Urinary π-glutathione S-transferase Predicts Advanced Acute Kidney Injury Following Cardiovascular Surgery

Kai-Hsiang Shu1,2,3, Chih-Hsien Wang2,4, Che-Hsiung Wu2,5, Tao-Min Huang2,6, Pei-Chen Wu2,7, Chien-Heng Lai2,4, Li-Jung Tseng2,4, Pi-Ru Tsai2,4, Rory Connolly8 & Vin-Cent Wu1,2

Urinary biomarkers augment the diagnosis of acute kidney injury (AKI), with AKI after cardiovascular surgeries being a prototype of prognosis scenario. Glutathione S-transferases (GST) were evaluated as biomarkers of AKI. Urine samples were collected in 141 cardiovascular surgical patients and analyzed for urinary alpha-(α-) and pi-(π-) GSTs. The outcomes of advanced AKI (KDIGO stage 2, 3) and all-cause in-patient mortality, as composite outcome, were recorded. Areas under the receiver operator characteristic (ROC) curves and multivariate generalized additive model (GAM) were applied to predict outcomes. Thirty-eight (26.9%) patients had AKI, while 12 (8.5%) were with advanced AKI. Urinary π-GST differentiated patients with/without advanced AKI or composite outcome after surgery (p < 0.05 by generalized estimating equation). Urinary π-GST predicted advanced AKI at 3 hrs post-surgery (p = 0.033) and composite outcome (p = 0.009), while the corresponding ROC curve had AUC of 0.784 and 0.783. Using GAM, the cutoff value of 14.7 μg/L for π-GST showed the best performance to predict composite outcome. The addition of π-GST to the SOFA score improved risk stratification (total net reclassification index = 0.47). Thus, urinary π-GST levels predict advanced AKI or hospital mortality after cardiovascular surgery and improve in SOFA outcome assessment specific to AKI.

In a physician's daily clinical practice, the evaluation of kidney function is certainly part of the comprehensive care for patients, with collection of numerous data points for association of acute kidney injury (AKI) and adverse clinical outcome1. Creatinine is certainly a useful tool for evaluation of kidney excretory function. However, it tells little about whether structural damage coexists or precedes the functional change. Urinary biomarkers have been proposed to predict and/or augment the diagnosis of AKI in order to overcome the limitations inherent to creatinine and urine output criteria, i.e., inadequate sensitivity and delayed response. A variety of serum and urinary biomarkers have been under extensive investigation and validation2 since the pioneer study series of neutrophil gelatinase-associated lipocalin (NGAL) more than a decade ago3,4. Biomarkers of AKI could be used to evaluate the severity of AKI at an earlier time point in the course of the disease, defining those with high risk of progression to more advanced kidney injury. Moreover, there is urgent need to use novel biomarkers as a guide for clinical decision-making, and augmenting selection of effective management strategies for advanced AKI, e.g. dialysis. The list of candidate biomarkers is ever-growing5, including but not limited to neutrophil gelatinase-associated lipocalin (NGAL), liver fatty acid-binding protein (L-FABP), kidney injury molecule 1 (KIM1), interleukin-18 (IL-18), cystatin C, and the combination of tissue inhibitor of metalloproteinase-2 (TIMP-2) and insulin-like growth factor binding protein 7 (IGFBP7), either serum or urine concentration. However, the predictive performance is variable, while the heterogeneous end-point definitions complicate the comparison. In order to

1Division of Nephrology, Department of Internal Medicine, National Taiwan University Hospital, Taipei, Taiwan. 2NSARF group (National Taiwan University Hospital Study Group of ARF), Taipei, Taiwan. 3Division of Nephrology, Department of Internal Medicine, Far Eastern Memorial Hospital, New Taipei City, Taiwan. 4Department of Surgery, National Taiwan University Hospital, Taipei, Taiwan. 5Division of Nephrology, Department of Internal Medicine, Mackay Memorial Hospital, Taipei, Taiwan. 6EKF Diagnostics, Cardiff, Wales, UK. Correspondence and requests for materials should be addressed to V.-C.W. (email: q91421028@ntu.edu.tw)
determine the diagnostic yield of various biomarkers, evaluating AKI after cardiovascular surgery became the prototype scenario for both development and validation purposes. Glutathione S-transferase (GST), a constitutive cytoplasmic enzyme, enriches in the renal tubular epithelial cells and is detectable in the urine when the cell integrity of renal tubules is damaged. As a scavenger of free radicals, GST may help tubular epithelial cells withstand stressful conditions. GST was originally reported to be a tubular injury marker elevated under renal insults. Further immunohistochemistry examinations demonstrated the localization of α- and π-GST in the proximal and distal renal tubules, respectively. In the era of AKI urinary biomarkers, GST gained its position in several well-known kidney injury models, although the discriminating power, mostly expressed as area under the receiver operating characteristic (ROC) curve was variable, while the optimal timeframe of marker measurement was not well established. The performance and best cut-off value of both α- and π-GST are thus evaluated prospectively in this study, in order to elucidate their respective performance as critical AKI biomarkers, in terms of dialysis or mortality.

Materials and Methods

Study Population. This study was conducted by the biomarker investigators from NSARF (the National Taiwan University Study Group on Acute Renal Failure), a multi-center, prospectively constructed database of AKI. From 2009 to 2010, patients having cardiovascular surgery, including coronary bypass and valvular operations, were enrolled prospectively from tertiary centers in northern Taiwan and an affiliated local hospital in central Taiwan. Patients with the following conditions were excluded: those who had undergone renal replacement therapy, those who were diagnosed as AKI by KDIGO definitions during index hospitalization before operation, those who were younger than 18 years of age, had a history of renal transplantation or nephrectomy, and estimated glomerular filtration rate <30 ml/1.73 m² at the time of ICU enrollment. Availability of a baseline serum creatinine level 7 to 180 days prior to the index hospital admission was mandatory. The institutional research ethic review board of National Taiwan University Hospital approved the data collection of this study (201105040RC). This research was carried out in accordance with the approved guidelines. Written informed consent was obtained from subjects.

Clinical Data Collection. Relative clinical information was obtained, including demographic data, intervention procedures, and comorbidity status. Urine output and creatinine levels were recorded at each time point after operation as detailed in the study protocol. Post-operative inotropic agent use was quantified as per inotropic equivalents. Operation related parameters included operation method, clamping time of aorta, and cardiopulmonary bypass time. Lengths of total admission and ICU admission were also recorded. Disease severity was evaluated as Sequential Organ Failure Assessment (SOFA) score.

Sample Collection. Urine samples were obtained in the ICU at 3, 6, 9, 12, and 24 hours post-surgery. Other laboratory examinations were performed as indicated clinically. Urine specimens were collected by standardized procedure, centrifuged within 1 hour and the sediments discarded. The urine samples, collected in separate polypropylene tubes containing sodium azide, were stored at −80°C until required. Each specimen was centrifuged at 800 g at 4°C for 5 minutes, and the supernatant was collected for an ELISA assay.

Biomarker Measurements. The urinary α-GST and π-GST levels were determined by EKF Diagnostics Human Alpha and Pi GST EIA Test Kits (product BIO91 and BIO85, respectively). The results were expressed in μg/mL. The inter- and intra-assay coefficient of variation for α-GST was 6.3% and 2.7%, respectively, and for π-GST, 8.6% and 3.1%, respectively. The lower limit of detection for α-GST and for π-GST was 0.3 μg/mL. Assays were completed as described by the manufacturer’s protocol, and each measurement was performed in duplicate. Urine creatinine levels were measured using the Jaffe assay, with standardization to isotope dilution mass spectrometry (IDMS)-traceable reference.

Outcome Definitions. Clinical endpoint was defined as the occurrence of advanced AKI (stage 2 or 3, specified as the Kidney Disease: Improving Global Outcomes KDIGO classification). Both urine and creatinine criteria applied. The baseline creatinine measurement was defined as pre-admission baseline creatinine, where available, or the first creatinine measurement in the emergency department. The composite outcome of developing stage 2 or 3 AKI and in-hospital mortality was also analyzed. Patients were followed until the time of death or hospital discharge, whichever earlier.

Statistics Analyses. All analyses were performed with SPSS software, version 19 (IBM, Armonk, NY) and R software. To compare continuous variables, a two-tailed unpaired t test was utilized; for categorical variables, the χ² or Fisher’s exact test were applied. To measure performance of urinary GST, a receiver-operating characteristic (ROC) curve was generated, with further calculation of the area under the curve (AUC). To assess the performance of the biomarker over the clinical model, we calculated the improvement in Harrell C statistic, accounting for censoring in survival biomarker concentrations, and transformed using natural logarithm if necessary before adding individually to the clinical model.

To examine the effect of AKI on various time-dependent GST variables over the sequential changes of time, marginal linear regression models were fitted to these repeatedly measured responses using the generalized estimating equation (GEE) method. In fitting logistic regression model using the GEE method, the first-order autocorrelation was chosen as the working correlation structure, and then the robust standard error was chosen to compute the p value. Standardized regression coefficients and their 95% confidence intervals (CI) were calculated. All the univariate significant and non-significant relevant covariates (listed in Table 1) and some of their interactions were put on the variable list to be selected. The significance levels for entry (SLE) and for stay (SLS).
were set to 0.15 for being conservative. Then, with the aid of substantive knowledge, the best candidate final logistic regression model was identified manually by dropping the covariates with p value > 0.05 one at a time until all regression coefficients were significantly different from 0.

In order to display the implications of GST for individual patients, a generalized additive model (GAM) (with spline) incorporating the subject-specific (longitudinal) random effects was plotted and adjusted for other clinical parameters to predict the outcomes. Simple and multiple generalized additive models (GAMs) were fitted to detect nonlinear effects of continuous covariates and identify appropriate cut-off point(s) for discretizing continuous covariates, if necessary, during the stepwise variable selection procedure. We defined the optimal cut-off value as log odd equals to zero. The vgam function (with the default values of smoothing parameters) of the VGAM package was used to fit GAMs for the binary responses in R.

Moreover, the ability of π-GST to more accurately stratify individuals into higher or lower risk categories (reclassification) was investigated by net reclassification improvement (NRI) and integrated discrimination improvement (IDI). An increase in NRI was calculated in a model with sequential organ failure assessment score (SOFA) and π-GST, compared with SOFA risk factors alone. We reclassified the patients who died from all-causes or who developed advanced AKI using 0–8%, 8–30%, and >30% for the risk categories. A p-value less than 0.05 was considered significant.

Results

Clinical Characteristics. A total of 141 patients receiving cardiovascular surgery were eligible for the study, with subsequent time-varying sample collection. Thirty-eight patients suffered from AKI among 141 subjects (26.9%), with 26 (18.4%) having stage 1 AKI. Seven patients (5.0%) developed stage 2 AKI, and 5 patients (3.5%) progressed to stage 3. The clinical characteristics of patients with and without advanced AKI (defined as KDIGO stage 2 or 3) are described in Table 1. The underlying comorbidities were comparable in both groups. Advanced AKI patients had lower hemoglobin and albumin levels. Mean serum creatinine level and estimated glomerular
filtration rate were not different between patients with or without advanced AKI. The in-hospital mortality rate in this cohort was 5.7%. The advanced AKI group was highlighted by higher clinical severity measured by the 24-hour SOFA score, where this critical condition translated into a longer ICU stay and higher in-hospital mortality rate. Those without any AKI had a mean π-GST level of 35.9 μg/L at 3 hours after operation, while the mean in those with any AKI was 131.1 μg/L (p = 0.084 by two-tailed t test).

Model Prediction of Stage 2 or 3 AKI. The respective π-GST levels are shown in Fig. 1, grouped as with/without advanced AKI (KDIGO stage 2 or 3), at various post-operative timing points and with/without normalization to urinary creatinine. There was a trend toward π-GST elevation post operation in the advanced AKI group (by GEE, p < 0.001) but the difference disappeared if creatinine-adjusted π-GST was analyzed. Table 2 summarizes the AUCs for ROC curves generated for the diagnosis of advanced AKI by π-GST. The AUCs for π-GST at 3 hours and 9 hours post surgery were 0.784 (95% CI 0.673–0.895, p < 0.005) and 0.767 (95% CI 0.610–0.924, p = 0.005), respectively. At 3 hours, patients without advanced AKI had a mean π-GST level of 59.1 μg/L, while the mean in those with advanced AKI was 88.8 μg/L (p = 0.467 by two-tailed unpaired t test). Figure 2 illustrates the corresponding ROC curves for π-GST.

Compared to the positive result of π-GST, α-GST levels failed to show diagnostic accuracy by ROC curve or statistical significance between advanced AKI and non-advanced AKI groups by GEE model.

In the multivariable risk model, the independent risk factors to stage 2 or 3 AKI were urinary π-GST level at 3 hrs post surgery, inotropic equivalents, hemoglobin, and BMI (Table 3). The generalized additive model (GAM) plot was generated thereafter and demonstrated the positive correlation between increased urinary π-GST levels at 3 hrs post surgery and the risk of developing advanced AKI (Fig. 3). After adjusting the nonlinear effects of the variables, π-GST = 16.5 μg/L (ln [π-GST (μg/L)] = 2.8) independently predicted postoperative stage 2 or 3 AKI.
The corresponding sensitivity and specificity were 75.0% and 68.2%, with positive and negative predictive values (PPV and NPV) being 18.0% and 96.7%, respectively.

In the present study, we also examined the performance of adjusted GST concentration, i.e. urinary GST concentration divided by urinary creatinine concentration. Nonetheless, adjustment by urinary creatinine failed to improve the diagnostic accuracy, although with a borderline significance for prediction (for urinary \( \pi \)-GST divided by urinary creatinine \{adjusted \( \pi \)-GST\} at 3 hrs post surgery, AUC of ROC curve 0.675, 95% CI 0.546 to 0.804, \( p < 0.05 \)).

Table 3. Multivariable risk model for advanced AKI or composite outcome (advanced AKI or in-hospital death). AKI, acute kidney injury; BMI, body mass index, GST, glutathione S-transferase, CI, confidence interval. *at 3 hours post surgery.

![Figure 2. ROC curves for urinary \( \pi \)-GST levels predicting advanced AKI. Curves for different time points labeled as 3 hours [T3], 6 hours [T6], 9 hours [T9], 12 hours [T12], and 24 hours post-surgery [T24].](image)

![Figure 3. GAM plot for the probability of advanced AKI for urinary \( \pi \)-GST levels at T3 (3 hours post surgery). The model incorporates the subject-specific (longitudinal) random effects, expressed as the logarithm of the odd (logit). The probability of outcome events was constructed with \( \pi \)-GST level and was centered to have an average of zero over the range of the data as constructed with the GAM. Log \([\pi \text{-GST (\mu g/L)}]\) = 2.8 was an independent factor predicting postoperative stage 2 or 3 AKI.](image)

The corresponding sensitivity and specificity were 75.0% and 68.2%, with positive and negative predictive values (PPV and NPV) being 18.0% and 96.7%, respectively.

In the present study, we also examined the performance of adjusted GST concentration, i.e. urinary GST concentration divided by urinary creatinine concentration. Nonetheless, adjustment by urinary creatinine failed to improve the diagnostic accuracy, although with a borderline significance for prediction (for urinary \( \pi \)-GST divided by urinary creatinine \{adjusted \( \pi \)-GST\} at 3 hrs post surgery, AUC of ROC curve 0.675, 95% CI 0.546 to 0.804, \( p < 0.05 \)).

**Model Prediction of Composite Outcome of Stage 2 or 3 AKI or In-hospital Mortality.** Figure 4 depicts the urinary \( \pi \)-GST levels for those with/without composite outcome, i.e., advanced AKI of KDIGO stage.
2 or 3 or in-hospital mortality. The π-GST levels differ significantly in the two aforementioned groups by the analysis of GEE (over time, \( p < 0.001 \)). The AUCs (Table 2) of ROC curves (Fig. 5) were remarkable for the significant predictive power of π-GST levels at various time points: AUC 0.783 at 3 hrs post surgery (95% CI 0.682 to 0.884, \( p < 0.005 \)), AUC 0.675 for data at 6 hrs post surgery (95% CI 0.512 to 0.837, \( p < 0.05 \)), and AUC 0.763 at 9 hrs post surgery (95% CI 0.626 to 0.900, \( p < 0.005 \)). At 3 hours, patients without composite outcome had a mean π-GST level of 54.3 μg/L, while the mean in those with composite was 123 μg/L (\( p = 0.231 \) by two-tailed unpaired t test).

For cases of advanced AKI, the adjustment of urinary creatinine level failed to increase the diagnostic accuracy of urinary π-GST. Neither α-GST nor creatinine-adjusted α-GST levels were able to differentiate those with/without composite events.

Table 3 demonstrates the multivariable model generated for predicting the composite outcome of advanced AKI or death by π-GST. Independent variables were applied to generate the GAM graph (Fig. 6), in which π-GST levels at 3 hrs post surgery were plotted against the log of the odd of the probability for composite outcome. The cutoff value in the GAM model, i.e., π-GST = 14.7 μg/L, \( \ln [\pi\text{-GST} (\mu g/L)] = 2.7 \), translated into a sensitivity of 73.3% and a specificity of 66.7%, while PPV and NPV in our studied population were 20.8% and 95.5%, respectively.

In relation to composite outcome, we examined reclassification for detection of advanced AKI or death. The addition of π-GST to the SOFA score led to a significant increase in risk stratification (total NRI = 0.47; 95% CI, 0.22 to 0.73; \( p < 0.001 \)). The majority of this effect came from those with death or dialysis events (NRI event = 0.29; 95% CI, 0.05 to 0.52; \( p < 0.001 \)), whereas the NRI non-event was 0.19 (95% CI, 0.10 to 0.28; \( p = 0.38 \)). Similarly, the total integrated discrimination improvement (IDI) was significant at 0.16 (95% CI, 0.07 to 0.24; \( p < 0.001 \)).
During AKI periods34. Moreover, due to the significant augmentation of tubular creatinine secretion under the
creatinine level to be in a steady state. Contradictory to this, the serum creatinine is not under steady state
ratio to the urine creatinine level reflects the hope to extrapolate the value to 24-hour excretion amount, requiring
accuracy to predict progression to severe outcome, such as death or advanced AKI. These results may be useful in
progression to severe outcomes. Our results indicate that addition of

tubulation enzyme 34,36. This could be valuable for clinical use to identify the timing regarding the
contrast agents in the kidneys 34,36. This could be valuable for clinical use to identify the timing regarding the
hypoperfusion, with early release even after mild injury 35. Due to its rapid increase in urinary concen-
tration, its applicability might be more appropriate in a setting that monitors the renal toxic effects of drugs and
renal support pending recovery, since the culprit cannot usually be known for sure in clinical practice. This
situation is also supported by the ability of

AKI at 3 hrs post surgery (AUC<sup>60</sup> mL/min/1.73 m<sup>2</sup> (AUC 0.61) and oliguria (AUC 0.72)<sup>15</sup>. The serial GST level change may be a trajectory of
hospital mortality.

**Discussion**

In the present study, we evaluated the proximal tubule enzyme α-GST and distal tubule enzyme π-GST as urini-
ary biomarkers for AKI prediction. After cardiovascular surgery, the π-GST level at 3 hrs post surgery showed a
higher discrimination power than α-GST level to predict advanced AKI and/or in-hospital mortality. We further
identified the optimal cutoff value for π-GST to predict advanced AKI and composite outcome according to GAM
plot, where the log odds for the outcome equals zero. The addition of π-GST to the SOFA score could lead to a
significant improvement in outcome stratification.

Considering the analysis for stage 2 or 3 AKI, significant AUCs were obtained by using π-GST level at 3 hrs
post surgery. This more severe kidney injury may reflect early extensive tubular dysfunction in the distal renal
tube or damage at AKI occurrence, further amplifying the discrepancy to milder forms of AKI or those without
clinical kidney injury. This supports the view that these markers are for early diagnosis of AKI and for “early”
diagnosis only. Susantitaphong et al. reported GST could predict advanced AKI at 2 hrs following cardiopulmo-
nary bypass (AUC = 0.76)<sup>16</sup>. In Koyner’s report<sup>34</sup>, π-GST could predict the progression to stage 3 AKI at 6 hr post
surgery in 123 adults undergoing cardiac surgery (AUC = 0.78). We further showed π-GST predicts stage 2 and 3
AKI at 3 hrs post surgery (AUC = 0.78). Collectively, these results supported the finding that GST is very sensitive
to renal hypoperfusion, with early release even after mild injury<sup>35</sup>. Due to its rapid increase in urinary concen-
tration, its applicability might be more appropriate in a setting that monitors the renal toxic effects of drugs and
contrast agents in the kidneys<sup>34,36</sup>. This could be valuable for clinical use to identify the timing regarding the
diagnosis initiation and, identifying those patients with AKI who have severe tubular injury and will require fur-
ther renal support pending recovery, since the culprit cannot usually be known for sure in clinical practice. This
notion is also supported by the ability of π-GST to predict dialysis requirement in patients with baseline eGFR
<60 mL/min/1.73 m<sup>2</sup> (AUC 0.61) and oliguria (AUC 0.72)<sup>15</sup>. The serial GST level change may be a trajectory of
renal function change.

The better predictive value of π-GST compared to α-GST for established AKI in this study is in accordance
with other work<sup>15,16</sup>. π-GST showed better performance than α-GST regarding advanced AKI and in hospital
mortality after cardiovascular surgery. α-GST was frequently elevated even though the subject did not meet the
diagnostic criteria of AKI. Whether this reflects false-positive results or a signal to subclinical AKI is not clear.
Similar discrepancy was noted in a sepsis-induced AKI study<sup>12</sup>. Renal GST was upregulated transcriptionally in
nephrotoxin-induced cell model<sup>37</sup> and AKI rat model<sup>38</sup>, demonstrating not only an injury model but also with
possible mechanistic relevance<sup>29</sup>. Whether the distal marker of π-GST outperformed the proximal marker of
α-GST in representing the underlying mechanism of AKI remained to be elucidated, but it may reflect the nature
of different etiologies of severe kidney insults.

In real world practice, we lack markers to determine which patients will develop advanced AKI and disease
progression to severe outcomes. Our results indicate that addition of π-GST to the SOFA score could improve the
accuracy to predict progression to severe outcome, such as death or advanced AKI. These results may be useful in
designing therapeutic trials for AKI by targeting patients most likely to benefit.

In our data, adjustment to urinary creatinine concentration only further attenuated the diagnostic power. The
ratio to the urine creatinine level reflects the hope to extrapolate the value to 24-hour excretion amount, requiring
the creatinine level to be in a steady state. Contradictory to this, the serum creatinine is not under steady state
during AKI periods<sup>34</sup>. Moreover, due to the significant augmentation of tubular creatinine secretion under the
setting of decreased GFR, the presence of tubular damage further impaired creatinine secretion. Thus, the role of
creatinine adjustment for urinary biomarkers is uncertain<sup>40</sup>. While it may occasionally enhance the diagnostic
efficiency for various biomarkers in rats, π-GST was characteristic of high sensitivity in our study and holds partic-
ular promise to facilitate improved clinical decision in patients at risk of developing advanced AKI.
The strengths of this study include multi-center design and homogenous (post-cardiovascular surgery) patient population. By enrolling patients with known baseline creatinine levels, we also reduced misclassification of early AKI status. This finding further substantiates a biological link between GST and clinical disease. Improving risk prediction may facilitate participation in AKI trials, enhance patient care in the setting of critical illness, and guide patient decision making and counseling with regard to treatment plans. However, the limitation of the current study is that we did not design to compare or combine these biomarkers with other novel AKI biomarkers. The existing data serves as a developmental set for the diagnostic test however, external validation in other populations should be considered. In the era of multiple biomarker development, panels of biomarkers, including GST, should be validated for additional value in combining more than one biomarker. This may help us to find better roles for GSTs and other biomarkers. Finally, although the performance of biomarkers under the setting of cardiovascular surgery is fairly well defined, extrapolating to other clinical scenarios may not be as successful. The generalizability to medical rather than surgical patients remains an ongoing issue.

Conclusion

In summary, our data demonstrated that \( \pi \text{-GST} \) may be a valuable tool for early diagnosis of advanced AKI and prediction of prognosis after cardiovascular surgery, whereas \( \alpha \text{-GST} \) did not. \( \pi \text{-GST} \) could improve the predictive ability of SOFA for composite outcome. Further external validation in a larger population is necessary to confirm this finding.

References

1. Lafrance, J. P. & Miller, D. R. Acute kidney injury associates with increased long-term mortality. *J Am Soc Nephrol* 21, 345–352 (2010).
2. Siew, E. D., Ware, L. B. & Ikizler, T. A. Biological markers of acute kidney injury. *J Am Soc Nephrol* 22, 810–820 (2011).
3. Mishra, J. Identification of Neutrophil Gelatinase-Associated Lipocalin as a Novel Early Urinary Biomarker for Ischemic Renal Injury. *J Am Soc Nephrol* 14, 2534–2543 (2003).
4. Mishra, J. et al. Neutrophil gelatinase-associated lipocalin (NGAL) as a biomarker for acute renal injury after cardiac surgery. *The Lancet* 365, 1231–1238 (2005).
5. Schley, G. et al. Comparison of Plasma and Urine Biomarker Performance in Acute Kidney Injury. *PloS One* 10, e0145042 (2015).
6. Belcher, J. M., Edelstein, C. L. & Parikh, C. R. Clinical applications of biomarkers for acute kidney injury. *Am J Kidney Dis* 57, 930–940 (2011).
7. Parikh, C. R. et al. Postoperative biomarkers predict acute kidney injury and poor outcomes after adult cardiac surgery. *J Am Soc Nephrol* 22, 1748–1757 (2011).
8. Feinfield, D. A., Fleischner, G. M. & Arians, I. M. Urinary ligandin and glutathione-S-transferase in gentamicin-induced nephrotoxicity in the rat. *Clin Sci (Lond)* 61, 123–125 (1981).
9. Branten, A. J., Mulder, T. P., Peters, W. H., Assmann, K. J. & Wetzels, J. F. Urinary excretion of glutathione S transferase alpha and pi in patients with proteinuria: reflection of the site of tubular injury. *Nephron* 85, 120–126 (2000).
10. Harrison, D. J., Kharbanda, R., Cunningham, D. S., McLelean, L. I. & Hayes, J. D. Distribution of glutathione S-transferase isoenzymes in human kidney: basis for possible markers of renal injury. *J Clin Pathol* 42, 624–628 (1989).
11. Shaw, M. The use of histologically defined specific biomarkers in drug development with special reference to the glutathione S-transferases. *Cancer Biomark* 1, 69–74 (2005).
12. Walshe, C. M., Odejayi, F., Ng, S. & Marsh, B. Urinary glutathione S-transferase as an early marker for renal dysfunction in patients admitted to intensive care with sepsis. *Crit Care Resusc* 11, 204–209 (2009).
13. Yavuz, I. et al. Importance of urinary measurement of glutathione S-transferase in renal dysfunction patients after on- and off-pump coronary artery bypass surgery. *Thorac Cardiovasc Surg* 57, 125–129 (2009).
14. McMahon, B. A., Koyner, J. L. & Murray, P. T. Urinary glutathione S-transferases in the pathogenesis and diagnostic evaluation of acute kidney injury following cardiac surgery: a critical review. *Curr Opin Crit Care* 16, 550–555 (2010).
15. Seabra, V. F. et al. Urinary alpha-GST and pi-GST for prediction of dialysis requirement or in-hospital death in established acute kidney injury. *Biomarkers* 16, 709–717 (2011).
16. Susantitaphong, P. et al. Urinary alpha- and pi-glutathione S-transferases for early detection of acute kidney injury following cardiopulmonary bypass. *Biomarkers* 18, 331–337 (2013).
17. Pencina, M. J., D’Agostino, R. B., Sr., D’Agostino, R. B., Jr. & Vasan, R. S. Evaluating the added predictive ability of a new marker: from area under the ROC curve to reclassification and beyond. *Stat Med* 27, 157–172, discussion 207–112 (2008).
18. Huang, T. M. et al. Preoperative Proteinuria Predicts Adverse Renal Outcomes after Coronary Artery Bypass Grafting. *J Am Soc Nephrol* 22, 156–163 (2011).
19. Wu, V. C. et al. Sustained low-efficiency dialysis versus continuous veno-venous hemofiltration for postsurgical acute renal failure. *Am J Surg* 199, 466–476 (2009).
20. Wu, V. C. et al. Early renal replacement therapy in patients with postoperative acute liver failure associated with acute renal failure: effect on postoperative outcomes. *J Am Coll Surg* 205, 266–276 (2007).
21. Wu, V. C. et al. Risk factors of early redialysis after weaning from postoperative acute renal replacement therapy. *Intensive Care Med* 34, 101–108 (2008).
22. Shiao, C. C. et al. Late initiation of renal replacement therapy is associated with worse outcomes in acute kidney injury after major abdominal surgery. *Crit Care* 13, R171 (2009).
23. Kidney Disease: Improving Global Outcomes (KDIGO) Glomerulonephritis Work Group. KDIGO Clinical Practice Guideline for Glomerulonephritis. *Kidney Int Suppl* 2, 139–274 (2012).
24. Wu, V. C. et al. The hemodynamic effects during sustained low-efficiency dialysis versus continuous veno-venous hemofiltration for uremic patients with brain hemorrhage: a crossover study. *J Neurosurg* 119, 1288–1295 (2013).
25. Vincent, J. L. et al. The SOFA (Sepsis-related Organ Failure Assessment) score to describe organ dysfunction/failure. On behalf of the Working Group on Sepsis-Related Problems of the European Society of Intensive Care Medicine. *Intensive Care Med* 22, 707–710 (2006).
26. R Core Team. *R: A language and environment for statistical computing*, R Foundation for Statistical Computing, Vienna, Austria. http://www.R-project.org/ (2014).
27. Zeiger, S. L. & Liang, K. Y. Longitudinal data analysis for discrete and continuous outcomes. *Biometrics* 42, 121–130 (1986).
28. Woo, M. J., Reiter, J. P. & Karr, A. F. Estimation of propensity scores using generalized additive models. *Stat Med* 27, 3805–3816 (2008).
29. Wu, V. C. et al. Endothelial progenitor cells in primary aldosteronism: a biomarker of severity for aldosterone vasculopathy and prognosis. *J Clin Endocrinol Metab* 96, 3175–3183 (2011).
30. Hin, L. Y., Lau, T. K., Rogers, M. S. & Chang, A. M. Dichotomization of continuous measurements using generalized additive modelling—application in predicting intrapartum caesarean delivery. Stat Med 18, 1101–1110 (1999).
31. Yee, T. W. VGAM: Vector generalized linear and additive models. R package, version 1.0–1, http://CRAN.R-project.org/package= VGAM (2014).
32. Yee, T. W. Reduced-rank vector generalized linear models with two linear predictors. Computational Statistics and Data Analysis 71, 889–902 (2014).
33. Yee, T. W. & Wild, C. J. Vector generalized additive models. Journal of the Royal Statistical Society, Series B, Methodological 58, 481–493 (1996).
34. Koyner, J. L. et al. Urinary biomarkers in the clinical prognosis and early detection of acute kidney injury. Clin J Am Soc Nephrol 5, 2154–2165 (2010).
35. Ralib, A. M. et al. The clinical utility window for acute kidney injury biomarkers in the critically ill. Crit Care 18, 601 (2014).
36. de Geus, H. R., Fortrie, G., Betjes, M. G., van Schaik, R. H. & Groeneveld, A. B. Time of injury affects urinary biomarker predictive values for acute kidney injury in critically ill, non-septic patients. BMC Nephrol 14, 273 (2013).
37. Zhu, H., Zhang, L., Amin, A. R. & Li, Y. Coordinated upregulation of a series of endogenous antioxidants and phase 2 enzymes as a novel strategy for protecting renal tubular cells from oxidative and electrophilic stress. Exp Biol Med (Maywood) 233, 753–765 (2008).
38. Swain, A. et al. Urinary biomarkers in hexachloro-1:3-butadiene-induced acute kidney injury in the female Hanover Wistar rat: correlation of alpha-glutathione S-transferase, albumin and kidney injury molecule-1 with histopathology and gene expression. J Appl Toxicol 31, 366–377 (2011).
39. Alge, J. L. & Arthur, J. M. Biomarkers of AKI: A Review of Mechanistic Relevance and Potential Therapeutic Implications. Clin J Am Soc Nephrol 10, 147–155 (2015).
40. Tonomura, Y., Uehara, T., Yamamoto, E., Torii, M. & Matsubara, M. Decrease in urinary creatinine in acute kidney injury influences diagnostic value of urinary biomarker-to-creatinine ratio in rats. Toxicology 290, 241–248 (2011).
41. Prowle, J. R. et al. Combination of biomarkers for diagnosis of acute kidney injury after cardiopulmonary bypass. Ren Fail 37, 408–416 (2015).

Acknowledgements
Assays were performed by Rory Connolly using EKF Diagnostics kits, Human Alpha GST (BIO91) and Human Pi GST EIA (BIO85).

Author Contributions
K.-H.S. drafted the manuscript. C.-H.W., C.-H.W., T.-M.H., P.-C.W. and R.C. helped revise the manuscript. C.-H.L., L.-J.T. and P.-R.T. participated in the acquisition of specimens and clinical data. R.C. carried out the immunoassays. V.-C.W. contributed to the design and coordination of study, performed the statistical analysis, and the revision of manuscript.

Additional Information
Competing financial interests: The human alpha GST and piGST assays were performed by Rory Connolly, R&D Scientist at EKF Diagnostics. No other potential conflicts of interest relevant to this article were reported.

How to cite this article: Shu, K.-H. et al. Urinary π-glutathione S-transferase Predicts Advanced Acute Kidney Injury Following Cardiovascular Surgery. Sci. Rep. 6, 26335; doi: 10.1038/srep26335 (2016).

This work is licensed under a Creative Commons Attribution 4.0 International License. The images or other third party material in this article are included in the article’s Creative Commons license, unless indicated otherwise in the credit line; if the material is not included under the Creative Commons license, users will need to obtain permission from the license holder to reproduce the material. To view a copy of this license, visit http://creativecommons.org/licenses/by/4.0/