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 Research Laboratory for INtelligent system

 in Agro-hydrological Monitoring and

 Management (INAMM)

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Simulation of Runoff Changes

based-on Land Use/Cover in Lam Pachi Basin by CA-Markov and SWAT Models

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Outlines

- Introduction
- Study area: Lam Phachi Basin, western Thailand
- Methodology and Results
 - Land use projection by CA-Markov model
 - Runoff simulation by SWAT model



Introduction

- Surface runoff is a key component in hydrological cycle
- The quantification of runoff indicates water resources availabilities in a basin and provides important information for water management
- Land use/land cover are one of main factors affecting runoff processes
- This work is part of a research program
 - Integrated Water Resource Management of Lam Pachi Basin
 - Land use change, runoff, water use, drought, flood, erosion



Objectives

 This work aimed to projecting the effects of land use/land cover change (LULCC) on runoff yield in Lam Pachi basin, Western Thailand



Study Area: Lam Phachi Basin





Lam Pachi Basin, Thailand



- Watershed area 2543 km2
- Precipitation 1210 mm/yr (SW monsoon)
- Temperature 27.3°C (max-min: 35.9°C (Apr)- 19.8°C (Jan))





Methodology and Result Land use projection by CA-Markov model Runoff simulation by SWAT model



Methodology

- To develop CA-Markov model for spatially projection of future land use/land cover
- To develop SWAT-based hydrological model for evaluating the runoff response to the projected LULCC
- To simulate the impact of LULCC on runoff yield in the Lam Pachi basin by SWAT model.



Land Use Projection by CA-Markov

- Tool
 - MOLUSCE: Modules for Land Use Change Evaluation
 - QGIS plug-in
- Data
 - Land use data (Department of Land Development, DLD)
 - Auxiliary data



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Modules for Land Use Change Evaluation



CA/Markov model

- Convert vector land use data into raster format with grid size of 60 m
- Reclassify into 5 main group: agriculture, forest, built-up area, water, miscellaneous
- Driven forces: topography (slope), transportation (distance from road), neighbor (distance from existing built-up area)
- Transition probability matrix : land 2008 vs land 2015 (step of 7 year)
- Transitional potential modeling by logistic regression: 3 driven-force
- Future projection of land use change by Monte-Carlo Cellular Automata technique with 5 iterations: 2015, 2022, 2029, 2036, 2043, 2050





Land use map of Lam Phachi basin

Reclassified into 5 main classes





Area Changes

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Transition probability matrix or Markov matrix

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Land Use Change 2008-2015

Land Use	2008	2015	Diff.	% Area	% Area	%
	(km2)	(km2)	(km2)	2008	2015	Different
Agriculture	1092.51	1087.29	-5.22	39.04%	38.85%	-0.19%
Forest	1591.44	1536.75	-54.69	56.87%	54.91%	-1.95%
Misc.	74.57	73.75	-0.82	2.66%	2.64%	-0.03%
Built-up	18.34	75.52	57.18	0.66%	2.70%	2.04%
Water	21.75	25.31	3.56	0.78%	0.90%	0.13%



Transition Matrix

			2015		
Land use	Agriculture	Forest	Misc.	Built-up	Water
Agriculture	91.50%	1.06%	2.19%	4.95%	0.31%
Forest	3.82%	95.74%	0.11%	0.29%	0.04%
Misc.	30.82%	1.66%	63.53%	1.98%	2.01%
Built-up	14.39%	0.80%	1.79%	82.92%	0.10%
Water	5.81%	1.09%	1.31%	0.76%	91.03%





Driving factor data



Define Samples			ciulai Automata Sinu		messayes	
Method Logistic Regression		-				
Maximum iterations	100	Coef	ficients Standard	deviations P-va	lues	
Neighbourhood	1 px	₽ [1.0 → 1.0	1.0 → 2.0	1.0 → 3.0	1.0 → 4.0
Pseudo R-squared (count)	0.97400	β0	-0.604624673927	-0.147692130093	1.31562546169	1.3288893410
		β1	1.25298726961	0.315335410075	-0.159223390383	-0.415713120
	Fit model	β2	-0.883911323421	-0.74575860349	1.02770989878	0.9985011866
		β3	-0.0806739199222	1.1397055326	2.0394494421	1.0522085279
	1	β4	0.529572889642	0.236792246021	0.70042555212	1.5765403893
f(z) =		β5	-2.75658218777	-1.49387267774	-1.84376681674	-0.804887687
-(-)	$1 + e^{-z}$	<u>β6</u>	-0.567231682837	1.42400568535	1.26335541893	1.1036821508
		β7	1.33035681316	1.20308667648	1.91224280034	1.5186022789
		β8	0.54778007966	0.920446202301	0.780914727632	2.3723854043
$= \beta_0 + \beta_1 X_1$	$+ \beta_2 X_2 + \cdots + \beta_n X_n$	ρ ^{β9}	0.896591885834	-1.27470119325	-1.40251295077	-0.209603394
		β1(-0.687113849402	0.20970776532	0.885007076179	0.8772365339
		•	-1 47741060504	1 20157061901	1.71720670024	1.640040552

Transitional Potential Modeling Logistic regression



CA : Cellular Automata

Monte-Carlo Cellular Automata simulation







Projected Land Use (%)

Year	Land use				
	Agriculture	Forest	Misc.	Built-up	Water
2008	39.75%	56.34%	2.68%	0.42%	0.81%
2015	39.39%	54.35%	2.65%	2.66%	0.95%
2022	41.79%	52.78%	1.91%	2.73%	0.80%
2029	42.99%	51.87%	1.90%	2.75%	0.49%
2036	43.55%	50.84%	1.83%	2.69%	1.09%
2043	42.04%	51.27%	3.17%	2.69%	0.83%
2050	47.03%	46.74%	2.35%	2.79%	1.08%
2015-2050	+7.64%	-7.60%	-0.30%	+0.13%	+0.13%



Projected Land Use

- The result revealed that more than half of the Lam Pachi Basin covered by forest area.
- The CA-Markov model projected the future land use in the next 45 years.
 - Forest area would decrease about 7.6%
 - Agricultural area would increase at the similar rate.



Runoff Simulation by SWAT model

- Tool
 - SWAT model
- Data
 - Land use Department of Land Development (DLD)
 - + projected land use maps
 - Soil Department of Land Development (DLD)
 - DEM CGIAR-CSI SRTM 90m Digital Elevation Data
 - http://srtm.csi.cgiar.org/
 - Climate Thai Meteorological Department (TMD)
 - Streamflow Royal Irrigation Department (RID)





Observation stations



Soil texture



Model calibration and validation

- Calibration : monthly streamflow, 2010-2015
- Validation : 2005-2009

Parameter	Unit	Method	Initial Value	Range	Fitted Value
1. CN2.mgt	-	Relative	varied	-10% +10%	-9.55%
2. GWQMN.gw	mm	Absolute	1000	+0 +4000	+3582
3. SOL_AWC.sol	mm _{H2O} /mm _{soil}	Absolute	varied	+0 +0.4	+0.1455

Note:

- 1. CN2.mgt = Initial SCS runoff curve number for moisture condition II
- 2. GWQMN.gw = Threshold water level in shallow aquifer for base flow
- 3. SOL_AWC.sol = Available water capacity of the soil layer





Hydrometric	Calibration	(2010-2015)	Validation (2005-2009)		
Station	R2	NSE	R2	NSE	
K25A	0.818	0.647	0.821	0.504	
K17	0.880	0.772	0.846	0.711	
K61	0.894	0.792	N/A	N/A	
K 62	0.936	0.863	N/A	N/A	

Note: N/A = Observed data not available

Model performance indicators





Projected Basin Water Yield (mm)

	LU2015	LU2022	LU2029	LU2036	LU2043	LU2050
January	34.74	34.42	34.3	34.19	34.37	33.81
February	21.71	21.5	21.42	21.34	21.45	21.07
March	20.94	20.96	20.99	20.94	20.98	21.08
April	29.57	30.21	30.48	30.45	30.38	31.53
May	34.02	34.28	34.52	34.27	34.54	35.21
June	40.65	40.91	41.16	40.91	41.26	41.99
July	44.09	44.13	44.23	43.96	44.3	44.59
August	46.19	46.35	46.56	46.29	46.54	46.98
September	183.59	186.29	187.85	187.99	186.91	193.35
October	368.92	373.42	375.81	376.14	373.73	383.94
November	162.17	163.3	164.08	163.93	163.79	166.53
December	59 <mark>.6</mark> 4	59.16	59.11	58.85	59.27	58.64
Annual	1046.23	1054.93	1060.51	1059.26	1057.52	1078.72

Surface Runoff

- Minor changes in term of water yield at the basin scale.
- More changes could be observed at subbasin level.
 - 1. Agricultural land yielded more water than forest in high land
 - 2. Land conversion from forest to agriculture in sloped high-land resulted in increase of water yield
 - 3. Whereas the same conversion in flat low-land resulted in reduction of yield







Acknowledgement

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Thank you



