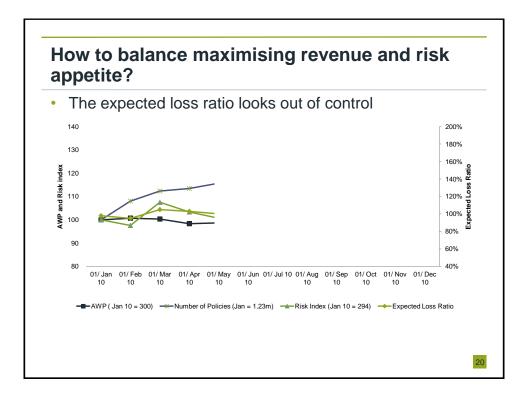
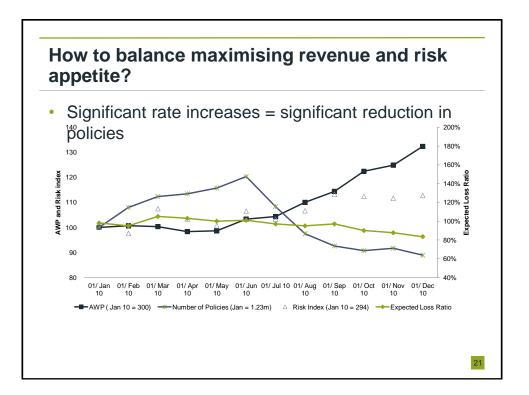
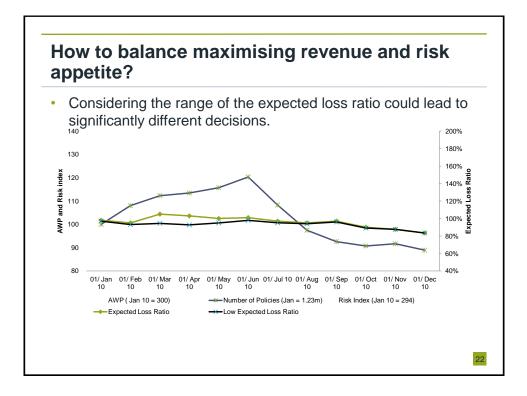
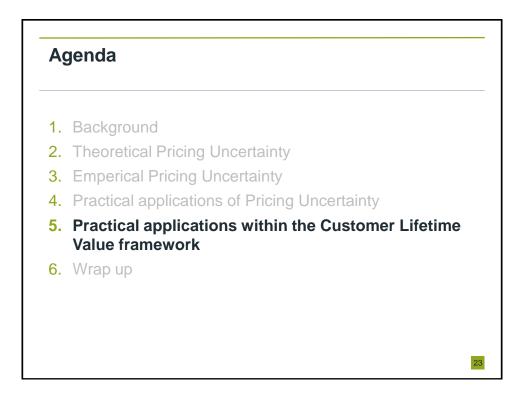


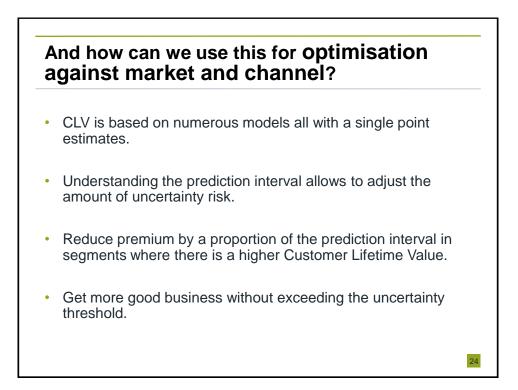
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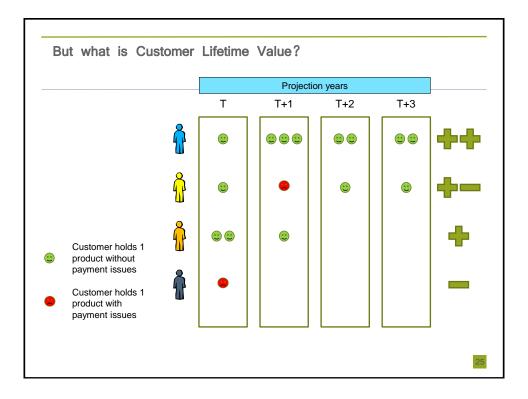


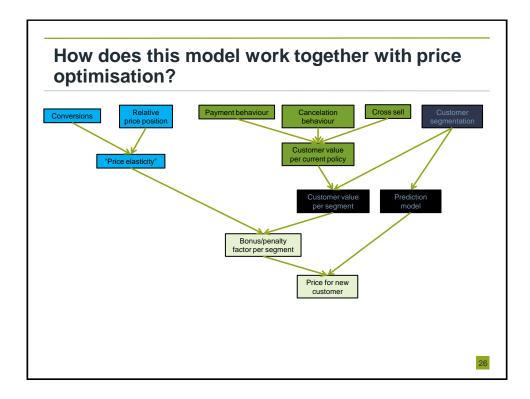












The Customer Lifetime Value model is very intuitive Using the company's client data to model client behaviour: • A higher lapse rate results in a lower customer value.

- A higher probability of non-payment results in a lower customer value.
- A higher rate of cross selling results in a higher customer value.
- A higher product yield results in a higher customer value.

Assign a "pound value" to predicted client behaviour

- Administrative costs (e.g. dept collectors)
- Fixed cost per policy
- Profit per policy

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