes, A.E.; Fukuda, K.; Noble, L.J. Cellular response in the cerebellum after midline traumatic brain injury in the rat. *Neurosci. Lett.* 1996, 214, 95–98. [CrossRef]
49. Anderson, T.N.; Hinson, H.E. Damaged: Elevated GFAP and UCH-L1 as the Black Flag of Brain Injury. *Resuscitation* 2020, 154, 110–111. [CrossRef]
50. Middleton, J. UCH-L1 and GFAP Testing (i-STAT TBI Plasma) for the Detection of Intracranial Injury Following Mild Traumatic Brain Injury. *Am. Fam. Physician* 2022, 105, 313–314.
51. Papa, L.; Brophy, G.M.; Welch, R.D.; Lewis, L.M.; Braga, C.F.; Tan, C.N.; Ameli, N.J.; Lopez, M.A.; Haeussler, C.A.; Mendez Giordano, D.I.; et al. Time Course and Diagnostic Accuracy of Glial and Neuronal Blood Biomarkers GFAP and UCH-L1 in a Large Cohort of Trauma Patients with and without Mild Traumatic Brain Injury. *JAMA Neurol*. 2016, 73, 551–560. [CrossRef] [PubMed]
52. Papa, L.; Zonfrillo, M.R.; Welch, R.D.; Lewis, L.M.; Braga, C.F.; Tan, C.N.; Ameli, N.J.; Lopez, M.A.; Haeussler, C.A.; Mendez Giordano, D.; et al. Evaluating glial and neuronal blood biomarkers GFAP and UCH-L1 as gradients of brain injury in concussive, subconcussive and non-concussive trauma: A prospective cohort study. *BMJ Paediatr. Open* 2019, 3, e000473. [CrossRef] [PubMed]
53. Rhine, T.; Babcock, L.; Zhang, N.; Leach, J.; Wade, S.L. Are UCH-L1 and GFAP promising biomarkers for children with mild traumatic brain injury? *Brain Inj*. 2016, 30, 1231–1238. [CrossRef] [PubMed]
54. Shahim, P.; Politis, A.; van der Merwe, A.; Moore, B.; Ekanayake, V.; Lippa, S.M.; Chou, Y.Y.; Pham, D.L.; Butman, J.A.; Diaz-Arrastia, R.; et al. Time course and diagnostic utility of NF-L, tau, GFAP, and UCH-L1 in subacute and chronic TBI. *Neurology* 2020, 95, e623–e636. [CrossRef]
55. Yang, Z.; Xu, H.; Sura, L.; Arja, R.D.; Patterson, R.L.; Rossignol, C.; Albyaram, M.; Rajderkar, D.; Ghosh, S.; Wang, K.; et al. Combined GFAP, NFL, Tau, and UCH-L1 panel increases prediction of outcomes in neonatal encephalopathy. *Pediatr. Res*. 2022, Online ahead of print. [CrossRef]
56. Diaz-Arrastia, R.; Wang, K.K.; Papa, L.; Sorani, M.D.; Yue, J.K.; Puccio, A.M.; McMahon, P.J.; Inoue, T.; Yuh, E.L.; Lingsma, H.F.; et al. Acute biomarkers of traumatic brain injury: Relationship between plasma levels of ubiquitin C-terminal hydrolase-L1 and glial fibrillary acidic protein. *J. Neurotrauma*. 2014, 31, 19–25. [CrossRef]
57. Perez-Barcena, J.; Crespi, C.; Frontera, G.; Llompart-Pou, J.A.; Salazar, O.; Golinye, V.; Ibanez, J.; Bullock, M.R.; de Rivera Vaccari, J.P. Levels of caspase-1 in cerebrospinal fluid of patients with traumatic brain injury: Correlation with intracranial pressure and outcome. *J. Neurosurg*. 2020, 134, 1644–1649. [CrossRef]
58. Csuka, E.; Morganti-Kossmann, M.C.; Lenzlinger, P.M.; Joller, H.; Trentz, O.; Kossmann, T. IL-10 levels in cerebrospinal fluid after brain injury is associated with blood-brain barrier dysfunction and nerve growth factor production. *Brain Inj*. 2013, 27, 419–425. [CrossRef] [PubMed]
59. Rossi, S.A.; Halliday, M.I.; Campbell, G.C.; Byrnes, D.P.; Rowlands, B.J. The presence of tumour necrosis factor in CSF and plasma after severe head injury. *Br. J. Neurosurg.* 2014, 28, 419–425. [CrossRef] [PubMed]
60. Tobinick, E.; Kim, N.M.; Reyzin, G.; Rodriguez-Romanacce, H.; DePuy, V. Selective TNF inhibition for chronic stroke and traumatic brain injury: An observational study involving 629 consecutive patients treated with perispinal etanercept. *CNS Drugs* 2012, 26, 1051–1070. [CrossRef]
61. Aisiku, I.P.; Yamal, J.M.; Doshi, P.; Benoit, J.S.; Gopinath, S.; Goodman, J.C.; Robertson, C.S. Plasma cytokines IL-6, IL-8, and IL-10 are associated with the development of acute respiratory distress syndrome in patients with severe traumatic brain injury. *Crit. Care* 2016, 20, 288. [CrossRef] [PubMed]
62. Kossmann, T.; Stahel, P.F.; Lenzlinger, P.M.; Redl, H.; Dubs, R.W.; Trentz, O.; Schlag, G.; Morganti-Kossmann, M.C. Interleukin-8 released into the cerebrospinal fluid after brain injury is associated with blood-brain barrier dysfunction and nerve growth factor production. *J. Cereb. Blood Flow Metab.* 1997, 17, 280–289. [CrossRef] [PubMed]
63. Rhodes, J.; Sharkey, J.; Andrews, P. Serum IL-8 and MCP-1 concentration do not identify patients with enlarging contusions after traumatic brain injury. *J. Traum.* 2009, 66, 1591–1597. [CrossRef] [PubMed]
64. Timmerman, K.L.; Amonette, W.E.; Markofski, M.M.; Ansinelli, H.A.; Gleason, E.A.; Rasmussen, B.B.; Mossberg, K.A. Blunted IL-6 and IL-10 response to maximal aerobic exercise in patients with traumatic brain injury. *Eur. J. Appl. Physiol.* 2015, 115, 111–118. [CrossRef]
65. Dalla Libera, A.L.; Regner, A.; de Paoli, J.; Centenaro, L.; Martins, T.T.; Simon, D. IL-6 polymorphism associated with fatal outcome in patients with severe traumatic brain injury. *Brain Inj.* 2011, 25, 365–369. [CrossRef] [PubMed]
66. Kerr, N.; Garcia-Contreras, M.; Abbassi, S.; Mejias, N.H.; Desousa, B.R.; Ricordi, C.; Dietrich, W.D.; Keane, R.W.; de Rivero Vaccari, J.P. Inflammasome Proteins in Serum and Serum-Derived Extracellular Vesicles as Biomarkers of Stroke. *Front Mol. Neurosci.* 2018, 11, 309. [CrossRef]

67. Scott, X.O.; Stephens, M.E.; Desir, M.C.; Dietrich, W.D.; Keane, R.W.; de Rivero Vaccari, J.P. The Inflammasome Adaptor Protein ASC in Mild Cognitive Impairment and Alzheimer’s Disease. *Int. J. Mol. Sci.* 2020, 21, 4674. [CrossRef]

68. Keane, R.W.; Dietrich, W.D.; de Rivero Vaccari, J.P. Inflammasome Proteins As Biomarkers of Multiple Sclerosis. *Front Neurol.* 2018, 9, 135. [CrossRef]

69. Weaver, C.; Cyr, B.; de Rivero Vaccari, J.C.; de Rivero Vaccari, J.P. Inflammasome Proteins as Inflammatory Biomarkers of Age-Related Macular Degeneration. *Transl. Vis. Sci. Technol.* 2020, 9, 27. [CrossRef]

70. Forouzandeh, M.; Besen, J.; Keane, R.W.; de Rivero Vaccari, J.P. The Inflammasome Signaling Proteins ASC and IL-18 as Biomarkers of Psoriasis. *Front Pharmacol.* 2020, 11, 1238. [CrossRef]

71. Cyr, B.; Keane, R.W.; de Rivero Vaccari, J.P. ASC, IL-18 and Galectin-3 as Biomarkers of Non-Alcoholic Steatohepatitis: A Proof of Concept Study. *Int. J. Mol. Sci.* 2020, 21, 8580. [CrossRef] [PubMed]

72. de Rivero Vaccari, J.C.; Brand, F.J., 3rd; Berti, A.F.; Alonso, O.F.; Bullock, M.R.; de Rivero Vaccari, J.P. Mincle signaling in the innate immune response after traumatic brain injury. *J. Neurotraum.* 2015, 32, 228–236. [CrossRef] [PubMed]

73. Woodcock, T.; Morganti-Kossmann, M.C. The role of markers of inflammation in traumatic brain injury. *Front Neurol.* 2013, 4, 18. [CrossRef] [PubMed]

74. Xia, J.; Broadhurst, D.I.; Wilson, M.; Wishart, D.S. Translational biomarker discovery in clinical metabolomics: An introductory tutorial. *Metabolomics* 2013, 9, 280–299. [CrossRef] [PubMed]

75. Garcia, J.M.; Stillings, S.A.; Leclerc, J.L.; Phillips, H.; Edwards, N.J.; Robicsek, S.A.; Hoh, B.L.; Blackburn, S.; Dore, S. Role of Interleukin-10 in Acute Brain Injuries. *Front Neurol.* 2017, 8, 244. [CrossRef] [PubMed]

76. Zhu, H.; Hu, S.; Li, Y.; Sun, Y.; Xiong, X.; Hu, X.; Chen, J.; Qiu, S. Interleukins and Ischemic Stroke. *Front. Immunol.* 2022, 13, 828447. [CrossRef] [PubMed]

77. Alam, A.; Thelin, E.P.; Tajsic, T.; Khan, D.Z.; Khellaf, A.; Patani, R.; Helmy, A. Cellular infiltration in traumatic brain injury. *J. Neuroinflamm.* 2020, 17, 328. [CrossRef]

78. Rodney, T.; Osier, N.; Gill, J. Pro- and anti-inflammatory biomarkers and traumatic brain injury outcomes: A review. *Cytokine* 2018, 110, 248–256. [CrossRef]

79. Miao, W.; Zhao, Y.; Huang, Y.; Chen, D.; Luo, C.; Su, W.; Gao, Y. IL-13 Ameliorates Neuroinflammation and Promotes Functional Recovery after Traumatic Brain Injury. *J. Immunol.* 2020, 204, 1486–1498. [CrossRef]

80. Abraham, E.; Regan, R.F. The effects of hemorrhage and trauma on interleukin 2 production. *Arch. Surg.* 1985, 120, 1341–1344. [CrossRef]

81. Fierz, W.; Bossuyt, X. Likelihood Ratio Approach and Clinical Interpretation of Laboratory Tests. *Front. Immunol.* 2021, 12, 655262. [CrossRef] [PubMed]

82. Hanley, J.A.; McNeil, B.J. A method of comparing the areas under receiver operating characteristic curves derived from the same cases. *Radiology* 1983, 148, 839–843. [CrossRef] [PubMed]