

Seasonal forecasting of water resources

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Mediterranean water scarcity



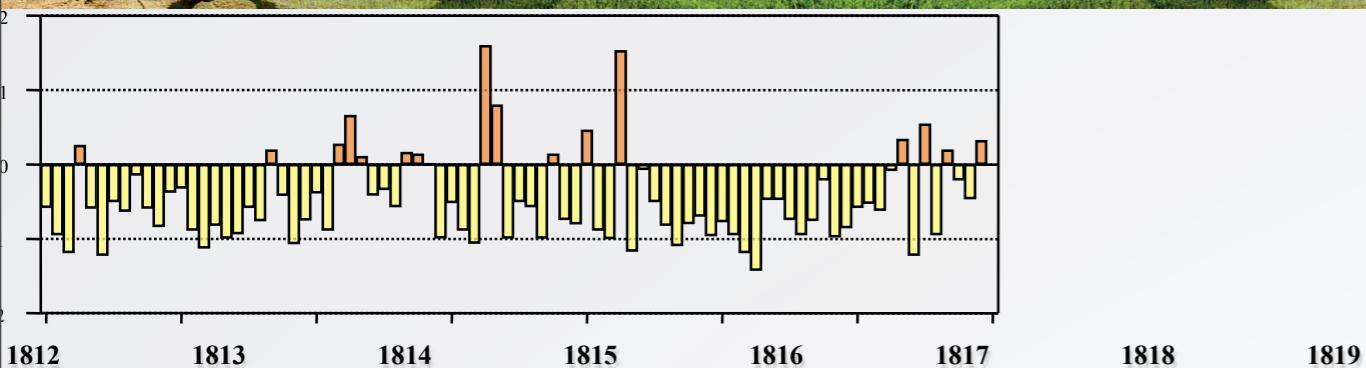
Mediterranean water scarcity

- ♦ Water scarcity is a recurrent problem in the Mediterranean



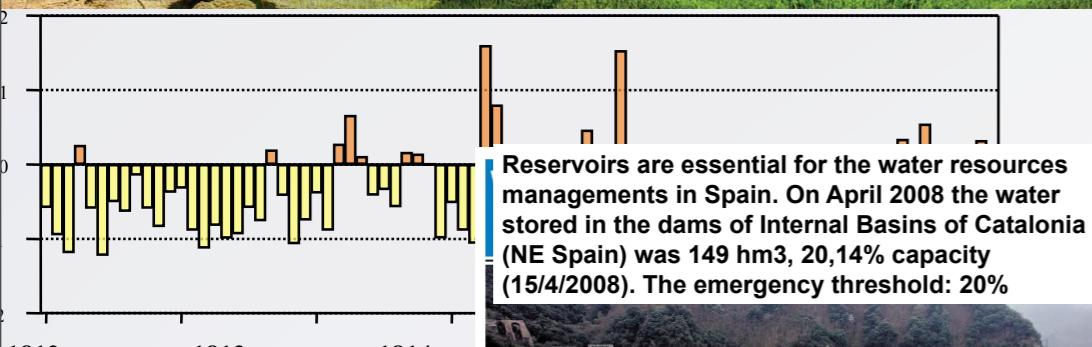
Mediterranean water scarcity

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Mediterranean water scarcity

♦ Water scarcity is a recurrent problem in the Mediterranean



1818 1819

Emergencia nacional

• El Govern califica en términos dramáticos la sequía y trata de buscar un pacto con CiU que aliviente la guerra del agua

RAMÓN SOLÁ

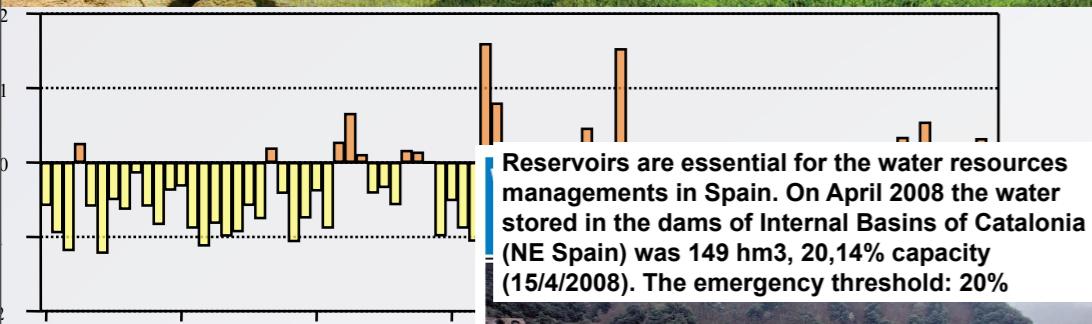
Catalunya vive un estado de "emergencia nacional" por culpa de la sequía. La Generalitat de Medi Ambient i Habitatge, presidida por Artur Mas, hizo ayer esta apelación ante el Parlament. Los habitantes del área metropolitana de Barcelona, "miserables" un encuestado

más alto nivel, entre el presidente de la Generalitat y el líder de la oposición, se quejaron de las restricciones al consumo doméstico poniendo sobre la mesa la posibilidad de imponer restricciones a los habitantes de las ciudades más grandes de Catalunya, "miserables" un encuestado

CONTINUA EN LA PÁGINA SIGUIENTE →

Mediterranean water scarcity

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Emergencia nacional

• El Govern califica en términos dramáticos la sequía y trata de buscar un pacto con CiU que aliviente la guerra del agua

RAMON SOPENA

Catalunya vive un estado de "emergencia nacional" por culpa de la sequía. La Generalitat, presidida por Artur Mas, declaró el 27 de marzo la situación que está más que justificado un acuerdo político al

más alto nivel, entre el presidente de la Generalitat y el líder de la coalición independentista "Som Nostres i Habitatges". Artur Mas, hizo ayer esta apelación en su intervención en el pleno ordinario del área metropolitana de Barcelona, "mercorde" un encuestado

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ta el país con una serie amenaza

de restricciones al consumo doméstico pendiente sobre las ca-

tas de la población

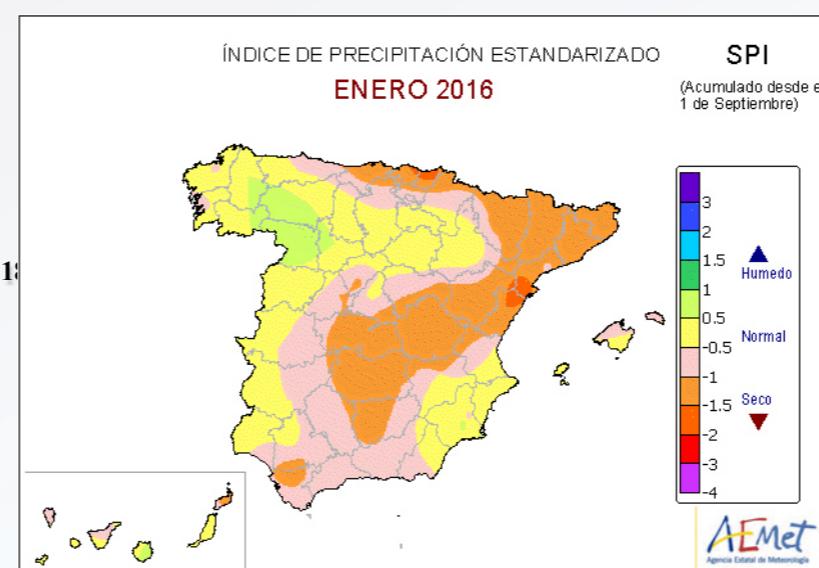
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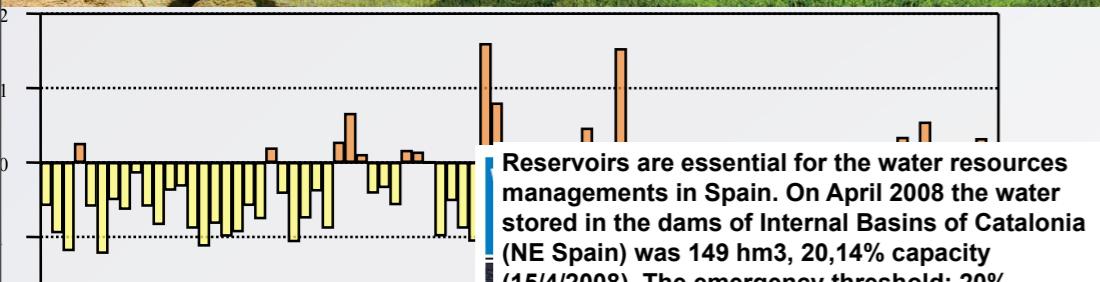
ta el problema que afro-

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Emergencia nacional

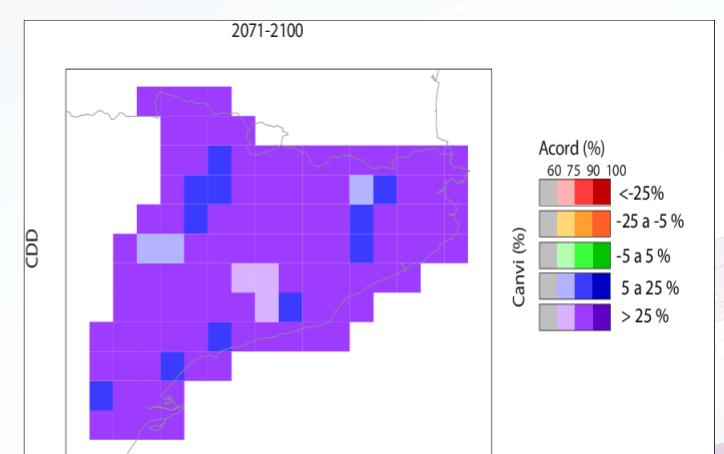
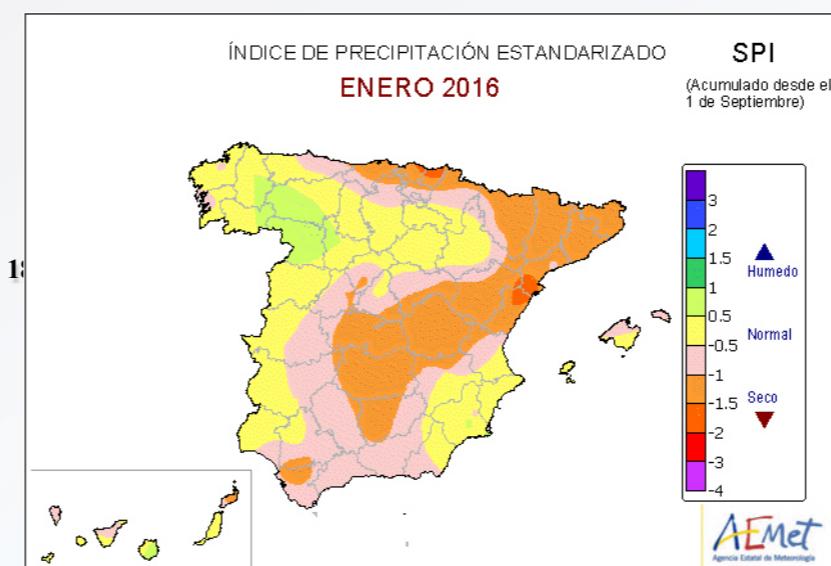
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más alto nivel, entre el presidente de la Generalitat y el líder de la coalición independentista de Mossos d'Esquadra i Habitatges, Joan Ortega. Bébas, hizo ayer esta apelación en su intervención en el pleno ordinario del área metropolitana de Barcelona, "mercorio" un encuestado

CONTINUA EN LA PÁGINA SIGUIENTE →



Mediterranean water scarcity

Mediterranean water scarcity

- ◆ Water scarcity is a recurrent problem in the Mediterranean



- ◆ Linked to the relationship between:



Mediterranean water scarcity

- ◆ Water scarcity is a recurrent problem in the Mediterranean

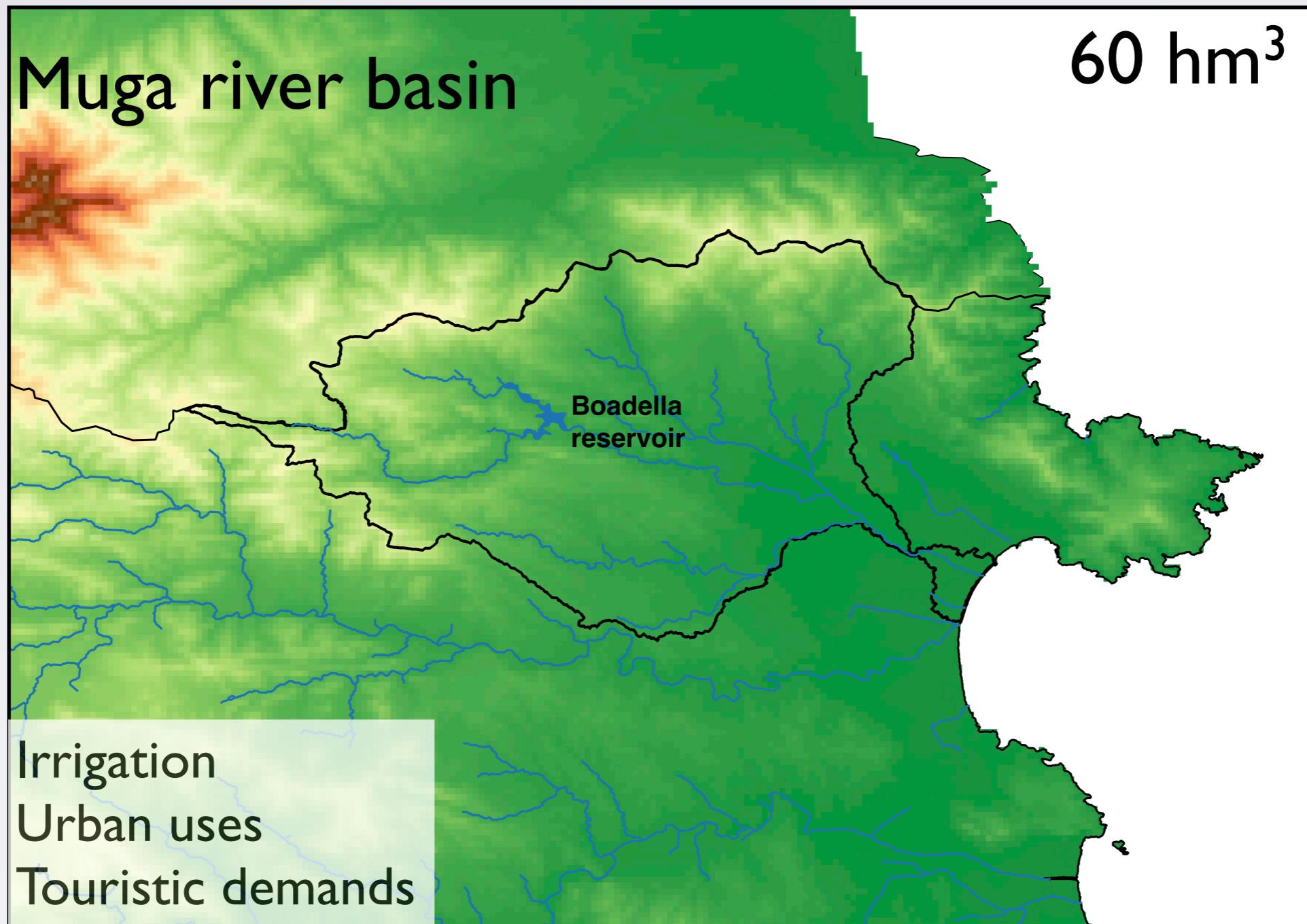
In this context foretelling the behaviour of dam **supplies** with months **in advance** can be used to optimize water management

- ◆ Linked to the relationship between

Water supplies



Boadella reservoir

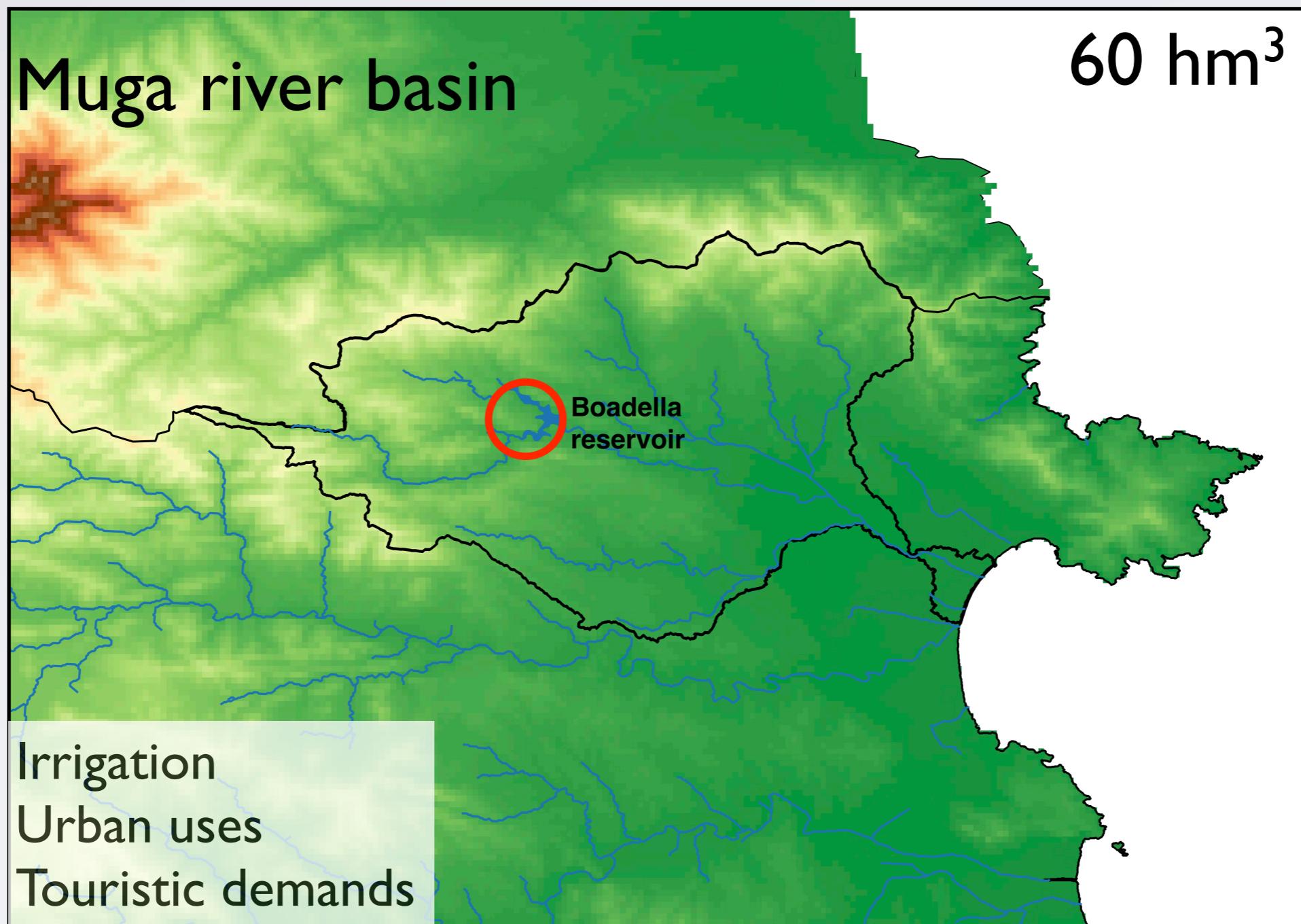


Case study description

MLR construction

Predictability

Boadella reservoir



Case study description

MLR construction

Predictability

ECMWF System 4 data

Case study description

MLR construction

Predictability

ECMWF System 4 data

- ◆ Monthly **ECMWF System-4** precipitation and temperature **15**-member ensemble re-forecasts ($0.75^\circ \times 0.75^\circ$) for the period **1981-2010**. Each initialization starts in the 1st day of each month encompassing a **7-month** time integration

Case study description

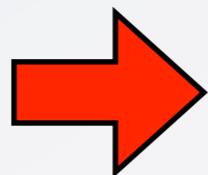
MLR construction

Predictability

ECMWF System 4 data

- ♦ Monthly **ECMWF System-4** precipitation and temperature **15**-member ensemble re-forecasts ($0.75^\circ \times 0.75^\circ$) for the period **1981-2010**. Each initialization starts in the 1st day of each month encompassing a **7-month** time integration

Interpretation example



April S4 Forecasts

Issued 1st Apr.
Lead 1 (m-0)

Issued 1st Mar.
Lead 2 (m-1)

Issued 1st Feb.
Lead 3 (m-2)

Issued 1st Jan.
Lead 4 (m-3)

Issued 1st Dec.
Lead 5 (m-4)

Issued 1st Nov.
Lead 6 (m-5)

Issued 1st Oct.
Lead 7 (m-6)

Case study description

MLR construction

Predictability

E-OBS data

Case study description

MLR construction

Predictability

E-OBS data

- ◆ **E-OBS** daily **precipitation** and **temperature** high-resolution ($0.25^\circ \times 0.25^\circ$) gridded dataset over the period **1981-2010**

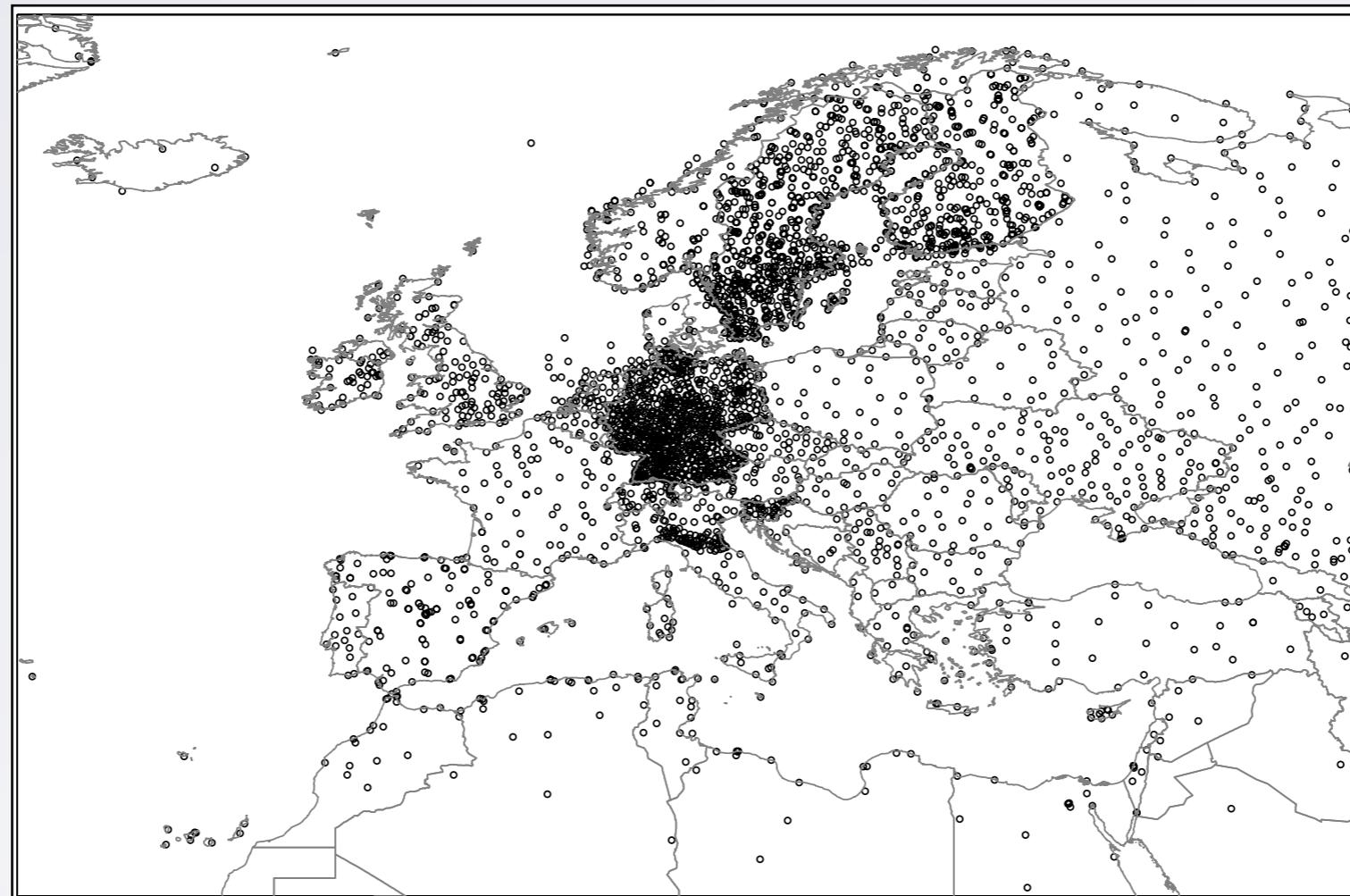
Case study description

MLR construction

Predictability

E-OBS data

- ◆ E-OBS daily **precipitation** and **temperature** high-resolution ($0.25^\circ \times 0.25^\circ$) gridded dataset over the period **1981-2010**



E-OBS v8.0 temperature station cover map for Europe.

Case study description

MLR construction

Predictability

Boadella data

Case study description

MLR construction

Predictability

Boadella data

- ◆ Monthly in-flow, out-flow and volume data **observed** at the **Boadella** reservoir for the period **1981-2010**



Boadella reservoir.

Case study description

MLR construction

Predictability

Multiple Linear Regression strategy

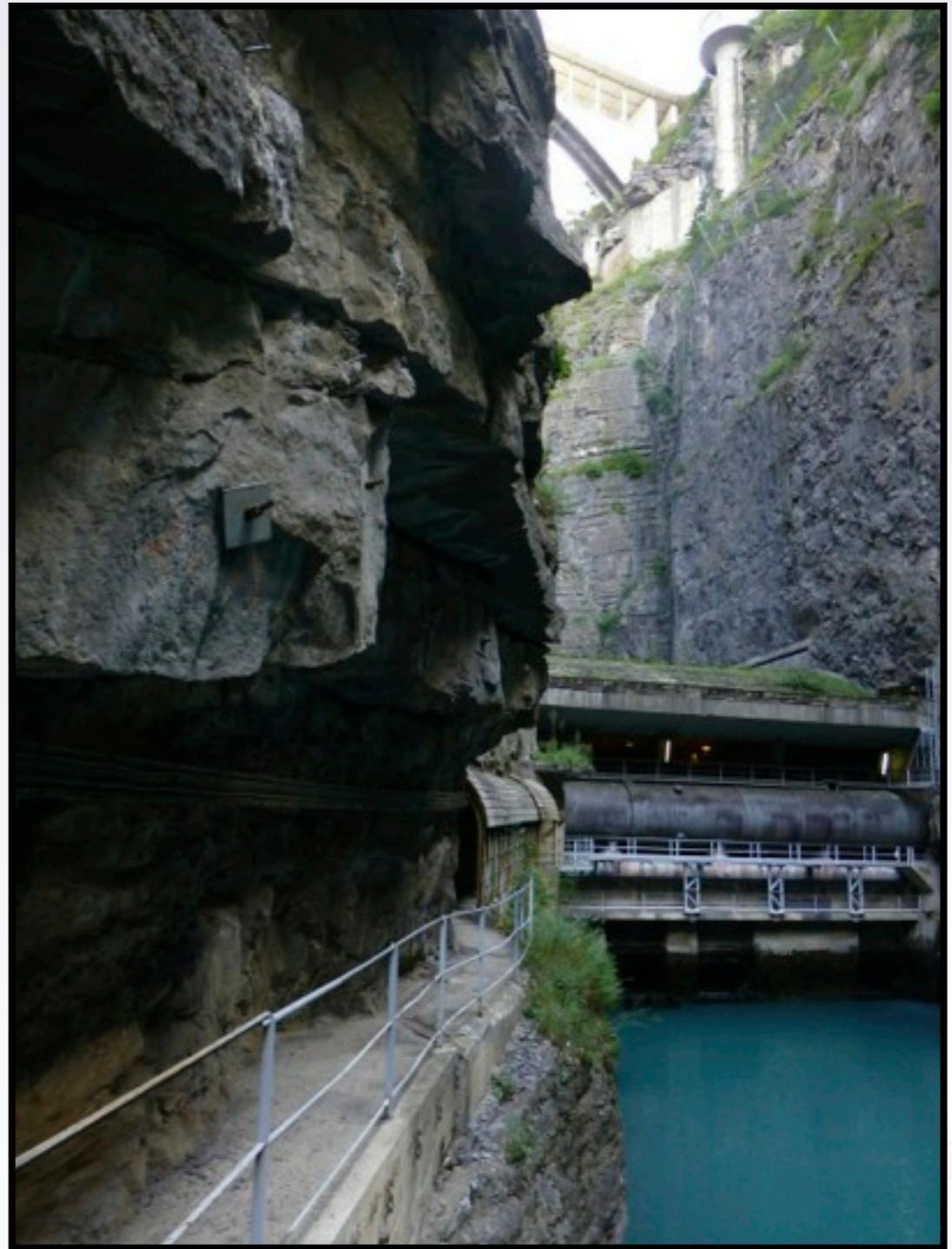
Case study description

MLR construction

Predictability

Multiple Linear Regression strategy

Predictors



Case study description

MLR construction

Predictability

Multiple Linear Regression strategy

Predictors

◆ In-flow

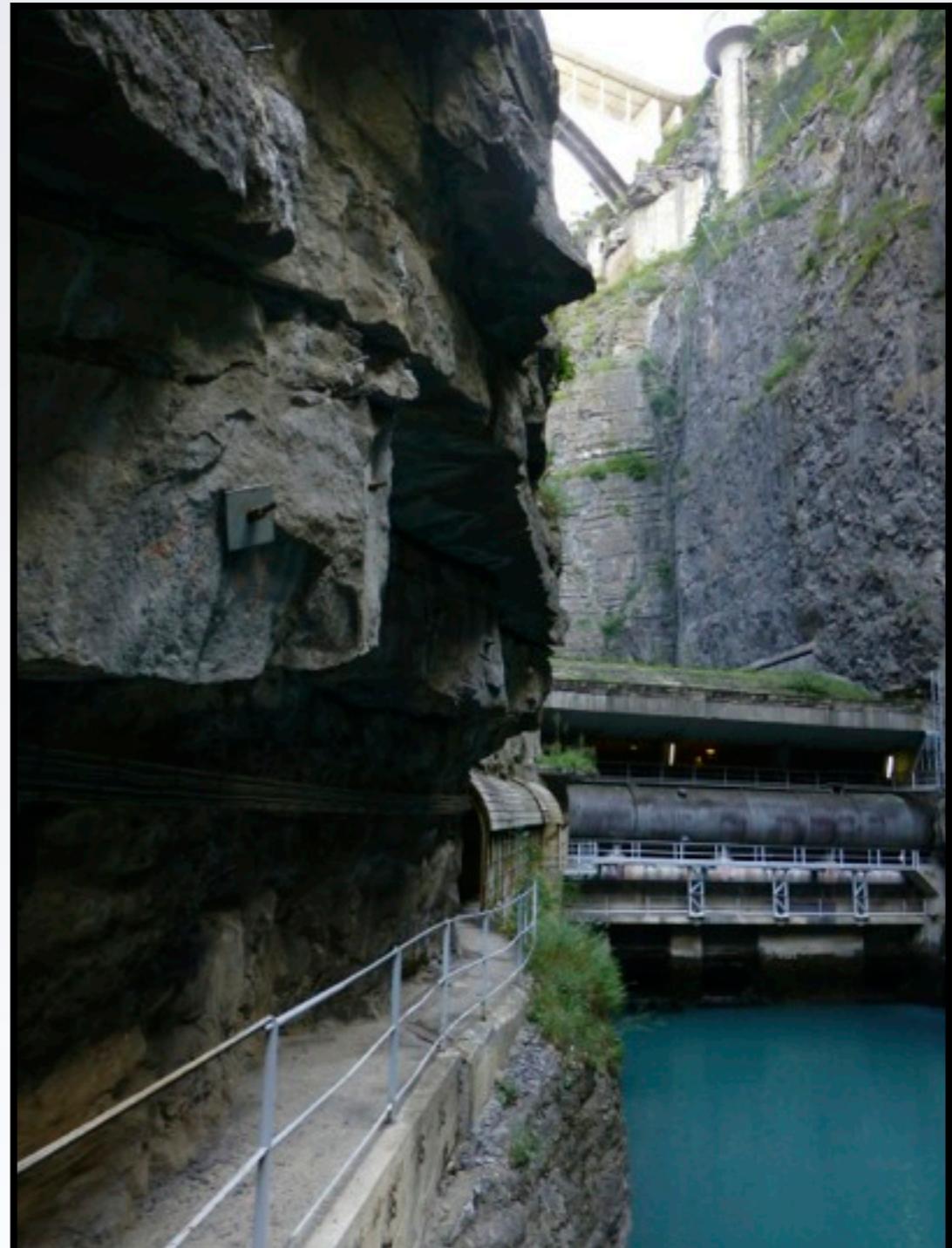
- Precip -Tmax - Tmin

◆ Volume

- In-flow - Tmax - Tmin

◆ Out-flow

- Vol - Tmax - Tmin



Case study description

MLR construction

Predictability

Multiple Linear Regression strategy

Predictors

◆ In-flow

- Precip - T_{max} - T_{min}

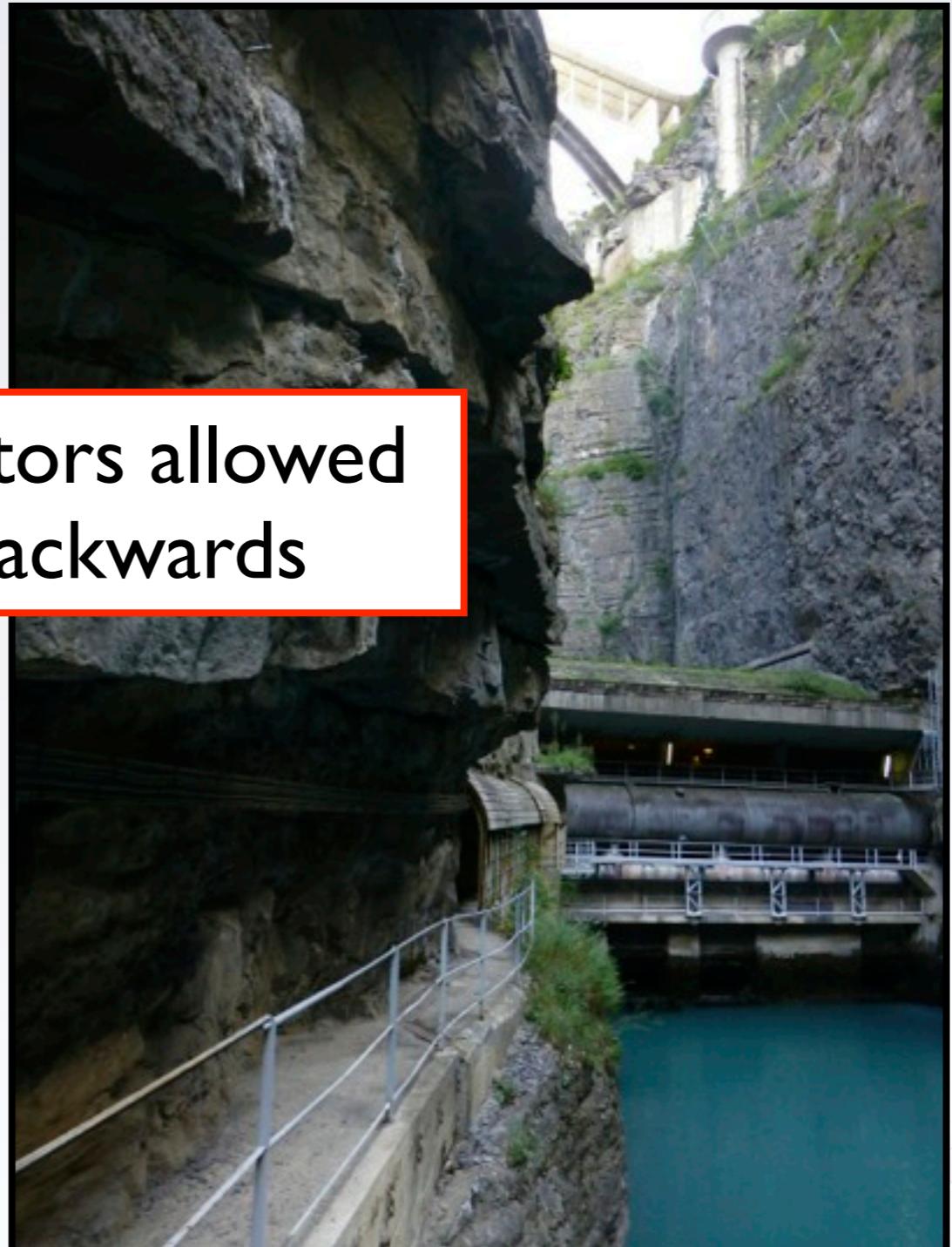
Antecedent predictors allowed
up-to **1 year** backwards

◆ Volume

- In-flow - T_{max} - T_{min}

◆ Out-flow

- Vol - T_{max} - T_{min}



Case study description

MLR construction

Predictability

Multiple Linear Regression strategy

Predictors

◆ In-flow

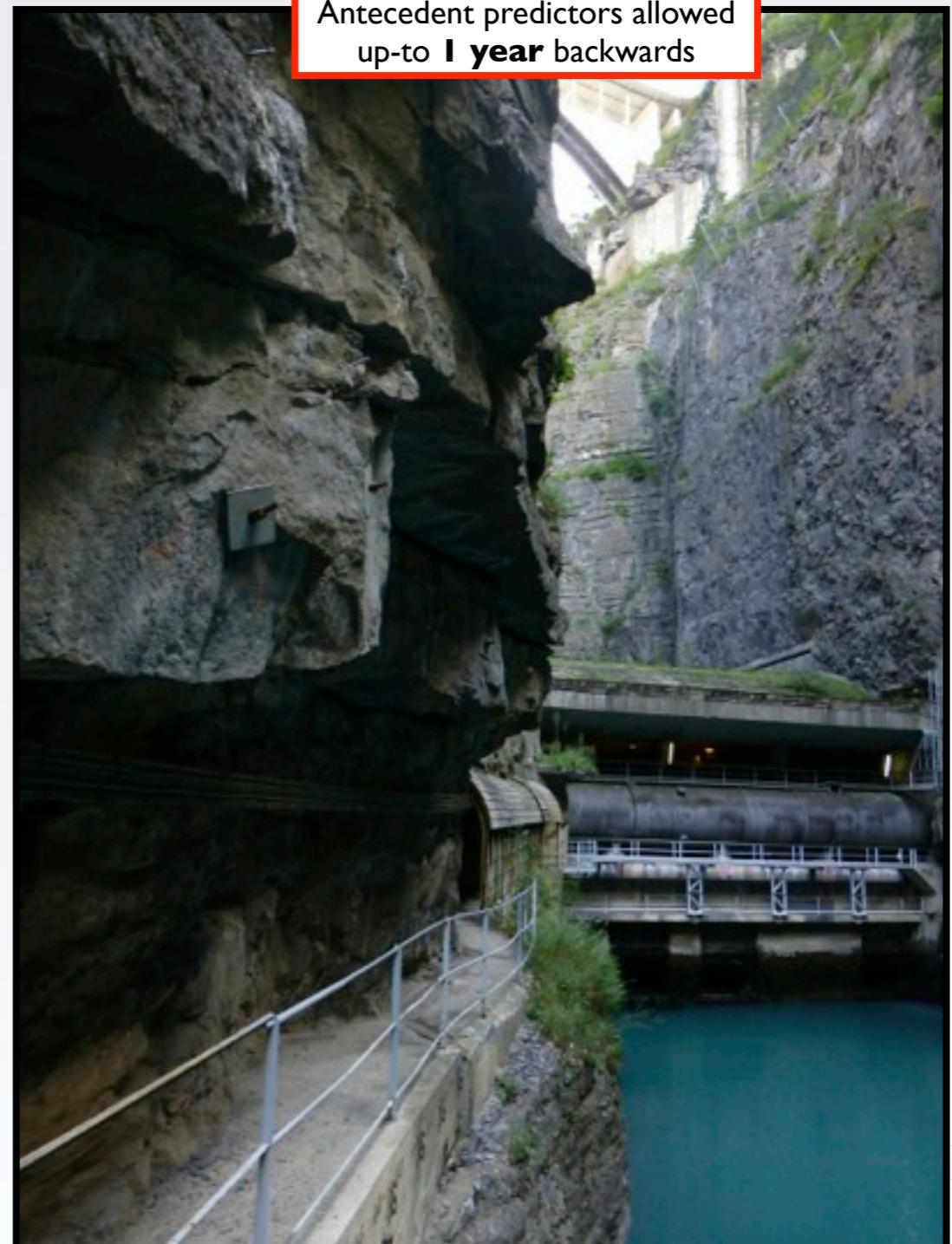
- Precip -Tmax - Tmin

◆ Volume

- In-flow - Tmax - Tmin

◆ Out-flow

- Vol - Tmax - Tmin



Case study description

MLR construction

Predictability

MLR perfect prognosis

Case study description

MLR construction

Predictability

MLR perfect prognosis

In-flow

	Best Predictor Combination	R ²
Jan	$\{rr_{(12-12)}, rr_{(1-1)}, Tx_{(1-1)}, Tn_{(10-1)}\}$	0.81
Feb	$\{rr_{(12-12)}, rr_{(1-2)}, Tx_{(6-9)}, Tn_{(11-1)}\}$	0.77
Mar	$\{rr_{(5-8)}, rr_{(1-2)}, Tx_{(7-7)}\}$	0.41
Apr	$\{rr_{(4-4)}, rr_{(12-3)}, Tx_{(6-6)}, Tn_{(5-7)}, Tn_{(8-10)}\}$	0.66
May	$\{rr_{(4-5)}, rr_{(7-7)}, rr_{(10-10)}\}$	0.73
Jun	$\{rr_{(6-6)}, Tn_{(9-9)}, Tn_{(11-11)}\}$	0.63
Jul	$\{rr_{(6-7)}, Tx_{(10-1)}, Tn_{(11-12)}\}$	0.79
Aug	$\{rr_{(7-8)}, Tx_{(1-7)}, Tn_{(10-1)}\}$	0.51
Sep	$\{rr_{(6-9)}, Tx_{(3-3)}, Tn_{(8-8)}\}$	0.31
Oct	$\{rr_{(9-10)}, Tx_{(11-3)}, Tn_{(1-2)}\}$	0.76
Nov	$\{rr_{(11-11)}, rr_{(7-7)}, Tn_{(4-5)}\}$	0.57
Dec	$\{rr_{(10-12)}, Tn_{(5-7)}, Tn_{(8-9)}\}$	0.55

Case study description

MLR construction

Predictability

MLR perfect prognosis

Volume

	Best Predictor Combination	R ²
Jan	$\{flwin_{(5-1)}, Tx_{(9-9)}, Tn_{(11-11)}, Tn_{(2-4)}\}$	0.76
Feb	$\{flwin_{(5-2)}, Tx_{(10-10)}, Tn_{(10-11)}\}$	0.69
Mar	$\{flwin_{(5-12)}, flwin_{(2-3)}, Tn_{(9-9)}, Tx_{(6-7)}\}$	0.68
Apr	$\{flwin_{(2-4)}, flwin_{(5-12)}, Tx_{(1-1)}, Tx_{(6-9)}, Tn_{(8-9)}\}$	0.66
May	$\{flwin_{(3-4)}, flwin_{(6-12)}, Tn_{(8-9)}, Tx_{(1-1)}\}$	0.60
Jun	$\{flwin_{(3-4)}, flwin_{(11-12)}, Tn_{(9-9)}, Tx_{(12-1)}, Tx_{(9-9)}\}$	0.66
Jul	$\{flwin_{(3-6)}, flwin_{(7-7)}, Tn_{(6-7)}, Tn_{(9-9)}, Tx_{(9-9)}, Tx_{(7-7)}\}$	0.67
Aug	$\{flwin_{(3-5)}, flwin_{(6-8)}, Tn_{(6-7)}, Tx_{(1-1)}, Tx_{(9-9)}\}$	0.79
Sep	$\{flwin_{(3-6)}, flwin_{(7-8)}, Tn_{(6-7)}, Tx_{(7-8)}, Tx_{(9-9)}\}$	0.76
Oct	$\{flwin_{(3-4)}, flwin_{(6-10)}, Tx_{(1-1)}, Tx_{(10-10)}, Tx_{(9-9)}\}$	0.82
Nov	$\{flwin_{(4-4)}, flwin_{(7-10)}, Tn_{(12-4)}, Tx_{(2-2)}\}$	0.84
Dec	$\{flwin_{(5-12)}, Tn_{(6-7)}, Tx_{(2-2)}, Tx_{(9-9)}\}$	0.83

Case study description

MLR construction

Predictability

MLR perfect prognosis

Out-flow

	Best Predictor Combination	R ²
Jan	$\{rr_{(10-1)}, Tx_{(11-1)}\}$	0.60
Feb	$\{rr_{(1-2)}, Tx_{(7-9)}, Tn_{(1-1)}\}$	0.86
Mar	$\{rr_{(7-8)}, vl_{(12-2)}\}$	0.42
Apr	$\{rr_{(10-2)}\}$	0.32
May	$\{rr_{(4-5)}, vl_{(10-4)}\}$	0.55
Jun	$\{vl_{(3-4)}\}$	0.33
Jul	$\{Tx_{(4-4)}, Tn_{(10-11)}, vl_{(3-5)}\}$	0.75
Aug	$\{vl_{(3-5)}, vl_{(6-6)}, vl_{(7-7)}\}$	0.90
Sep	$\{Tx_{(5-6)}, vl_{(7-7)}, vl_{(8-8)}\}$	0.85
Oct	$\{rr_{(1-4)}, rr_{(6-6)}, Tn_{(7-7)}\}$	0.51
Nov	$\{rr_{(1-6)}, Tn_{(4-5)}\}$	0.18
Dec	$\{rr_{(10-12)}, Tx_{(9-11)}, Tn_{(10-12)}\}$	0.18

Case study description

MLR construction

Predictability

Forecasting strategies

Case study description

MLR construction

Predictability

Forecasting strategies

◆ Climatology

Case study description

MLR construction

Predictability

Forecasting strategies

◆ Climatology

◆ Persistence

Case study description

MLR construction

Predictability

Forecasting strategies

- ◆ Climatology
- ◆ Persistence
- ◆ Antecedent + Climatology (MLR)
- ◆ Antecedent + BC S4 (MLR)
- ◆ Antecedent + MOS-analog S4 (MLR)
- ◆ Antecedent + LR S4 (MLR)

Case study description

MLR construction

Predictability

Forecasting strategies

◆ Climatology



Current **operational**
approach

◆ Persistence

◆ Antecedent + Climatology (MLR)

◆ Antecedent + BC S4 (MLR)

◆ Antecedent + MOS-analog S4 (MLR)

◆ Antecedent + LR S4 (MLR)

Case study description

MLR construction

Predictability

Economic Value

Case study description

MLR construction

Predictability

Economic Value

		Ocurrence	
		Yes	No
Preventive Action	Yes	αC	βC
	No	γL	0

$$EV = \frac{TE - TE_{clim}}{TE_{perf} - TE_{clim}}$$

Case study description

MLR construction

Predictability

Economic Value

		Ocurrence
		Yes No
Preventive Action	Yes	αC
	No	γL
		βC
		0

$$R = C/L$$

$$EV = \frac{H p_c R + F (1 - p_c) R + (1 - H)p_c - \min \{R, p_c\}}{p_c R - \min \{R, p_c\}}$$

$$EV = \frac{TE - TE_{clim}}{TE_{perf} - TE_{clim}}$$

Case study description

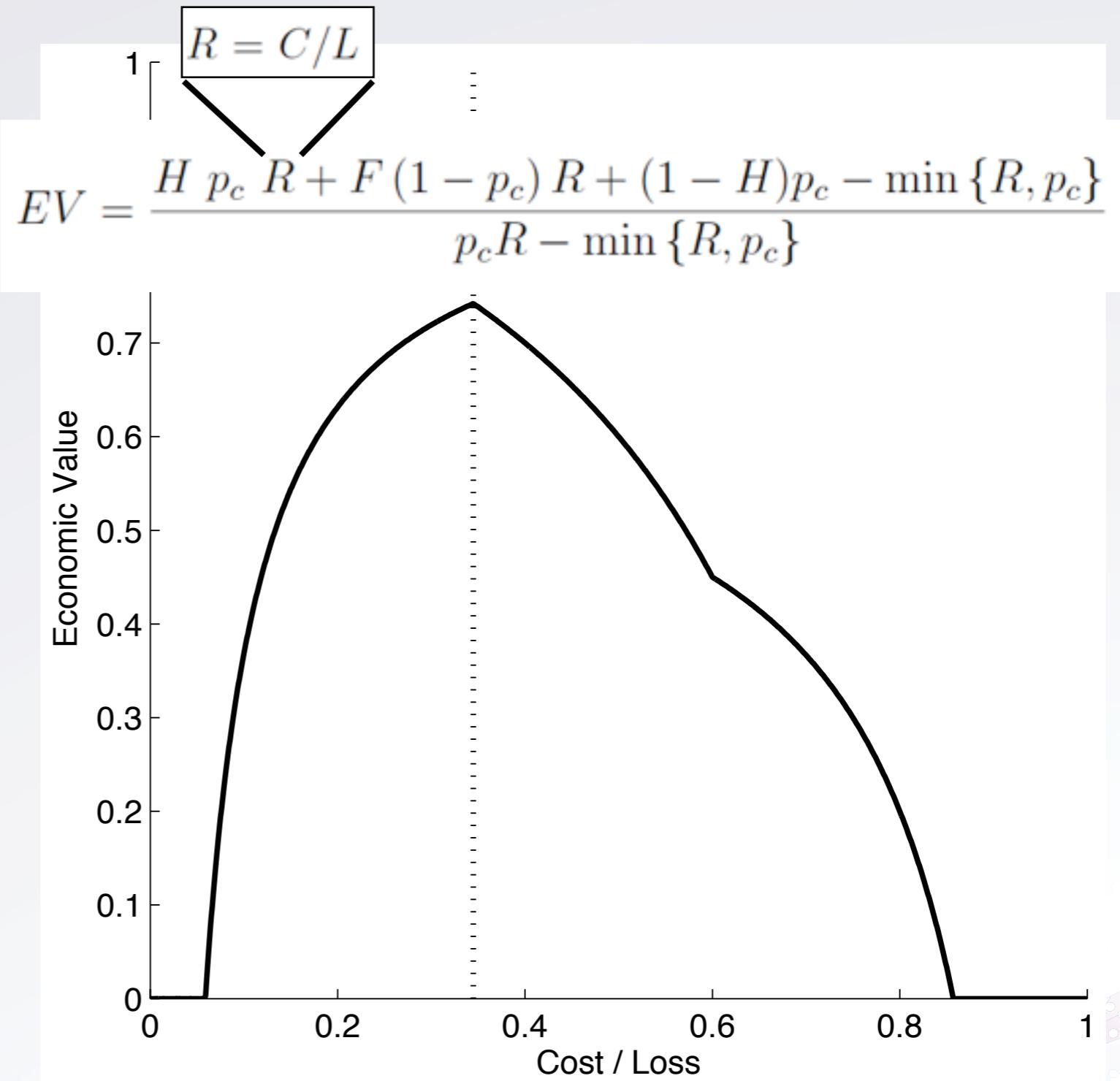
MLR construction

Predictability

Economic Value

		Ocurrence	
		Yes	No
Preventive Action	Yes	αC	βC
	No	γL	0

$$EV = \frac{TE - TE_{clim}}{TE_{perf} - TE_{clim}}$$



Case study description

MLR construction

Predictability

Seasonal forecast (volume)

Case study description

MLR construction

Predictability

10

Seasonal forecast (volume)

Economic Value (July - Lead 5 - Humid conditions)

Case study description

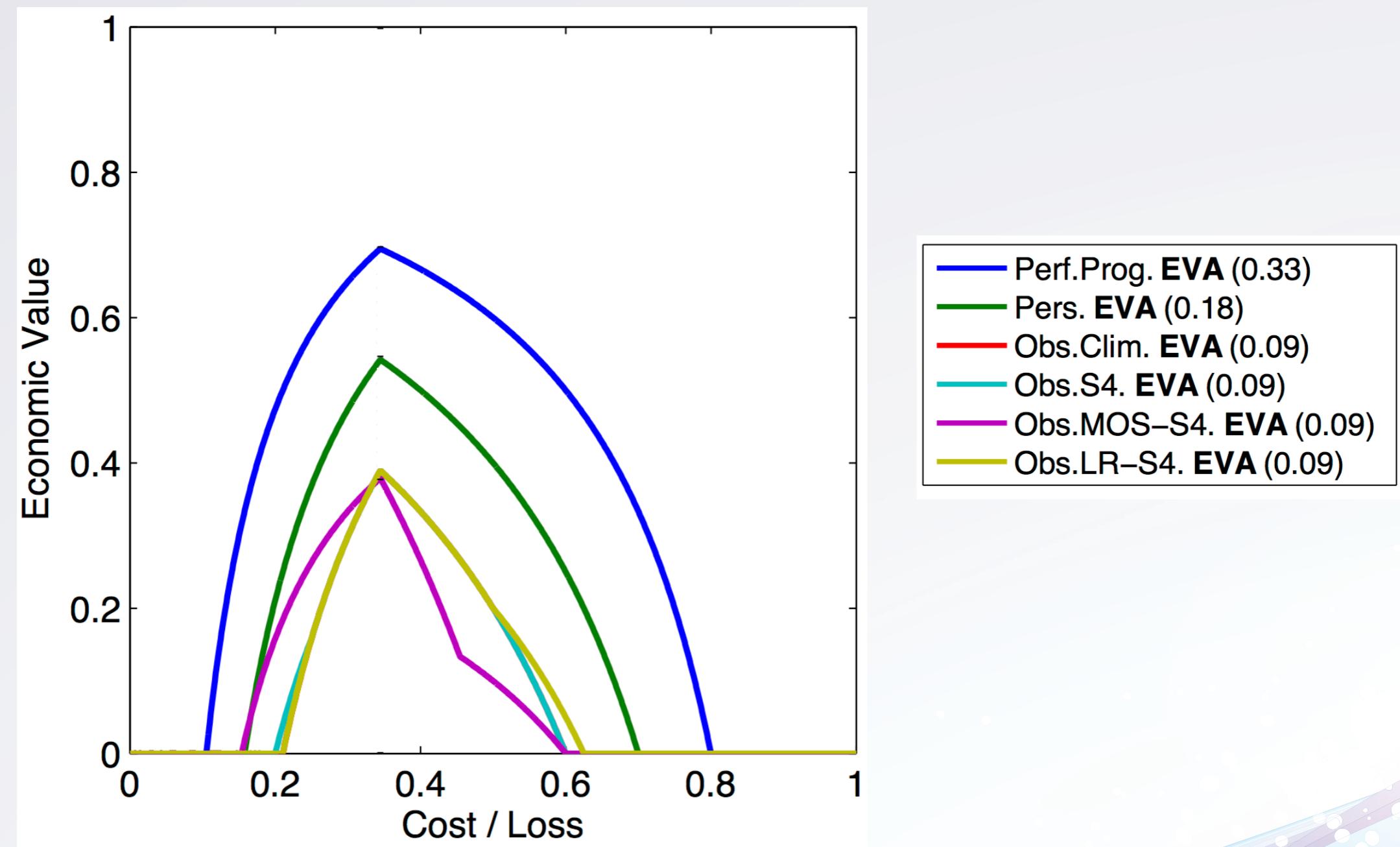
MLR construction

Predictability

10

Seasonal forecast (volume)

Economic Value (July - Lead 5 - Humid conditions)



Case study description

MLR construction

Predictability

10

Seasonal forecast (volume)

Case study description

MLR construction

Predictability

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Seasonal forecast (volume)

EV - 12 month - 7 lead

Case study description

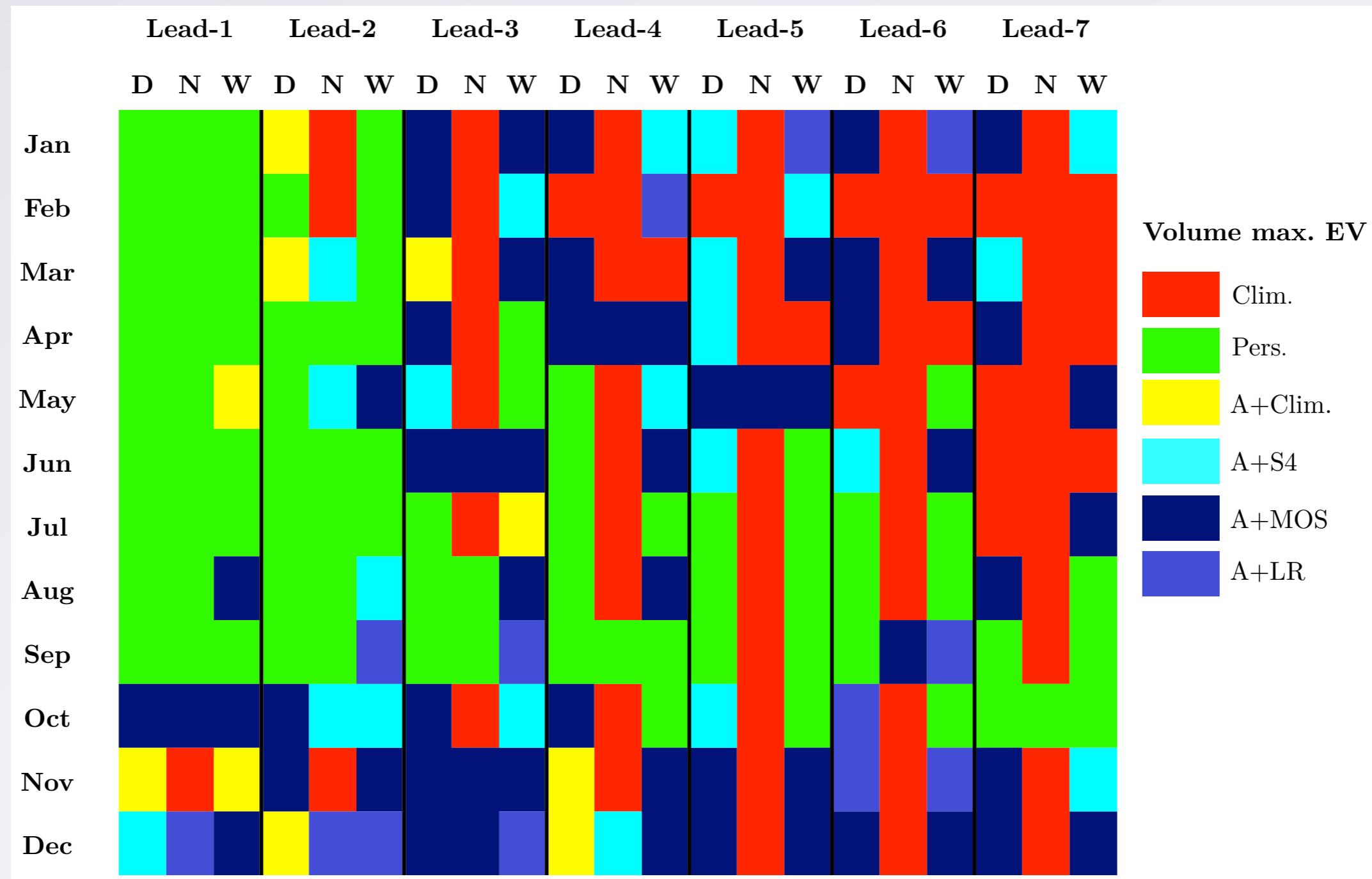
MLR construction

Predictability

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Seasonal forecast (volume)

EV - 12 month - 7 lead



Case study description

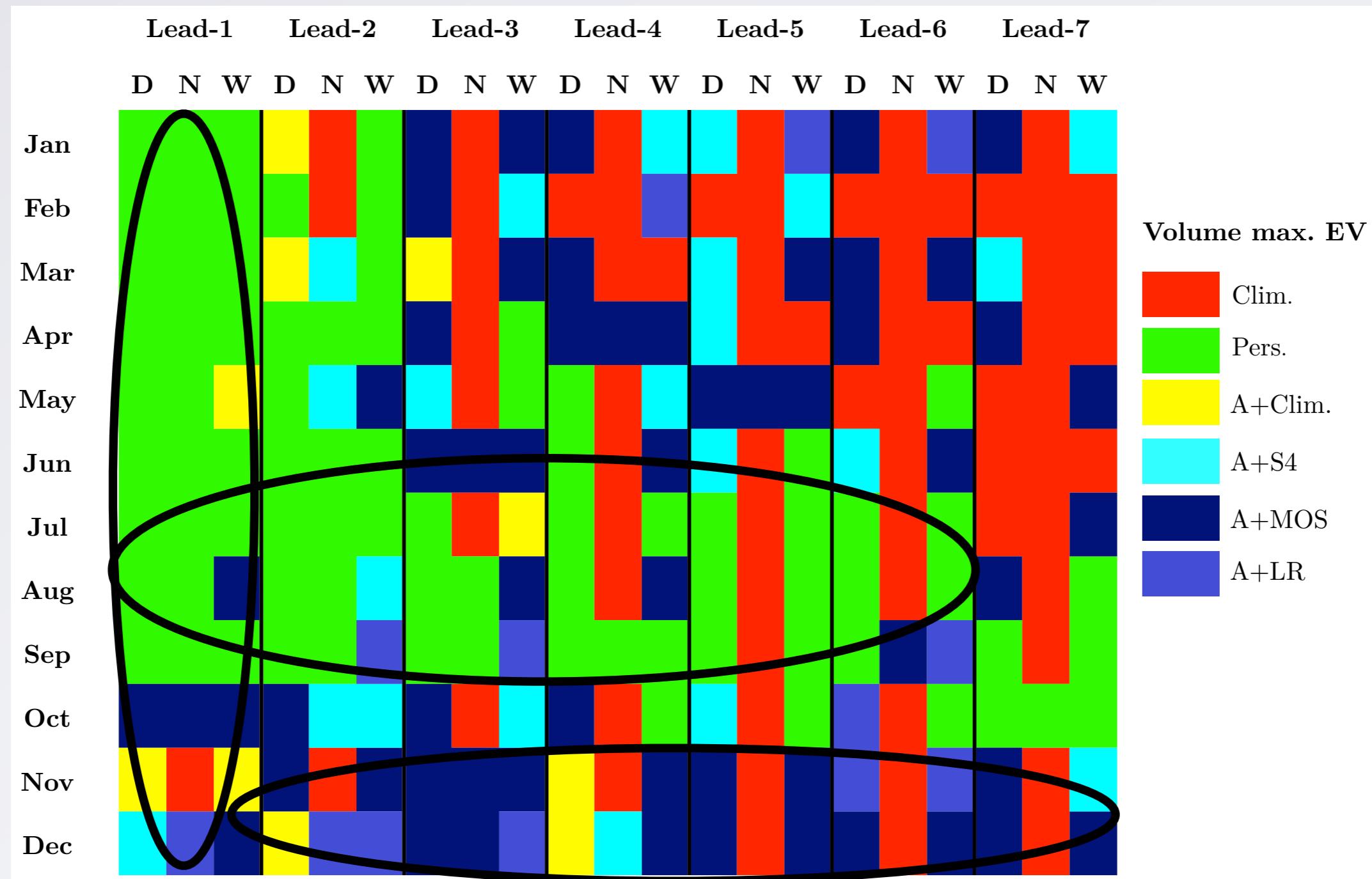
MLR construction

Predictability

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Seasonal forecast (volume)

EV - 12 month - 7 lead

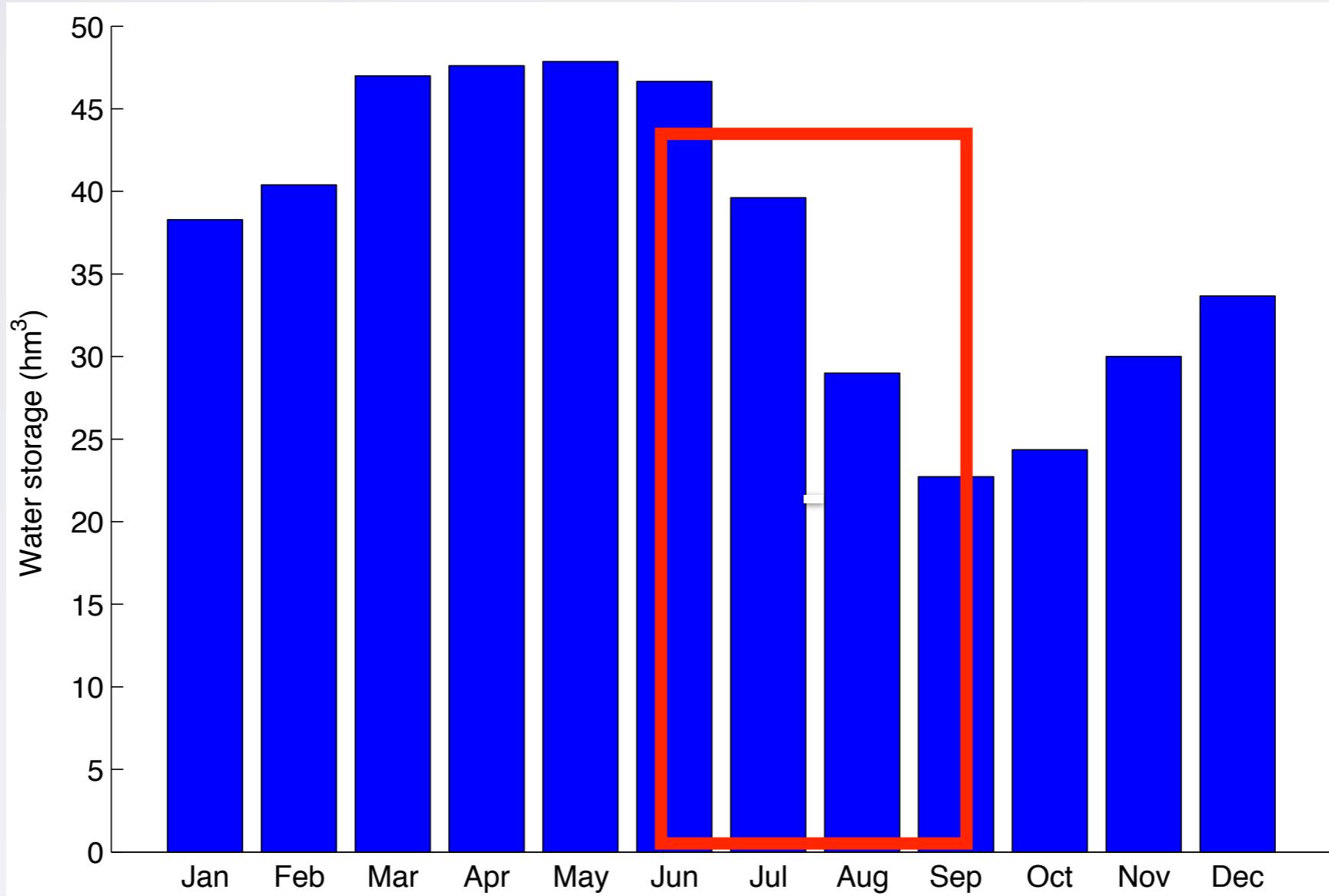


Case study description

MLR construction

Predictability

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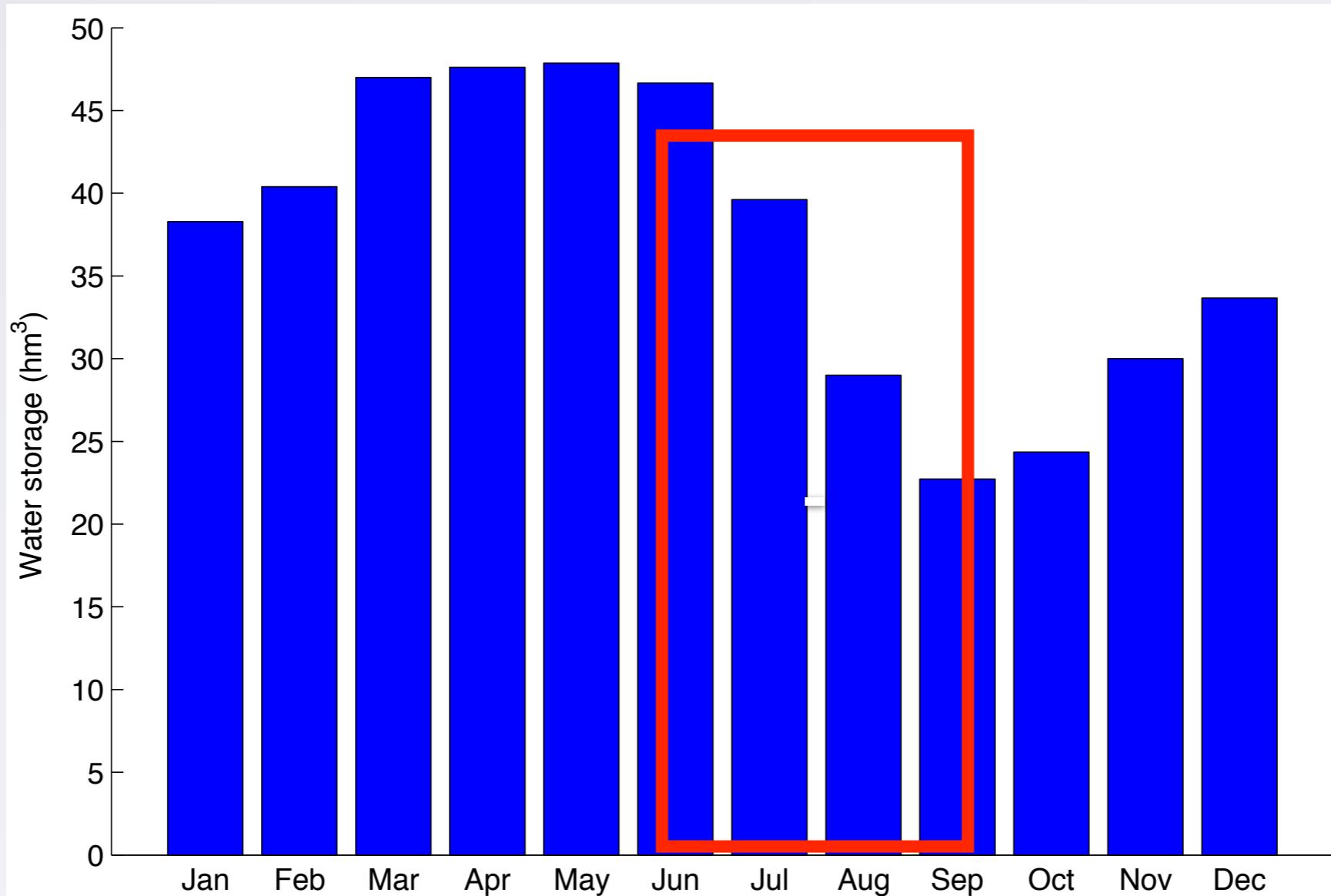
Case study description

MLR construction

Predictability

I2

Climatology: volume



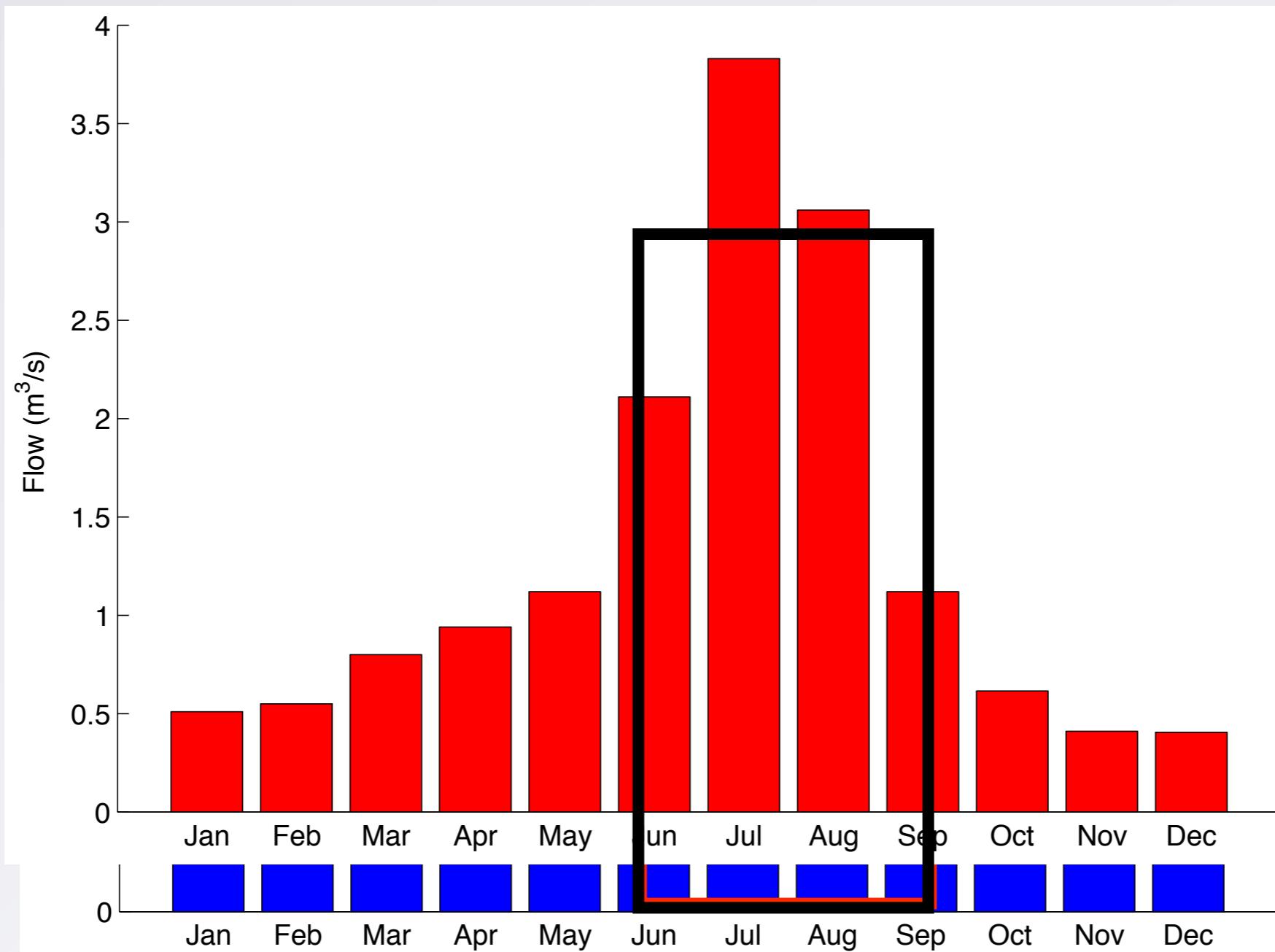
Case study description

MLR construction

Predictability

I2

Climatology: out-flow



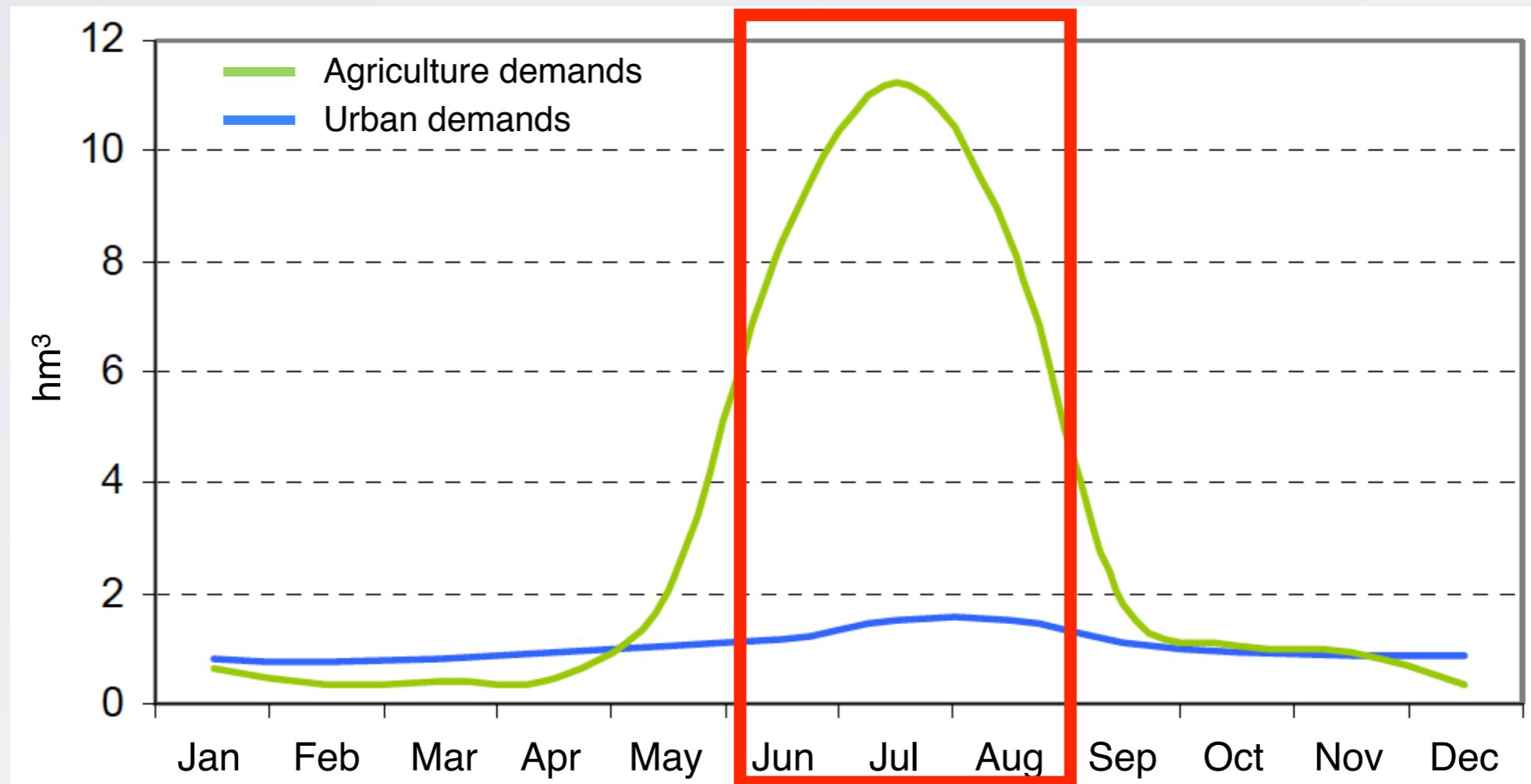
Case study description

MLR construction

Predictability

I2

Climatology: demands



Case study description

MLR construction

Predictability

I2

Synthesis:

Seasonal forecast of water resources

Synthesis:

Seasonal forecast of water resources

♦ In-flow

- Predictability beyond climatology generally restricted to **lead one**

Synthesis:

Seasonal forecast of water resources

♦ In-flow

- Predictability beyond climatology generally restricted to **lead one**

♦ Volume

- Predictability beyond climatology up to **lead four** in all months but February
- Positive Economic Value up to **lead seven** from August to January

Synthesis:

Seasonal forecast of water resources

♦ In-flow

- Predictability beyond climatology generally restricted to **lead one**

♦ Volume

- Predictability beyond climatology up to **lead four** in all months but February
- Positive Economic Value up to **lead seven** from August to January

♦ Out-flow

- Predictability beyond climatology at **lead one** for almost all months
- From August to October enhanced predictabilities up to **lead seven**

Take home messages:

Seasonal forecast of water resources

Take home messages:

Seasonal forecast of water resources

- a) Generally, all the three variables can be **successfully** modelled with **MLR** in perfect prognosis conditions. **Volume** is the best modelled variable, followed by **in-flow** and **out-flow**.
- b) **Summer** seasonal forecasts with skill beyond climatology can be issued from a minimum of **four months** in advance for volume and out-flow variables.

Thank you for your attention!

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DAM



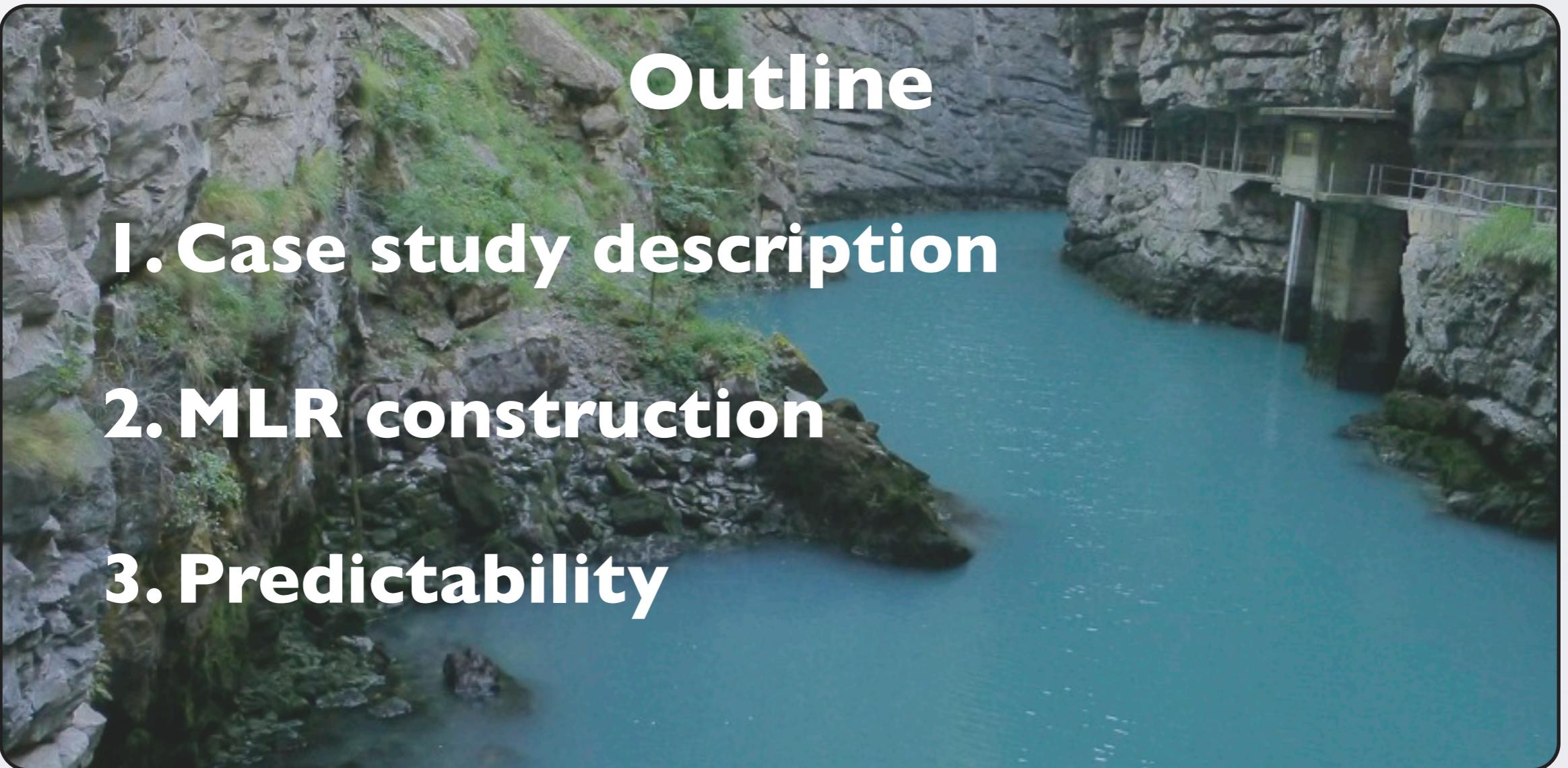
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Outline

- 1. Case study description**
- 2. MLR construction**
- 3. Predictability**