

A Drug Development Tool for Trial Enrichment in Patients With Autosomal Dominant Polycystic Kidney Disease



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Introduction: Total kidney volume (TKV) is a promising imaging biomarker for tracking and predicting the natural history of patients with autosomal dominant polycystic kidney disease.

Methods: A drug development tool was developed by linking longitudinal TKV measurements to the probability of a 30% decline of estimated glomerular filtration rate (eGFR) or end-stage renal disease. Drug development tools were developed based on observational data collected over multiple decades for an eGFR decline and end-stage renal disease in 641 and 866 patients with autosomal dominant polycystic kidney disease, respectively.

Results: The statistical association between predicted TKV at the time of a 30% decline of eGFR and that at the time of end-stage renal disease were both highly significant (P < 0.0001). The drug development tool was applied to demonstrate the utility of trial enrichment according to prespecified baseline TKV, age, and eGFR as enrollment criteria in hypothetical clinical trials. Patients with larger TKV (\geq 1000 ml) displayed steeper slopes of hazard, which translated into a higher risk of a 30% decline of eGFR within each baseline age (< or \geq 40 years) or baseline eGFR (< or \geq 50 ml/min per 1.73 m²) subgroups.

Discussion: These results suggest that, when eGFR is preserved, patients with larger TKV are more likely to progress to a 30% decline of eGFR within the course of a clinical trial, whereas eGFR and age displayed limited predictive value of disease progression in early disease. Pharmaceutical sponsors and academic investigators are encouraged to prospectively employ the above drug development tool to optimize trial designs in patients with autosomal dominant polycystic kidney disease.

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A utosomal dominant polycystic kidney disease (ADPKD) is the most common hereditary kidney disease. There is an increasing body of evidence demonstrating that the kidneys of patients with ADPKD progressively increase in size from birth throughout life, and the clinical symptoms and signs of ADPKD including hypertension, gross hematuria, flank and abdominal pain, and declining glomerular filtration rate (GFR) are associated with increased kidney volume.^{1,2} The clinical course of ADPKD is marked by a decades-long period of stable kidney function, as measured by GFR, despite the relentless expansion of total kidney volume (TKV) due to growth of cysts. There is evidence in the literature from both animal and human studies to support TKV as a prognostic endpoint for use in clinical trials for ADPKD.^{1,3–5}

Medical imaging is gaining an important role in clinical trials. This has been driven by significant improvements in medical imaging technology and quality and the increasing need to leverage these technologies to reduce drug development time. The US Food and Drug Administration's (FDA's) Critical Path Initiative acknowledges the potential value of imaging as a research tool in drug development. In addition, the recent FDA

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Guidance for Industry on the Qualification of Drug Development Tools (http://www.fda.gov/downloads/ drugs/guidancecomplianceregulatoryinformation/guida nces/ucm230597.pdf) acknowledges that biomarkers may assess many different types of biological characteristics or parameters including radiographic or other imaging-based measurements. The Polycystic Kidney Disease Outcomes Consortium (PKDOC) has identified TKV as an imaging biomarker that is most relevant for tracking and predicting the natural history of ADPKD. The PKDOC has developed the first-ever Clinical Data Interchange Standards Consortium therapeutic-area-specific data standards for ADPKD to allow for the mapping and integration of observational data from both patient registries and Consortium for Radiologic Imaging Studies of Polycystic Kidney Disease (CRISP) cohort studies into a common dataset.⁶ This rich and robust dataset has allowed the PKDOC to develop a statistical model linking longitudinal TKV measurements in concert with age and estimated glomerular filtration rate (eGFR) to the probability of a 30% decline of eGFR or end-stage renal disease (ESRD) and ultimately leverage the model as a drug development tool (DDT) for trial enrichment in patients with ADPKD.

METHODS

The PKDOC has developed the first-ever Clinical Data Interchange Standards Consortium data standards for ADPKD to allow for the mapping and pooling of available data into a common dataset that has been used to support the regulatory qualification of TKV by the FDA and European Medicines Agency.⁶ This common dataset is one of the largest ever datasets of patients with ADPKD, with a total of 2355 patients. The PKDOC dataset includes 1182 subjects who have at least 2 images for the measurement of TKV taken at least 6 months apart. Observational, prospectively obtained data from 5 sources were aggregated into a common database in a standard Clinical Data Interchange Standards Consortium structure: (i) University of Colorado - Denver, (ii) Mayo Clinic, (iii) Emory University, (iv) CRISP1, and (v) CRISP2. The content of these databases is described elsewhere and in Supplementary Table S1.⁶

Construction of DDT

Joint models linking longitudinal TKV measurements, in combination with other prognostic factors, were constructed for the probability of a 30% decline of eGFR as well as ESRD.^{7–9} In a first step, linear mixedeffect models with a random intercept were used to fit ln-transformed TKV values for all datasets.⁹ Patients with at least 2 TKV measurements separated by at least 6 months were included in the analysis. Baseline TKV was defined as the first TKV measurement for a subject, irrespective of modality including computerized tomography (CT), magnetic resonance imaging (MRI), or ultrasound (US), whereas baseline age was the age associated with the first TKV measurement. The population for the assessment of a 30% decline of eGFR included 1182 patients with at least 2 images of TKV.

For the time-to-event model of a 30% decline of eGFR, the association parameter between predicted TKV at the time of a 30% decline of eGFR was modeled using a piecewise linear model (12 knots). Baseline eGFR was calculated from the first valid serum creatinine measurement on or within 365 days of the baseline TKV. Because many of the creatinine measurements were made using older colorimetric methods, eGFR was derived using the original Modification of Diet in Renal Disease equation for creatinine methods that are not calibrated to an isotope dilution mass spectrometry reference method.¹⁰ For creatinine methods calibrated to an isotope dilution mass spectrometry reference method, the isotope dilution mass spectrometrytraceable Modification of Diet in Renal Disease study equation was used to derive eGFR.¹¹ Because the goal of this project was to determine whether TKV, along with other prognostic factors such as baseline age and eGFR, can accurately predict the risk of a decline of eGFR, only endpoint measurements that occurred after the first baseline TKV measurement were considered. A subsequent (confirmatory) measurement within any timeframe was required to confirm that the original 30% decline was not transient.^{12,13} Data rules are summarized in Supplementary Table S2. Of the 1182 patients with at least 2 images of TKV, a total of 541 patients did not have eGFR at baseline or a confirmatory eGFR (Supplementary Figure S1). As a result, a total of 641 patients were used in the joint analysis of TKV and the probability of a 30% decline of eGFR. Baseline eGFR and baseline age were statistically significant parameters for joint modeling of TKV and the probability of a 30% decline of eGFR.

For the ESRD model, the association parameter between predicted TKV at the time of ESRD was modeled using a Weibull model. ESRD was defined as a patient with either dialysis or transplant. Of the 1182 patients with at least 2 images of TKV, a total of 316 patients did not have eGFR at baseline or a missing date of ESRD. Data rules are summarized in Supplementary Table S2 and data flow is displayed in Supplementary Figure S1. As a result, a total of 866 patients were used in the joint analysis of TKV and the probability of ESRD. Baseline age, baseline eGFR, and interaction terms were found statistically significant for the ESRD model. Data utilized for analysis came from registries as opposed to clinical trials and patients were seen on an irregular basis. For example, the endpoint of an eGFR decline could only be defined after a minimum of 2 measurements of TKV. There is a slight possibility of bias due to missing data, but this is thought to be unlikely because data missingness was due to the randomness of the available registry data.

Data Splitting and Interval Validation

Cross-validation was performed using a 5-fold or 10fold cross-validation approach.¹⁴ Data were split into 5 parts with roughly equal number of subjects. Splitting was stratified to maintain a similar proportion of patients from the CRISP and registry datasets in the reference and test datasets. Each fold served as a test dataset in the following steps, whereas the rest of the data consisted of the training dataset (i.e., the 4 other folds). The joint model (including prognostic factors) was fitted to the training dataset (4/5 of the folds). Prediction of disease outcomes for the test dataset was performed by simulating from the joint model using each individual prognostic factor (longitudinal TKV data, baseline age, and baseline eGFR) from the test dataset. Model-based predicted probabilities in the test dataset were compared with observed disease outcomes in the test dataset. Predictive performance of the joint model was assessed by computing descriptive statistics of observed versus predicted probability of disease outcomes (precision and accuracy). The above steps were repeated for each fold.

Software

Joint modeling and cross-validation was performed using the JM package in R 3.0.2 (64-bit).

RESULTS

Baseline Characteristics

A total of 1182 patients with ADPKD and at least 2 TKV data collected over 30 years of follow-up were available in the database. Figure 1 provides a highlevel visual perspective of the mix of imaging modalities for subjects with at least 2 TKV measurements from 5 clinical sites. Several imaging modalities have been utilized to determine TKV in patients with ADPKD. These included US as determined using the ellipsoid method, MRI, and CT scan.



Years

Figure 1. Time span and modalities of images for subjects with 2 or more images (n = 1182). CRISP, Consortium for Radiologic Imaging Studies of Polycystic Kidney Disease; CT, computerized tomography; MRI, magnetic resonance imaging; TKV, total kidney volume; US, ultrasound.

Baseline characteristics of subjects with ADPKD included in the joint model analysis are presented in Table 1. A total of 541 patients were excluded for the 30% decline of eGFR endpoint because of missing baseline eGFR or if the endpoint was observed between the first and second TKV measurement. For the 30% decline of the eGFR model, a total of 641 patients with at least 2 TKV measurements separated by at least 6 months were included in the analysis 192 of whom presented a 30% decline of eGFR (30.0%). The populations used in modeling the probability of a 30% decline of eGFR consisted of 1140 subjects with mean baseline TKV, age, and eGFR of 1141 ml, 33.2 years, and 84.3 ml/min per 1.73 m², respectively. For ESRD, a total of 316 patients were excluded because of missing baseline eGFR or a missing date of ESRD. For the ESRD model, a total of 866 patients with at least 2 TKV measurements separated by at least 6 months were included

 Table 1. Baseline characteristics of subjects with ADPKD included in the joint model analysis

	Population for the analysis of 30 decline of eGFR $(N = 641)$	Population for the analysis of ESRD (N = 866)				
Baseline characteristics	Mean (SD)					
TKV (ml)	1141 (1186.1)	1214 (1249.4)				
Age (yr)	33.2 (15.64)	34.7 (15.91)				
Age, N (%)						
0 to <20 yr	142 (22.2%)	175 (20.2%)				
20 to $<$ 40 yr	272 (42.4%)	351 (40.5%)				
40 to $<\!\!60$ yr	193 (30.1%)	289 (33.4%)				
60 to $<\!\!80$ yr	33 (5.1%)	49 (5.7%)				
80 to 100 yr	1 (0.2%)	2 (0.2%)				
eGFR (ml/min per 1.73 m ²) CKD stages	84.3 (35.54)	80.9 (36.70)				
1	227 (35.4%)	281 (32.4%)				
2	263 (41.0%)	335 (38.7%)				
3	127 (19.8%)	203 (23.4%)				
4	24 (3.7%)	46 (5.3%)				
5	0 (0.0%)	1 (0.1%)				
	Count	t (%)				
Sex						
Male	245 (38.2)	333 (38.5)				
Female	396 (61.8)	533 (61.5)				
Race						
White	595 (92.8)	798 (92.1)				
Black	29 (4.5)	37 (4.3)				
Other	17 (2.7)	31 (3.6)				
Genotype						
PKD1	386 (60.2)	470 (54.3)				
PKD2	45 (7.0)	51 (5.9)				
Missing	196 (30.6)	328 (37.9)				
No mutation detected	14 (2.2)	17 (2.0)				

ADPKD, autosomal dominant polycystic kidney disease; CKD, chronic kidney disease; eGFR, estimated glomerular filtration rate; ESRD, end-stage renal disease; PKD, polycystic kidney disease; TKV, total kidney volume. in the analysis 147 of whom developed ESRD (17.0%). Baseline characteristics of subjects used for the joint modeling of TKV and ESRD were consistent with those observed for the 30% eGFR decline analysis. Overall, baseline characteristics were consistent to those reported in randomized controlled clinical trials of ADPKD.^{15,16}

Construction of DDT

Joint models were developed to simultaneously assess longitudinal TKV values and the probability of avoiding a 30% decline of eGFR and ESRD. In a first step, linear mixed-effect models were used to fit lntransformed TKV values for all datasets. Imaging modalities (US, MRI, and CT) were pooled together. Final parameters derived with the joint model are presented in Table 2. For the time-to-event models of a 30% decline of eGFR, the association parameter between predicted TKV at the time of a 30% decline of eGFR was modeled using a piecewise linear model (12 knots). Baseline eGFR and baseline age were statistically significant parameters for the 30% decline of the eGFR model. For the ESRD model, the association parameter between predicted TKV at the time of ESRD was modeled using a Weibull model. Baseline age, baseline eGFR, and interaction terms were found to be statistically significant for the ESRD model. For a 30% decline of eGFR, the rate of growth of TKV was 0.0516 (corresponding to 5.16% per year). The association between the predicted TKV at the time of a 30% decline of eGFR was highly statistically significant (P < 0.0001). Baseline eGFR was statistically significant (P = 0.0014), whereas the effect of baseline age was not statistically significant (P = 0.9586). Baseline age was retained in the final joint model because of the statistically significant interaction with baseline TKV and baseline eGFR and to allow flexibility in exploring trial

Table 2. Final parameters of the joint model for the probability of avoiding a 30% decline of eGFR and ESRD

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Endpoints	Parameters	Coeffici	ent (SE)	z value	P value	
30% decline	TKV model					
of eGFR	Intercept	6.6684	(0.0348)	191.9	< 0.0001	
	Rate of growth Event model	0.0516	(0.0024)	21.8	< 0.0001	
	Association $(TKV \rightarrow event)$	0.8457	(0.1204)	7.02	< 0.0001	
	Baseline eGFR	0.0101	(0.0032)	3.19	0.0014	
	Baseline age	0.0004	(0.0077)	0.0519	0.9586	
ESRD	TKV model					
	Intercept	6.712	(0.0301)	223.0	< 0.0001	
	Rate of growth	0.0560	(0.0020)	28.4	< 0.0001	
	Event model					
	Association	1.2365	(0.1631)	7.58	<0.0001	
	$(TKV \rightarrow event)$					
	Baseline eGFR	-0.0391	(0.0135)	-2.90	0.0037	
	Baseline age	0.0355	(0.0169)	2.10	0.0357	
	eGFR:Age	-0.0012	0.0003	-3.65	0.0003	

 $\mathsf{eGFR},$ estimated glomerular filtration rate; ESRD, end-stage renal disease; TKV, total kidney volume.

enrichment strategies according to this baseline characteristic.

For the ESRD model, the rate of growth of TKV was 0.056 (corresponding to 5.60% per year). The association between the predicted TKV value at the time of ESRD (Weibull function) was highly statistically significant (P < 0.0001). Baseline eGFR (P = 0.0037) and baseline age (P = 0.0357) were also statistically significant in addition to a statistically significant interaction between baseline eGFR and age.

Adequacies of TKV models for the 30% decline of eGFR and ESRD endpoints are presented in Figure 2. Overall, subject-level model-predicted versus observed log-transformed TKV and standardized residuals were very well fitted with the model (Figure 2a). This was demonstrated by the locally weighted scatter-plot smoother (LOESS, red dashed line) between subjectlevel prediction and observed values which were very close to the identity curve (i.e., black line). Furthermore, the distribution of standardized subjectlevel residuals was homogeneously distributed



Figure 2. Subject-level model-predicted versus (a) observed log transformed TKV and (b) standardized residuals for a 30% decline of eGFR and ESRD endpoints. eGFR, estimated glomerular filtration rate; ESRD, end-stage renal disease.

around 0, suggesting no residual effect (Figure 2b). Adequacies of event models, linking individual subject level TKV data to a 30% decline of eGFR and ESRD endpoints over follow-up time, are presented in Figure 3. Time to a 30% decline of eGFR and ESRD endpoints were well fitted with the joint model as demonstrated by the model-predicted value (red line) relative to observed probabilities (black lines) and 95% confidence intervals. A 5-fold data splitting method was used to evaluate the predictive performance of the final joint models. The predictive performance of the model was assessed by deriving mean prediction error and root-mean-square error values of individual observed versus predicted values for a 30% decline of eGFR and ESRD endpoints. Results of the crossvalidation are presented in Supplementary Tables S3 and S4. Mean prediction error values for a 30% decline of eGFR over 1, 3, 5, and 10 years of follow-up were 0.0849%, 1.34%, 1.70% and -1.35%, respectively. Mean prediction error values for avoiding ESRD over 1, 3, 5, and 10 years of follow-up were 0.228%, 1.62%, 5.24% and 13.5%, respectively.

Application of DDT for Trial Enrichment

The DDT was applied to demonstrate the utility of trial enrichment according to prespecified baseline TKV, age, and eGFR as enrollment criteria. The joint model was used to perform trial simulations to ultimately derive the predicted probabilities of a 30% decline of eGFR according to specific inclusion criteria of baseline TKV (\geq 1000 ml), baseline age (< or \geq 40 years), and baseline eGFR (< or \geq 50 ml/min per 1.73 m²). The predicted probabilities of avoiding a 30% decline of eGFR according to baseline TKV are presented in Figure 4. Patients with larger TKV (\geq 1000 ml) displayed steeper slopes of hazard, which translated into a higher risk of a 30% decline of eGFR within each



Figure 3. Model-predicted versus observed probabilities for avoiding a 30% decline of eGFR and ESRD over time. eGFR, estimated glomerular filtration rate; ESRD, end-stage renal disease.



Figure 4. Trial enrichment example—mean (95%) predicted probabilities for avoiding a 30% decline of eGFR as a function of baseline TKV categories. eGFR, estimated glomerular filtration rate; TKV, total kidney volume.

baseline age (< or \geq 40 years) or baseline eGFR (< or \geq 50 ml/min per 1.73 m²) subgroups. Patients with TKV \geq 1000 ml, eGFR <50 ml/min per 1.73 m²,

and age \geq 40 years displayed the highest risk of a 30% decline in eGFR. These results suggest that patients with larger TKV are more likely to progress to a 30%



Figure 5. Trial enrichment example—mean (95%) predicted probabilities for avoiding a 30% decline of estimated glomerular filtration rate (eGFR) as a function of baseline eGFR categories.

decline of eGFR, independent of other baseline characteristics.

In Figure 5, the predicted probabilities of avoiding a 30% decline of eGFR in patients with baseline eGFR < 50 ml/min per 1.73 m² were similar to those in patients with baseline eGFR \geq 50 ml/min per 1.73 m² in all subgroups, with the exception of patients \geq 40 years of age and baseline TKV \geq 1 liter (lower-right panel). The predicted probabilities of avoiding a 30% decline of eGFR according to baseline age are presented in Figure 6. Overall, predicted probabilities of avoiding a 30% decline of eGFR in patients < 40 years were similar to those in patient \geq 40 years of age.

Numerical values of predicted probabilities of a 30% decline of eGFR over time as a function of TKV, eGFR, or age are presented in Table 3. At year 3, median predicted probabilities of a 30% decline in eGFR in patients with TKV \geq 1000 and <1000 ml were 9.88% and 5.09%, respectively. At year 5, median predicted probabilities of a 30% decline in eGFR in patients with TKV \geq 1000 and <1000 ml were 20.8% and 11.3%, respectively. Finally, median predicted probabilities of a 30% decline in eGFR at year 10 in patients with TKV \geq 1000 and <1000 ml were 53.1% and 34.3%, respectively. Conversely, baseline eGFR and age did not provide a discrimination in probabilities of a 30% decline in eGFR. Numerical values of

predicted probabilities of a 30% decline of eGFR by TKV within each category of eGFR and age are presented in Table 4. Patients with larger TKV (\geq 1000 ml) displayed steeper slopes of hazard, which translated into a higher risk of a 30% decline of eGFR within each baseline age (< or \geq 40 years) or baseline eGFR (< or \geq 50 ml/min per 1.73 m²) subgroups. These results suggest that, when eGFR is preserved, patients with larger TKV are more likely to progress to a 30% decline of eGFR and age displayed limited predictive value of disease progression in early disease.

DISCUSSION

The current work presents evidence that supported the regulatory qualification of TKV as a prognostic biomarker according to its Context of Use, based on definitions set forth by the FDA's Guidance on the Qualification Process for Drug Development Tools (Guidance Compliance Regulatory Information), and the European Medicines Agency's "Qualification of novel methodologies for drug development."^{17,18} The DDT consists of a joint model linking the probability of a 30% decline of eGFR and ESRD as a function of longitudinal TKV. Joint modeling is considered as the gold standard method for assessing the effect of longitudinal time-varying covariates (e.g., TKV) in a



Figure 6. Trial enrichment example—mean (95%) predicted probabilities for avoiding a 30% decline of estimated glomerular filtration rate (eGFR) as a function of baseline age categories.

 Table 3. Trial enrichment example—tabulation of predicted

 probabilities of observing a 30% decline of eGFR as a function of

 baseline TKV, eGFR, or age

Predicted probabilities of a 30% decline of eGFR						
	Baseline TKV (ml)		Baseline eGFR (ml/min per 1.73 m ²)		Baseline age (yr)	
Time (yr)	≥1000	<1000	<50	≥50	≥40	≥40
1	1.75%	0.85%	1.41%	1.20%	1.32%	1.29%
2	3.97%	1.97%	3.20%	2.74%	3.01%	2.93%
3	9.88%	5.09%	8.03%	6.94%	7.61%	7.36%
4	15.8%	8.40%	12.9%	11.3%	12.3%	11.9%
5	20.8%	11.3%	17.1%	15.0%	16.3%	15.8%
6	25.2%	14.1%	20.8%	18.4%	20.0%	19.3%
7	34.8%	20.3%	29.0%	26.1%	28.0%	27.1%
8	42.9%	26.2%	36.2%	33.0%	35.0%	34.1%
9	48.0%	30.1%	40.7%	37.4%	39.5%	38.6%
10	53.1%	34.3%	45.3%	42.0%	44.1%	43.2%

These results suggest that, when eGFR is preserved, patients with larger TKV are more likely to progress to a 30% decline of eGFR within the course of a clinical trial, whereas eGFR and age displayed limited predictive value of disease progression in early disease. eGFR, estimated glomerular filtration rate: TKV, total kidnev volume.

time-to-event analysis of clinical endpoint.^{7–9} The DDT was applied to demonstrate the utility of trial enrichment according to prespecified baseline TKV, age, and eGFR as enrollment criteria in hypothetical clinical trials.

Several imaging modalities have been utilized to determine TKV in patients with ADPKD. The current joint modeling effort included US as determined using the ellipsoid method, MRI, and CT scan. The accuracy and precision of US in assessing TKV in ADPKD compared with MRI was previously determined as part of the CRISP studies.¹⁹ Overall, results from this study demonstrated a high correlation between US and MRI volumes to US (0.88 and 0.89, respectively). For the ellipsoid method, US TKV was 11% greater than MRI TKV, with an SD of 34%. For

the direct method, the mean difference was 9%, with an SD of 27%. In addition, previous sensitivity using multivariate Cox methods demonstrated that the imaging modality of TKV (US vs. MRI/CT) did not have an impact on the prediction of a 30% decline of eGFR or progression to ESRD.²⁰

The measured GFR was not available for the vast majority of patients. Although the measured GFR has frequently been thought of as a "gold standard" for clinical trial outcomes, recent analyses indicate that eGFR is comparable to mGFR for association with adverse outcomes such as ESRD and cardiovascular events.²¹

Data utilized for analysis came from registries as opposed to clinical trials and patients were seen on an irregular basis. There is a slight possibility of bias due to missing data, but this is thought to be unlikely because data missingness was due to the randomness of the available registry data.

Overall, the qualification of TKV as an imaging biomarker for tracking and predicting the natural history of ADPKD represents a significant, innovative step forward to establishing the commitment of health authorities, clinicians, and patients to address the unmet needs for this debilitating condition, thereby encouraging researchers and the pharmaceutical industry to develop promising new therapies for these patients. When used to assess the effect of interventional therapy, TKV will need to be assessed in the context of the investigational agent's mechanism of action and its potential for nephrotoxicity or other side effects that could impact renal function. The DDT tools (model codes) are available at https://c-path.org/. Pharmaceutical sponsors and academic investigators are encouraged to prospectively employ the above joint model to optimize trial designs in patients with ADPKD

Table 4. Trial enrichment example—tabulation of mean predicted probabilities of a 30% decline of eGFR as a function of baseline TKV within eGFR and age categories

		Predicted probabilities of a 30% decline of eGFR						
		TKV ≥ 1000 ml			TKV < 1000 ml			
	eGFR < 50 (ml/min per 1.73 m ²)		$eGFR \ge 50$ (ml/min per 1.73 m ²)		eGFR < 50 (ml/min per 1.73 m ²)		$eGFR \ge 50$ (ml/min per 1.73 m ²)	
Time (yr)	Age \geq 40 yr	Age $<$ 40 yr	Age \geq 40 yr	Age $<$ 40 yr	Age \geq 40 yr	Age $<$ 40 yr	Age \geq 40 yr	Age $<$ 40 yr
1	2.10%	1.80%	1.60%	1.50%	0.90%	0.80%	0.83%	0.85%
2	4.70%	4.10%	3.70%	3.40%	2.10%	2.00%	1.91%	1.96%
3	11.6%	10.1%	9.30%	8.50%	5.30%	5.10%	4.93%	5.01%
4	18.5%	16.1%	14.8%	13.7%	8.70%	8.40%	8.24%	8.31%
5	24.3%	21.1%	19.5%	18.2%	11.6%	11.2%	11.1%	11.3%
6	29.2%	25.6%	23.6%	22.3%	14.5%	13.9%	13.8%	14.0%
7	40.0%	35.2%	32.9%	31.0%	20.9%	19.9%	20.1%	20.4%
8	48.4%	43.6%	41.0%	38.7%	26.9%	25.6%	25.9%	26.2%
9	53.7%	48.7%	46.0%	43.6%	31.0%	29.4%	29.8%	30.3%
10	59.0%	53.8%	51.0%	48.5%	35.2%	33.4%	33.9%	34.6%

eGFR, estimated glomerular filtration rate; TKV, total kidney volume.

similar to other models that have been made publicly available by the FDA. $^{22-24}$

DISCLOSURE

Support for this project was provided by the PKD Foundation. Industry supporters of the PKDF include Otsuka Pharmaceuticals, Amgen, Pfizer Inc., Sanofi-Genzyme, and Novartis. RDP is an investigator and member of the Steering Committee for several Otsuka studies on ADPKD; he has been a consultant to Sanofi-Genzyme, Novartis, Vertex, and Mitsubishi Tanabe Pharmaceuticals. ABC is an investigator and member of the Steering Committee for several Otsuka studies on ADPKD; she has been a consultant to Sanofi-Genzyme. VET is an investigator and Chair of the Steering Committee for several Otsuka studies on ADPKD and is an investigator in a clinical trial for ADPKD sponsored by Novartis Pharmaceuticals. FSC is an employee of Otsuka Pharmaceuticals, which is developing treatments for ADPKD. JFM and M-SM are consultants paid by Pharsight Canada. All the other authors declared no competing interests.

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SUPPLEMENTARY MATERIAL

Table S1. Description of timeframe of enrollment, number

 of subjects enrolled, and process of each study/registry.

Table S2. Data rules.

 Table S3. Cross-validation of the joint model – predicted

 probabilities for avoiding a 30% worsening of eGFR.

Table S4. Cross-validation of the joint model—predicted probabilities of avoiding eSRD.

Figure S1. Flowchart for data flow.

Supplementary material is linked to the online version of the paper at http://www.kireports.org/.

REFERENCES

- Torres VE, Harris PC. Polycystic kidney disease: genes, proteins, animal models, disease mechanisms and therapeutic opportunities. J Intern Med. 2007;261:17–31.
- Wilson PD. Polycystic kidney disease. N Engl J Med. 2004;350:151–164.
- Chapman AB, Bost JE, Torres VE, et al. Kidney volume and functional outcomes in autosomal dominant polycystic kidney disease. *Clin J Am Soc Nephrol.* 2012;7:479–486.
- Irazabal MV, Rangel LJ, Bergstralh EJ, et al. Imaging classification of autosomal dominant polycystic kidney disease: a simple model for selecting patients for clinical trials. J Am Soc Nephrol. 2015;26:160–172.
- Grantham JJ, Cook LT, Torres VE, et al. Determinants of renal volume in autosomal-dominant polycystic kidney disease. *Kidney Int.* 2008;73:108–116.
- Perrone RD, Neville J, Chapman AB, et al. Therapeutic area data standards for autosomal dominant polycystic kidney disease: a report from the Polycystic Kidney Disease Outcomes Consortium (PKDOC). Am J Kidney Dis. 2015;66:583–590.
- Tsiatis AA, Davidian M. Joint modeling of longitudinal and timeto-event data: an overview. *Statistica Sinica*. 2004;14:809–834.
- Sweeting MJ, Thompson SG. Joint modelling of longitudinal and time-to-event data with application to predicting abdominal aortic aneurysm growth and rupture. *Biom J.* 2011;53:750–763.
- 9. Cai T, Pepe MS, Zheng Y, et al. The sensitivity and specificity of markers for event times. *Biostatistics*. 2006;7:182–197.
- National Kidney Foundation. K/DOQI clinical practice guidelines for chronic kidney disease: evaluation, classification, and stratification. Am J Kidney Dis. 2002;39(Suppl 1):S1–S266.
- 11. Miller WG. Estimating glomerular filtration rate. *Clin Chem Lab Med.* 2009;47:1017–1019.
- Coresh J, Turin TC, Matsushita K, et al.; CKD Prognosis Consortium. Decline in estimated glomerular filtration rate and subsequent risk of end-stage renal disease and mortality. *JAMA*. 2014;311:2518–2531.
- Chang WX, Asakawa S, Toyoki D, et al. Predictors and the subsequent risk of end-stage renal disease—usefulness of 30% decline in estimated GFR over 2 years. *PLoS One.* 2015;10: e0132927.
- Breiman L, Spector P. Submodel selection and evaluation in regression. The X-random case. Int Stat Rev. 1992;60:291–319.
- Schrier RW, Abebe KZ, Perrone RD, et al., for the HALT-PKD Trial Investigators. Blood pressure in early autosomal dominant polycystic kidney disease. *New Engl J Med*. 2014;371:2255–2266.
- Serra AL, Poster D, Kistler AD, et al. Sirolimus and kidney growth in autosomal dominant polycystic kidney disease. *N Engl J Med.* 2010;363:820–829.

- Food and Drug Administration. Qualification Process for Drug Development Tools, 2014. Available at: http://www.fda. gov/downloads/drugs/guidancecomplianceregulatoryinfor mation/guidances/ucm230597.pdf. Accessed December 30, 2016.
- European Medicines Agency. Qualification of novel methodologies for drug development: guidance to applicants, 2014. Available at: http://www.ema.europa.eu/docs/en_GB/document_ library/Regulatory_and_procedural_guideline/2009/10/WC500004 201.pdf. Accessed December 30, 2016.
- O'Neill WC, Robbin ML, Bae KT, et al. Sonographic assessment of the severity and progression of autosomal dominant polycystic kidney disease: the Consortium of Renal Imaging Studies in Polycystic Kidney Disease (CRISP). Am J Kidney Dis. 2005;46:1058–1064.
- Perrone RD, Mouksassi M-S, Romero K, et al. Total kidney volume is a prognostic biomarker of renal function decline and progression to end-stage renal disease in patients with

autosomal dominant polycystic kidney disease. *Kidney Int. Rep.* 2017;2:442–450.

- Ku E, Xie D, Shlipak M, et al.; CRIC Study Investigators. Change in measured GFR versus eGFR and CKD outcomes. *J Am Soc Nephrol.* 2016;27:2196–2204.
- Food and Drug Administration. Biomarker Qualification Program. Available at: http://www.fda.gov/Drugs/Development ApprovalProcess/DrugDevelopmentToolsQualificationProgram/ ucm284076.htm. Accessed December 30, 2016.
- Guidance for Industry. Qualification of Biomarker Total Kidney Volume in Studies for Treatment of Autosomal Dominant Polycystic Kidney Disease. Available at: http://www.fda.gov/ downloads/Drugs/GuidanceComplianceRegulatoryInformation/ Guidances/UCM458483.pdf. Accessed December 30, 2016.
- 24. Romero K, de Mars M, Frank D, et al. The coalition against major diseases: developing tools for an integrated drug development process for Alzheimer's and Parkinson's diseases. *Clin Pharmacol Ther.* 2009;86:365–367.