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Petro-physical Characterization of Sedimentary Environments Using Collaborative New Method in Wireline logs Interpretation

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Content

- Background and object
- Reservoir evaluation and Field develop plan
- Preliminary results of QI Machine Learning
- Concluding Remarks



Purpose of the project

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General Geology

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(After Genik, 1993)

General Geology



(IHS Energy Group, 2001)



Field Development Plan Optimization



Lithology Characterization and sedimentary facies



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Core-Log lithofacies integration



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Reservoir Sand

Log: GR is around 90 gAPI, lower density, lower neutron, high Sonic velocity, higher resistivity; porosity>10%; shale content<40%.

Core: Fining up medium grained, minor coarse grained, occasionally very coarse grained, moderately sorted, clear cross bedding, good porosity, heavy oil odor.

Tight Sand

Log: GR>100 gAPI, higher density, higher neutron, lower resistivity; porosity<10%; shale content<60%.

Core: Siltstone with very fine sand stripes in part, common argillaceous matrix, poor porosity, no shows.

Calcareous Sand

Log: Lower GR, high density, density>2.5, high resistivity; porosity<10%; calcite content>10%, shale content<60%.

Core: Medium grained, trace very coarse grained, sub-angular to sub-rounded, moderately sorted, trace kaolinitic cement, calcareous cement.

Shale

Log: Higher GR, wide neutron-density cross, low Sonic velocity, lower resistivity; shale content>60%.



В

С

	Clay Volume (Vcl)	Porosity (Ф)	Density (ρ)	Calcite Volume (Vclc)	Lith Classification
	Vcl ≤ 0.4	Φ≥0.1	/	/	Reservoir Sand
		Φ < 0.1	/	/	Tight Sand
	0.4 < Vcl < 0.6	Φ≥0.03	/	/	
		Φ < 0.03	ρ > 2.5	Vclc > 0.02	
			/	/	Shale
	Vcl ≥ 0.6	/	/	/	



Structure Modeling

- :1090 20 - :2010 20 - :2050 20 - :2150 20 - :2150 20 - :2150 20 - :2250 20



Fault Framework





Model Construction

Model Construction



Structure gridding



Well data upscaling

Property Modeling









Upscale



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Property Modeling



Decision Tree

- Machine Learning Module
 in Petrel
- Case Study
 - Porosity curve prediction
 - Shear wave curve prediction
 - porosity cube prediction







Porosity Log prediction

- Conventional logging curves of 8 wells are used for training
 - Gamma Ray (GR), Deep Resistivity (RT), Medium Resistivity (RLA3), Invaded Formation Resistivity (RXOZ), Density (RHOZ), Neutron (TNPH), Sonic (DT), photoelectric effect (PEFZ)
- Target: Effective Porosity (PHIE)
- Results curves are mostly consistent with manual interpretation





Shear wave prediction

- Conventional logging curves are used for training
 - Gamma Ray (GR), Deep Resistivity (RT), \succ Density (RHOZ), Neutron (TNPH), Sonic (DT)
- Target: Shear Slowness (DTSH)
- The predicted curve matches well with actual logged curve.
- discrimination of reservoir fluid

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Porosity Cube prediction

<u>Training model :</u>

- Prediction:
- Input: P-impedance, Density and Vp/Vs cube.
- Output: Porosity Cube
- <u>Result</u>
 - Match well with actual logging curve



Data integration & collaboration

Challenges:

- Data storage in PC or external disk
- Time-consuming for data searching, preparation and transfer.
- Human errors in data import (CRS, datum, units)
- Duplicated Data
- Research result display (color bar, template, well-section, workflow)



High Performance Computing Storage



Workflow

- Data index, search, access
- Data share, collaboration
- Data filtering and management
- Resource evaluation workflow, reservoir simulation workflow
- Play Chance Mapping





Concluding Remark

- The sedimentary facies and reservoir characterization was analyzed by core-Log lithofacies integrating.
- The resource evaluation and field development plan were completed based on the geological model from deterministic inversion.
- Case study of QL Machine Learning suggested the predicted porosity and shear wave match well with manual interpretation and actual log data, respectively.
- The Petrel E&P platform and Studio improved efficiency in data management and collaboration among researchers.



Thank you for listening

