

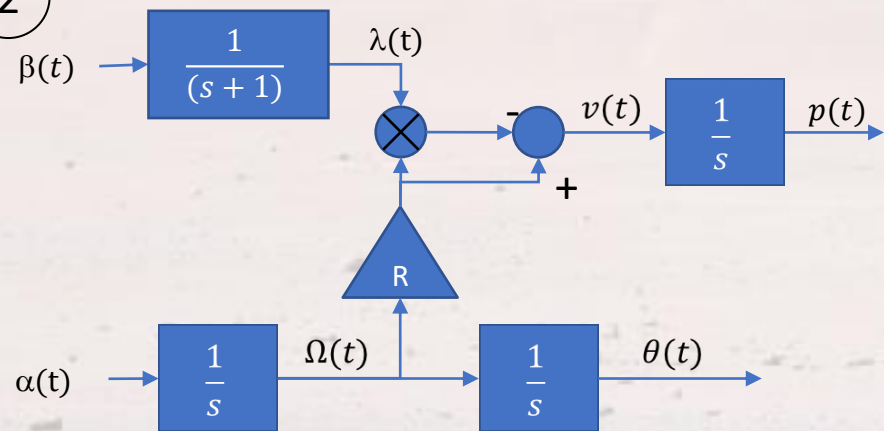
Power-aware Fusion of Visual and Wheel Odometry for Mobile Platform

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1 Given a typical localisation mode on a Mars rover...

	Measurement Frequency	Energy Cost	Localisation Accuracy	Requires to stop?	Prone to slip?
Wheel Odometry (WO)	High	Low	Low	No	Yes
Visual Odometry (VO)	Low	High	High	Yes	No

2 ...let's define a model for an adaptive filter...



State vector $x(t) = [p(t) \theta(t) \Omega(t) \lambda(t) m(t)]^T$

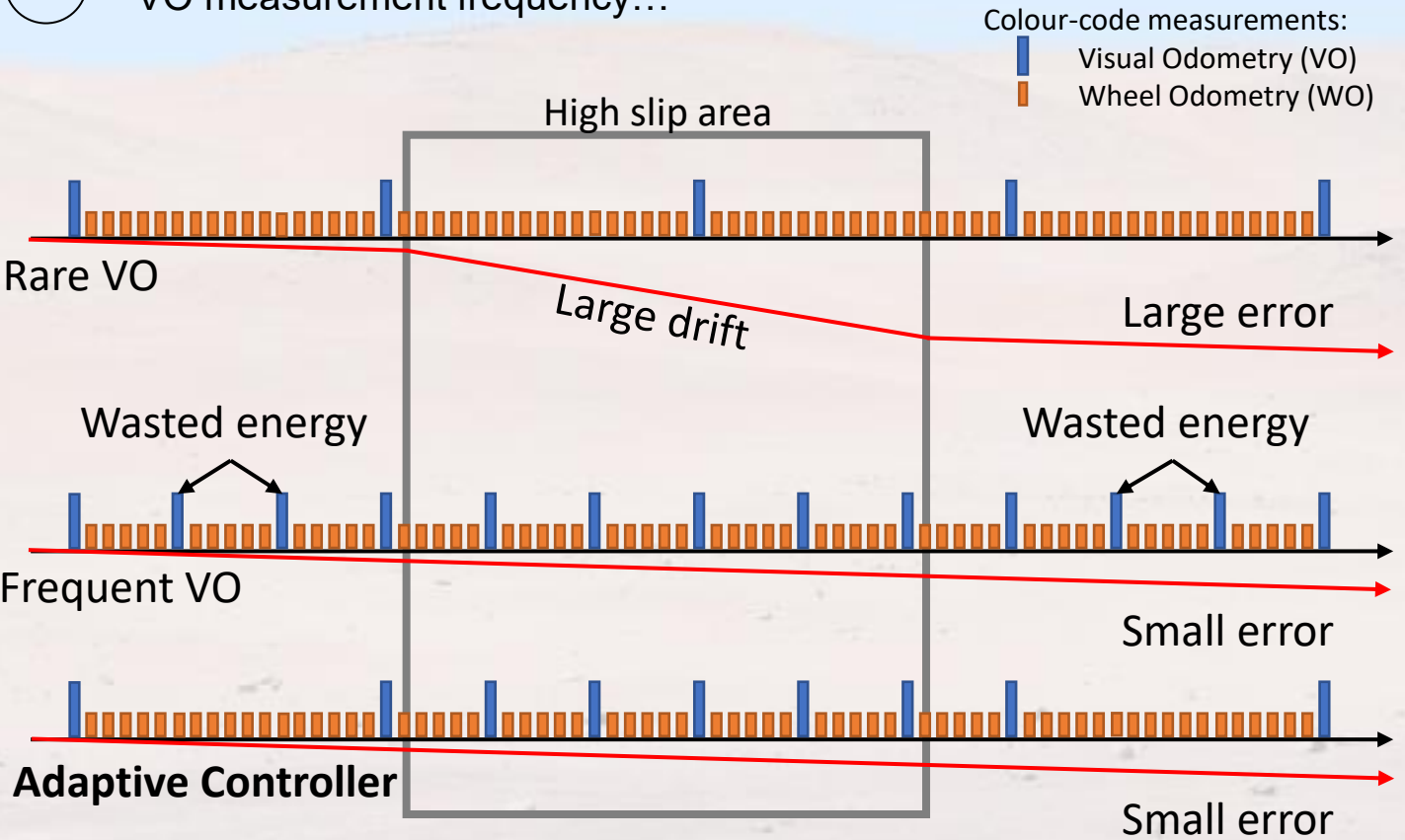
Augmented state variable $m(t^+) = \begin{cases} p(t) & \text{if VO measurement at time } t \\ m(t) & \text{otherwise} \end{cases}$

WO measurement $y_{WO}(t) = \theta(t) + \varepsilon_{WO}(t)$

VO measurement $y_{VO}(t) = p(t) - m(t) + \varepsilon_{VO}(t)$

Slip definition $\lambda(t) = 1 - v(t) / [R * \Omega(t)]$

3 ...which would use slip (λ) estimate to control VO measurement frequency...



- 4 ...to operate more efficiently:
- Less frequent VO on no slip ground = less energy used (fewer image acquisitions, fewer stops, fewer executions of complex algorithm)
 - More frequent VO in slip areas = maintain low position estimation error